# Package 'Lock5Data'

July 21, 2025

Title Datasets for ``Statistics: UnLocking the Power of Data"

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# Description

Datasets for first, second, and third editions of Statistics: Unlocking the Power of Data by Lock^5

6 ACS

# **Details**

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ACS

American Community Survey

# **Description**

Data from a sample of individuals in the American Community Survey

#### **Format**

A data frame with 2000 observations on the following 9 variables.

Sex 0=female and 1=male

Age Age (years)

Married 0=not married and 1=married

Income Wages and salary for the past 12 months (in \$1,000's)

HoursWk Hours of work per week

Race asian, black, other, or white

USCitizen 1=citizen and 0=noncitizen

HealthInsurance 1=have health insurance and 0=no health insurance

Language 1=English spoken at home and 0=other

# **Details**

The American Community Survey, administered by the US Census Bureau, is given every year to a random sample of about 3.5 million households (about 3% of all US households). Data on a random sample of 1% of all US residents are made public (after ensuring anonymity), and we have selected a random sub-sample of n = 2000 from the 2017 data for this dataset.

<sup>\*\*</sup> Updated for 3e (earlier version is ACS2010). \*\*

ACS2010 7

#### Source

The full public dataset can be downloaded at https://www.census.gov/programs-surveys/acs/microdata.html, and the full list of variables are at https://www.census.gov/programs-surveys/acs/microdata/documentation.html

ACS2010

American Community Survey - 2010

#### **Description**

Data from a sample of individuals in the 2010 American Community Survey

#### **Format**

A dataset with 1000 observations on the following 9 variables.

Sex 0=female and 1=male

Age (years)

Married 0=not married and 1=married

Income Wages and salary for the past 12 months (in \$1,000's)

Hours Wk Hours of work per week

Race asian, black, white, or other

USCitizen 1=citizen and 0=noncitizen

HealthInsurance 1=have health insurance and 0=no health insurance

Language 1=native English speaker and 0=other

#### **Details**

The American Community Survey, administered by the US Census Bureau, is given every year to a random sample of about 3.5 million households (about 3% of all US households). Data on a random sample of 1% of all US residents are made public (after ensuring anonymity), and we have selected a random sub-sample of n = 1000 from the 2010 data for this dataset.

\*\* From 2e - dataset has been updated for 3e \*\*

### Source

The full public dataset can be downloaded at

http://www.census.gov/acs/www/data documentation/pums data/,

and the full list of variables are at

http://www.census.gov/acs/www/Downloads/data documentation/pums/DataDict/PUMSDataDict10.pdf.

8 AllCountries

AllCountries

All Countries

# **Description**

Data on the countries of the world

#### **Format**

A data frame with 217 observations on the following 26 variables.

Country Country name

Code Three-letter code for country

LandArea Size in 1000 sq. km.

Population Population in millions

Density Number of people per square kilometer

GDP Gross Domestic Product (in \$US) per capita

Rural Percentage of population living in rural areas

CO2 CO2 emissions (metric tons per capita)

PumpPrice Price for a liter of gasoline (\$US)

Military Percentage of government expenditures directed toward the military

Health Percentage of government expenditures directed towards healthcare

ArmedForces Number of active duty military personnel (in 1,000's)

Internet Percentage of the population with access to the internet

Cell Cell phone subscriptions (per 100 people)

HIV Percentage of the population with HIV

Hunger Percent of the population considered undernourished

Diabetes Percent of the population diagnosed with diabetes

BirthRate Births per 1000 people

DeathRate Deaths per 1000 people

ElderlyPop Percentage of the population at least 65 years old

LifeExpectancy Average life expectancy (years)

FemaleLabor Percent of females 15 - 64 in the labor force

Unemployment Percent of labor force unemployed

Energy Kilotons of oil equivalent

Electricity Electric power consumption (kWh per capita)

Developed Categories for kilowatt hours per capita, 1= under 2500, 2=2500 to 5000, 3=over 5000

AllCountries1e 9

#### **Details**

Data for each variable were collected for 2018 (or most recently available year). Within a variable all country measurements are from the same year, but the year may vary between different variables depending on availability.

\*\* This dataset is updated from an earlier versions (now Allcountries1e and AllCountries2e) \*\*

#### Source

The data were gathered online from https://data.worldbank.org/. Accessed June 2019.

AllCountries1e

AllCountries - 1e

# **Description**

Data on the countries of the world

#### **Format**

A dataset with 213 observations on the following 18 variables.

```
Country
                 Name of the country
           Code
                  Three letter country code
      LandArea
                  Size in sq. kilometers
    Population
                 Population in millions
                  Energy usage (kilotons of oil)
         Energy
                 Percentage of population living in rural areas
          Rural
      Military
                  Percentage of government expenditures directed toward the military
         Health Percentage of government expenditures directed towards healthcare
            HIV Percentage of the population with HIV
      Internet Percentage of the population with access to the internet
     Developed
                 Categories for kilowatt hours per capita, 1= under 2500, 2=2500 to 5000, 3=over 5000
     BirthRate Births per 1000 people
    ElderlyPop Percentage of the population at least 65 years old
LifeExpectancy Average life expectancy (years)
                  CO2 emissions (metric tons per capita)
            GDP
                  Gross Domestic Product (per capita)
           Cell phone subscriptions (per 100 people)
                 Electric power consumption (kWh per capita)
   Electricity
```

# **Details**

Most data from 2008 to avoid many missing values in more recent years.

\*\* From 1e - dataset has been updated for 2e \*\*

### Source

Data collected from the World Bank website, worldbank.org.

10 AllCountries2e

### **Description**

Data on the countries of the world

#### **Format**

A dataset with 215 observations on the following 25 variables.

```
Country
                   Name of the country
                   Size in 1000 sq. kilometers
       LandArea
    Population
                   Population in millions
        Density
                   Number of people per square kilometer
                   Gross Domestic Product (in $US) per capita
            GDP
                   Percentage of population living in rural areas
          Rural
            C02
                  CO2 emissions (metric tons per capita)
     PumpPrice
                   Price for a liter of gasoline ($US)
      Military
                   Percentage of government expenditures directed toward the military
                   Percentage of government expenditures directed towards healthcare
         Health
   ArmedForces
                  Number of active duty military personnel (in 1,000's)
       Internet
                  Percentage of the population with access to the internet
           Cell
                  Cell phone subscriptions (per 100 people)
            HIV
                  Percentage of the population with HIV
         Hunger Percent of the population considered undernourished
      Diabetes Percent of the population diagnosed with diabetes
     BirthRate
                  Births per 1000 people
     DeathRate Deaths per 1000 people
    ElderlyPop
                  Percentage of the population at least 65 years old
LifeExpectancy
                  Average life expectancy (years)
                   Percent of females 15 - 64 in the labor force
   FemaleLabor
                  Percent of labor force unemployed
  Unemployment
         Energy
                  Energy usage (kilotons of oil equivalent)
   Electricity
                   Electric power consumption (kWh per capita)
                  Categories for kilowatt hours per capita, 1= under 2500, 2=2500 to 5000, 3=over 5000
     Developed
```

#### **Details**

Data for each variable were collected for years between 2012 and 2014. Within a variable all country measurements are from the same year, but the year may vary between different variables depending on availability.

#### Source

Data collected from the World Bank website, worldbank.org.

<sup>\*\*</sup> From 2e - dataset has been updated for 3e \*\*

APMultipleChoice 11

APMultipleChoice

AP Multiple Choice

# Description

Correct responses on Advanced Placement multiple choice exams

#### **Format**

A dataset with 400 observations on the following variable.

Answer Correct response: A, B, C, D, or E

# **Details**

Correct responses from multiple choice sections for a sample of released Advanced Placement exams

#### Source

Sample exams from several disciplines at http://apcentral.collegeboard.com

April14Temps

April 14th Temperatures

# **Description**

Temperatures in Des Moines, IA and San Francisco, CA on April 14th

# **Format**

A data frame with 25 observations on the following 3 variables.

Year 1995 to 2019

DesMoines Temperature in Des Moines (degrees F)

SanFrancisco Temperature in San Francisco (degrees F)

# **Details**

Average temperature for the day of April 14th in each of 25 years from 1995-2019

\*\* Data set updated for 3e (earlier versions are now April14Temps1e and April14Temps2e) \*\*

# Source

The University of Dayton Average Daily Temperature Archive at https://academic.udayton.edu/kissock/http/Weather/citylistUS.htm

12 April14Temps2e

April14Temps1e

April 14th Temperatures -1e

# **Description**

Temperatures in Des Moines, IA and San Francisco, CA on April 14th

#### **Format**

A dataset with 16 observations on the following 3 variables.

Year 1995-2010
DesMoines Temperature in Des Moines (degrees F)
SanFrancisco Temperature in San Francisco (degrees F)

#### **Details**

Average temperature for the day of April 14th in each of 16 years from 1995-2010 \*\* From 1e - dataset has been updated for 2e \*\*

#### **Source**

The University of Dayton Average Daily Temperature Archive at <a href="http://academic.udayton.edu/kissock/http/Weather/citylistUS.htm">http://academic.udayton.edu/kissock/http/Weather/citylistUS.htm</a>

April14Temps2e

April 14th Temperatures - 2e

# **Description**

Temperatures in Des Moines, IA and San Francisco, CA on April 14th

### **Format**

A dataset with 21 observations on the following 3 variables.

Year 1995 to 2015
DesMoines Temperature in Des Moines (degrees F)
SanFrancisco Temperature in San Francisco (degrees F)

#### **Details**

Average temperature for the day of April 14th in each of 21 years from 1995-2015 \*\* From 2e - dataset has been updated for 3e \*\*

# Source

The University of Dayton Average Daily Temperature Archive at <a href="http://academic.udayton.edu/kissock/http/Weather/citylistUS.htm">http://academic.udayton.edu/kissock/http/Weather/citylistUS.htm</a>

BaseballHits1e 13

BaseballHits1e

Baseball Hits

# **Description**

Number of hits, wins, and other stats for MLB teams - 2011

# **Format**

A dataset with 30 observations on the following 14 variables.

Team Name of baseball team

Either American AL or National NL League League

Wins Number of wins for the season

Number of runs scored Runs

Hits Number of hits

Doubles Number of doubles

Triples Number of triples

HomeRuns Number of home runs RBI Number of runs batted in

StolenBases Number of stolen bases

Number of times caught stealing CaughtStealing

> Number of walks Walks

Strikeouts Number of strikeouts

BattingAvg Team batting average

# **Details**

Data from the 2010 Major League Baseball regular season.

\*\* From 1e - dataset has been updated for 2e \*\*

# **Source**

http://www.baseball-reference.com/leagues/MLB/2011-standard-batting.shtml

BaseballHits2014

Baseball Hits - 2014

# **Description**

Number of hits, wins, and other stats for MLB teams - 2014

14 BaseballHits2019

# **Format**

A dataset with 30 observations on the following 14 variables.

Team Name of baseball team (3-character code)

League Either AL or NL

Wins Number of wins for the season

Runs Number of runs scored

Hits Number of hits
Doubles Number of doubles
Triples Number of triples
HomeRuns Number of home runs

RBI Number of runs batted in

StolenBases Number of stolen bases

CaughtStealing Number of times caught stealing

Walks Number of walks
Strikeouts Number of strikeouts

BattingAvg Team batting average

#### **Details**

Data from the 2014 Major League Baseball regular season.

\*\* From 2e - dataset has been updated for 3e \*\*

# Source

http://www.baseball-reference.com/leagues/MLB/2014-standard-batting.shtml

BaseballHits2019

Baseball Team Statistics (2019)

# Description

Number of hits, wins, and other stats for MLB teams in 2019

#### **Format**

A data frame with 30 observations on the following 14 variables.

Team Name of baseball team (3-character code)

League Either AL or NL

Wins Number of wins for the season

Runs Number of runs scored

Hits Number of hits

Doubles Number of doubles

BaseballSalaries2015

Triples Number of triples

HomeRuns Number of home runs

RBI Number of runs batted in

StolenBases Number of stolen bases

CaughtStealing Number of times caught stealing

Walks Number of walks

Strikeouts Number of strikeouts

BattingAvg Team batting average

# **Details**

Offensive team statistics for the 2019 Major League Baseball regular season.

\*\* Updated for 3e (earlier versions are now BaseballHits2014 and BaseballHits1e)

#### Source

https://www.baseball-reference.com/leagues/MLB/2019-standard-batting.shtml

BaseballSalaries2015 MLB Player Salaries in 2015

# **Description**

Opening Day salaries for all Major League Baseball players in 2015

# **Format**

A dataset with 868 observations on the following 4 variables.

Name Player's name

Salary 2015 season salary (in millions)

Team Abbreviated team name

Position Code for player's main position

# **Details**

Yearly salary (in millions of dollars) for all players on the rosters of Major League Baseball teams at the start of the 2015 season.

\*\* From 2e - dataset has been updated for 3e \*\*

#### Source

http://www.usatoday.com/sports/mlb/salaries

16 BaseballTimes

BaseballSalaries2019 MLB Player Salaries in 2019

# **Description**

Opening Day salaries for all Major League Baseball players in 2019

#### **Format**

A data frame with 877 observations on the following 4 variables.

Name Player's name

Salary 2019 season salary (in millions)

Team Abbreviated team name

POS Code for player's main position

# **Details**

Yearly salary (in millions of dollars) for all players on the rosters of Major League Baseball teams at the start of the 2019 season.

\*\* Updated for 3e (earlier version for 2015 is at BaseballSalaries2015). \*\*

### **Source**

https://databases.usatoday.com/mlb-salaries/

**BaseballTimes** 

Baseball Game Times

# **Description**

Information for a sample of 30 Major League Baseball games played during the 2011 season

# **Format**

A dataset with 30 observations on the following 9 variables.

Away Away team name Home Home team name

Runs Total runs scored (both teams)

Margin Margin of victory

Hits Total number of hits (both teams)

Errors Total number of errors (both teams)

Pitchers Total number of pitchers used (both teams)

Walks Total number of walks (both teams)

Benford 17

Time Elapsed time for game (in minutes)

# **Details**

Data from a sample of boxscores for Major League Baseball games played in August 2011.

# **Source**

http://www.baseball-reference.com/boxes/2011.shtml

Benford	Benford data
Defit of a	Benjora aara

# Description

Two examples to test Benford's Law

# **Format**

A dataset with 9 observations on the following 4 variables.

Digit	Leading digit (1-9)
BenfordP	Expected proportion according to Benford's law
Address	Frequency as a first digit in an address
Invoices	Frequency as the first digit in invoice amounts

# **Details**

Leading digits from 1188 addresses sampled from a phone book and 7273 amounts from invoices sampled at a company.

# **Source**

Thanks to Prof. Richard Cleary for providing the data

18 BodyFat

BikeCommute Bike Commute
--------------------------

# Description

Commute times for two kinds of bicycle

#### **Format**

A dataset with 56 observations on the following 9 variables.

Bike Type of material Carbon or Steel Date of the bike commute Date Distance Length of commute (in miles) Time Total commute time (hours:minutes:seconds) Minutes Time converted to minutes AvgSpeed Average speed during the ride (miles per hour) Maximum speed (miles per hour) TopSpeed Time converted to seconds Seconds Categories: 1Jan 2Feb 3Mar 4Apr 5May 6June 7July Month

# **Details**

Data from a personal experiment to compare commuting time based on a randomized selection between two bicycles made of different materials.

# Source

Thanks to Dr. Groves for providing his data.

# References

Bicycle weight and commuting time: randomised trial, in British Medical Journal, BMJ 2010;341:c6801.

|--|

# **Description**

Percent fat and other body measurements for a sample of men

BodyTemp50 19

#### **Format**

A dataset with 100 observations on the following 10 variables.

Bodyfat Percent body fat

Age Age in years

Weight Weight in pounds

Height Height in inches

Neck Neck circumference in cm.

Chest Chest circumference in cm.

Abdomen Abdomen circumference in cm.

Ankle Circumference in cm.

Biceps Extended biceps circumference in cm.

Wrist Wrist circumference in cm.

# **Details**

This is a subset of a larger sample of men who each had a percent body fat estimated by an underwater weighing technique. Other measurements were taken to see how they might be used to predict the body fat percentage.

#### Source

These data were contributed by Roger Johnson, then at Carleton University, to the Datasets Archive at the Journal of Statistics Education.

https://ww2.amstat.org/publications/jse/v4n1/datasets.johnson.html

The data were originally supplied by Dr. A. Garth Fisher, Human Performance Research Center, Brigham Young University, Provo, Utah 84602.

BodyTemp50 Body Temperatures

# **Description**

Sample of 50 body temperatures

# **Format**

A data frame with 50 observations on the following 3 variables.

BodyTemp Body temperature in degrees F Pulse Pulse rates (beat per minute) Sex F=Female, M=Male

#### **Details**

Body temperatures and pulse rates for a sample of 50 healthy adults. Note the Sex variable was labeled as Gender in earlier versions of this dataset. We acknowledge that this binary dichotomization is not a complete or inclusive representation of reality.

20 Caffeine Taps

# **Source**

Shoemaker, "What's Normal: Temperature, Gender and Heartrate", Journal of Statistics Education, Vol. 4, No. 2 (1996)

http://jse.amstat.org/v4n2/datasets.shoemaker.html

BootAtlantaCorr

Bootstrap Correlations for Atlanta Commutes

# **Description**

Bootstrap correlations between Time and Distance for 500 commuters in Atlanta

# **Format**

A dataset with 1000 observations on the following variable.

CorrTimeDist Correlation between Time and Distance for a bootstrap sample of Atlanta commuters

# **Details**

Correlations for bootstrap samples of Time vs. Distance for the data on Atlanta commuters in CommuteAtlanta.

#### **Source**

Computer simulation

CaffeineTaps

Caffeine Taps

# Description

Finger tap rates with and without caffeine

#### **Format**

A dataset with 20 observations on the following 2 variables.

Taps Number of finger taps in one minute
Group Treatment with levels Caffeine NoCaffeine

CAOSExam 21

#### **Details**

Results from a double-blind experiment where a sample of male college students were asked to tap their fingers at a rapid rate. The sample was then divided at random into two groups of ten students each. Each student drank the equivalent of about two cups of coffee, which included about 200 mg of caffeine for the students in one group but was decaffeinated coffee for the second group. After a two hour period, each student was tested to measure finger tapping rate (taps per minute). The goal of the experiment was to determine whether caffeine produces an increase in the average tap rate.

#### **Source**

Hand, Daly, Lund, McConway and Ostrowski, Handbook of Small Data Sets, Chapman and Hall, London (1994), pp. 40

CAOSExam

CAOS Exam Scores

# **Description**

Scores on a pre-test and post-test of basic statistics concepts

# **Format**

A dataset with 10 observations on the following 3 variables.

Student ID code for student Pretest CAOS Pretest score Posttest CAOS Posttest score

#### **Details**

The CAOS (Comprehensive Assessment of Outcomes in First Statistics Course) exam is designed to measure comprehension of basic statistical ideas in an introductory statistics course. This dataset has scores for ten students who took the CAOS pre-test at the start of a course and the post-test during the course itself. Each exam consists of 40 multiple choice questions and the score is the percentage correct.

### **Source**

A sample of 10 students from an introductory statistics course. Find out more about the CAOS exam at http://app.gen.umn.edu/artist/caos.html

22 CarbonDioxide2e

CarbonDioxide

Carbon Dioxide Levels

# **Description**

Atmospheric carbon dioxide levels by year

### **Format**

A data frame with 12 observations on the following 2 variables.

Year Every five years from 1960 to 2015

C02 Carbon dioxide level in parts per million

# **Details**

Carbon dioxide levels in the atmosphere over a 55 year span from 1960-2015.

\*\* Updated for 3e (earlier version is now CarbonDioxide2e) \*\*

#### **Source**

Dr. Pieter Tans, NOAA/ESRL. Values recorded at the Mauna Loa Observatory in Hawaii. https://gml.noaa.gov/ccgg/trends/

CarbonDioxide2e

Carbon Dioxide Levels - 2e

# **Description**

Atmospheric carbon dioxide levels by year

#### **Format**

A dataset with 11 observations on the following 2 variables.

Year Every five years from 1960 to 2010
Carbon dioxide level in parts per million

# **Details**

Carbon dioxide levels in the atmosphere over a 50 year span from 1960-2010.

\*\* From 2e - dataset has been updated for 3e \*\*

# Source

Dr. Pieter Tans, NOAA/ESRL (www.esrl.noaa.gov/gmd/ccgg/trends/). Values recorded at the Mauna Loa Observatory in Hawaii.

CarDepreciation 23

|--|

# Description

Depreciation for 20 car models.

#### **Format**

A dataset with 20 observations on the following 4 variables.

Car Name of the car model New Price of a new car

Used Value after new car leaves the lot after purchase

Depreciation Drop in value when a new car is driven away

# **Details**

Twenty car models were selected at random from *kellybluebook.com*. Original price (in dollars) and value after the car has been driven 10 miles were recorded for each model. The depreciation is the difference (New-Used).

# Source

New and used automobile costs determined using 2015 models selected from kellybluebook.com.

Cars2015	2020 Car Models

# **Description**

Information about new car models in 2020

# **Format**

A dataset with 110 observations on the following 24 variables.

```
Make Manufacturer (e.g. Chevrolet, Toyota, etc.)

Model Car model (e.g. Impala, Prius, ...)

Type Vehicle category (Small, Hatchback, Sedan, Sporty, Wagon, SUV, 7Pass)

LowPrice Lowest MSRP (in $1,000)

HighPrice Highest MSRP (in $1,000)

Drive Type of drive (FWD, RWD, AWD)

CityMPG City miles per gallon (EPA)
```

24 Cars 2020

Highway miles per gallon (EPA) HwyMPG FuelCap Fuel capacity (in gallons) Length Length (in inches) Width Width (in inches) Height Height (in inches) Wheelbase (in inches) UTurn Diameter (in feet) needed for a U-turn Weight Curb weight (in pounds) Acc030 Time (in seconds) to go from 0 to 30 mph Acc060 Time (in seconds) to go from 0 to 60 mph QtrMile Time (in seconds) to go ¼ mile Page number in the Consumer Reports New Car Buying Guide PageNum Size Small, Midsized, or Large

#### **Details**

Data for a set of 110 new car models in 2015 based on information in the Consumer Reports. \*\* From 2e - dataset has been updated for 3e \*\*

# Source

Data on new car models in 2020 accessed from Consumer Reports website. https://www.consumerreports.org/cars/

Cars2020 2020 Car Models

# **Description**

Information about new car models in 2020

# **Format**

A data frame with 110 observations on the following 21 variables.

Make Manufacturer (e.g. Chevrolet, Toyota, etc.)

Model Car model (e.g. Impala, Highlander, ...)

Type Vehicle category (Hatchback, Minivan, Sedan, Sporty, SUV, or Wagon)

LowPrice Lowest MSRP (in \$1,000)

HighPrice Highest MSRP (in \$1,000)

CityMPG City miles per gallon (EPA)

HwyMPG Highway miles per gallon (EPA)

Seating Seating capacity

Drive Type of drive (AWD, FWD, or RWD)

Cereal 25

```
Acc030 Time (in seconds) to go from 0 to 30 mph
Acc060 Time (in seconds) to go from 0 to 60 mph
QtrMile Time (in seconds) to go ¼ mile
Braking Distance to stop from 60 mph (dry pavement)
FuelCap Fuel capacity (in gallons)
Length Length (in inches)
Width Width (in inches)
Height Height (in inches)
Wheelbase Wheelbase (in inches)
UTurn Diameter (in feet) needed for a U-turn
Weight Curb weight (in pounds)
Size Large, Midsized, or Small
```

#### **Details**

Data for a set of 110 new car models in 2020 based on information in the Consumer Reports. \*\* Updated for 3e (an earlier version from 2015 is at Cars2015). \*\*

#### **Source**

Data on new car models in 2020 accessed from Consumer Reports website. https://www.consumerreports.org/cars/

Cereal	Breakfast Cereals	

# **Description**

Nutrition information for a sample of 30 breakfast cereals

# **Format**

A dataset with 30 observations on the following 10 variables.

Name	Brand name of cereal
Company	Manufacturer coded as G=General Mills, K=Kellog's or Q=Quaker
Serving	Serving size (in cups)
Calories	Calories (per cup)
Fat	Fat (grams per cup)
Sodium	Sodium (mg per cup)
Carbs	Carbohydrates (grams per cup)
Fiber	Dietary Fiber (grams per cup)
Sugars	Sugars (grams per cup)

26 CityTemps

Protein Protein (grams per cup)

#### **Details**

Nutrition contents for a sample of breakfast cereals, derived from nutrition labels. Values are per cup of cereal (rather than per serving).

#### Source

Cereal data obtained from nutrition labels at http://www.nutritionresource.com/foodcomp2.cfm?id=0800

CityTemps

City Temperatures

# Description

Mean monthly temperature in Moscow, Melbourne, and San Francisco for 2017 and 2018

#### **Format**

A data frame with 24 observations on the following 5 variables.

Year 2017 or 2018

Month 1=January through 12=December

Moscow Monthly temperatures in Moscow (Russia)

Melbourne (Australia)

San.Francisco Monthly temperatures in San Francisco (United States)

#### **Details**

Mean monthly temperatures in degrees C for the years 2017 and 2018 in each of three cities. \*\* Updated for 3e (an earlier version for 2014 and 2015 is at CityTemps2e). \*\*

### Source

Source: KNMI Climate Explorer at https://climexp.knmi.nl/selectstation.cgi?id=someone@somewhere Use station codes 94866 (Melbourne), 72494 (San Francisco), 27612 (Moscow).

CityTemps2e 27

City Tempsze City Temperatures - 2	CityTemps2e	City Temperatures -	2e
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# **Description**

Mean monthly temperature in Moscow, Melbourne, and San Francisco for 2014 and 2015

#### **Format**

A dataset with 24 observations on the following 5 variables.

Year 2014 or 2015

Month 1=January to 12=December

Moscow Monthly temperatures in Moscow (Russia)

Melbourne Monthly temperatures in Melbourne (Australia)

SanFrancisco Monthly temperatures in San Francisco (United States)

# **Details**

Mean monthly temperatures in degrees Celsius for the years 2014 and 2015 in each of three cities. \*\* From 2e - dataset has been updated for 3e \*\*

#### Source

KNMI Climate Explorer at https://climexp.knmi.nl/selectstation.cgi?id=someone@somewhere

neTreatment Cocaine Treatment	
Telreatment Cocaine Treatment	

# Description

Relapse/no relapse responses to three different treatments for cocaine addiction

#### **Format**

A dataset with 72 observations on the following 2 variables.

Drug Treatment drug: Desipramine, Lithium, or Placebo Relapse Did the patient relapse? no or yes

#### **Details**

Data from an experiment to investigate the effectiveness of the two drugs, desipramine and lithium, in the treatment of cocaine addiction. Subjects (cocaine addicts seeking treatment) were randomly assigned to take one of the treatment drugs or a placebo. The response variable is whether or not the subject relapsed (went back to using cocaine) after the treatment.

28 CollegeScores

# Source

Gawin, F., et.al., "Desipramine Facilitation of Initial Cocaine Abstinence", Archives of General Psychiatry, 1989; 46(2): 117 - 121.

ColaCalcium

Cola Calcium

# **Description**

Calcium excretion with diet cola and water

# **Format**

A dataset with 16 observations on the following 2 variables.

Drink Type of drink: Diet cola or Water Calcium Amount of calcium excreted (in mg.)

### **Details**

A sample of 16 healthy women aged 18 - 40 were randomly assigned to drink 24 ounces of either diet cola or water. Their urine was collected for three hours after ingestion of the beverage and calcium excretion (in mg.) was measured. The researchers were investigating whether diet cola leaches calcium out of the system, which would increase the amount of calcium in the urine for diet cola drinkers.

# Source

Larson, Amin, Olsen, and Poth, Effect of Diet Cola on Urine Calcium Excretion, Endocrine Reviews, 31[3]: S1070, June 2010. These data are recreated from the published summary statistics, and are estimates of the actual data.

CollegeScores

College Scorecard

# Description

Information on all US post-secondary schools collected by the Department of Education for the College Scorecard

CollegeScores 29

#### **Format**

A data frame with 6141 observations on the following 37 variables.

Name Name of the school

State State where school is located

ID ID number for school

Main Main campus? (1=yes, 0=branch campus)

Accred Accreditation agency

MainDegree Predominant undergrad degree (0=not classified, 1=certificate, 2=associate, 3=bachelors,4=only graduate)

HighDegree Highest degree (0=no degrees, 1=certificate, 2=associate, 3=bachelors, 4= graduate)

Control Control of school (Private, Profit, Public)

Region Region of country (Midwest, Northeast, Southeast, Territory, West)

Locale (City, Rural, Suburb, Town)

Latitude Latitude

Longitude Longitude

AdmitRate Admission rate

MidACT Median of ACT scores

AvgSAT Average combined SAT scores

Online Only online (distance) programs

Enrollment Undergraduate enrollment

White Percent of undergraduates who report being white

Black Percent of undergraduates who report being black

Hispanic Percent of undergraduates who report being Hispanic

Asian Percent of undergraduates who report being Asian

Other Percent of undergraduates who don't report one of the above

PartTime Percent of undergraduates who are part-time students

NetPrice Average net price (cost minus aid)

Cost Average total cost for tuition, room, board, etc.

TuitionIn In-state tuition and fees

TuitonOut Out-of-state tuition and fees

TuitionFTE Net Tuition revenue per FTE student

InstructFTE Instructional spending per FTE student

FacSalary Average monthly salary for full-time faculty

FullTimeFac Percent of faculty that are full-time

Pell Percent of students receiving Pell grants

CompRate Completion rate (percent who finish program within 150% of normal time)

Debt Average debt for students who complete program

Female Percent of female students

FirstGen Percent of first-generation students

MedIncome Median family income (in \$1,000)

30 CollegeScores2yr

#### **Details**

The US Department of Education maintains a database through its College Scorecard project of demographic information from all active postsecondary educational institutions that participate in Title IV. This dataset contains a small subsets of the variables in the full College Scorecard.

#### Source

Data downloaded from the US Department of Education's College Scorecard at https://collegescorecard.ed.gov/data/(November 2019)

CollegeScores2yr

College Scorecard - Two Year

# **Description**

Information on all US colleges and universities that primarily grant associate's degrees, collected by the Department of Education for the College Scoreboard.

#### **Format**

A data frame with 1141 observations on the following 37 variables.

Name Name of the school

State State where school is located

ID ID number for school

Main Main campus? (1=yes, 0=branch campus)

Accred Accreditation agency

MainDegree Predominant undergrad degree (2=associate)

HighDegree Highest degree (0=no degrees, 1=certificate, 2=associate, 3=bachelors, 4= graduate)

Control Control of school (Private, Profit, Public)

Region Region of country (Midwest, Northeast, Southeast, Territory, West)

Locale Locale (City, Rural, Suburb, Town)

Latitude Latitude

Longitude Longitude

AdmitRate Admission rate

MidACT Median of ACT scores

AvgSAT Average combined SAT scores

Online Only online (distance) programs

Enrollment Undergraduate enrollment

White Percent of undergraduates who report being white

Black Percent of undergraduates who report being black

CollegeScores4yr 31

Hispanic Percent of undergraduates who report being Hispanic

Asian Percent of undergraduates who report being Asian

Other Percent of undergraduates who don't report one of the above

PartTime Percent of undergraduates who are part-time students

NetPrice Average net price (cost minus aid)

Cost Average total cost for tuition, room, board, etc.

TuitionIn In-state tuition and fees

TuitonOut Out-of-state tuition and fees

TuitionFTE Net Tuition revenue per FTE student

InstructFTE Instructional spending per FTE student

FacSalary Average monthly salary for full-time faculty

FullTimeFac Percent of faculty that are full-time

Pell Percent of students receiving Pell grants

CompRate Completion rate (percent who finish program within 150% of normal time)

Debt Average debt for students who complete program

Female Percent of female students

FirstGen Percent of first-generation students

MedIncome Median family income (in \$1,000)

# **Details**

The US Department of Education maintains a database through its College Scorecard project of demographic information from all active postsecondary educational institutions that participate in Title IV. This dataset contains a small subset of the variables in the full College Scorecard and only the schools that primarily grant associate's degrees (MainDegree=2). The CollegeScores dataset contains these and other schools with other degree types.

#### Source

Data downloaded from the US Department of Education's College Scorecard at https://collegescorecard.ed.gov/data/(November 2019)

CollegeScores4yr

College Scorecard - Four Year

# Description

Information on all US colleges and universities that primarily grant bachelor's degrees, collected by the Department of Education for the College Scoreboard 32 CollegeScores4yr

#### **Format**

A data frame with 2012 observations on the following 37 variables.

Name Name of the school

State State where school is located

ID ID number for school

Main Main campus? (1=yes, 0=branch campus)

Accred Accreditation agency

MainDegree Predominant undergrad degree (3=bachelors)

HighDegree Highest degree (0=no degrees, 1=certificate, 2=associate, 3=bachelors, 4= graduate)

Control Control of school (Private, Profit, Public)

Region Region of country (Midwest, Northeast, Southeast, Territory, West)

Locale (City, Rural, Suburb, Town)

Latitude Latitude

Longitude Longitude

AdmitRate Admission rate

MidACT Median of ACT scores

AvgSAT Average combined SAT scores

Online Only online (distance) programs

Enrollment Undergraduate enrollment

White Percent of undergraduates who report being white

Black Percent of undergraduates who report being black

Hispanic Percent of undergraduates who report being Hispanic

Asian Percent of undergraduates who report being Asian

Other Percent of undergraduates who don't report one of the above

PartTime Percent of undergraduates who are part-time students

NetPrice Average net price (cost minus aid)

Cost Average total cost for tuition, room, board, etc.

TuitionIn In-state tuition and fees

TuitonOut Out-of-state tuition and fees

TuitionFTE Net Tuition revenue per FTE student

InstructFTE Instructional spending per FTE student

FacSalary Average monthly salary for full-time faculty

FullTimeFac Percent of faculty that are full-time

Pell Percent of students receiving Pell grants

CompRate Completion rate (percent who finish program within 150% of normal time)

Debt Average debt for students who complete program

Female Percent of female students

FirstGen Percent of first-generation students

MedIncome Median family income (in \$1,000)

CommuteAtlanta 33

# **Details**

The US Department of Education maintains a database through its College Scorecard project of demographic information from all active postsecondary educational institutions that participate in Title IV. This dataset contains a small subset of the variables in the full College Scorecard and only the schools that primarily grant bachelor's degrees (MainDegree=3). The CollegeScores dataset contains these and other schools with other degree types.

#### Source

Data downloaded from the US Department of Education's College Scorecard at https://collegescorecard.ed.gov/data/(November 2019)

CommuteAtlanta

Commute Atlanta

# **Description**

Commute times and distances for a sample of 500 people in Atlanta

# **Format**

A data frame with 500 observations on the following 5 variables.

City Atlanta

Age Age of the respondent (in years)

Distance Commute distance (in miles)

Time Commute time (in minutes)

Sex F or M

# **Details**

Data from the US Census Bureau's American Housing Survey (AHS) which contains information about housing and living conditions for samples from certain metropolitan areas. These data were extracted from respondents in the Atlanta metropolitan area. They include only cases where the respondent worked somewhere other than home. Values show the time (in minutes) and distance (in miles) that respondents typically traveled on their commute to work each day as well as age and sex.

# Source

Sample chosen using DataFerret at http://www.thedataweb.org/index.html.

34 CompassionateRats

CommuteStLouis

Commute Times in St. Louis

# **Description**

Commute times and distances for a sample of 500 people in St. Louis

#### **Format**

A dataset with 500 observations on the following 5 variables.

City St. Louis

Age Age of the respondent (in years)
Distance Commute distance (in miles)

Time Commute time (in minutes)

Sex F or M

#### **Details**

Data from the US Census Bureau's American Housing Survey (AHS) which contains information about housing and living conditions for samples from certain metropolitan areas. These data were extracted from respondents in the St. Louis metropolitan area. They include only cases where the respondent worked somewhere other than home. Values show the time (in minutes) and distance (in miles) that respondents typically traveled on their commute to work each day as well as age and sex.

# Source

Sample chosen using DataFerret at http://www.thedataweb.org/index.html.

 ${\tt CompassionateRats}$ 

Compassionate Rats

# **Description**

Would a rat attempt to free a trapped rat?

### **Format**

A dataset with 30 observations on the following 2 variables.

Sex Sex of the rat: coded as F or M
Empathy Freed the trapped rat? no or yes

CricketChirps 35

# **Details**

In a recent study, some rats showed compassion by freeing another trapped rat, even when chocolate served as a distraction and even when the rats would then have to share the chocolate with their freed companion.

#### **Source**

Bartal I.B., Decety J., and Mason P., "Empathy and Pro-Social Behavior in Rats," Science, 2011; 224(6061):1427-1430.

CricketChirps

Cricket Chirps

# Description

Cricket chirp rate and temperature

# **Format**

A dataset with 7 observations on the following 2 variables.

Temperature Air temperature in degrees F
Chirps Cricket chirp rate (chirps per minute)

# **Details**

The data were collected by E.A. Bessey and C.A. Bessey who measured chirp rates for crickets and temperatures during the summer of 1898.

# **Source**

From E.A Bessey and C.A Bessey, Further Notes on Thermometer Crickets, American Naturalist, (1898) 32, 263-264.

36 DecemberFlights

DDS

**Developmental Services** 

# **Description**

Funding for individuals by the California Department of Developmental Services (DDS),

#### **Format**

A dataset with 1000 observations on the following 6 variables.

ID ID code for subject

AgeCohort Age group (0-5, 6-12, 13-17, 18-21, 22-50, 50+)

Age Age in years

Expenditures Annual expenditures in dollars

Ethnicity Ethnic group

#### **Details**

The California Department of Developmental Services (DDS) allocates funds to support developmentally disabled California residents (such as those with autism, cerebral palsy, or intellectual disabilities) and their families. We refer to those supported by DDS as DDS consumers. The dataset DDS includes data on annual expenditure (in \$), ethnicity, age, and gender for 1000 DDS consumers.

#### Source

Taylor, S.A. and Mickel, A. E. (2014). "Simpson's Paradox: A Data Set and Discrimination Case Study Exercise," Journal of Statistics Education, 22(1). The dataset has been altered slightly for privacy reasons, but is based on actual DDS consumers.

DecemberFlights

December Flights

# **Description**

Difference between actual and scheduled arrival for United and Delta flights in December 2018.

# **Format**

A data frame with 2000 observations on the following 2 variables.

Airline Delta or United

Difference Actual - Scheduled arrival times (in minutes)

DecemberFlights2e 37

### **Details**

For a sample of 1000 December flights (in 2018) from each airline, we find the difference between actual and scheduled arrival times. A negative value indicates the flight arrived early.

\*\* Updated for 3e (earlier version from 2014 is in DecemberFlights2e.)

### **Source**

Downloaded from the Bureau of Transportation Statistics (https://www.transtats.bts.gov/).

DecemberFlights2e

December Flights - 2e

## Description

Difference between actual and scheduled arrival for a sample of United and Delta flights in December 2014.

### **Format**

A dataset with 2000 observations on the following 2 variables.

Airline Delta or United
Difference (Actual - Scheduled arrival times)

## **Details**

For a sample of 1000 December flights (in 2014) from each airline, we find the difference between actual and scheduled arrival times. A negative value indicates the flight arrived early.

\*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

Downloaded from the Bureau of Transportation Statistics (https://www.bts.gov/). More specific URL is https://www.transtats.bts.gov/DL\_SelectFields.asp?Table\_ID=236&DB\_Short\_Name=On-Time.

38 DietDepression

DietDepression

Diet and Depression

# **Description**

Results from a study of a short-term diet intervention on depression.

### **Format**

A data frame with 75 observations on the following 10 variables.

Group Control or Diet

CESD1 CESD depression score on Day 1

CESD21 CESD depression score on Day 21

CESDDiff Change in CESD depression score

DASS1 DASS depression score on Day 1

DASS21 DASS depression score on Day 21

DASSDiff Change in DASS depression score

BMI1 Body Mass Index on Day 1

BMI21 Body Mass Index on Day 21

BMIDiff Change in Body Mass Index

### **Details**

A group of researchers in Australia conducted a short (three-week) dietary intervention in a randomized controlled experiment. In the study, 75 college-age students with elevated depression symptoms and relatively poor diet habits were randomly assigned to either a healthy diet intervention group or a control group. The researchers recorded the change over the three-week period on two different numeric scales of depression (the CESD scale and the DASS scale). The CESD (Centre for Epidemiological Studies Depression) score is based more on clinical observations, while the DASS (Depression, Anxiety, and Stress Scale) depends more on self-reported information. They also recorded body mass index (BMI) at the start and end of the 21 day period.

### Source

Francis HM, et al., "A brief diet intervention can reduce symptoms of depression in young adults - A randomised controlled trial," PLoS ONE, 14(10), October 2019.

Digits 39

Digits	Digit Counts		
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# Description

Digits from social security numbers and student selected "random numbers"

## **Format**

A dataset with 150 observations on the following 7 variables.

Random	Four digit random numbers given by a sample of students
RND1	First digit
RND2	Second digit
RND3	Third digit
RND4	Fourth digit
SSN8	Eighth digit of social security number
SSN9	Last digit of social security number

### **Details**

A sample of students were asked to give a random four digit number. The numbers are given in the dataset, along with separate columns for each of the four digits. The data also show the last two digits of each student's social security number (SSN).

# Source

In-class student surveys from several classes.

DogOwner	Dog/Owner matches	

# Description

Experiment to match dogs with owners

## **Format**

A dataset with 25 observations on the following variable.

Match Was the dog correctly paired with it's owner? no or yes

40 DrugResistance

### **Details**

Pictures were taken of 25 owners and their purebred dogs, selected from dog parks. Study participants were shown a picture of an owner together with pictures of two dogs (the owner's dog and another random dog from the study) and asked to choose which dog most resembled the owner. Each dog-owner pair was viewed by 28 naive undergraduate judges, and the pairing was deemed "correct" (yes) if the majority of judges (more than 14) chose the correct dog to go with the owner. \*\* In first edition, but not as dataset in 2e \*\*

### **Source**

Roy and Christenfeld, Do Dogs Resemble their Owners?, Psychological Science, Vol. 15, No. 5, 2004, pp. 361 - 363.

DrugResistance

Drug Resistance

## **Description**

Effect on drug resistance by level of treatment in mice.

### **Format**

A dataset with 72 observations on the following 5 variables.

Treatment Untreated, Light, Moderate, or Aggressive

Mouse weight in grams Weight Red blood cell density RBC

Density of resistant parasites

ResistantDensity DaysInfectious Days infectious with resistant parasites

### **Details**

In an experiment to study drug resistance in mice, groups of 18 mice were injected with a mixture of drug-resistant and drug-susceptible malaria parasites. One group received no treatment while the others got limited, moderate, or aggressive amounts of anti-malarial treatment. The weight and red blood cell density reflect the initial health of the mice. Density of resistant parasites and number of days infectious measure the effectiveness of the treatment.

### Source

Huijben S, Bell AS, Sim DG, Tomasello D, Mideo N, Day T, Read AF (2013) Aggressive chemotherapy and the selection of drug resistant pathogens. PLoS Pathogens 9(9): e1003578.

http://dx.doi.org/10.1371/journal.ppat.1003578

Huijben S, et al., (2013). Data from: Aggressive chemotherapy and the selection of drug resistant pathogens. Dryad Digital

Repository. http://dx.doi.org/10.5061/dryad.09qc0

EducationLiteracy 41

EducationLiteracy

Education and Literacy

## **Description**

Education spending and literacy rates for countries.

### **Format**

A data frame with 170 observations on the following 4 variables.

Country Name of country

Code Three-letter code for country

Education Education spending (as a percentage of GDP)

Literacy Literacy rate

### **Details**

For each country, we have public spending on education (as a percentage of GDP) and literacy rate (percentage of the population who can read and write).

\*\* Updated for 3e (an earlier version is at EducationLiteracy2e). \*\*

# Source

Most recent data (as of 2019) for each country obtained from https://www.worldbank.org/en/home.

EducationLiteracy2e

Education Literacy - 2e

### **Description**

Education spending and literacy rates for countries.

## **Format**

A dataset with 188 observations on the following 3 variables.

Country Name of country

Education Education spending (as a percentage of GDP)

Literacy Literacy rate

42 ElectionMargin

### **Details**

For each country, we have public spending on education (as a percentage of GDP) and literacy rate (percentage of the population who can read and write).

\*\* From 2e - dataset has been updated for 3e \*\*

### Source

Most recent data (as of 2015) for each country obtained from worldbank.org and http://www.knoema.com

ElectionMargin

Election Margin

## **Description**

Approval rating and election margin for recent presidential elections

### **Format**

A dataset with 12 observations on the following 5 variables.

Year Certain election years from 1940-2012

Candidate Incumbent US president

Approval Presidential approval rating at time of election Margin Margin of victory/defeat (as a percentage)

Result Outcome of the election for the incumbent: Lost or Won

## **Details**

Data include US Presidential elections since 1940 in which an incumbent was running for president. The approval rating for the sitting president is compared to the margin of victory/defeat in the election.

\*\* Updated for 2e (original is now ElectionMargin1e) \*\*

## Source

Silver, Nate, "Approval Ratings and Re-Election Odds", *fivethirtyeight.com*, posted January 28, 2011 and *http:\realclearpolitics.org* 

EmployedACS 43

EmployedACS Employed in American Community Survey	EmployedACS	Employed in American Community Survey
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## **Description**

Employed individuals from the American Community Survey (ACS) dataset

### **Format**

A data frame with 1287 observations on the following 9 variables.

Sex 0=female and 1=male

Age Age (years)

Married 0=not married and 1=married

Income Wages and salary for the past 12 months (in \$1,000's)

HoursWk Hours of work per week

Race asian, black, other, white

USCitizen 1=citizen and 0=noncitizen

HealthInsurance 1=have health insurance and 0= no health insurance

Language 1=native English speaker and 0=other

### **Details**

This is a subset of the ACS dataset including only 1287 individuals who were employed. (HoursWk>0)

\*\* Updated for 3e (an earlier version is at EmployedACS2010). \*\*

## Source

The full public dataset can be downloaded at https://www.census.gov/programs-surveys/acs/microdata/access.html, and the full list of variables is at https://www.census.gov/programs-surveys/acs/microdata.html

EmployedACS2010 Employed in American Community Survey - 2010

## Description

Employed individuals from the American Community Survey (ACS) dataset in 2010

44 ExerciseHours

### **Format**

A dataset with 431 observations on the following 9 variables.

Sex 0=female and 1=male

Age (years)

Married 0=not married and 1=married

Income Wages and salary for the past 12 months (in \$1,000's)

Hours Wk Hours of work per week

Race asian, black, white, or other

USCitizen 1=citizen and 0=noncitizen

HealthInsurance 1=have health insurance and 0= no health insurance

Language 1=native English speaker and 0=other

### **Details**

This is a subset of the ACS dataset including only 431 individuals who were employed.

\*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

The full public dataset can be downloaded at

http://www.census.gov/acs/www/data documentation/pums data/,

and the full list of variables are at

http://www.census.gov/acs/www/Downloads/data documentation/pums/DataDict/PUMSDataDict10.pdf

ExerciseHours

Exercise Hours

### **Description**

Amount of exercise per week for students (and other variables)

### **Format**

A data frame with 50 observations on the following 7 variables.

Year Year in school (1=First year,..., 4=Senior)

Sex F or M

Hand Left (1) or Right (r) handed?

Exercise Hours of exercise per week

TV Hours of TV viewing per week

Pulse Resting pulse rate (beats per minute)

Pierces Number of body piercings

FacebookFriends 45

### **Details**

Data from an in-class survey of statistics students asking about amount of exercise, TV viewing, handedness, sex, pulse rate, and number of body piercings. Note the Sex variable was labeled as Gender in earlier versions of this dataset. We acknowledge that this binary dichotomization is not a complete or inclusive representation of reality.

#### Source

In-class student survey.

FacebookFriends

Facebook Friends

## **Description**

Data on number of Facebook friends and grey matter density in brain regions related to social perception and associative memory.

### **Format**

A dataset with 40 observations on the following 2 variables.

GMdensity Normalized z-scores of grey matter density in certain brain regions FBfriends Number of friends on Facebook

### **Details**

A recent study in Great Britain examines the relationship between the number of friends an individual has on Facebook and grey matter density in the areas of the brain associated with social perception and associative memory. The study included 40 students at City University London.

### Source

Kanai, R., Bahrami, B., Roylance, R., and Rees, G., "Online social network size is reflected in human brain structure," Proceedings of the Royal Society, 7 April 2012; 279(1732): 1327-1334. Data approximated from information in the article.

46 FireAnts

FatMice18	Fat Mice 18	

## **Description**

Weight gain for mice with different nighttime light conditions

### **Format**

A dataset with 18 observations on the following 2 variables.

```
Light Light treatment: LD= normal light/dark cycle OR LL=bright light at night WgtGain4 Weight gain (grams over a four week period)
```

### **Details**

This is a subset of the LightatNight dataset, showing body mass gain in mice after 4 weeks for two of the treatment conditions: a normal light/dark cycle (LD) or a bright light on at night (LL).

\*\* In first edition, but not 2e \*\*

### **Source**

Fonken, L., et. al., "Light at night increases body mass by shifting time of food intake," Proceedings of the National Academy of Sciences, October 26, 2010; 107(43): 18664-18669.

# Description

Reactions of lizards to the presence of fire ants.

### **Format**

A dataset with 80 observations on the following 3 variables.

Invasion	Coded as Uninvaded or Invaded, depending on if the lizard comes from a region with fire ants
Twitches	Number of twitches the lizard makes when encountering fire ants
Flee	Time for the lizard to flee in seconds (more than one minute is recorded as 61).

FisherIris 47

### **Details**

The red imported fire ant, *Solenopsis invicta*, is native to South America, but has an expansive invasive range, including much of the southern United States (invasion of this ant is predicted to go global). In the United States, these ants occupy similar habitats as fence lizards. The ants eat the lizards and the lizards eat the ants, and in either scenario the venom from the fire ant can be fatal to the lizard. The study explored the question of whether lizards learn to adapt their behavior if their environment has been invaded by fire ants by taking lizards from an uninvaded habitat (eastern Arkansas) and lizards from an invaded habitat (southern Alabama, which has been invaded for more than 70 years), exposing them to fire ants, and measuring how long it takes each lizard to flee and the number of twitches each lizard does.

### **Source**

Langkilde, T. (2009). "Invasive fire ants alter behavior and morphology of native lizards"", Ecology, 90(1): 208-217. Thanks to Dr. Langkilde for providing the data.

FisherIris

Fisher's Iris Data

## **Description**

Measurements of three iris species

### **Format**

A dataset with 150 observations on the following 5 variables.

Type Species of iris, Setosa, Virginica, or Versicolor calLength Petal length in mm.

PetalLength Petal length in mm.
PetalWidth Petal width in mm.
SepalLength Sepal length in mm.
SepalWidth Sepal width in mm.

### **Details**

Data used in Fisher's 1936 paper, this famous dataset looks at measurements for samples of three different species of iris. The petal is part of the flower itself and the sepals are green leaves, directly under the petals, providing support.

### Source

R. A. Fisher (1936). "The use of multiple measurements in taxonomic problems". Annals of Eugenics 7 (2): 179–188. doi:10.1111/j.1469-1809.1936.tb02137.x.

48 FishGills3

FishGills12

Fish Respiration and Calcium - Full Data

### **Description**

An experiment to look at fish respiration rates in water with different levels of calcium.

### **Format**

A dataset with 360 observations on the following 2 variables.

Calcium Amount of calcium in the water (mg/L)
GillRate Respiration rate (beats per minute)

### **Details**

Fish were randomly assigned to twelve tanks with different levels (measured in mg/L) of calcium. Respiration rate was measured as number of gill beats per minute.

### **Source**

Thanks to Prof. Brad Baldwin for supplying the data.

FishGills3

Fish Respiration and Calcium

## **Description**

Respiration rate for fish in three levels of calcium.

### **Format**

A dataset with 90 observations on the following 2 variables.

Calcium Level of calcium Low 0.71 mg/L, Medium 5.24 mg/L, or High 18.24 mg/L GillRate Respiration rate (beats per minute)

### **Details**

Fish were randomly assigned to three tanks with different levels (low, medium and high) of calcium. Respiration rate was measured as number of gill beats per minute.

## Source

Thanks to Prof. Brad Baldwin for supplying the data.

Flight179 49

## Description

Flight times for Flight 179 (Boston-SF) and Flight 180 (SF-Boston).

### **Format**

A dataset with 36 observations on the following 3 variables.

Date	Date of the flight (5th, 15th and 25th of each month in 2010
Flight179	Flying time (Boston-SF) in minutes
Flight180	Flying time (SF-Boston) in minutes

## **Details**

United Airlines Flight 179 was a daily flight from Boston to San Francisco. Flight 180 goes in the other direction (SF to Boston). The data show the airborne flying times for each flight on the three dates each month (5th, 15th and 25th) in 2010.

\*\* In first edition, but not in 2e - replaced by Flight433 \*\*

### Source

Data collected from the Bureau of Transportation Statistics website at <a href="http://www.bts.gov/xml/ontimesummarystatistics/src/dstat/OntimeSummaryAirtime.xml">http://www.bts.gov/xml/ontimesummarystatistics/src/dstat/OntimeSummaryAirtime.xml</a>

|--|

# Description

Flight times for Flight 433 (Boston-SF) in January 2019.

### **Format**

A data frame with 28 observations on the following variable.

AirTime Airborne flying time (in minutes) for Flight 433, Boston to San Francisco

## **Details**

United Airlines Flight 433 was a daily flight from Boston to San Francisco. The data show the airborne flying times for the flight on each day of January 2019.

\*\*Updated for 3e (earlier version from 2016 is in Flight433\_2e) \*\*

50 FloridaLakes

### **Source**

Data collected from the Bureau of Transportation Statistics website at <a href="https://www.transtats.bts.gov/">https://www.transtats.bts.gov/</a>

Flight433\_2e

Flight 433 - 2e

## **Description**

Flight times for Flight 433 (Boston-SF) in January 2016.

### **Format**

A dataset with 31 observations on the following 1 variable.

Airtime Airborne flying time (in minutes) for Flight 433, Boston to San Francisco

## **Details**

United Airlines Flight 433 was a daily flight from Boston to San Francisco. The data show the airborne flying times for the flight on each day of January 2016.

\*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

Data collected from the Bureau of Transportation Statistics website at <a href="http://www.bts.gov/xml/ontimesummarystatistics/src/dstat/OntimeSummaryAirtime.xml">http://www.bts.gov/xml/ontimesummarystatistics/src/dstat/OntimeSummaryAirtime.xml</a>

FloridaLakes

Florida Lakes

# Description

Water quality measurements for a sample of lakes in Florida

FootballBrain 51

### **Format**

A dataset with 53 observations on the following 12 variables.

ID An identifying number for each lake

Lake Name of the lake

Alkalinity Concentration of calcium carbonate (in mg/L)

pH Acidity

Calcium Amount of calcium in water
Chlorophyll Amount of chlorophyll in water

AvgMercury Average mercury level for a sample of fish (large mouth bass) from each lake

NumSamples Number of fish sampled at each lake
MinMercury Minimum mercury level in a sampled fish
MaxMercury Maximum mercury level in a sampled fish

ThreeYrStdMercury Adjusted mercury level to account for the age of the fish

AgeData Mean age of fish in each sample

#### **Details**

This dataset describes characteristics of water and fish samples from 53 Florida lakes. Some variables (e.g. Alkalinity, pH, and Calcium) reflect the chemistry of the water samples. Mercury levels were recorded for a sample of large mouth bass selected at each lake.

#### Source

Lange, Royals, and Connor, Transactions of the American Fisheries Society (1993)

FootballBrain Football Brain Measurements
---

## Description

Brain measurements for non-football players, football players with no concussion history, and football players with a concussion history.

### **Format**

A dataset with 75 observations on the following 5 variables.

Group Control=no football, FBNoConcuss=football player but no concussions, or FBConcuss=football player with concussion history

Hipp Total hippocampus volume, in microL LeftHipp Left hippocampus volume, in microL Years Number of years playing football

Cognition Cognitive testing composite reaction time score, given as a percentile

52 ForestFires

### **Details**

The study included 3 groups, with 25 cases in each group. The control group consisted of healthy individuals with no history of brain trauma who were comparable to the other groups in age, sex, and education. The second group consisted of NCAA Division 1 college football players with no history of concussion, while the third group consisted of NCAA Division 1 college football players with a history of concussion. High resolution MRI was used to collect brain hippocampus volume. Data were collected between June 2011 and August 2013. The data values given here are estimated from information given in the paper.

### **Source**

Singh R, Meier T, Kuplicki R, Savitz J, et al., "Relationship of Collegiate Football Experience and Concussion With Hippocampal Volume and Cognitive Outcome," JAMA, 311(18), 2014

ForestFires

Forest Fires

### **Description**

Characteristics of forest fires in Montesinho park (Portugal)

#### **Format**

A data frame with 517 observations on the following 13 variables.

X West to east coordinates for the site (1=farthest west to 9= farthest east)

Y North to south coordinates for the site (1=farthest north to 9=farthest south)

Month Month of the year (jan to dec)

Day Day of the week (sun to sat)

FFMC Fine fuel moisture code

DMC Duff moisture code

DC Drought code

ISI Initial spread index

Temp Outside temperature (in celsius)

RH Relative humidity (in %)

Wind Wind speed (in km/h)

Rain Rain in past 30 minutes (in mm/sq-m)

Area Total burned area (in hectares)

Genetic Diversity 53

### **Details**

Data were recorded for fires in the Montesinho natural park in Portugal between January 2000 and December 2003. A map of the park (see the pdf linked below) is divided into 9x9 grid sections (given by the x,y-coordinates in the first two columns of the dataset). There are four components of a Fire Weather Index that rate how weather conditions might increase fire danger. FFMC. DMC, and DC reflect various measures of moisture content, while the ISI score indicated how fast a fire might spread (for example, by wind). For all four measures larger values are associated with more fire danger. Fires that are less than 100 square meters in size (0.01 hectares) are recorded as Area=0.

#### Source

Data downloaded from the UCI Machine Learning Repository, <a href="https://archive.ics.uci.edu/ml/datasets/Forest+Fires">https://archive.ics.uci.edu/ml/datasets/Forest+Fires</a>

Original article: P. Cortez and A. Morais. "A Data Mining Approach to Predict Forest Fires using Meteorological Data", in New Trends in Artificial Intelligence, Proceedings of the 13th EPIA 2007 - Portuguese Conference on Artificial Intelligence (December 2007) http://www.dsi.uminho.pt/~pcortez/fires.pdf

GeneticDiversity

Genetic Diversity

### **Description**

Genetic diversity for different populations are compared to the distance from East Africa.

### **Format**

A dataset with 52 observations on the following 5 variables.

Population Identifier for each population

Country Main country where the population is found Continent Continent where the population is found

 $\label{lem:condition} \mbox{Genetic Diversity} \quad \mbox{A measure of genetic diversity in the population}$ 

Distance Distance by land to East Africa (in km)

### **Details**

The data give a measure of genetic diversity for different populations and the geographic distance of each population from East Africa (Addis Ababa, Ethiopia), as one would travel over the surface of the earth by land (migration long ago is thought to have happened by land).

### Source

Calculated using data from S Ramachandran, O Deshpande, CC Roseman, NA Rosenberg, MW Feldman, LL Cavalli-Sforza. "Support from the relationship of genetic and geographic distance in human populations for a serial founder effect originating in Africa,"" Proceedings of the National Academy of Sciences, 2005, 102: 15942-15947.

54 GlobalInternet2019

GlobalInternet2010

Global Internet Usage - 2010

## Description

Internet usage for several countries

### **Format**

A dataset with 9 observations on the following 3 variables.

Country Name of country

PercentFastConnection Percent of internet users with a fast connection

HoursOnline Average number of hours online in February 2011

### **Details**

The Nielsen Company measured connection speeds on home computers in nine different countries. Variables include the percent of internet users with a fast connection (defined as 2Mb/sec or faster) and the average amount of time spent online, defined as total hours connected to the web from a home computer during the month of February 2011.

\*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

NielsenWire, "Swiss Lead in Speed: Comparing Global Internet Connections", April 1, 2011

GlobalInternet2019

Global Internet Usage

### **Description**

Internet usage for several countries

### **Format**

A data frame with 9 observations on the following 3 variables.

Country Name of country

InternetSpeed Average download speed (in Mb)

HoursOnline Average hours online per day

GolfRound 55

## **Details**

The Worldwide Broadband Speed League tests internet speeds at millions of access points around the world. The average download speed for each country is derived from those data. The DataReportal site provides summaries of country level data on internet usage obtained from various sources. The average number of hours spent online for each country is based on survey data reported at that site.

\*\* Updated for 3e (earlier version from 2011 is at GlobalInternet2011).

### Source

Internet speeds for 2019 downloaded from https://www.cable.co.uk/broadband/speed/worldwide-speed-league/ Online hours for 2019 downloaded from https://datareportal.com/library

**GolfRound** 

Golf Round

## **Description**

Scorecard for 18 holes of golf

#### Format

A data frame with 18 observations on the following 4 variables.

Hole Hole number (1 to 18)

Distance Length of the hole (in yards)

Par Par for the hole

Score Actual number of stokes needed in this round

## **Details**

Data come from a scorecard for one round of golf at the Potsdam Country Club. Par is the expected number of strokes a good golfer should need to complete the hole.

### Source

Personal file

56 GSWarriors2016

# Description

Data from a survey of introductory statistics students.

### **Format**

A dataset with 343 observations on the following 6 variables.

Exercise	Hours of exercise (per week)
SAT	Combined SAT scores (out of 1600)
GPA	Grade Point Average (0.00-4.00 scale)
Pulse	Pulse rate (beats per minute)
Piercings	Number of body piercings
CodedSex	0=female or 1=male

### **Details**

This is a subset of the StudentSurvey dataset where cases with missing values have been dropped and sex is coded as a 0/1 indicator variable.

## **Source**

A first day survey over several different introductory statistics classes.

GSWarriors2016	Golden State Warriors Basketball - 2016

# Description

Game log data for the Golden State Warriors basketball team in 2015-2016

### **Format**

A dataset with 82 observations on the following 33 variables.

Game	ID number for each game
Date	Date the game was played
Location	Away or Home
Орр	Opponent team
Win	Game result: L or W
FG	Field goals made

GSWarriors2019 57

FGA Field goals attempted FG3 Three-point field goals made Three-point field goals attempted FG3A FT Free throws made FTA Free throws attempted Rebounds Total rebounds OffReb Offensive rebounds Assists Number of assists Steals Number of steals Blocks Number of shots blocked Turnovers Number of turnovers Number of fouls Fouls Points Number of points scored Opponent's field goals made 0ppFG Opponent's Field goals attempted OppFGA 0ppFG3 Opponent's Three-point field goals made OppFG3A Opponent's Three-point field goals attempted 0ppFT Opponent's Free throws made OppFTA Opponent's Free throws attempted Opponent's Total rebounds OppRebounds OppOffReb Opponent's Offensive rebounds OppAssists Opponent's assists OppSteals Opponent's steals Opponent's shots blocked OppBlocks

### **Details**

Information from online boxscores for all 82 regular season games played by the Golden State Warriors basketball team during the 2015-2016 season.

Opponent's turnovers

Opponent's points scored

Opponent's fouls

\*\* From 2e - dataset has been updated for 3e \*\*

OppTurnovers

OppFouls

OppPoints

### Source

Data for the 2015-2016 Golden State games downloaded from <a href="http://www.basketball-reference.com/teams/GSW/2016/gamelog/">http://www.basketball-reference.com/teams/GSW/2016/gamelog/</a>

GSWarriors2019

Golden State Warriors Basketball (2019)

## **Description**

Game log data for the Golden State Warriors basketball team in 2018-2019

58 GSWarriors2019

### **Format**

A data frame with 82 observations on the following 33 variables.

Game ID number for each game

Date Date the game was played (mm/dd/yyy)

Location Away or Home

Opp Opponent team

Win Game result: L or W

Points Number of points scored

FG Field goals made

FGA Field goals attempted

FG3 Three-point field goals made

FG3A Three-point field goals attempted

FT Free throws made

FTA Free throws attempted

Rebounds Total rebounds

OffReb Offensive rebounds

Assists Number of assists

Steals Number of steals

Blocks Number of shots blocked

Turnovers Number of turnovers

Fouls Number of fouls

OppPoints Opponent's points scored

OppFG Opponent's field goals made

OppFGA Opponent's field goals attempted

OppFG3 Opponent's three-point field goals made

OppFG3A Opponent's three-point field goals attempted

OppFT Opponent's free throws made

OppFTA Opponent's free throws attempted

OppRebounds Opponent's total rebounds

OppOffReb Opponent's offensive rebounds

OppAssists Opponent's assists

OppSteals Opponent's steals

OppBlocks Opponent's shots blocked

OppTurnovers Opponent's turnovers

OppFouls Opponent's fouls

HappyPlanetIndex 59

### **Details**

Information from online boxscores for all 82 regular season games played by the Golden State Warriors basketball team during the 2018-2019 season.

\*\* Updated for third edition (2e version is now GSWarriors2016, 1e version is MiamiHeat dataset) \*\*

### Source

Data for the 2018-2019 Golden State games downloaded from https://www.basketball-reference.com/teams/GSW/2019/gamelog/

HappyPlanetIndex Happy Planet Index

## **Description**

Measurements related to happiness and well-being for 143 countries.

#### **Format**

A dataset with 143 observations on the following 11 variables.

Name of country Country Region 1=Latin America, 2=Western nations, 3=Middle East, 4=Sub-Saharan Africa, 5=South Asia, 6=East Asia, 7=former Communist countries Happiness Score on a 0-10 scale for average level of happiness (10 is happiest) LifeExpectancy Average life expectancy (in years) Footprint Ecological footprint - a measure of the (per capita) ecological impact Happy Life Years - combines life expectancy with well-being HLY Happy Planet Index (0-100 scale) HPI HPIRank HPI rank for the country GDPperCapita Gross Domestic Product (per capita) Human Development Index HDI Population Population (in millions)

#### **Details**

Data for 143 countries from the Happy Planet Index Project that works to quantify indicators of happiness, well-being, and ecological footprint at a country level.

### **Source**

Marks, N., "The Happy Planet Index", www.TED.com/talks, August 29, 2010. Data downloaded from http://www.happyplanetindex.org/data/

60 HeightData

HeatCognition

Heat and Cognition

## Description

Effect of heat on cognitive ability

### **Format**

A data frame with 46 observations on the following 3 variables.

AC Whether the student had air conditioning on in the room, No or Yes

MathZRT Z-score of reaction time solving math problems

ColorsZRT Z-score of reaction time solving STROOP color problems

### **Details**

Forty-six college students were asked to solve cognitive problems first thing in the morning during a heat wave in their Northeastern city. Twenty of the students had air-conditioning in their rooms and twenty-six did not. Z-scores of reaction times are given for math problems and for color dissonance problems.

### Source

Cedeo Laurent JG, Williams A, Oulhote Y, Zanobetti A, Allen JG, Spengler JD "Reduced cognitive function during a heat wave among residents of non-air-conditioned buildings: An observational study of young adults in the summer of 2016." PLoS Med 15(7): e1002605, July 10, 2018. https://journals.plos.org/plosmedicine/article?id=10.1371/journal.pmed.1002605. (Dataset is simplified from the repeated measures design used in the original study.)

HeightData

Height Data

### **Description**

Heights measured for the same 94 children over 18 years.

### **Format**

A dataset with 94 observations on the following 33 variables.

ID Identification number)

Sex Mor F

Year\_1 Height (in cm.) at age 1 year

Year\_1.25 Height (in cm.) at age 1.25 years

HockeyPenalties2011 61

```
Year_1.5 Height (in cm.) at age 1.5 years
Year_1.75 Height (in cm.) at age 1.75 years
Year_2 Height (in cm.) at age 2 years
Year_3 Height (in cm.) at age 3 years
Year_4 Height (in cm.) at age 4 years
Year_5 Height (in cm.) at age 5 years
See below for full list of years...
Year_1.5 Height (in cm.) at age 17.5 years
Year_1.8 Height (in cm.) at age 18 years
```

### **Details**

In the 1940's and 1950's, the heights of 39 boys and 54 girls, in centimeters, were measured at 30 different time points between the ages of 1 and 18 years as part of the University of California Berkeley growth study. Ages for measurement are 1, 1,25, 1,5, 1,75, 2, 3, 4, 5, 6, 7, 8, 8.5, 9, 9.5, 10, 10.5, 11, 11,5, 12, 12.5, 13, 13.5, 14, 14.5, 15, 15.5, 16, 16.5, 17, 17.5, 18.

#### Source

Tuddenham, R. D., and Snyder, M. M. (1954) "Physical growth of California boys and girls from birth to age 18", University of California Publications in Child Development, 1, 183-364.

HockeyPenalties2011 Hockey Penalties - 2011

## **Description**

Penalty minutes (per game) for NHL teams in 2010-11

### **Format**

A dataset with 30 observations on the following 2 variables.

Team Name of the team

PIMperG Average penalty minutes per game

### **Details**

Data give the average number of penalty minutes for each of the 30 National Hockey League (NHL) teams during the 2010-11 regular season.

\*\* From 2e - dataset has been updated for 3e \*\*

## Source

Data obtained online at www.nhl.com

62 HollywoodMovies

HockeyPenalties2019 Hockey Penalties (2019)

## Description

Penalty minutes (per game) for NHL teams in 2018-2019

### **Format**

A data frame with 30 observations on the following 4 variables.

Team Name of the team

PIM Average penalty minutes per game

OppPIM Average opponent's penalty minutes per game

Playoff Did the team make the playoffs? (N or Y)

### **Details**

Data give the average number of penalty minutes for each of the 30 National Hockey League (NHL) teams (and their opponents) during the 2018-2019 regular season.

\*\* Updated for 3e (earlier version from 2010-11 is at HockeyPenalties2011). \*\*

### **Source**

Data obtained online at https://www.hockey-reference.com/leagues/NHL\_2019.html#all\_stats

HollywoodMovies

Hollywood Movies

### **Description**

Data on movies released in Hollywood between 2012 and 2018

## Format

A data frame with 1295 observations on the following 15 variables.

Movie Title of the movie

LeadStudio Primary U.S. distributor of the movie

RottenTomatoes Rotten Tomatoes rating (critics)

AudienceScore Audience rating (via Rotten Tomatoes)

Genre One of Action Adventure, Black Comedy, Comedy, Concert, Documentary, Drama, Horror, Musical, Romantic Comedy, Thriller, or Western

TheatersOpenWeek Number of screens for opening weekend

OpeningWeekend Opening weekend gross (in millions)

BOAvgOpenWeekend Average box office income per theater, opening weekend

Budget Production budget (in millions)

DomesticGross Gross income for domestic (U.S.) viewers (in millions)

WorldGross Gross income for all viewers (in millions)

ForeignGross Gross income for foreign viewers (in millions)

Profitability WorldGross as a percentage of Budget

OpenProfit Percentage of budget recovered on opening weekend

Year Year the movie was released

### **Details**

Information from 1295 movies released from Hollywood between 2012 and 2018.

\*\* Updated for 3e (earlier versions are HollywoodMovies2013 and HollywoodMovies2011). \*\*

#### Source

Movie data obtained from

https://www.boxofficemojo.com/ https://www.the-numbers.com/ https://www.rottentomatoes.com/

HollywoodMovies2011

ForeignGross

Hollywood Movies in 2011

## **Description**

Data on movies released in Hollywood in 2011

#### **Format**

A dataset with 136 observations on the following 14 variables.

Movie Title of movie
LeadStudio Studio that released the movie
RottenTomatoes Rotten Tomatoes rating (reviewers)
AudienceScore Audience rating (via Rotten Tomatoes)
Story General theme - one of 21 themes
Genre Action Adventure Animation Comedy Drama Fantasy Horror Romance Thriller
TheatersOpenWeek Number of screens for opening weekend
BOAverageOpenWeek Average opening week box office income (per theater)
DomesticGross Gross income for domestic viewers (in \$ millions)

Gross income for foreign viewers (in \$ millions)

Gross income for all viewers (in \$ millions) WorldGross

Production budget (in \$ millions) Budget Profitability WorldGross as a percentage of Budget Opening weekend gross (in \$ millions) OpeningWeekend

### **Details**

Information from 136 movies released from Hollywood in 2011. \*\* This dataset has been updated for 2e with more years of data (in HollywoodMovies) \*\*

### Source

McCandless, D., "Most Profitable Hollywood Movies" from "Information is Beautiful" at http://www.informationisbeautiful,net.data/ and http://bit.ly/hollywoodbudgets.

HollywoodMovies2013

Hollywood Movies - 2013

## **Description**

Data on movies released in Hollywood between 2007 and 2013

### **Format**

A dataset with 970 observations on the following 16 variables.

Title of movie Movie Studio that released the movie LeadStudio

RottenTomatoes Rotten Tomatoes rating (reviewers) AudienceScore Audience rating (via Rotten Tomatoes) General theme - one of 21 themes Story

One of 14 possible genres Genre

Number of screens for opening weekend

TheatersOpenWeek Opening weekend gross (in \$ millions) OpeningWeekend

Average opening week box office income (per theater) BOAverageOpenWeek DomesticGross Gross income for domestic viewers (in \$ millions) ForeignGross Gross income for foreign viewers (in \$ millions) WorldGross Gross income for all viewers (in \$ millions)

Production budget (in \$ millions) Budget

WorldGross as a percentage of Budget Profitability

OpenProfit Percentage of budget recovered on opening weekend

> Year Year the movie was released

HomesForSale 65

## **Details**

```
Information from 970 movies released from Hollywood between 2007 and 2013. ** From 2e - dataset has been updated for 3e **
```

### **Source**

McCandless, D., "Most Profitable Hollywood Movies" from "Information is Beautiful" at <a href="http://www.informationisbeautiful,net.data/">http://www.informationisbeautiful,net.data/</a> and <a href="http://bit.ly/hollywoodbudgets">http://bit.ly/hollywoodbudgets</a>.

HomesForSale

Homes For Sale (2019)

## **Description**

Data on homes for sale in four states in 2019

### **Format**

A data frame with 120 observations on the following 5 variables.

```
State Location of the home (CA, NJ, NY, or PA)
```

Price Asking price (in \$1,000's)

Size Area of all rooms (in 1,000's sq. ft.)

Beds Number of bedrooms

Baths Number of bathrooms

### **Details**

Data for samples of homes for sale in each state, selected from *zillow.com*. \*\* Updated for 3e (earlier version from 2010 is in HomesForSale2e). \*\*

## Source

Data collected from https://www.zillow.com/ in 2019.

66 HomesForSaleCA

HomesForSale2e

Home for Sale - 2e

## **Description**

Data on homes for sale in four states

### **Format**

A dataset with 120 observations on the following 5 variables.

State Location of the home: CA NJ NY PA

Price Asking price (in \$1,000's)

Size Area of all rooms (in 1,000's sq. ft.)

Beds Number of bedrooms
Baths Number of bathrooms

#### **Details**

Data for samples of homes for sale in each state, selected from zillow.com.

\*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

Data collected from www.zillow.com in 2010.

HomesForSaleCA

Homes For Sale in California (2019)

## **Description**

Data for a sample of homes offered for sale in California

## **Format**

A data frame with 30 observations on the following 5 variables.

State Location of the home (CA)

Price Asking price (in \$1,000's)

Size Area of all rooms (in 1,000's sq. ft.)

Beds Number of bedrooms

Baths Number of bathrooms

HomesForSaleCA2e 67

## **Details**

Data for sample of homes for sale in California, selected from *zillow.com*. This is a subset of the HomesForSale dataset.

\*\* Updated for 3e (earlier version from 2010 is in HomesForSaleCA2e). \*\*

## Source

Data collected from https://www.zillow.com/ in 2019.

HomesForSaleCA2e

Home for Sale in California -2e

# Description

Data for a sample of homes offered for sale in California

## **Format**

A dataset with 30 observations on the following 5 variables.

State Location of the home: CA Price Asking price (in \$1,000's)

Size Area of all rooms (in 1,000's sq. ft.)

Beds Number of bedrooms
Baths Number of bathrooms

### **Details**

Data for samples of homes for sale in California, selected from zillow.com.

\*\* From 2e - dataset has been updated for 3e \*\*

## **Source**

Data collected from www.zillow.com in 2010.

68 HomesForSaleCanton2e

HomesForSaleCanton

Homes For Sale in Canton, NY (2019)

### **Description**

Data for a sample of homes offered for sale in Canton, NY

#### **Format**

A data frame with 30 observations on the following 4 variables.

Price Asking price (in \$1,000's)

Size Area of all rooms (in 1,000's sq. ft.)

Beds Number of bedrooms

Baths Number of bathrooms

## **Details**

Data for a sample of homes for sale in Canton, NY, selected from zillow.com. \*\* Updated for 3e (earlier version from 2010 is in HomesForSaleCanton2e). \*\*

### Source

Data collected from <a href="https://www.zillow.com/">https://www.zillow.com/</a> in 2019.

HomesForSaleCanton2e Homes for sale in Canton, NY - 2e

# **Description**

Prices of homes for sale in Canton, NY

### **Format**

A dataset with 10 observations on the following variable.

Asking price for the home (in \$1,000's) Price

## **Details**

Data for samples of homes for sale in Canton, NY, selected from zillow.com.

\*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

Data collected from www.zillow.com in 2010.

HomesForSaleNY 69

HomesForSaleNY	Homes For Sale in New York (2019)	

# Description

Data for a sample of homes offered for sale in New York (state)

### **Format**

A data frame with 30 observations on the following 5 variables.

State Location of the home (NY)

Price Asking price (in \$1,000's)

Size Area of all rooms (in 1,000's sq. ft.)

Beds Number of bedrooms

Baths Number of bathrooms

### **Details**

Data for a sample of homes for sale in New York, selected from *zillow.com*. This is a subset of the HomesForSale dataset.

\*\* Updated for 3e (earlier version from 2010 is in HomesForSaleNY2e). \*\*

### **Source**

Data collected from https://www.zillow.com/ in 2019.

Home for Sale in New York - 2
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## **Description**

Data for a sample of homes offered for sale in New York State

## **Format**

A dataset with 30 observations on the following 5 variables.

State Location of the home: NY
Price Asking price (in \$1,000's)
Size Area of all rooms (in 1,000's sq. ft.)

Beds Number of bedrooms
Baths Number of bathrooms

70 HomingPigeons

## **Details**

Data for samples of homes for sale in New York, selected from *zillow.com*.

\*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

Data collected from www.zillow.com in 2010.

HomingPigeons

Homing Pigeons

# Description

Results from the 2019 Midwest Classic Homing Pigeon race

### **Format**

A data frame with 1412 observations on the following 5 variables.

Position Finishing position in the race

Loft Name of the pigeon's home loft

Sex C=cock (male) or H=hen (female)

Distance Distance (in miles) from release point to home loft

Speed (in yards per minute)

## **Details**

Finishing results from 1412 pigeons completing the 2019 Midwest Classic race for homing pigeons on June 30, 2019. Each loft may enter multiple pigeons.

#### Source

Final race report from the Midwest Homing Pigeon Association, downloaded from <a href="http://www.midwesthpa.com/MIDFinalReports.htm">http://www.midwesthpa.com/MIDFinalReports.htm</a>

Honeybee 71

Honeybee C	Colonies
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# Description

Number of honeybee colonies (1995-2012)

### **Format**

A dataset with 18 observations on the following 2 variables.

```
Year Year
Colonies Estimated number of honeybee colonies in the US (in thousands)
```

### **Details**

Data collected from the USDA on the estimated number of honeybee colonies in the US for the years 1995 through 2012.

## Source

USDA National Agriculture and Statistical Services,

http://usda.mannlib.cornell.edu/MannUsda/viewDocumentInfo.do?documentID=1191 Accessed September 2015.

HoneybeeCircuits 1	Honeybee Circuits

## **Description**

Number of circuits for honeybee dances and nest quality

# **Format**

A dataset with 78 observations on the following 2 variables.

Circuits Number of waggle dance circuits for a returning scout bee Quality Quality of the nest site: High or Low

72 HoneybeeWaggle

### **Details**

When honeybees are looking for a new home, they send out scouts to explore options. When a scout returns, she does a "waggle dance" with multiple circuit repetitions to tell the swarm about the option she found. The bees then decide between the options and pick the best one. Scientists wanted to find out how honeybees decide which is the best option, so they took a swarm of honeybees to an island with only two possible options for new homes: one of very high honeybee quality and one of low quality. They then kept track of the scouts who visited each option and counted the number of waggle dance circuits each scout bee did when describing the option.

### Source

Seeley, T., Honeybee Democracy, Princeton University Press, Princeton, NJ, 2010, p. 128

HoneybeeWaggle

Honeybee Waggle

## **Description**

Honeybee dance duration and distance to nesting site

### **Format**

A dataset with 7 observations on the following 2 variables.

Distance Distance to the potential nest site (in meters)
Duration Duration of the waggle dance (in seconds)

## **Details**

When honeybee scouts find a food source or a nice site for a new home, they communicate the location to the rest of the swarm by doing a "waggle dance." They point in the direction of the site and dance longer for sites farther away. The rest of the bees use the duration of the dance to predict distance to the site.

### Source

Seeley, T., Honeybee Democracy, Princeton University Press, Princeton, NJ, 2010, p. 128

HotDogs1e 73

HotDogs1e

Hot Dog Eating Contest

### Description

Winning number of hot dogs consumed in an eating contest

#### **Format**

A dataset with 10 observations on the following 2 variables.

Year Year of the contest: 2002-2011

HotDogs Winning number of hot dogs consumed

#### **Details**

Every Fourth of July, Nathan's Famous in New York City holds a hot dog eating contest, in which contestants try to eat as many hot dogs (with buns) as possible in ten minutes. The winning number of hot dogs are given for each year from 2002-2011.

\*\* From 1e - dataset has been updated for 2e \*\*

### Source

Downloaded from https://en.wikipedia.org/wiki/Nathan's\_Hot\_Dog\_Eating\_Contest

HotDogs2015

Hot Dog Eating Contest - 2015

## Description

Winning number of hot dogs consumed in an eating contest

# **Format**

A dataset with 14 observations on the following 2 variables.

Year Year of the contest: 2002-2015
HotDogs Winning number of hot dogs consumed

### **Details**

Every Fourth of July, Nathan's Famous in New York City holds a hot dog eating contest, in which contestants try to eat as many hot dogs (with buns) as possible in ten minutes. The winning number of hot dogs are given for each year from 2002-2015.

\*\* From 2e - dataset has been updated for 3e \*\*

74 HouseStarts2015

### Source

Downloaded from https://en.wikipedia.org/wiki/Nathan's\_Hot\_Dog\_Eating\_Contest

HotDogs2019

Hot Dog Eating Contest

# Description

Winning number of hot dogs consumed in an eating contest (2002-2019)

#### **Format**

A data frame with 18 observations on the following 2 variables.

Year Year of the contest: 2002 to 2019

HotDogs Winning number of hot dogs consumed

### **Details**

Every Fourth of July, Nathan's Famous in New York City holds a hot dog eating contest, in which contestants try to eat as many hot dogs (with buns) as possible in ten minutes. The winning number of hot dogs are given for each year from 2002-2019.

\*\* Data set updated for 3e (earlier versions are HotDogs2015 and HotDogs1e) \*\*

### **Source**

Downloaded from https://en.wikipedia.org/wiki/Nathan's\_Hot\_Dog\_Eating\_Contest

HouseStarts2015

Housing Starts - 2015

# Description

Quarterly housing starts in the United States from 2000-2015

### **Format**

A dataset with 64 observations on the following 3 variables.

Year (2000 to 2015) Year

Q1=Jan-Mar, Q2=Apr-June, Q3=July-Sept, Q4=Oct-Dec Quarter New US residential house construction starts (in thousands) Houses

HouseStarts2018 75

# **Details**

Number of new homes started in the US for each quarter from 2000-2015.

\*\* From 2e - dataset has been updated for 3e \*\*

#### **Source**

Census.gov website https://www.census.gov/econ/currentdata/ https://www.census.gov/econ/currentdata/dbsearch?program=RESCONST&startYear=2000 &endYear=2016&categories=2000

HouseStarts2018

Housing Starts (2000-2018)

# **Description**

Quarterly housing starts in the United States from 2000-2018

## **Format**

A data frame with 76 observations on the following 3 variables.

Year Year (2000 to 2018)

Quarter Q1=Jan-Mar, Q2=Apr-June, Q3=July-Sept, Q4=Oct-Dec

Houses New US residential house construction starts (in thousands)

### **Details**

Number of new homes started in the US for each quarter from 2000-2018. Updated for 3e (earlier version is in HouseStarts2015)

### Source

Census.gov website https://www.census.gov/econ/currentdata/

https://www.census.gov/econ/currentdata/dbsearch?program=RESCONST&startYear=2000&endYear=2018&categories=STARTS&dataType=SINGLE&geoLevel=US&notAdjusted=1&submit=GET+DATA&releaseScheduleId=

76 HumanTears50

HumanTears25

Human Tears -Sadness and Sexual Arousal

### **Description**

Differences in sadness and sexual arousal ratings for 25 men sniffing female tears or a placebo in a matched pairs experiment.

#### **Format**

A data frame with 25 observations on the following 2 variables.

SexDiff Difference in sexual arousal rating (placebo rating - tears rating)

SadDiff Difference in sadness rating (placebo rating - tears rating)

### **Details**

Twenty-five men had a pad attached to their upper lip that contained either female tears collected from women who watched a sad film or a salt solution (as a placebo) that had been trickled down the same women's faces. The data were collected following a double-blind matched pairs design, where the order was randomized. The men were shown pictures of female faces and asked "To what extent is this face sad?" or "To what extent is this face sexually arousing?" Men's answers were input using a Visual Analog Scale, which were then converted to a scale with results between about 200 and 800. The data show the difference in rating (placebo rating minus sadness rating) for each man for the sad question (SadDiff) or the sexual arousal question (SexDiff). Data are approximated from information given in the article.

#### Source

Gelstein, S, et al., "Human Tears Contain a Chemosignal," Science, 331(6014), 226-230, January 14, 2011.

HumanTears50

Human Tears - Testosterone

# Description

Differences in testosterone levels for 50 men in a matched pairs experiment, where the differences are between sniffing female tears and sniffing a placebo

### **Format**

A data frame with 50 observations on the following 3 variables.

Placebo Testosterone level after sniffing a placebo

Tears Testosterone level after sniffing female tears

Difference Difference in testosterone level (Placebo - Tears)

Hurricanes2014 77

#### **Details**

Fifty men had a pad attached to their upper lip that contained either female tears collected from women who watched a sad film or a salt solution (as a placebo) that had been trickled down the same women's faces. The data were collected following a double-blind matched pairs design, where the order was randomized and the data were collected on consecutive days. After sniffing each substance (placebo or tears), men had their salivary testosterone levels measured, in pg/ml. Data are approximated from information given in the article.

#### Source

Gelstein, S, et al., "Human Tears Contain a Chemosignal," Science, 331(6014), 226-230, January 14, 2011.

Hurricanes2014

Hurricanes - 2014

### Description

Hurricanes making landfall on the US east coast each year (1914-2014)

### **Format**

A dataset with 64 observations on the following 3 variables.

Year Year (1914 to 2014)
Hurricanes Number of hurricanes making landfall on US East coast

## Details

Number of hurricanes making landfall on the East coast of the United States - yearly 1914-2014. \*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

Weather Underground website at https://www.wunderground.com/hurricane/hurrarchive.asp

78 ICUAdmissions

Hurricanes2018

Hurricanes (1914 to 2018)

#### **Description**

Hurricanes in the North Atlantic each year (1914-2018)

#### **Format**

A data frame with 105 observations on the following 2 variables.

Year Year (1914 to 2018)

Hurricanes Number of North Atlantic hurricanes

# **Details**

```
Number of North Atlantic hurricanes - yearly 1914-2018.

** Updated for 3e (earlier version through 2014 is in Hurricanes 2014). **
```

#### **Source**

Weather Underground website at https://www.wunderground.com/hurricane/archive

**ICUAdmissions** 

Intensive Care Unit Admissions

### **Description**

Data from patients admitted to an intensive care unit

### **Format**

A dataset with 200 observations on the following 21 variables.

ID Patient ID number Patient status: 0=lived or 1=died Status Patient's age (in years) Age 0=male or 1=female Sex Race Patient's race: 1=white, 2=black, or 3=other Service Type of service: 0=medical or 1=surgical Is cancer involved? 0=no or 1=yes Cancer Renal Is chronic renal failure involved? 0=no or 1=yes Is infection involved? 0=no or 1=yes Infection CPR Patient gets CPR prior to admission? 0=no or 1=yes Systolic Systolic blood pressure (in mm of Hg)

ImmuneTea 79

HeartRate Pulse rate (beats per minute)

Previous Previous admission to ICU within 6 months? 0=no or 1=yes

Type Admission type: 0=elective or 1=emergency Fracture Fractured bone involved? 0=no or 1=yes

PO2 Partial oxygen level from blood gases under 60? 0=no or 1=yes

PH pH from blood gas under 7.25? 0=no or 1=yes

PCO2 Partial carbon dioxide level from blood gas over 45? 0=no or 1=yes

Bicarbonate Bicarbonate from blood gas under 18? 0=no or 1=yes
Creatinine Creatinine from blood gas over 2.0? 0=no or 1=yes
Consciousness Level: 0=conscious, 1=deep stupor, or 2=coma

#### **Details**

Data from a sample of 200 patients following admission to an adult intensive care unit (ICU).

# Source

DASL dataset downloaded from http://lib.stat.cmu.edu/DASL/Datafiles/ICU.html

ImmuneTea Immune Tea

## **Description**

Interferon gamma production and tea drinking

#### **Format**

A dataset with 21 observations on the following 2 variables.

InterferonGamma Measure of interferon gamma production
Drink Type of drink: Coffee or Tea

#### Details

Eleven healthy non-tea-drinking individuals were asked to drink five or six cups of tea a day, while ten healthy non-tea and non-coffee-drinkers were asked to drink the same amount of coffee, which has caffeine but not the L-theanine that is in tea. The groups were randomly assigned. After two weeks, blood samples were exposed to an antigen and production of interferon gamma was measured.

### Source

Adapted from Kamath, et.al., "Antigens in tea-Beverage prime human V 2V2 T cells in vitro and in vivo for memory and non-memory antibacterial cytokine responses", Proceedings of the National Academy of Sciences, May 13, 2003.

**Inkjet Printers** 

### **Description**

Data from online reviews of inkjet printers

### **Format**

A dataset with 20 observations on the following 6 variables.

Model name of printer
PPM Printing rate (pages per minute) for a benchmark set of print jobs
PhotoTime Time (in seconds) to print 4x6 color photos
Price Typical retail price (in dollars)
CostBW Cost per page (in cents) for printing in black & white
CostColor Cost per page (in cents) for printing in color

#### **Details**

Information from reviews of inkjet printers at PCMag.com in August 2011.

#### **Source**

Inkjet printer reviews found at http://www.pcmag.com/reviews/printers, August 2011.

LifeExpectancyVehicles

*Life Expectancy and Vehicle Registrations (2017)* 

# **Description**

Yearly US life expectancy and number of registered vehicles (1970-2017)

## Format

A data frame with 48 observations on the following 3 variables.

Year Year (1970 to 2017)

LifeExpectancy Average life expectancy (in years) for babies born in the year

Vehicles Number of motor vehicles registered in the US (in millions)

### **Details**

Life expectancy (in years for babies born each year) and number of vehicles registered in the US for each year from 1970 to 2017.

\*\* Updated for 3e (earlier versions are LifeExpectancyVehicles2e and LifeExpectancyVehicles1e) \*\*

#### Source

Vehicle registrations from the Federal Highway Administration, https://www.fhwa.dot.gov/policyinformation/statistics.cfm.

Lifetime data from the Centers for Disease Control and Prevention, National Center for Health Statistics https://www.cdc.gov/nchs/hus/contents2019.htm?search=Life\_expectancy,.

LifeExpectancyVehicles1e

*Life Expectancy and Vehicle Registrations - 1e* 

### **Description**

Yearly US life expectancy and number of registered vehicles (1970-2009)

#### Format

A dataset with 40 observations on the following 3 variables.

Year Year

LifeExpectancy Average life expectancy (in years) for babies born in the year Vehicles Number of motor vehicles registered in the US (in millions)

# **Details**

Life expectancy (in years for babies born each year) and number of vehicles registered in the US for each year from 1970 to 2009.

\*\* From 1e - dataset has been updated for 2e \*\*

#### Source

Vehicle registrations from US Census Bureau, <a href="http://www.census.gov/compendia/statab/cats/transportation.html">http://www.census.gov/compendia/statab/cats/transportation.html</a> Lifetime data from the Centers for Disease Control and Prevention, National Center for Health Statistics, Health Data Interactive, <a href="https://www.cdc.gov/nchs/hdi.htm">www.cdc.gov/nchs/hdi.htm</a>

82 LightatNight

LifeExpectancyVehicles2e

Life Expectancy and Vehicle Registrations - 2e

### **Description**

Yearly US life expectancy and number of registered vehicles (1970-2013)

## **Format**

A dataset with 44 observations on the following 3 variables.

Year Year

LifeExpectancy Average life expectancy (in years) for babies born in the year Vehicles Number of motor vehicles registered in the US (in millions)

### **Details**

Life expectancy (in years for babies born each year) and number of vehicles registered in the US for each year from 1970 to 2013.

\*\* From 2e - dataset has been updated for 3e \*\*

### Source

Vehicle registrations from US Census Bureau, <a href="http://www.census.gov/compendia/statab/cats/transportation.html">http://www.census.gov/compendia/statab/cats/transportation.html</a> Lifetime data from the Centers for Disease Control and Prevention, National Center for Health Statistics, Health Data Interactive, <a href="https://www.cdc.gov/nchs/hdi.htm">www.cdc.gov/nchs/hdi.htm</a>

Light at Night for Mice

# Description

Data on body mass gain from an experiment with mice having different nighttime light conditions

# **Format**

A dataset with 18 observations on the following 2 variables.

Group Light=dim light at night or Dark=dark at night
BMGain Body mass gain (in grams over a three week period)

LightatNight4Weeks 83

#### **Details**

In this study, 18 mice were randomly split into two groups. One group was on a normal light/dark cycle (Dark) and the other group had light during the day and dim light at night (Light). The dim light was equivalent to having a television set on in a room. The mice in darkness ate most of their food during their active (nighttime) period, matching the behavior of mice in the wild. The mice with dim light at night, however, consumed much of their food during the well-lit rest period, when most mice are usually sleeping. The change in body mass was recorded after three weeks.

\*\* See also LightatNight4Weeks or LightatNight8Weeks for more variables measured at other points in the same experiment, with a third experimental condition which had 9 additional mice with a bright light on all the time. \*\*

#### Source

Fonken, L., et. al., "Light at night increases body mass by shifting time of food intake," Proceedings of the National Academy of Sciences, October 26, 2010; 107(43): 18664-18669.

Light at Night for Mice - After 4 Weeks

### **Description**

Data from an experiment with mice having different nighttime light conditions

#### **Format**

A dataset with 27 observations on the following 9 variables.

DM=dim light at night, LD=dark at night, or LL=bright light at night Light Body mass gain (in grams over a four week period) BMGain Blood corticosterone level (a measure of stress) Corticosterone DavPct Percent of calories eaten during the day Consumption Daily food consumption (grams) GlucoseInt Glucose intolerant? No or Yes GTT15 Glucose level in the blood 15 minutes after a glucose injection GTT120 Glucose level in the blood 120 minutes after a glucose injection Activity A measure of physical activity level

#### Details

In this study, 27 mice were randomly split into three groups. One group was on a normal light/dark cycle (LD), one group had bright light on all the time (LL), and one group had light during the day and dim light at night (DM). The dim light was equivalent to having a television set on in a room. The mice in darkness ate most of their food during their active (nighttime) period, matching the behavior of mice in the wild. The mice in both dim light and bright light, however, consumed more than half of their food during the well-lit rest period, when most mice are sleeping. Values in this dataset are recorded after four weeks in the experimental condition.

84 LightatNight8Weeks

- \*\* This dataset was named LightatNight in the first edition \*\*
- \*\* See also LightatNight8Weeks for the same data after 8 weeks or LightatNight with just BMGain after 3 weeks for the DM and LD groups. \*\*

#### Source

Fonken, L., et. al., "Light at night increases body mass by shifting time of food intake," Proceedings of the National Academy of Sciences, October 26, 2010; 107(43): 18664-18669.

Light at Night for Mice - After 8 Weeks

### **Description**

Data from an experiment with mice having different nighttime light conditions

#### **Format**

A dataset with 27 observations on the following 9 variables.

DM=dim light at night, LD=dark at night, or LL=bright light at night Light Body mass gain (in grams over an eight week period) BMGain Corticosterone Blood corticosterone level (a measure of stress) DayPct Percent of calories eaten during the day Daily food consumption (grams) Consumption GlucoseInt Glucose intolerant? No or Yes Glucose level in the blood 15 minutes after a glucose injection GTT15 GTT120 Glucose level in the blood 120 minutes after a glucose injection Activity A measure of physical activity level

### Details

In this study, 27 mice were randomly split into three groups. One group was on a normal light/dark cycle (LD), one group had bright light on all the time (LL), and one group had light during the day and dim light at night (DM). The dim light was equivalent to having a television set on in a room. The mice in darkness ate most of their food during their active (nighttime) period, matching the behavior of mice in the wild. The mice in both dim light and bright light, however, consumed more than half of their food during the well-lit rest period, when most mice are sleeping. Values in this dataset are recorded after eight weeks in the experimental condition.

\*\* See also LightatNight4Weeks for the same data after 4 weeks or LightatNight with just BMGain after 3 weeks for just the DM and LD groups. \*\*

#### Source

Fonken, L., et. al., "Light at night increases body mass by shifting time of food intake," Proceedings of the National Academy of Sciences, October 26, 2010; 107(43): 18664-18669.

MalevolentUniformsNFL 85

MalevolentUniformsNFL Malevolent Uniforms NFL

### **Description**

Perceived malevolence of uniforms and penalties for National Football League (NFL) teams

#### **Format**

A dataset with 28 observations on the following 3 variables.

NFLTeam Team name

ZPenYds Z-score for penalty yards

### **Details**

Participants with no knowledge of the teams rated the jerseys on characteristics such as timid/aggressive, nice/mean and good/bad. The averages of these responses produced a "malevolence" index with higher scores signifying impressions of more malevolent uniforms. To measure aggressiveness, the authors used the amount of penalty yards converted to z-scores and averaged for each team over the seasons from 1970-1986.

### Source

Frank and Gilovich, "The Dark Side of Self- and Social Perception: Black Uniforms and Aggression in Professional Sports", Journal of Personality and Social Psychology, Vol. 54, No. 1, 1988, p. 74-85.

MalevolentUniformsNHL Malevolent Uniforms NHL

# Description

Perceived malevolence of uniforms and penalties for National Hockey League (NHL) teams

### **Format**

A dataset with 28 observations on the following 3 variables.

NHLTeam Team name

NHL\_Malevolence Score reflecting the "malevolence" of a team's uniform

ZPenMin Z-score for penalty minutes

86 MammalLongevity

### **Details**

Participants with no knowledge of the teams rated the jerseys on characteristics such as timid/aggressive, nice/mean and good/bad. The averages of these responses produced a "malevolence" index with higher scores signifying impressions of more malevolent uniforms. To measure aggressiveness, the authors used the amount of penalty minutes converted to z-scores and averaged for each team over the seasons from 1970-1986.

### **Source**

Frank and Gilovich, "The Dark Side of Self- and Social Perception: Black Uniforms and Aggression in Professional Sports", Journal of Personality and Social Psychology, Vol. 54, No. 1, 1988, p. 74-85.

MammalLongevity

Mammal Longevity

### **Description**

Longevity and gestation period for mammals

#### **Format**

A dataset with 40 observations on the following 3 variables.

Animal Species of mammal
Gestation Time from fertilization until birth (in days)
Longevity Average lifespan (in years)

#### **Details**

Dataset with average lifespan (in years) and typical gestation period (in days) for 40 different species of mammals.

# **Source**

2010 World Almanac, pg. 292.

Manhattan Apartments 87

ManhattanApartments

Manhattan Apartment Prices (2019)

### **Description**

Apartment prices for sale in Manhattan in 2019

#### **Format**

A data frame with 20 observations on the following variable.

Rent Monthly rent (in dollars)

### **Details**

Monthly rents for a sample of 20 one-bedroom apartments in Manhattan, NY that were advertised on Craig's List in November, 2019.

#### **Source**

Apartments newly advertised on Craig's List at https://newyork.craigslist.org/, November, 2019.

ManhattanApartments2011

Manhattan Apartment Prices - 2011

# **Description**

Monthly rent for one-bedroom apartments in Manhattan, NY

#### **Format**

A dataset with 20 observations on the following variable.

Rent Montly rent in dollars

### **Details**

Monthly rents for a sample of 20 one-bedroom apartments in Manhattan, NY that were advertised on Craig's List in July, 2011.

\*\* From 2e - dataset has been updated for 3e \*\*

## Source

Apartments advertised on Craig's List at newyork.craigslist.org, July 5, 2011.

88 MastersGolf

MarriageAges

Marriage Ages

# Description

Ages for husbands and wives from marriage licenses

### **Format**

A dataset with 100 observations on the following 2 variables.

Husband Age of husband at marriage
Wife Age of wife at marriage

#### **Details**

Data from a sample of 100 marriage licenses in St. Lawrence County, NY gives the ages of husbands and wives for newly married couples.

### **Source**

Thanks to Linda Casserly, St. Lawrence County Clerk's Office

MastersGolf

Masters Golf Scores

# **Description**

Scores from the 2011 Masters golf tournament

### **Format**

A dataset with 20 observations on the following 2 variables.

First First round score (in relation to par)
Final Final four round score (in relation to par)

# **Details**

Data for a random sample of 20 golfers who made the cut at the 2011 Masters golf tournament.

### **Source**

2011 Masters tournament results at http://www.masters.com/en\_US/discover/past\_winners.html

MateChoice 89

MateChoice	Fruitfly Survival - by Mate Choice	

# **Description**

Number of fruitflies surviving depending on number of mating choices.

#### **Format**

A dataset with 50 observations on the following 3 variables.

Choice	Number of surviving larvae (out of 200) when female had a choice of mates
NoChoice	Number of surviving larvae (out of 200) when female had only one choice for a mate
Difference	Choice - NoChoice

### **Details**

In an experiment, two hundred larvae from female fruitflies that were exposed to many male fruitflies were tracked to see how many survived. This was compared to a different set of 200 larvae from females that were exposed to only one male each. Values in the dataset give how many of the 200 larvae survived. This process was replicated 50 times, so each row of the dataset corresponds to the survival counts (and difference) for one run, starting with 200 larvae of each type.

### Source

Patridge, L. (1980). "Mate choice increases a component of offspring fitness in fruit flies," Nature, 283:290-291, 1/17/80.

|--|--|--|

# **Description**

Comparing actual movements to mental imaging movements

### **Format**

A dataset with 32 observations on the following 3 variables.

Action	Treatment: Actual motions or Mental imaging motions
PreFatigue	Time (in seconds) to complete motions before fatigue
PostFatigue	Time (in seconds) to complete motions after fatigue

90 MiamiHeat

#### **Details**

In this study, participants were asked to either perform actual arm pointing motions or to mentally imagine equivalent arm pointing motions. Participants then developed muscle fatigue by holding a heavy weight out horizontally as long as they could. After becoming fatigued, they were asked to repeat the previous mental or actual motions. Eight participants were assigned to each group, and the time in seconds to complete the motions was measured before and after fatigue.

#### Source

Data approximated from summary statistics in: Demougeot L. and Papaxanthis C., "Muscle Fatigue Affects Mental Simulation of Action," The Journal of Neuroscience, July 20, 2011, 31(29):10712-10720.

MiamiHeat

Miami Heat Basketball

## **Description**

Game log data for the Miami Heat basketball team in 2010-11

#### **Format**

A dataset with 82 observations on the following 33 variables.

OppFG3A

Game ID number for each game Date Date the game was played Location Away or Home Opp Opponent team Win Game result: L or W FG Field goals made FGA Field goals attempted FG3 Three-point field goals made FG3A Three-point field goals attempted FT Free throws made FTA Free throws attempted Rebounds Total rebounds OffReb Offensive rebounds Assists Number of assists Number of steals Steals Blocks Number of shots blocked Turnovers Number of turnovers Fouls Number of fouls Points Number of points scored Opponent's field goals made 0ppFG OppFGA Opponent's Field goals attempted Opponent's Three-point field goals made OppFG3

Opponent's Three-point field goals attempted

MindsetMatters 91

0ppFT Opponent's Free throws made OppFTA Opponent's Free throws attempted OppOffReb Opponent's Offensive rebounds OppRebounds Opponent's Total rebounds OppAssists Opponent's assists OppSteals Opponent's steals OppBlocks Opponent's shots blocked OppTurnovers Opponent's turnovers OppFouls Opponent's fouls OppPoints Opponent's points scored

### **Details**

Information from online boxscores for all 82 regular season games payed by the Miami Heat basketball team during the 2010-11 season.

\*\* This is from the first edition, updated in second edition to GSWarriors dataset \*\*

### **Source**

Data for the 2010-11 Miami games downloaded from http://www.basketball-reference.com/teams/MIA/2011/gamelog/

indsetMatters Mindset Matters
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# **Description**

Data from a study of perceived exercise with maids

### **Format**

A dataset with 75 observations on the following 14 variables.

Cond	Treatment condition: 0=uninformed or 1=informed
Age	Age (in years)
Wt	Original weight (in pounds)
Wt2	Weight after 4 weeks (in pounds)
BMI	Original body mass index
BMI2	Body mass index after 4 weeks
Fat	Original body fat percentage
Fat2	Body fat percentage after 4 weeks
WHR	Original waist to hip ratio
WHR2	Waist to hip ratio after 4 weeks
Syst	Original systolic blood pressure
Syst2	Systolic blood pressure after 4 weeks
Diast	Original diastolic blood pressure

### Diast2 Diastolic blood pressure after 4 weeks

#### **Details**

In 2007 a Harvard psychologist recruited 75 female maids working in different hotels to participate in a study. She informed 41 maids (randomly chosen) that the work they do satisfies the Surgeon General's recommendations for an active lifestyle (which is true), giving them examples for how and why their work is good exercise. The other 34 maids were told nothing (uninformed). Various characteristics (weight, body mass index, ...) were recorded for each subject at the start of the experiment and again four weeks later. Maids with missing values for weight change have been removed.

#### Source

Crum, A.J. and Langer, E.J. (2007). Mind-Set Matters: Exercise and the Placebo Effect, Psychological Science, 18:165-171. Thanks to the authors for supplying the data.

MustangPrice

Mustang Prices

# **Description**

Price, age, and mileage for used Mustang cars at an internet website

### **Format**

A dataset with 25 observations on the following 3 variables.

Age Age of the car (in years)

Miles Mileage on the car (in 1,000's)

Price Asking price (in \$1,000's)

## **Details**

A statistics student, Gabe McBride, was interested in prices for used Mustang cars being offered for sale on an internet site. He sampled 25 cars from the website and recorded the age (in years), mileage (in thousands of miles) and asking price (in \$1,000's) for each car in his sample.

#### **Source**

Student project with data collected from autotrader.com in 2008.

NBAPlayers2011 93

NBAPlayers2011

NBA Players Data for 2010-11 Season

## Description

Data from the 2010-2011 regular season for 176 NBA basketball players.

### **Format**

A dataset with 176 observations on the following 25 variables.

Player Name of player Age (in years) Age Team Team name Games played (out of 82) Games Starts Games started Mins Minutes played MinPerGame Minutes per game FGMade Field goals made Field goals attempted FGAttempt FGPct Field goal percentage FG3Made Three-point field goals made FG3Attempt Three-point field goals attempted FG3Pct Three-point field goal percentage FTMade Free throws made FTAttempt Free throws attempted FTPct Free throw percentage OffRebound Offensive rebounds DefRebound Defensive rebounds Rebounds Total rebounds Assists Number of assists Steals Number of steals Blocks Number of blocked shots Turnovers Number of turnovers Fouls Number of personal fouls Points Number of points scored

#### **Details**

Data for 176 NBA basketball players from the 2010-2011 regular season. Includes all players who averaged more than 24 minutes per game.

\*\* From 1e - dataset has been updated (in (NBAPlayers2015) for 2e \*\*

## Source

Data downloaded from http://www.basketball-reference.com/leagues/NBA\_2011\_stats.html

94 NBAPlayers2015

NBAPlayers2015

NBA Players Data for 2014-15 Season

### **Description**

Data from the 2014-2015 regular season for 182 NBA basketball players.

#### **Format**

A dataset with 182 observations on the following 25 variables.

```
Name of player
  Position
             PG=point guard, SG=shooting guard, PF=power forward, SF=small forward, C=center
             Age (in years)
       Age
      Team
            Team name
     Games
             Games played (out of 82)
    Starts
             Games started
      Mins Minutes played
MinPerGame
             Minutes per game
    FGMade
             Field goals made
FGAttempt Field goals attempted
     FGPct
            Field goal percentage
   FG3Made
             Three-point field goals made
             Three-point field goals attempted
FG3Attempt
    FG3Pct Three-point field goal percentage
    FTMade Free throws made
FTAttempt Free throws attempted
     FTPct
            Free throw percentage
OffRebound Offensive rebounds
DefRebound Defensive rebounds
  Rebounds Total rebounds
   Assists
             Number of assists
    Steals Number of steals
    Blocks Number of blocked shots
 Turnovers Number of turnovers
     Fouls Number of personal fouls
    Points Number of points scored
```

### **Details**

Data for 182 NBA basketball players from the 2014-2015 regular season. Includes all players who averaged more than 24 minutes per game that season.

\*\* From 2e - dataset has been updated for 3e \*\*

### Source

http://www.basketball-reference.com/leagues/NBA\_2015\_stats.html

NBAPlayers2019 95

NBAPlayers2019

NBA Players Data for 2018-19 Season

# Description

Data from the 2018-2019 regular season for 193 NBA basketball players.

#### **Format**

A data frame with 193 observations on the following 26 variables.

Player Name of player

Pos PG=point guard, SG=shooting guard, PF=power forward, SF=small forward, C=center

Age (in years)

Team Team name

Games Games played (out of 82)

Starts Games started

Mins Minutes played

MinPerGame Minutes per game

FGMade Field goals made

FGAttempt Field goals attempted

FGPct Field goal percentage

FG3Made Three-point field goals made

FG3Attempt Three-point field goals attempted

FG3Pct Three-point field goal percentage

FTMade Free throws made

FTAttempt Free throws attempted

FTPct Free throw percentage

OffRebound Offensive rebounds

DefRebound Defensive rebounds

Rebounds Total rebounds

Assists Number of assists

Steals Number of steals

Blocks Number of blocked shots

Turnovers Number of turnovers

Fouls Number of personal fouls

Points Number of points scored

96 NBAStandings2011

### **Details**

Data for 193 NBA basketball players from the 2018-2019 regular season. Includes all players who averaged more than 24 minutes per game that season.

\*\* Data set updated for 3e (earlier versions are NBAPlayers2015 and NBAPlayers2011). \*\*

### **Source**

https://www.basketball-reference.com/leagues/NBA\_2019\_totals.html

NBAStandings2011

NBA 2010-11 Regular Season Standings

# **Description**

Won-Loss record and statistics for NBA Teams in 2010-2011

#### **Format**

A dataset with 30 observations on the following 6 variables.

Team Team name

Wins Number of wins in an 82 game regular season

Losses Number of losses

WinPct Proportion of games won

PtsFor Average points scored per game
PtsAgainst Average points allowed per game

#### **Details**

Won-Loss record and regular season statistics for 30 teams in the National Basketball Association for the 2010-2011 season.

\*\* From 1e - dataset has been updated for 2e and 3e\*\*

# **Source**

Data downloaded from http://www.basketball-reference.com/leagues/NBA\_2011\_games.html

NBAStandings2016 97

MRAStan	dings2016	
INDA S LAH	UTHESZELO	

NBA 2015-2016 Regular Season Standings

### **Description**

Won-Loss record and statistics for NBA Teams in 2015-2016

#### **Format**

A dataset with 30 observations on the following 6 variables.

Team Team name

Wins Number of wins in an 82 game regular season

Losses Number of losses

WinPct Proportion of games won

PtsFor Average points scored per game
PtsAgainst Average points allowed per game

#### **Details**

Won-Loss record and regular season statistics for 30 teams in the National Basketball Association for the 2015-2016 season.

\*\* From 2e - dataset has been updated for 3e \*\*

### Source

Data downloaded from http://www.basketball-reference.com/leagues/NBA\_2016\_games.html

NBAStandings2019

NBA 2018-2019 Regular Season Standings

## **Description**

Won-Loss record and statistics for NBA Teams in 2018-2019

#### **Format**

A data frame with 30 observations on the following 6 variables.

Team Team name

Wins Number of wins in an 82 game regular season

Losses Number of losses

WinPct Proportion of games won

PtsFor Average points scored per game

PtsAgainst Average points allowed per game

98 NFLContracts2015

### **Details**

Won-Loss record and regular season statistics for 30 teams in the National Basketball Association for the 2018-2019 season.

\*\* Data set updated for 3e (earlier version are NBAStandings2016 and NBAStandings1e) \*\*

#### Source

Data downloaded from http://www.basketball-reference.com/leagues/NBA\_2019\_games.html

NFLContracts2015

NFL Contracts in 2015

### **Description**

Dollar size of contracts for all NFL players in 2015

#### **Format**

A dataset with 2099 observations on the following 5 variables.

Player Player's name

Position Code for the primary position of the player (QB=quarterback, etc.)

Team Nickname of the team

TotalMoney Total value of the contract (in millions of dollars)
YearlySalary Salary (in millions of dollars) for the 2015 season

### **Details**

This dataset contains salary information for all National Football League (NFL) players under contract for the 2015 season. Many contracts extend over multiple years, so TotalMoney gives the overall size of the contract and YearlySalary indicates how much of that is to be paid for the 2015 season. All amounts are in millions of dollars.

\*\* From 2e - dataset has been updated for 3e \*\*

### Source

Contract data collected from http://OverTheCap.com, accessed September 16, 2015.

NFLContracts2019 99

NFLContracts2019	NFL Contracts in 2019
IN ECONTINUE COLOR	

# **Description**

Dollar size of contracts for all NFL players in 2019

### **Format**

A data frame with 1988 observations on the following 5 variables.

Player Player's name

Position Code for the primary position of the player (QB=quarterback, etc.)

Team Nickname of the team

Total Money Total value of the contract (in millions of dollars)

YearlySalary Salary (in millions of dollars) for the 2019 season

#### **Details**

This dataset contains salary information for all National Football League (NFL) players under contract for the 2019 season. Many contracts extend over multiple years, so TotalMoney gives the overall size of the contract and YearlySalary indicates how much of that is to be paid for the 2019 season. All amounts are in millions of dollars.

\*\* Updated for 3e (earlier version is NFLContracts2015). \*\*

#### Source

Contract data collected from https://overthecap.com, accessed September, 2019.

NFLPreSeason2014	Wins for NFL Teams (2005-2014)	
------------------	--------------------------------	--

# **Description**

Number of preseason and regular season wins for NFL teams, each year from 2005 to 2014.

### Format

A dataset with 320 observations on the following 4 variables.

Team Code for one of 32 NFL teams Season Year between 2005 and 2014

Preseason Number of preseason wins (out of 4 games)
RegularWins Number of regular season wins (out of 16 games)

NFLPreseason2019

#### **Details**

Number of wins in the preseason (out of 4 preseason games) and regular season (out of 16 regular season games) for each of the 32 National Football (NFL) teams over a ten year period from 2005 to 2014

\*\* From 2e - dataset has been updated for 3e \*\*

#### Source

Data available at http://www.pro-football-reference.com/.

NFLPreseason2019

Wins for NFL Teams (2005-2019)

# **Description**

Number of preseason and regular season wins for NFL teams, each year from 2005 to 2019.

#### **Format**

A data frame with 480 observations on the following 4 variables.

Team Code for one of 32 NFL teams

Season Year between 2005 and 2019

Preseason Number of preseason wins (out of 4 games)

RegularWins Number of regular season wins (out of 16 games)

## **Details**

Number of wins in the preseason (out of 4 preseason games) and regular season (out of 16 regular season games) for each of the 32 National Football (NFL) teams over a fifteen year period from 2005 to 2019.

\*\* Updated for 3e (earlier version is now NFLPreseason2014). \*\*

#### Source

Data available at https://www.pro-football-reference.com/.

NFLScores2011 101

NFLScores2011

NFL Game Scores in 2011

# **Description**

Results for all NFL games for the 2011 regular season

### **Format**

A dataset with 256 observations on the following 11 variables.

Week Week of the season (1 through 17) HomeTeam Home team name AwayTeam Visiting team name HomeScore Points scored by the home team AwayScore Points scored by the visiting team HomeYards Yards gained by the home team AwayYards Yards gained by the visiting team HomeT0 Turnovers lost by the home team Turnovers lost by the visiting team AwayT0 Date Of the game Day Day of the week: Mon, Sat, Sun, or Thu

### **Details**

Data for all 256 regular season games in the National Football League (NFL) for the 2011 season. \*\* From 2e - dataset has been updated for 3e \*\*

### **Source**

NFL scores and game statistics found at <a href="http://www.pro-football-reference.com/years/2011/games.htm">http://www.pro-football-reference.com/years/2011/games.htm</a>.

NFLScores2018

NFL Scores in 2018

# **Description**

Results for all NFL games for the 2018 regular season

102 NHANES

#### **Format**

A data frame with 256 observations on the following 11 variables.

Week Week of the season (1 through 17)

HomeTeam Home team name

AwayTeam Visiting team name

HomeScore Points scored by the home team

AwayScore Points scored by the visiting team

HomeYards Yards gained by the home team

AwayYards Yards gained by the visiting team

HomeTO Turnovers lost by the home team

AwayT0 Turnovers lost by the visiting team

Date Date of the game

Day Day of the week (Mon, Sat, Sun, or Thu)

#### **Details**

Data for all 256 regular season games in the National Football League (NFL) for the 2018 season. \*\* Updated for 3e (earlier version is NFLScores2011). \*\*

# Source

NFL scores and game statistics found at https://www.pro-football-reference.com/years/2018/games.htm.

**NHANES** 

National Health and Nutrition Examination Survey (NHANES) Subset

# **Description**

A subset of the 2009-2010 National Health and Nutrition Examination Survey (NHANES).

### **Format**

A data frame with 4716 observations on the following 5 variables.

Case Case ID number

Organic Buy any food labeled organic (past 30 days)? (No or Yes)

Health Self-rating of health (Excellent, Very good, Fair, Good, or Poor)

HealthBinary Health with two categories: Poor / Fair / Good or Very good / Excellent

Income Monthly income? (dollars)

NutritionStudy 103

#### **Details**

This dataset is a subset of the 2009-2010 National Health and Nutrition Examination Survey (NHANES). NHANES is a national survey conducted by the Centers for Disease Control and Prevention (CDC) on a random sample of Americans. This subset contains data on select variables for the subset of people with responses to the questions about buying organic food and self-reported health status.

#### Source

The data were downloaded from https://www.cdc.gov/nchs/nhanes/index.htm.

### **Description**

Variables related to nutrition and health for 315 individuals

#### **Format**

A dataset with 315 observations on the following 17 variables.

```
TD
                 ID number for each subject in this sample
           Age
                 Subject's age (in years)
                 Smoker? coded as No or Yes
        Smoke
                Weight/(Height^2)
     Quetelet
      Vitamin Vitamin use: coded as 1=Regularly, 2=Occasionally, or 3=No
     Calories Number of calories consumed per day
           Fat Grams of fat consumed per day
        Fiber
                 Grams of fiber consumed per day
      Alcohol Number of alcoholic drinks consumed per week
  Cholesterol
                Cholesterol consumed (mg per day)
     BetaDiet Dietary beta-carotene consumed (mcg per day)
  RetinolDiet Dietary retinol consumed (mcg per day)
   BetaPlasma
                 Plasma beta-carotene (ng/ml)
RetinolPlasma
                 Plasma retinol (ng/ml)
                 Coded as Female or Male
           Sex
                 Coded as No Occasional Regular
   VitaminUse
   PriorSmoke
                 Smoking status: coded as 1=Never, 2=Former, or 3=Current
```

### **Details**

Data from a cross-sectional study to investigate the relationship between personal characteristics and dietary factors, and plasma concentrations of retinol, beta-carotene and other carotenoids. Study subjects were patients who had an elective surgical procedure during a three-year period to biopsy or remove a lesion of the lung, colon, breast, skin, ovary or uterus that was found to be non-cancerous.

### Source

Nierenberg, Stukel, Baron, Dain, and Greenberg, "Determinants of plasma levels of beta-carotene and retinol", American Journal of Epidemiology (1989). Data downloaded from <a href="http://lib.stat.cmu.edu/datasets/Plasma\_Retinol">http://lib.stat.cmu.edu/datasets/Plasma\_Retinol</a>.

OlympicMarathon2008 2008 Olympic Men's Marathon

# Description

Times for all finishers in the men's marathon at the 2008 Olympics

### **Format**

A data frame with 76 observations on the following 5 variables.

Rank Order of finish
Athlete Name of marathoner
Nationality Country of marathoner
Time Time as H:MM:SS
Minutes Time in minutes

#### **Details**

Results for all finishers in the 2008 Men's Olympic marathon in Beijing, China. \*\* This 1e version has been updated for 2e and 3e\*\*

# Source

http://2008olympics.runnersworld.com/2008/08/mens-marathon-results.html

OlympicMarathon2012 2012 Olympic Men's Marathon

# **Description**

Times for all finishers in the men's marathon at the 2012 Olympics

OlympicMarathon2016

#### 105

# **Format**

A data frame with 85 observations on the following 4 variables.

Athlete Name of marathoner

Country Nationality of marathoner (3 letter country code)

Time Time as H:MM:SS
Minutes Time in minutes

#### **Details**

Results for all finishers in the 2012 Men's Olympic marathon in London, England.

\*\* From 2e - dataset has been updated for 3e \*\*

#### Source

http://www.olympic.org/olympic-results/london-2012/athletics/marathon-m, accessed October 2015.

OlympicMarathon2016

2016 Olympic Men's Marathon

# Description

Times for all finishers in the men's marathon at the 2016 Olympics

#### **Format**

A data frame with 140 observations on the following 4 variables.

Athlete Name of marathoner

Country Nationality of marathoner (3 letter country code)

Time Time as H:MM:SS

Minutes Time in minutes

#### **Details**

Results for all finishers in the 2016 Men's Olympic marathon in Rio de Janeiro, Brazil.

\*\* Updated for 3e (earlier versions are now in OlympicMarathon2012 and OlympicMarathon2008)

\*\*

### Source

https://olympics.com/en/olympic-games/rio-2016/results/athletics/marathon-men

106 OttawaSenators

OrganicEffect Eating Organic Foods	fect Eating Organic Foods	OrganicEffect	OrganicEffect Eating Organic Foods

### **Description**

Data comparing pesticide levels in family members when eating non-organic vs organic food

### **Format**

A dataset with 160 observations on the following 6 variables.

Person	Code for family member, Father, Mother, GirlA, GirlB, Boy	
Pesticide	One of eight different pesticides measured	
Day	Day of the measurement (Day1, Day3, Day4, or Day6)	
NonOrganic	Level of the pesticide after eating a non-organic diet	
Organic	Level of the pesticide after eating an organic diet	
Diff	Difference = NonOrganic - Organic	

### **Details**

A study looked at a Swedish family that ate a conventional diet (non-organic), and then had them eat only organic for two weeks. Pesticide concentrations for several different pesticides were measured in micrograms/g creatinine by testing morning urine. Multiple measurements were taken for each person before the switch to organic foods, and then again after participants had been eating organic for at least one week.

#### **Source**

Magner, J., Wallberg, P., Sandberg, J., and Cousins, A.P. (2015). "Human exposure to pesticides from food: A pilot study," IVL Swedish Environmental Research Institute. https://www.coop.se/PageFiles/429812/Coop%20Ekoeffekten\_Report%20ENG.pdf, January 2015

Ottawa Senators Hockey Team (2014-2015)	OttawaSenators	Ottawa Senators Hockey Team (2014-2015)	
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# **Description**

Data for 24 players on the 2014-2105 Ottawa Senators NHL team

OttawaSenators2010 107

#### **Format**

A dataset with 24 observations on the following 10 variables.

Player Players name Position D=defense, C=center, RW=right wing, LW=left wing Age (in years) Age Games played in the 2014-15 NHL season (out of 82) Games Goals Goals Assists Assists Points Goals + Assists PlusMinus Difference between (even strength) goals for and against while on ice PenMins Number of penalty minutes

MinPerGame Average minutes on the ice per game

#### **Details**

Data for all players (except goalies) who played at least 10 games with the Ottawa Senators hockey team in the 2014-15 NHL season.

\*\* This is an updated version (previous version is now in OttawaSenators1e) \*\*

#### Source

http://www.hockey-reference.com/teams/OTT/2015.html, accessed October 2015.

OttawaSenators2010 Ottawa Senators Hockey Team - 2010

# **Description**

Data for 24 players on the 2009-10 Ottawa Senators

### **Format**

A dataset with 24 observations on the following 2 variables.

Number of points (goals + assists) scored PenMins Number of penalty minutes

#### **Details**

Points scored and penalty minutes for 24 players (excluding goalies) playing ice hockey for the Ottawa Senators during the 2009-10 NHL regular season.

\*\* From 1e - dataset has been updated for 2e and 3e \*\*

#### Source

Data obtained from http://senators.nhl.com/club/stats.htm.

108 PASeniors

OttawaSenators2019

Ottawa Senators Hockey Team (2018-2019)

# Description

Data for 26 players on the 2018-2109 Ottawa Senators NHL team

#### **Format**

A data frame with 26 observations on the following 10 variables.

Player Players name

Position D=defense, C=center, RW=right wing, LW=left wing

Age (in years)

Games Games played in the 2018-19 NHL season (out of 82)

Goals Goals

Assists Assists

Points Goals + Assists

PlusMinus Difference between (even strength) goals for and against while on ice

PenMins Number of penalty minutes

MinPerGame Average minutes on the ice per game

## **Details**

Data for all players (except goalies) who played at least 10 games with the Ottawa Senators hockey team in the 2018-2019 NHL season.

\*\* Updated for 3e (previous versions are now OttawaSenators2015 and OttawaSenators1e) \*\*

### **Source**

https://www.hockey-reference.com/teams/OTT/2019.html

**PASeniors** 

Pennsylvania High School Seniors

# **Description**

Information on a sample of high school seniors from the state of Pennsylvania between 2010 and 2019.

PASeniors 109

#### **Format**

A data frame with 457 observations on the following 36 variables.

Year Year student submitted data

Gender Female or Male

Age Age (in years)

Hand Dominant hand (Left, Right, or Both)

Height Height (in cm)

Foot Foot length (in cm)

Armspan (in cm)

Languages Languages spoken

GetToSchool Main mode of transportation to school (Bus, Car, or Walk - Walk includes bicycle)

TravelTime Travel time to school (in minutes)

ReactionTime Time (in seconds) to click when a color changes

MemoryScore Score in an online memory game

Activity Favorite physical activity

Music Favorite genre of music

BirthMonth Birth month

Season Favorite season

Allergies Have allergies? (No or Yes)

Vegetarian Vegetarian? (No or Yes)

FavFood Favorite food

Drink Beverage used most often during the day

FavSubject Favorite subject in school

Sleep1 Typical hours of sleep on a school night

Sleep2 Typical hours of sleep on a non-school night

Occupants Number of occupants at home

Communicate Most often method to communicate with friends

TextsSent Number of texts sent (previous day)

HangHours Hours last week spent hanging out with friends

HWHours Hours last week spent doing homework

SportsHours Hours last week spent playing sports or outdoor activities

VideoGameHours Hours last week spent playing computer/video games

ComputerHours Hours last week spent using a computer

TVHours Hours last week spent watching TV

WorkHours Hours last week spent working at a paid job

SchoolPressure Amount of pressure due to schoolwork

Superpower Most desired superpower (Fly, Freeze time, Invisibility, Super strength, or Telepathy)

Preference Prefers to be Famous, Happy, Healthy, or Rich

PumpkinBeer PumpkinBeer

## **Details**

The dataset gives responses for a random sample of high school seniors in Pennsylvania who participated in the Census at Schools project.

## **Source**

Data from U.S. Census at School (https://ww2.amstat.org/censusatschool/) downloaded and used with the permission of the American Statistical Association.

PizzaGirl

Pizza Girl Tips

# **Description**

Data on tips for pizza deliveries

## **Format**

A dataset with 24 observations on the following 2 variables.

Tip Amount of tip (in dollars)
Shift Data collected over three different shifts

#### **Details**

"Pizza Girl" collected data on her deliveries and tips over three different evening shifts.

# Source

Pizza Girl: Statistical Analysis at

http://slice.seriouseats.com/archives/2010/04/statistical-analysis-of-a-pizza-delivery-shift-20100429.html.

PumpkinBeer Pumpkin Beer

# **Description**

Ratings of different kinds of pumpkin beer by a wife and husband

QuizPulse10

## **Format**

A data frame with 18 observations on the following 8 variables.

Name Name of pumpkin beer

Brewer Name of brewery that produced the beer

WifeRating Rating on a 0-10 scale by the wife

HusbandRating Rating on a 0-10 scale by the husband

WifeComments Text of comments by the wife

HusbandComments Text of comments by the husband

Average Average of the two ratings (wife and husband)

Year Year the ratings were done (2011 to 2019)

#### **Details**

A Lock wife and husband are fans of pumpkin flavored beer, so they have each rated a variety of different brands of pumpkin beer over the years.

#### **Source**

Personal records

OuizPulse10

Quiz vs Lecture Pulse Rates

# **Description**

Paired data with pulse rates in a lecture and during a quiz for 10 students

#### **Format**

A dataset with 10 observations on the following 3 variables.

Student ID number for the student

Quiz Pulse rate (beats per minute) during a quiz Lecture Pulse rate (beats per minute) during a lecture

## **Details**

Ten students in an introductory statistics class measured their pulse rate (beats per minute) in two settings: first, in the middle of a regular class lecture and second, while taking an in-class quiz.

## Source

In-class data collection

112 RestaurantTips

RandomP50N200	Simulated proportions	
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# Description

Counts and proportions for 5000 simulated samples with n=200 and p=0.50

#### **Format**

A dataset with 5000 observations on the following two variables

Count Number of simulated "yes" responses in 200 trials
Phat Sample proportion (Count/200)

#### **Details**

Results from 5000 simulations of samples of size n=200 from a population with proportion of "yes" responses at p=0.50.

#### Source

Computer simulation

# **Description**

Tip data from the First Crush Bistro

## **Format**

A dataset with 157 observations on the following 7 variables.

```
Bill Size of the bill (in dollars)
Tip Size of the tip (in dollars)
Credit Paid with a credit card? n or y
Guests Number of people in the group
Day of the week: m=Monday, t=Tuesday, w=Wednesday, th=Thursday, or f=Friday
Server Code for specific waiter/waitress: A, B, or C
PctTip Tip as a percentage of the bill
```

RetailSales 113

## **Details**

The owner of a bistro called First Crush in Potsdam, NY was interested in studying the tipping patterns of his customers. He collected restaurant bills over a two week period that he believes provide a good sample of his customers. The data recorded from 157 bills include the amount of the bill, size of the tip, percentage tip, number of customers in the group, whether or not a credit card was used, day of the week, and a coded identity of the server.

## **Source**

Thanks to Tom DeRosa at First Crush for providing the tipping data.

RetailSales

Retail Sales (2009-2019)

# **Description**

Monthly U.S. Retail Sales from 2009 to 2019

# Format

A data frame with 129 observations on the following 3 variables.

Month (Jan through Dec)

Year Years from 2009 to 2019

Sales Monthly U.S. retail sales (in billions of dollars)

# **Details**

Data show the monthly retail sales (in billions) for the U.S. economy in each month from January 2009 through September 2019.

\*\* Updated for 3e (earlier versions are RetailSales2e and RetailSales1e). \*\*

## **Source**

Data downloaded from https://www.census.gov/retail/.

114 RockandRoll2012

RetailSales2011 Retail Sales (2000-2011)

# Description

Monthly U.S. Retail Sales

#### **Format**

A dataset with 136 observations on the following 3 variables.

Month Month of the year Year Years from 2000 to 2011

Sales U.S. retail sales (in billions of dollars)

## **Details**

Data show the monthly retail sales (in billions) for the U.S. economy in each month from January 2000 through April 2011.

\*\* From 1e - dataset has been updated for 2e and 3e \*\*

## **Source**

Data downloaded from http://www.census.gov/retail/

RockandRoll2012 Rock & Roll Hall of Fame (2012)

# Description

Groups and Individuals in the Rock and Roll Hall of Fame (2012)

# Format

A dataset with 273 observations on the following 4 variables.

Inductee Name of the group or individual

FemaleMembers Yes if individual or member of the group is female, otherwise No

Category Type of individual or group: Performer, Non-performer, Early Influence,

Lifetime Achievement, Sideman

People Number of people in the group

RockandRoll2015

## **Details**

```
All inductees of the Rock & Roll Hall of Fame as of 2012. ** From 1e - dataset has been updated for 2e and 3e **
```

# **Source**

Rock & Roll Hall of Fame website, http://rockhall.com/inductees/alphabetical/

RockandRoll2015

Rock & Roll Hall of Fame (2015)

# **Description**

Groups and Individuals in the Rock and Roll Hall of Fame (2015)

## **Format**

A dataset with 303 observations on the following 4 variables.

Inductee Name of the group or individual

FemaleMembers Yes if individual or member of the group is female, otherwise No

Category Type of individual or group: Performer, Non-performer, Early Influence,

Lifetime Achievement, Sideman

People Number of people in the group

# **Details**

All inductees of the Rock & Roll Hall of Fame as of 2015.

\*\* From 2e - dataset has been updated for 3e \*\*

## Source

Rock & Roll Hall of Fame website, http://rockhall.com/inductees/alphabetical/

SalaryGender SalaryGender

RockandRoll2019

Rock & Roll Hall of Fame (2019)

# **Description**

Groups and Individuals in the Rock and Roll Hall of Fame as of 2019

## **Format**

A data frame with 329 observations on the following 4 variables.

Inductee Name of the group or individual

FemaleMembers Yes if individual or member of the group is female, otherwise No

Category Type of individual or group: Early Influence, Lifetime Achievement, Non-performer, Performer, or Sideman

People Number of people in the group

## **Details**

All inductees of the Rock & Roll Hall of Fame as of 2019.

\*\* Updated for 3e (earlier versions are now RockandRoll2015 and RockandRoll1e) \*\*

# Source

Rock & Roll Hall of Fame website, https://www.rockhall.com/inductees/a-z

SalaryGender

Salary and Gender

## **Description**

Salaries for college teachers

## **Format**

A dataset with 100 observations on the following 4 variables.

Salary Annual salary in \$1,000's Gender 0=female or 1=male

Age Age in years

PhD 1=have PhD or 0=no PhD

SampColleges 117

#### **Details**

A random sample of college teachers taken from the 2010 American Community Survey (ACS) 1-year Public Use Microdata Sample (PUMS).

#### Source

Downloaded from https://www.census.gov/programs-surveys/acs/data/pums.html

SampColleges

Sample of US Post-secondary Schools

## **Description**

Information for a sample of 50 US post-secondary schools from the Department of Education's College Scorecard

## **Format**

A data frame with 50 observations on the following 37 variables.

Name Name of the school

State State where school is located

ID ID number for school

Main Main campus? (1=yes, 0=branch campus)

Accred Accreditation agency

MainDegree Predominant undergrad degree (0=not classified, 1=certificate, 2=associate, 3=bachelors,4=only graduate)

HighDegree Highest degree (0=no degrees, 1=certificate, 2=associate, 3=bachelors, 4= graduate)

Control Control of school (Private, Profit, Public)

Region Region of country (Midwest, Northeast, Southeast, Territory, West)

Locale (City, Rural, Suburb, Town)

Latitude Latitude

Longitude Longitude

AdmitRate Admission rate

MidACT Median of ACT scores

AvgSAT Average combined SAT scores

Online Only online (distance) programs

Enrollment Undergraduate enrollment

White Percent of undergraduates who report being white

Black Percent of undergraduates who report being black

Hispanic Percent of undergraduates who report being Hispanic

118 SampColleges2yr

Asian Percent of undergraduates who report being Asian

Other Percent of undergraduates who don't report one of the above

PartTime Percent of undergraduates who are part-time students

NetPrice Average net price (cost minus aid)

Cost Average total cost for tuition, room, board, etc.

TuitionIn In-state tuition and fees

TuitonOut Out-of-state tuition and fees

TuitionFTE Net Tuition revenue per FTE student

InstructFTE Instructional spending per FTE student

FacSalary Average monthly salary for full-time faculty

FullTimeFac Percent of faculty that are full-time

Pell Percent of students receiving Pell grants

CompRate Completion rate (percent who finish program within 150% of normal time)

Debt Average debt for students who complete program

Female Percent of female students

FirstGen Percent of first-generation students

MedIncome Median family income (in \$1,000)

## **Details**

The US Department of Education maintains a database through its College Scorecard project of demographic information from all active postsecondary educational institutions that participate in Title IV. This dataset contains information from a sample of the 50 schools selected from CollegeScores.

#### Source

Data downloaded from the US Department of Education's College Scorecard at https://collegescorecard.ed.gov/data/(November 2019)

SampColleges2yr

Sample of College Scorecard - Two Year

# Description

Information for a sample of 50 US post-secondary schools that primarily grant associate's degrees, from the Department of Education's College Scorecard

SampColleges2yr 119

#### **Format**

A data frame with 50 observations on the following 31 variables.

Name Name of the school

State State where school is located

ID ID number for school

Main Main campus? (1=yes, 0=branch campus)

Accred Accreditation agency

MainDegree Predominant undergrad degree (0=not classified, 1=certificate, 2=associate, 3=bachelors,4=only graduate)

HighDegree Highest degree (0=no degrees, 1=certificate, 2=associate, 3=bachelors, 4= graduate)

Control Control of school (Private, Profit, Public)

Region Region of country (Midwest, Northeast, Southeast, Territory, West)

Locale Locale (City, Rural, Suburb, Town)

Enrollment Undergraduate enrollment

White Percent of undergraduates who report being white

Black Percent of undergraduates who report being black

Hispanic Percent of undergraduates who report being Hispanic

Asian Percent of undergraduates who report being Asian

Other Percent of undergraduates who don't report one of the above

PartTime Percent of undergraduates who are part-time students

NetPrice Average net price (cost minus aid)

Cost Average total cost for tuition, room, board, etc.

TuitionIn In-state tuition and fees

TuitonOut Out-of-state tuition and fees

TuitionFTE Net Tuition revenue per FTE student

InstructFTE Instructional spending per FTE student

FacSalary Average monthly salary for full-time faculty

FullTimeFac Percent of faculty that are full-time

Pell Percent of students receiving Pell grants

CompRate Completion rate (percent who finish program within 150% of normal time)

Debt Average debt for students who complete program

Female Percent of female students

FirstGen Percent of first-generation students

MedIncome Median family income (in \$1,000)

#### **Details**

Details The US Department of Education maintains a database through its College Scorecard project of demographic information from all active postsecondary educational institutions that participate in Title IV. This dataset contains information from a sample of the two-year colleges selected from all two-year colleges in CollegeScores2yr.

120 SampColleges4yr

#### Source

Data downloaded from the US Department of Education's College Scorecard at https://collegescorecard.ed.gov/data/(November 2019)

SampColleges4yr

Sample of College Scorecard - Four Year

# Description

Information on a sample of 50 US four-year colleges and universities from the Department of Education's College Scoreboard

#### **Format**

A data frame with 50 observations on the following 37 variables.

Name Name of the school

State State where school is located

ID ID number for school

Main Main campus? (1=yes, 0=branch campus)

Accred Accreditation agency

MainDegree Predominant undergrad degree (3=bachelors)

HighDegree Highest degree (0=no degrees, 1=certificate, 2=associate, 3=bachelors, 4= graduate)

Control Control of school (Private, Profit, Public)

Region Region of country (Midwest, Northeast, Southeast, Territory, West)

Locale (City, Rural, Suburb, Town)

Latitude Latitude

Longitude Longitude

AdmitRate Admission rate

MidACT Median of ACT scores

AvgSAT Average combined SAT scores

Online Only online (distance) programs

Enrollment Undergraduate enrollment

White Percent of undergraduates who report being white

Black Percent of undergraduates who report being black

Hispanic Percent of undergraduates who report being Hispanic

Asian Percent of undergraduates who report being Asian

Other Percent of undergraduates who don't report one of the above

PartTime Percent of undergraduates who are part-time students

NetPrice Average net price (cost minus aid)

SampCountries 121

Cost Average total cost for tuition, room, board, etc.

TuitionIn In-state tuition and fees

TuitonOut Out-of-state tuition and fees

TuitionFTE Net Tuition revenue per FTE student

InstructFTE Instructional spending per FTE student

FacSalary Average monthly salary for full-time faculty

FullTimeFac Percent of faculty that are full-time

Pell Percent of students receiving Pell grants

CompRate Completion rate (percent who finish program within 150% of normal time)

Debt Average debt for students who complete program

Female Percent of female students

FirstGen Percent of first-generation students

MedIncome Median family income (in \$1,000)

#### **Details**

The US Department of Education maintains a database through its College Scorecard project of demographic information from all active postsecondary educational institutions that participate in Title IV. This dataset contains information from a sample of the four-year colleges and universities selected from all four-year colleges in CollegeScores4yr.

#### Source

Data downloaded from the US Department of Education's College Scorecard at https://collegescorecard.ed.gov/data/(November 2019)

SampCountries

Sample of Countries

# **Description**

Data on a sample of fifty countries of the world (2018)

## **Format**

A data frame with 50 observations on the following 25 variables.

Country Country name

LandArea Size in 1000 sq. km.

Population Population in millions

Density Number of people per square kilometer

GDP Gross Domestic Product (in \$US) per capita

122 SampCountries1e

Rural Percentage of population living in rural areas

CO2 CO2 emissions (metric tons per capita)

PumpPrice Price for a liter of gasoline (\$US)

Military Percentage of government expenditures directed toward the military

Health Percentage of government expenditures directed towards healthcare

ArmedForces Number of active duty military personnel (in 1,000's)

Internet Percentage of the population with access to the internet

Cell Cell phone subscriptions (per 100 people)

HIV Percentage of the population with HIV

Hunger Percent of the population considered undernourished

Diabetes Percent of the population diagnosed with diabetes

BirthRate Births per 1000 people

DeathRate Deaths per 1000 people

ElderlyPop Percentage of the population at least 65 years old

LifeExpectancy Average life expectancy (years)

FemaleLabor Percent of females 15 - 64 in the labor force

Unemployment Percent of labor force unemployed

EnergyUse Kilotons of oil equivalent

Electricity Electric power consumption (kWh per capita)

Developed Categories for kilowatt hours per capita, 1= under 2500, 2=2500 to 5000, 3=over 5000

## Details

Data from AllCountries for a random sample of 50 countries. Data for 2016-2018 to avoid many missing values in more recent years.

\*\* Updated for 3e (earlier versions are now SampCountries2e and SampCountries1e). \*\*

#### Source

Data collected from the World Bank website, http://www.worldbank.org.

SampCountries1e

Sample of Countries - 1e

## Description

Data on a sample of fifty countries of the world (2008)

SampCountries2e 123

## **Format**

A dataset with 50 observations on the following 13 variables.

```
Country
LandArea
Size in sq. kilometers
Population
Population in millions
Energy
Rural
Percentage of population living in rural areas
Military
Health
Percentage of government expenditures directed toward the military
Percentage of government expenditures directed towards healthcare
HIV
Percentage of the population with HIV
Internet
Percentage of the population with access to the internet
Developed
BirthRate
Births per 1000 people
ElderlyPop
Percentage of the population at least 65 years old
LifeExpectancy
Name of the country
Size in sq. kilometers
Population in millions
Energy
Energy usage (kilotons of oil)
Percentage of government expenditures directed toward the military
Percentage of the population with HIV
Internet
Percentage of the population with access to the internet
Developed
BirthRate
Births per 1000 people
ElderlyPop
Percentage of the population at least 65 years old
Average life expectancy (in years)
```

## **Details**

A subset of data from AllCountries for a random sample of 50 countries in 2008.

\*\* From 1e - dataset has been updated for 2e and 3e \*\*

#### **Source**

Data collected from the World Bank website, http://www.worldbank.org.

# **Description**

Data on a sample of fifty countries of the world (2014)

## **Format**

A dataset with 50 observations on the following 25 variables.

Country	Name of the country
LandArea	Size in 1000 sq. kilometers
Population	Population in millions
Density	Number of people per square kilometer
GDP	Gross Domestic Product (in \$US) per capita
Rural	Percentage of population living in rural areas
C02	CO2 emissions (metric tons per capita)
PumpPrice	Price for a liter of gasoline (\$US)

124 SandP500

Percentage of government expenditures directed toward the military Military Health Percentage of government expenditures directed towards healthcare ArmedForces Number of active duty military personnel (in 1,000's) Percentage of the population with access to the internet Internet Cell Cell phone subscriptions (per 100 people) HIV Percentage of the population with HIV Hunger Percent of the population considered undernourished Diabetes Percent of the population diagnosed with diabetes BirthRate Births per 1000 people DeathRate Deaths per 1000 people ElderlyPop Percentage of the population at least 65 years old Average life expectancy (years) LifeExpectancy Female Labor Percent of females 15 - 64 in the labor force Percent of labor force unemployed Unemployment Energy usage (kilotons of oil equivalent) Energy Electricity Electric power consumption (kWh per capita) Developed Categories for kilowatt hours per capita, 1= under 2500, 2=2500 to 5000, 3=over 5000

#### **Details**

Data from AllCountries for a random sample of 50 countries. Data for 2012- -2014 to avoid many missing values in more recent years.

\*\* From 2e - dataset has been updated for 3e \*\*

#### Source

Data collected from the World Bank website, http://www.worldbank.org.

SandP500 *S&P 500 Prices* 

## **Description**

Daily data for S&P 500 Stock Index

#### **Format**

A data frame with 251 observations on the following 6 variables.

Date Trading date (mm/dd/yyy)

Open Opening value

High High point for the day

Low Low point for the day

Close Closing value

Volume Shares traded (in millions)

SandP5001e 125

# **Details**

Daily prices for the S&P 500 Stock Index for trading days in 2018.

\*\* Updated for 3e (earlier versions are SandP5002e from 2014 and SandP5001e from 2010). \*\*

#### **Source**

Downloaded from https://finance.yahoo.com/quote/^GSPC/history?ltr=1

SandP5001e

S&P 500 Prices

# **Description**

Daily data for S&P 500 Stock Index

## **Format**

A dataset with 252 observations on the following 6 variables.

Date Trading date
Open Opening value
High Point for t

High High point for the day Low Down for the day

Close Closing value

Volume Shares traded (in millions)

# **Details**

Daily prices for the S&P 500 Stock Index for trading days in 2010. \*\* From 1e - dataset has been updated for 2e and 3e \*\*

# **Source**

Downloaded from http://finance.yahoo.com/q/hp?s=^GSPC+Historical+Prices

126 SandwichAnts

# **Description**

Daily data for S&P 500 Stock Index

## **Format**

A dataset with 252 observations on the following 6 variables.

Date Trading date
Open Opening value
High High point for the day
Low Low point for the day
Close Closing value

Volume Shares traded (in millions)

#### **Details**

```
Daily prices for the S&P 500 Stock Index for trading days in 2014. ** From 2e - dataset has been updated for 3e **
```

## **Source**

Downloaded from http://finance.yahoo.com/q/hp?s=^GSPC+Historical+Prices

# Description

Ant counts on samples of different sandwiches

## **Format**

A dataset with 24 observations on the following 5 variables.

Butter	Butter on the sandwich? no (Cases with Butter=yes are in SandwichAnts2)
Filling	Type of filling: Ham & Pickles, Peanut Butter, or Vegemite
Bread	Type of bread: Multigrain, Rye, White, or Wholemeal
Ants	Number of ants on the sandwich
0rder	Trial number

SandwichAnts2

#### **Details**

As young students, Dominic Kelly and his friends enjoyed watching ants gather on pieces of sandwiches. Later, as a university student, Dominic decided to study this with a more formal experiment. He chose three types of sandwich fillings (vegemite, peanut butter, and ham & pickles), four types of bread (multigrain, rye, white, and wholemeal), and put butter on some of the sandwiches.

To conduct the experiment he randomly chose a sandwich, broke off a piece, and left it on the ground near an ant hill. After several minutes he placed a jar over the sandwich bit and counted the number of ants. He repeated the process, allowing time for ants to return to the hill after each trial, until he had two samples for each combination of the factors.

This dataset has only sandwiches with no butter. The data in SandwichAnts2 adds information for samples with butter.

#### Source

Margaret Mackisack, "Favourite Experiments: An Addendum to What is the Use of Experiments Conducted by Statistics Students?", Journal of Statistics Education (1994) http://www.amstat.org/publications/jse/v2n1/mackisack.supp.html

SandwichAnts2

Sandwich Ants - Part 2

## **Description**

Ant counts on samples of different sandwiches

## Format

A dataset with 48 observations on the following 5 variables.

Butter Butter on the sandwich? no or yes

Filling Type of filling: Ham & Pickles, Peanut Butter, or Vegemite

Bread Type of bread: Multigrain, Rye, White, or Wholemeal

Ants Number of ants on the sandwich

Order Trial number

## Details

As young students, Dominic Kelly and his friends enjoyed watching ants gather on pieces of sand-wiches. Later, as a university student, Dominic decided to study this with a more formal experiment. He chose three types of sandwich fillings (vegemite, peanut butter, and ham & pickles), four types of bread (multigrain, rye, white, and wholemeal), and put butter on some of the sandwiches.

To conduct the experiment he randomly chose a sandwich, broke off a piece, and left it on the ground near an ant hill. After several minutes he placed a jar over the sandwich bit and counted the number of ants. He repeated the process, allowing time for ants to return to the hill after each trial, until he had two samples for each combination of the three factors.

128 SleepCaffeine

## **Source**

Margaret Mackisack, "Favourite Experiments: An Addendum to What is the Use of Experiments Conducted by Statistics Students?", Journal of Statistics Education (1994) http://www.amstat.org/publications/jse/v2n1/mackisack.supp.html

SkateboardPrices

Skateboard Prices

# Description

Prices of skateboards for sale online

#### **Format**

A dataset with 20 observations on the following variable.

Price Selling price in dollars

## **Details**

Prices for skateboards offered for sale on eBay.

## **Source**

Random sample taken from all skateboards available for sale on eBay on February 12, 2012.

SleepCaffeine

Sleep Caffeine

# Description

Experiment to compare word recall after sleep or caffeine

#### **Format**

A dataset with 24 observations on the following 2 variables.

Group Treatment: Caffeine or Sleep Words Number of words recalled SleepStudy 129

#### **Details**

A random sample of 24 adults were divided equally into two groups and given a list of 24 words to memorize. During a break, one group takes a 90 minute nap while another group is given a caffeine pill. The response variable is the number of words participants are able to recall following the break.

#### Source

Mednick, Cai, Kanady, and Drummond, "Comparing the benefits of caffeine, naps and placebo on verbal, motor and perceptual memory", Behavioural Brain Research, 193 (2008), 79-86.

SleepStudy

Sleep Study

## Description

Data from a study of sleep patterns for college students.

#### **Format**

A dataset with 253 observations on the following 27 variables.

Gender 1=male, 0=female ClassYear Year in school, 1=first year, ..., 4=senior LarkOwl Early riser or night owl? Lark, Neither, or Owl NumEarlyClass Number of classes per week before 9 am EarlyClass Indicator for any early classes Grade point average (0-4 scale) GPA ClassesMissed Number of classes missed in a semester CognitionZscore Z-score on a test of cognitive skills PoorSleepQuality Measure of sleep quality (higher values are poorer sleep) DepressionScore Measure of degree of depression Measure of amount of anxiety AnxietyScore StressScore Measure of amount of stress DepressionStatus Coded depression score: normal, moderate, or severe AnxietyStatus Coded anxiety score: normal, moderate, or severe Stress Coded stress score: normal or high Combined score for depression, anxiety and stress DASScore Happiness Measure of degree of happiness AlcoholUse Self-reported: Abstain, Light, Moderate, or Heavy Drinks Number of alcoholic drinks per week WeekdayBed Average weekday bedtime (24.0=midnight) WeekdayRise Average weekday rise time (8.0=8 am) WeekdaySleep Average hours of sleep on weekdays WeekendBed Average weekend bedtime (24.0=midnight) WeekendRise Average weekend rise time (8.0=8 am)

Smiles

WeekendSleep Average weekend bedtime (24.0=midnight)
AverageSleep Average hours of sleep for all days

AllNighter Had an all-nighter this semester? 1=yes, 0=no

#### **Details**

The data were obtained from a sample of students who did skills tests to measure cognitive function, completed a survey that asked many questions about attitudes and habits, and kept a sleep diary to record time and quality of sleep over a two week period.

#### Source

Onyper, S., Thacher, P., Gilbert, J., Gradess, S., "Class Start Times, Sleep, and Academic Performance in College: A Path Analysis," April 2012; 29(3): 318-335. Thanks to the authors for supplying the data.

Smiles Smiles

# Description

Experiment to study effect of smiling on leniency in judicial matters

#### **Format**

A dataset with 68 observations on the following 2 variables.

Leniency Score assigned by a judgment panel (higher is more lenient)

Group Treatment group: neutral or smile

## **Details**

Hecht and LeFrance conducted a study examining the effect of a smile on the leniency of disciplinary action for wrongdoers. Participants in the experiment took on the role of members of a college disciplinary panel judging students accused of cheating. For each suspect, along with a description of the offense, a picture was provided with either a smile or neutral facial expression. A leniency score was calculated based on the disciplinary decisions made by the participants.

#### Source

LaFrance, M., & Hecht, M. A., "Why smiles generate leniency", Personality and Social Psychology Bulletin, 21, 1995, 207-214.

SpeedDating 131

|--|

## **Description**

Data from a sample of four minute speed dates.

## **Format**

A dataset with 276 observations on the following 22 variables.

```
Would the male like another date? 1=yes 0=no
        DecisionM
        DecisionF
                      Would the female like another date? 1=yes 0=no
                     How much the male likes his partner (1-10 scale)
             LikeM
             LikeF
                     How much the female likes her partner (1-10 scale)
     PartnerYesM
                     Male's estimate of chance the female wants another date (1-10 scale)
     PartnerYesF
                     Female's estimate of chance the male wants another date (1-10 scale)
              AgeM
                     Male's age (in years)
                     Females age (in years)
              AgeF
                     Male's race: Asian Black Caucasian Latino Other
             RaceM
                     Female's race: Asian Black Caucasian Latino Other
             RaceF
     AttractiveM
                     Male's rating of female's attractiveness (1-10 scale)
     AttractiveF
                     Female's rating of male's attractiveness (1-10 scale)
                     Male's rating of female's sincerity (1-10 scale)
         SincereM
                     Female's rating of male's sincerity (1-10 scale)
         SincereF
    IntelligentM
                     Male's rating of female's intelligence (1-10 scale)
    IntelligentF
                     Female's rating of male's intelligence (1-10 scale)
              FunM
                     Male's rating of female as fun (1-10 scale)
              FunF
                     Female's rating of male as fun (1-10 scale)
       AmbitiousM
                     Male's rating of female's ambition (1-10 scale)
       AmbitiousF
                     Female's rating of male's ambition (1-10 scale)
                     Male's rating of female's shared interests (1-10 scale)
SharedInterestsM
SharedInterestsF
                     Female's rating of male's shared interests (1-10 scale)
```

# **Details**

Participants were students at Columbia's graduate and professional schools, recruited by mass email, posted fliers, and fliers handed out by research assistants. Each participant attended one speed dating session, in which they met with each participant of the opposite sex for four minutes. Order and session assignments were randomly determined. After each four minute "speed date," participants filled out a form rating their date on a scale of 1-10 on various attributes. Only data from the first date in each session is recorded here.

## Source

Gelman, A. and Hill, J., Data analysis using regression and multilevel/hierarchical models, Cambridge University Press: New York, 2007

132 StatGrades

SplitBill

Split Bill vs Individual Meal Costs

# **Description**

Meal costs when ordering individually vs splitting a bill

#### **Format**

A dataset with 48 observations on the following 4 variables.

Payment Payment method: Individual or Split

Sex F = female or M = male

Items Number of items ordered

Cost Cost of items ordered in Israeli new shekel's (ILS)

#### **Details**

Subjects were 48 Israeli students who were randomly assigned to eat in groups of six (three males and three females) at a restaurant. Half the groups were told that they would pay for meals individually and half were told that the group would split the bill equally. The number of items ordered and cost (in Israeli new shekels) was recorded for each individual.

## Source

Gneezy, U., Haruvy, E., and Yafe, H. "The Inefficiency of Splitting the Bill," The Economic Journal, 2004; 114, 265-280.

StatGrades

Statistics Exam Grades

# Description

Grades on statistics exams

## **Format**

A dataset with 50 observations on the following 3 variables.

Exam1 Score (out of 100 points) on the first exam

Exam2 Score (out of 100 points) on the second exam

Final Score (out of 100 points) on the final exam

StockChanges 133

## **Details**

Exam scores for a sample of students who completed a course using Statistics: Unlocking the Power of Data as a text. The dataset contains scores on Exam1 (Chapters 1 to 4), Exam2 (Chapters 5 to 8), and the Final exam (entire book).

#### **Source**

Random selection of students in an introductory statistics course.

StockChanges

Stock Changes

# **Description**

Stock price change for a sample of stocks from the S&P 500 (August 2-6, 2010)

#### **Format**

A dataset with 50 observations on the following variable.

SPChange Change in stock price (in dollars)

#### **Details**

A random sample of 50 companies from Standard & Poor's index of 500 companies was selected. The change in the price of the stock (in dollars) over the 5-day period from August 2 - 6, 2010 was recorded for each company in the sample.

# Source

Data obtained from http://money.cnn.com/data/markets/sandp/

StorySpoilers

Story Spoilers

## Description

Ratings for stories with and without spoilers

StressedMice

#### **Format**

A dataset with 12 observations on the following 3 variables.

Story ID for story
Spoiler Average (0-10) rating for spoiler version
Original Average (0-10) rating for original version

#### **Details**

This study investigated whether a story spoiler that gives away the ending early diminishes suspense and hurts enjoyment. For twelve different short stories, the study's authors created a second version in which a spoiler paragraph at the beginning discussed the story and revealed the outcome. Each version of the twelve stories was read by at least 30 people and rated on a 1 to 10 scale to create an overall rating for the story, with higher ratings indicating greater enjoyment of the story. Stories 1 to 4 were ironic twist stories, stories 5 to 8 were mysteries, and stories 9 to 12 were literary stories.

#### Source

Leavitt, J. and Christenfeld, N., "Story Spoilers Don't Spoil Stories," Psychological Science, published OnlineFirst, August 12, 2011.

StressedMice Stressed Mice

## **Description**

Time in darkness for mice in different environments

#### **Format**

A dataset with 14 observations on the following 2 variables.

Time Time spent in darkness (in seconds)
Environment Type of environment: Enriched or Standard

#### **Details**

In the study, mice were randomly assigned to either an enriched environment where there was an exercise wheel available, or a standard environment with no exercise options. After three weeks in the specified environment, for five minutes a day for two weeks, the mice were each exposed to a "mouse bully" - a mouse who was very strong, aggressive, and territorial. One measure of mouse anxiety is amount of time hiding in a dark compartment, with mice who are more anxious spending more time in darkness. The amount of time spent in darkness is recorded for each of the mice.

StudentSurvey 135

## **Source**

Data approximated from summary statistics in: Lehmann and Herkenham, "Environmental Enrichment Confers Stress Resiliency to Social Defeat through an Infralimbic Cortex-Dependent Neuroanatomical Pathway", The Journal of Neuroscience, April 20, 2011, 31(16):61596173.

StudentSurvey

Student Survey Data

## **Description**

Data from a survey of students in introductory statistics courses

#### **Format**

A data frame with 362 observations on the following 17 variables.

Year Year in school

Sex code F=female or M=male

Smoke Smoker? No or Yes

Award Preferred award: Academy, Nobel, or Olympic

HigherSAT Which SAT is higher? Math or Verbal

Exercise Hours of exercise per week

TV Hours of TV viewing per week

Height (in inches)

Weight (in pounds)

Siblings Number of siblings

BirthOrder Birth order, 1=oldest

VerbalSAT VerbalSAT score

MathSAT Math SAT scorer

SAT Combined Verbal + Math SAT

GPA College grade point average

Pulse Pulse rate (beats per minute)

Piercings Number of body piercings

#### **Details**

Data from an in-class survey given to introductory statistics students over several years. Note the Sex variable was labeled as Gender in earlier versions of this dataset. We acknowledge that this binary dichotomization is not a complete or inclusive representation of reality.

# Source

In-class student survey

Synchronized Movement Synchronized Movement

## **Description**

Effects of synchronized movement activities

#### **Format**

A dataset with 264 observations on the following 11 variables.

Sex f = female or m = male

Group Type of activity. Coded as HS+HE, HS+LE, LS+HE, or LS+LE

for High/Low Synchronization + High/Low Exertion

Synch Synchronized activity? yes or no

Exertion Exertion level: high or low

PainToleranceBefore Measure of pain tolerance (mm Hg) before activity

PainTolerance Measure of pain tolerance (mm Hg) after activity

PainTolDiff Difference (after - before) in pain tolerance

MaxPressure Reached the maximum pressure (300 mm Hg) when testing pain tolerance (after)

CloseBefore Rating of closeness to the group before activity (1=least close to 7=most close)

Rating of closeness to the group after activity (1=least close to 7=most close)

CloseDiff Change on closeness rating (after - before)

## **Details**

From a study of 264 high school students in Brazil to examine the effect of doing synchronized movements (such as marching in step or doing synchronized dance steps) and the effect of exertion on variables, such as pain tolerance and attitudes towards others. Students were randomly assigned to activities that involved synchronized or non-synchronized movements involving high or low levels of exertion. Pain tolerance was measured with a blood pressure cuff, going to a maximum possible reading of 300 mmHg.

#### **Source**

Tarr B, Launay J, Cohen E, and Dunbar R, "Synchrony and exertion during dance independently raise pain threshold and encourage social bonding," Biology Letters, 11(10), October 2015.

TenCountries 137

TenCountries

Ten Countries

## **Description**

A subset of the AllCountries data for a random sample of ten countries

## **Format**

A data frame with 10 observations on the following 4 variables.

Country Country name

Code Three-letter country code

Area Size in 1000 sq. kilometers

PctRural Percentage of population living in rural areas

#### **Details**

Area and percent rural for a sample of ten countries from AllCountries dataset.

\*\* Updated for 3e (earlier versions are now TenCountries2e and TenCountries1e) \*\*

#### Source

Data collected from the World Bank website, https://www.worldbank.org/en/home

TenCountries1e

Ten Countries - 1e

# **Description**

A subset of the AllCountries data for a random sample of ten countries

#### **Format**

A dataset with 10 observations on the following 4 variables.

Country Country name

Code Three-letter country code
Area Size in 1000 sq. kilometers

PctRural Percentage of population living in rural areas

## **Details**

Area and percent rural for a sample of ten countries from AllCountries dataset.

\*\* From 1e - dataset has been updated for 2e and 3e \*\*

138 TextbookCosts

## **Source**

Data collected from the World Bank website, http://www.worldbank.org.

TenCountries2e

Ten Countries - 2e

# **Description**

A subset of the AllCountries data for a random sample of ten countries

#### **Format**

A dataset with 10 observations on the following 4 variables.

Country Country name

Code Three-letter country code
Area Size in 1000 sq. kilometers

PctRural Percentage of population living in rural areas

## **Details**

Area and percent rural for a sample of ten countries from AllCountries dataset.

\*\* From 2e - dataset has been updated for 3e \*\*

## Source

Data collected from the World Bank website, http://www.worldbank.org.

TextbookCosts

Textbook Costs

# **Description**

Prices for textbooks for different courses

## **Format**

A data frame with 40 observations on the following 3 variables.

Field General discipline of the course: Arts, Humanities, NaturalScience, or SocialScience

Books Number of books required

Cost Total cost (in dollars) for required books

ToenailArsenic 139

# **Details**

Data are from samples of ten courses in each of four disciplines at a liberal arts college. For each course the bookstore's website lists the required texts(s) and costs. Data were collected for the Fall 2011 semester.

#### **Source**

Bookstore online site

ToenailArsenic

Toenail Arsenic

# Description

Arsenic in toenails of 19 people using private wells in New Hampshire

#### **Format**

A dataset with 19 observations on the following variable.

Arsenic Level of arsenic found in toenails (ppm)

# **Details**

Level of arsenic was measured in toenails of 19 subjects from New Hampshire, all with private wells as their main water source.

#### **Source**

Adapted from Karagas, et.al.,"Toenail Samples as an Indicator of Drinking Water Arsenic Exposure", Cancer Epidemiology, Biomarkers and Prevention 1996;5:849-852.

TrafficFlow

Traffic Flow

# **Description**

Traffic flow times from a simulation with timed and flexible traffic lights

140 USStates

#### **Format**

A dataset with 24 observations on the following 3 variables.

Timed Delay time (in minutes) for fixed timed lights

Flexible Delay time (in minutes) for flexible communicating lights

Difference (Timed-Flexible) for each simulation

#### **Details**

Engineers in Dresden, Germany were looking at ways to improve traffic flow by enabling traffic lights to communicate information about traffic flow with nearby traffic lights. The data show results of one experiment where they simulated buses moving along a street and recorded the delay time (in seconds) for both a fixed time and a flexible system of lights. The process was repeated under both conditions for a sample of 24 simulated scenarios.

#### Source

Lammer and Helbing, "Self-Stabilizing decentralized signal control of realistic, saturated network traffic", Santa Fe Institute working paper \# 10-09-019, September 2010.

USStates

US State Data

# Description

Various data for all 50 US States.

#### **Format**

A data frame with 50 observations on the following 22 variables.

State State name

HouseholdIncome Median household income (in \$1,000's)

Region MW=Midwest, NE=Northeast, S=South, W=West

Population Number of residents (in millions for 2014)

EighthGradeMath Average score NAEP mathematics for 8th-grade students

HighSchool % of residents (ages 25-34) who are high school graduates

College % of residents (ages 25-34) who are college graduates

IQ Estimated mean IQ score of residents

GSP Gross state product (in \$1,000's per capita)

Vegetables % of residents eating vegetables at least once per day

Fruit % of residents eating fruit at least once per day

Smokers % of residents who smoke

USStates1e 141

Physical Activity % who do 150+ minutes of aerobic physical activity per week

Obese % obese residents (BMI 30+)

NonWhite % nonwhite residents

HeavyDrinkers % heavy drinkers (men: 14+ drinks/week, women 7+ drinks/week)

Electoral Number of state votes in the presidential electoral college

ClintonVote Proportion of votes for Democrat Clinton in 2016 presidential election

Elect2016 State winner in 2016 presidential election (D=Clinton, R=Trump)

TwoParents % of children living in two-parent households

StudentSpending School spending (in \$1,000 per pupil)

Insured % of adults (ages 19-64) who have any kind of health coverage

#### **Details**

Information from each of the 50 states of the United States. Years vary from 2013 to 2018 depending on data availability.

\*\* Updated for 3e (earlier versions are now USStates2e and USStates1e) \*\*

#### Source

U.S. Census Bureau, 2013-2017 5-Year American Community Survey

http://factfinder.census.gov/faces/tableservices/jsf/pages/productview.xhtml?pid= ACS\_17\_5YR\_DP03&src=pt

 $http://factfinder.census.gov/faces/tableservices/jsf/pages/productview.xhtml?pid=ACS\_17\_5YR\_S1501\&src=pt$ 

http://factfinder.census.gov/faces/tableservices/jsf/pages/productview.xhtml?pid=ACS\_17\_5YR\_B02001&prodType=table

http://factfinder.census.gov/faces/nav/jsf/pages/index.xhtml (Table C23008)

https://factfinder.census.gov/faces/tableservices/jsf/pages/productview.xhtml?pid=ACS\_17\_5YR\_S2701&prodType=table

USStates1e

US State Data - 1e

# **Description**

Various data for all 50 US States

142 USStates2e

#### **Format**

A dataset with 50 observations on the following 17 variables.

Name of state State HouseholdIncome Mean household income (in dollars) Mean IQ score of residents ΙQ McCainVote Percentage of votes for John McCain in 2008 Presidential election Area of the country: MW=Midwest, NE=Northeast, S=South, or W=West Region ObamaMcCain Which 2008 Presidential candidate won state? M=McCain or 0=Obama Population Number of residents (in millions) EighthGradeMath Average score NAEP mathematics for 8th-grade students Percentage of high school graduates HighSchool Gross State Product (dollars per capita) GSP FiveVegetables Percentage of residents who eat at least five servings of fruits/vegetables per day Percentage of residents who smoke Smokers Percentage of residents who have competed in a physical activity in past month PhysicalActivity 0bese Percentage of residents classified as obese

Obese Percentage of residents classified as obese
College Percentage of residents with college degrees
NonWhite Percentage of residents who are not white
HeavyDrinkers Percentage of residents who drink heavily

## **Details**

Information from each of the 50 states of the United States. \*\* From 1e - dataset has been updated for 2e and 3e \*\*

## Source

Various online sources, mostly at www.census.gov

USState	es2e	US State Data - 2e

## **Description**

Various data for all 50 US States in 2014.

#### Format

A dataset with 50 observations on the following 22 variables.

State	State name
HouseholdIncome	Median household income (in \$1,000's)
Region	MW=Midwest, NE=Northeast, S=South, W=West
Population	Number of residents (in millions for 2014)
EighthGradeMath	Average score NAEP mathematics for 8th-grade students (2013)

WaterStriders 143

HighSchool Percent of residents (ages 25-34) who are high school graduates Percent of residents (ages 25-34) who are college graduates College Estimated mean IO score of residents ΙQ Gross state product (in \$1,000's per capita in 2013) GSP Vegetables Percent of residents eating vegetables at least once per day Fruit Percent of residents eating fruit at least once per day Smokers Percent of residents who smoke PhysicalActivity Percent who do 150+ minutes of aerobic physical activity per week 0bese Percent obese residents (BMI 30+) NonWhite Percent nonwhite residents (in 2013) HeavyDrinkers Percent heavy drinkers (men: 3+ drinks/day, women 2+ drinks/day) Number of state votes in the presidential electoral college Electoral Proportion of votes for Obama in 2012 presidential election ObamaVote ObamaRomney State winner in 2012 presidential election (O=Obama, R=Romney) Percent of children living in two-parent households TwoParents StudentSpending School spending (in \$1,000 per pupil in 2013) Insured Percent of adults (ages 18-64) who have any kind of health coverage

#### **Details**

Information from each of the 50 states of the United States (from 2013 or 2014). \*\* From 2e - dataset has been updated for 3e \*\*

## Source

U.S. Census Bureau, 2009-2013 5-Year American Community Survey http://factfinder.census.gov/faces/tableservices/jsf/pages/productview.xhtml?pid=ACS\_13\_5YR\_DP03&src=pt http://factfinder.census.gov/faces/tableservices/jsf/pages/productview.xhtml?pid=ACS\_13\_5YR\_S1501&src=pt http://factfinder.census.gov/faces/tableservices/jsf/pages/productview.xhtml?pid=ACS\_13\_5YR\_B02001&prodType=table http://factfinder.census.gov/faces/nav/jsf/pages/index.xhtml (Table C23008)

WaterStriders Water Striders

## **Description**

Mating activity for water striders

#### **Format**

A dataset with 10 observations on the following 3 variables.

AggressiveMale Hyper-aggressive male in group? No or Yes

144 WaterTaste

FemalesHiding Proportion of time the female water striders were in hiding

MatingActivity Measure of mean mating activity (higher numbers meaning more mating)

## **Details**

Water striders are common bugs that skate across the surface of water. Water striders have different personalities and some of the males are hyper-aggressive, meaning they jump on and wrestle with any other water strider near them. Individually, because hyper-aggressive males are much more active, they tend to have better mating success than more inactive striders. This study examined the effect they have on a group. Four males and three females were put in each of ten pools of water. Half of the groups had a hyper-aggressive male as one of the males and half did not. The proportion of time females are in hiding was measured for each of the 10 groups, and a measure of mean mating activity was also measured with higher numbers meaning more mating.

#### Source

Sih, A. and Watters, J., "The mix matters: behavioural types and group dynamics in water striders," Behaviour, 2005; 142(9-10): 1423.

WaterTaste
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## **Description**

Blind taste test to compare brands of bottled water

#### **Format**

A dataset with 100 observations on the following 10 variables.

Gender	Gender of respondent: F=Female M=Male
Age	Age (in years)
Class	Year in school F=First year J=Junior 0=Other P S0=Sophomore SR=Senior
UsuallyDrink	Usual source of drinking water: Bottled, Filtered, or Tap
FavBotWatBrand	Favorite brand of bottled water
Preference	Order of preference: A=Sams Choice, B=Aquafina, C=Fiji, and D=Tap water
First	Top choice among Aquafina, Fiji, SamsChoice, or Tap
Second	Second choice
Third	Third choice
Fourth	Fourth choice

## **Details**

Result from a blind taste test comparing four different types of water (Sam's Choice, Aquafina, Fiji, and tap water). Participants rank ordered waters when presented in a random order.

Wetsuits 145

# Source

"Water Taste Test Data" by M. Leigh Lunsford and Alix D. Dowling Finch in the Journal of Statistics Education (Vol 18, No, 1) 2010

http://www.amstat.org/publications/jse/v18n1/lunsford.pdf

Wetsuits Wetsuits

## **Description**

Swim velocity (for 1500 meters) with and without wearing a wetsuit

#### **Format**

A dataset with 12 observations on the following 4 variables.

Wetsuit Maximum swim velocity (m/sec) when wearing a wetsuit

NoWetsuit Maximum swim velocity (m/sec) when wearing a regular bathing suit

Gender Gender of swimmer: F or M

Type Type of athlete: swimmer or triathlete

## **Details**

A study tested whether wearing wetsuits influences swimming velocity. Twelve competitive swimmers and triathletes swam 1500m at maximum speed twice each; once wearing a wetsuit and once wearing a regular bathing suit. The order of the trials was randomized. Each time, the maximum velocity in meters/sec of the swimmer was recorded.

## **Source**

de Lucas, R.D., Balildan, P., Neiva, C.M., Greco, C.C., Denadai, B.S. (2000). "The effects of wetsuits on physiological and biomechanical indices during swimming," Journal of Science and Medicine in Sport, 3 (1): 1-8.

YoungBlood Young Blood

## Description

Effects of transfusions of young blood on exercise endurance in mice

YoungBlood

# **Format**

A dataset with 30 observations on the following 2 variables.

Plasma Whether the blood came from a Young or Old mouse

Runtime Maximum treadmill run time (in minutes) in a 90-minute window

# **Details**

The data come from a study to see if transfusions of blood plasma from young mice (equivalent to about a 25-year-old person) can counteract or reverse brain aging in old mice (equivalent to about a 70-year-old person.) Old mice were randomly assigned to receive plasma from either a young mice or another old mouse, and exercise endurance was measured.

## Source

Data come from two references, and are estimated from summary statistics and graphs. Sanders L, "Young blood proven good for old brain," Science News, 185(11), May 31, 2014. Manisha S, et al., "Restoring Systemic GDF11 Levels Reverses Age-Related Dysfunction in Mouse Skeletal Muscle," Science, 9 May 2014.

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