Package 'shorts'

July 23, 2025

Type Package

Title Short Sprints

Version 3.2.0

Description Create short sprint acceleration-velocity (AVP) and force-velocity (FVP) profiles and predict kinematic and kinetic variables using the timing-gate split times, laser or radar gun data, tether devices data, as well as the data provided by the GPS and LPS monitoring systems. The modeling method utilized in this package is based on the works of Furusawa K, Hill AV, Parkinson JL (1927) <doi:10.1098/rspb.1927.0035>, Greene PR. (1986) <doi:10.1016/0025-5564(86)90063-5>, Chelly SM, Denis C. (2001) <doi:10.1097/00005768-200102000-00024>, Clark KP, Rieger RH, Bruno RF, Stearne DJ. (2017) <doi:10.1519/JSC.000000000002081>, Samozino P. (2018) <doi:10.1007/978-3-319-05633-3_11>, Samozino P. and Peyrot N., et al (2022) <doi:10.1111/sms.14097>, Clavel, P., et al (2023) <doi:10.1016/j.jbiomech.2023.111602>, Jovanovic M. (2023) <doi:10.1080/10255842.2023.2170713>, and Jovanovic M., et al (2024) <doi:10.3390/s24092894>.

URL https://mladenjovanovic.github.io/shorts/

BugReports https://github.com/mladenjovanovic/shorts/issues

License MIT + file LICENSE

Encoding UTF-8

LazyData true

RoxygenNote 7.2.3

Depends R (>= 2.10)

Imports stats, LambertW, tidyr, ggplot2, minpack.lm, purrr

Config/testthat/edition 3

NeedsCompilation no

Author Mladen Jovanović [aut, cre, cph] (ORCID: https://orcid.org/0000-0002-4013-6530>)

Maintainer Mladen Jovanović <coach.mladen.jovanovic@gmail.com>

Repository CRAN

Date/Publication 2024-05-22 11:40:02 UTC

Contents

coef.shorts_model	. 2
confint.shorts_model	. 3
convert_FVP	. 4
create_FVP	. 5
create_sprint_trace	. 6
create_timing_gates_splits	. 7
dynaspeed	. 8
find_functions	. 9
find_optimal_distance	. 12
fitted.shorts_model	. 13
format_splits	. 14
get_air_resistance	. 15
jb_morin	. 16
laser_gun_data	. 17
LPS_session	. 18
model_functions	. 18
optimal_functions	. 26
plot.shorts_model	. 29
predict.shorts_model	. 30
predict_kinematics	. 31
print.shorts_model	. 35
probe_functions	. 36
radar_gun_data	
residuals.shorts_model	. 39
split_times	. 39
summary.shorts_model	
vescovi	. 41
	43

Index

coef.shorts_modelS3 method for extracting model parameters from shorts_model object

Description

S3 method for extracting model parameters from shorts_model object

Usage

S3 method for class 'shorts_model'
coef(object, ...)

Arguments

object	<pre>shorts_model object</pre>
	Extra arguments. Not used

confint.shorts_model

Examples

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
   gates = split_distances,
   MSS = 10,
   MAC = 9,
   FD = 0.25,
   TC = 0
)
# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
coef(simple_model)</pre>
```

confint.shorts_model S3 method for providing confidence intervals for the shorts_model

Description

S3 method for providing confidence intervals for the shorts_model

Usage

```
## S3 method for class 'shorts_model'
confint(object, ...)
```

Arguments

object	<pre>shorts_model object</pre>
	Forwarded to generic confint() function

```
## Not run:
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
    gates = split_distances,
    MSS = 10,
    MAC = 9,
    FD = 0,
    TC = 0,
    noise = 0.01
)
# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
confint(simple_model)
## End(Not run)
```

convert_FVP

Description

This function converts back the Force-Velocity profile (FVP; F0 and V0 parameters) to Acceleration-Velocity profile (AVP; MSS and MAC parameters)

Usage

```
convert_FVP(
  F0,
  V0,
  bodymass = 75,
  inertia = 0,
  resistance = 0,
  wind_velocity = 0,
  ...
)
```

Arguments

F0, V0	Numeric vectors. FV profile parameters
bodymass	Body mass in kg. Used to calculate relative power and forwarded to get_air_resistance
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
wind_velocity	In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)
	Forwarded to predict_power_at_distance

Value

A list with calculated MSS and MAC parameters

```
FVP <- create_FVP(7, 8.3, inertia = 10, resistance = 50)
convert_FVP(FVP$F0, FVP$V0, inertia = 10, resistance = 50)</pre>
```

Description

Creates Force-Velocity Profile (FVP) modified using ideas by Pierre Samozino and JB-Morin, et al. (2016) and Pierre Samozino and Nicolas Peyror, et al (2021).

Usage

```
create_FVP(
   MSS,
   MAC,
   bodymass = 75,
   inertia = 0,
   resistance = 0,
   wind_velocity = 0,
   ...
)
```

Arguments

MSS, MAC	Numeric vectors. Model parameters
bodymass	Body mass in kg. Used to calculate relative power and forwarded to get_air_resistance
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
wind_velocity	In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)
	Forwarded to predict_power_at_distance

Value

List containing the following elements:

bodymass Returned bodymass used in FV profiling

F0 Horizontal force when velocity=0

F0_rel F0 divided by bodymass

V0 Velocity when horizontal force=0

Pmax Maximal horizontal power

Pmax_rel Pmax divided by bodymass

FV_slope Slope of the FV profile. See References for more info

References

Samozino P, Rabita G, Dorel S, Slawinski J, Peyrot N, Saez de Villarreal E, Morin J-B. 2016. A simple method for measuring power, force, velocity properties, and mechanical effectiveness in sprint running: Simple method to compute sprint mechanics. Scandinavian Journal of Medicine & Science in Sports 26:648–658. DOI: 10.1111/sms.12490.

Samozino P, Peyrot N, Edouard P, Nagahara R, Jimenez-Reyes P, Vanwanseele B, Morin J. 2022. Optimal mechanical force-velocity profile for sprint acceleration performance. Scandinavian Journal of Medicine & Science in Sports 32:559–575. DOI: 10.1111/sms.14097.

Examples

```
data("jb_morin")
m1 <- model_radar_gun(time = jb_morin$time, velocity = jb_morin$velocity)
fv_profile <- create_FVP(
    MSS = m1$parameters$MSS,
    MAC = m1$parameters$MAC,
    bodyheight = 1.72,
    bodymass = 120,
)
fv_profile</pre>
```

create_sprint_trace Create Sprint Trace

Description

This function creates sprint trace either using time or distance vectors

Usage

```
create_sprint_trace(
   MSS,
   MAC,
   time = NULL,
   distance = NULL,
   TC = 0,
   DC = 0,
   FD = 0,
   remove_leading = FALSE
)
```

Arguments

MSS, MAC	Numeric vector. Model parameters
time	Numeric vector.
distance	Numeric vector.
ТС	Numeric vector. Time-shift added to sprint times. Default is 0
DC	Numeric vector. Distance-shift added to sprint distance. Default is 0
FD	Numeric vector. Flying start distance. Default is 0
remove_leading	Should trace leading to sprint be removed? Default is FALSE

Value

Data-frame with following 6 columns

time Measurement-scale time vector in seconds. Equal to parameter time

distance Measurement-scale distance vector in meters. Equal to parameter distance

velocity Velocity vector in m/s

acceleration Acceleration vector in m/s/s

sprint_time Sprint scale time vector in seconds. Sprint always start at t=0s

sprint_distance Sprint scale distance vector in meters. Sprint always start at d=0m

See Also

create_timing_gates_splits

Examples

```
df <- create_sprint_trace(8, 7, time = seq(0, 6, by = 0.01))
df <- create_sprint_trace(8, 7, distance = seq(0, 40, by = 1))</pre>
```

Description

This function is used to generate timing gates splits with predetermined parameters

Usage

```
create_timing_gates_splits(
    MSS,
    MAC,
    gates = c(5, 10, 20, 30, 40),
    FD = 0,
    TC = 0,
    noise = 0
)
```

Arguments

MSS, MAC	Numeric vectors. Model parameters
gates	Numeric vectors. Distances of the timing gates
FD	Numeric vector. Flying start distance. Default is 0
ТС	Numeric vector. Time-correction added to split times (e.g., reaction time). Default is 0
noise	Numeric vector. SD of Gaussian noise added to the split times. Default is 0

See Also

create_sprint_trace

Examples

```
create_timing_gates_splits(
  gates = c(10, 20, 30, 40, 50),
  MSS = 10,
  MAC = 9,
  FD = 0.5,
  TC = 0
)
```

dynaspeed

DynaSpeed Single Sprint Data

Description

DynaSpeed(TM) data collected for a single athlete (female, 177cm, 64kg) and a single sprint over 40m. Sampling frequency is 1,000Hz. Additional time and distance shift is added to the dataset to provide a sandbox for potential issues during the analysis

Usage

data(dynaspeed)

find_functions

Format

Data frame with 4 variables and 7,251 observations:

time time in seconds

distance Distance in meters

velocity Smoothed velocity in meters per second

raw_velocity Velocity in meters per second

Author(s)

Håkan Andersson The High-Performance Center Växjö, Sweden <hakan.andersson@hpcsweden.com>

find_functions Find functions

Description

Family of functions that serve a purpose of finding maximal value and critical distances and times at which power, acceleration or velocity drops below certain threshold.

find_peak_power_distance finds peak power and distance at which peak power occurs

find_peak_power_time finds peak power and time at which peak power occurs

find_velocity_critical_distance finds critical distance at which percent of MSS is achieved

find_velocity_critical_time finds critical time at which percent of MSS is achieved

find_acceleration_critical_distance finds critical distance at which percent of MAC is reached

find_acceleration_critical_time finds critical time at which percent of MAC is reached

find_power_critical_distance finds critical distances at which peak power over percent is achieved

find_power_critical_time finds critical times at which peak power over percent is achieved

Usage

```
find_peak_power_distance(MSS, MAC, inertia = 0, resistance = 0, ...)
```

find_peak_power_time(MSS, MAC, inertia = 0, resistance = 0, ...)

find_velocity_critical_distance(MSS, MAC, percent = 0.9)

find_velocity_critical_time(MSS, MAC, percent = 0.9)

find_acceleration_critical_distance(MSS, MAC, percent = 0.9)

find_functions

```
find_acceleration_critical_time(MSS, MAC, percent = 0.9)
find_power_critical_distance(
 MSS,
 MAC,
 inertia = 0,
 resistance = 0,
 percent = 0.9,
  . . .
)
find_power_critical_time(
 MSS,
 MAC,
  inertia = 0,
 resistance = 0,
 percent = 0.9,
  . . .
```

```
)
```

Arguments

MSS, MAC	Numeric vectors. Model parameters
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
	Arguments passed on to get_air_resistance
	velocity Instantaneous running velocity in meters per second (m/s)
	bodymass In kilograms (kg). Default is 75kg
	bodyheight In meters (m). Default is 1.75m
	barometric_pressure In Torrs. Default is 760Torrs
	air_temperature In Celzius (C). Default is 25C
	wind_velocity In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)
percent	Numeric vector. Used to calculate critical distance. Default is 0.9

Value

find_peak_power_distance returns list with two elements: peak_power and distance at which peak power occurs

find_peak_power_time returns list with two elements: peak_power and time at which peak power occurs

find_functions

References

Haugen TA, Tønnessen E, Seiler SK. 2012. The Difference Is in the Start: Impact of Timing and Start Procedure on Sprint Running Performance: Journal of Strength and Conditioning Research 26:473–479. DOI: 10.1519/JSC.0b013e318226030b.

Samozino P. 2018. A Simple Method for Measuring Force, Velocity and Power Capabilities and Mechanical Effectiveness During Sprint Running. In: Morin J-B, Samozino P eds. Biomechanics of Training and Testing. Cham: Springer International Publishing, 237–267. DOI: 10.1007/978-3-319-05633-3_11.

```
dist <- seq(0, 40, length.out = 1000)
velocity <- predict_velocity_at_distance(</pre>
  distance = dist,
 MSS = 10,
  MAC = 9
)
acceleration <- predict_acceleration_at_distance(</pre>
  distance = dist,
  MSS = 10,
  MAC = 9
)
# Use ... to forward parameters to the shorts::get_air_resistance
pwr <- predict_power_at_distance(</pre>
  distance = dist,
 MSS = 10,
 MAC = 9
  # bodyweight = 100,
  # bodyheight = 1.9,
  # barometric_pressure = 760,
  # air_temperature = 25,
  # wind_velocity = 0
)
# Find critical distance when 90% of MSS is reached
plot(x = dist, y = velocity, type = "l")
abline(h = 10 * 0.9, col = "gray")
abline(v = find_velocity_critical_distance(MSS = 10, MAC = 9), col = "red")
# Find critical distance when 20% of MAC is reached
plot(x = dist, y = acceleration, type = "1")
abline(h = 9 * 0.2, col = "gray")
abline(v = find_acceleration_critical_distance(MSS = 10, MAC = 9, percent = 0.2), col = "red")
# Find peak power and location of peak power
plot(x = dist, y = pwr, type = "l")
peak_pwr <- find_peak_power_distance(</pre>
```

```
MSS = 10,
MAC = 9
# Use ... to forward parameters to the shorts::get_air_resistance
)
abline(h = peak_pwr$peak_power, col = "gray")
abline(v = peak_pwr$distance, col = "red")
# Find distance in which relative power stays over 75% of PMAX'
plot(x = dist, y = pwr, type = "l")
abline(h = peak_pwr$peak_power * 0.75, col = "gray")
pwr_zone <- find_power_critical_distance(MSS = 10, MAC = 9, percent = 0.75)
abline(v = pwr_zone$lower, col = "blue")
abline(v = pwr_zone$upper, col = "blue")
```

find_optimal_distance Function that finds the distance at which the sprint, probe, or FV profile is optimal (i.e., equal to 100 perc)

Description

Function that finds the distance at which the sprint, probe, or FV profile is optimal (i.e., equal to 100 perc)

Usage

```
find_optimal_distance(..., optimal_func = optimal_FV, min = 1, max = 60)
```

Arguments

	Forwarded to selected optimal_func
optimal_func	Selected profile optimization function. Default is optimal_FV
min, max	Distance over which to find optimal profile distance

Value

Distance

Examples

```
MSS <- 10
MAC <- 8
bodymass <- 75
fv <- create_FVP(MSS, MAC, bodymass)
find_optimal_distance(
  F0 = fv$F0,
        V0 = fv$V0,
```

fitted.shorts_model

```
bodymass = fv$bodymass,
optimal_func = optimal_FV,
method = "max"
)
find_optimal_distance(
   MSS = MSS,
   MAC = MAC,
   optimal_func = optimal_MSS_MAC
)
find_optimal_distance(
   MSS = MSS,
   MAC = MAC,
   optimal_func = probe_MSS_MAC
)
```

fitted.shorts_model S3 method for returning predictions of shorts_model

Description

S3 method for returning predictions of shorts_model

Usage

```
## S3 method for class 'shorts_model'
fitted(object, ...)
```

Arguments

object	<pre>shorts_model object</pre>
	Extra arguments. Not used

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
   gates = split_distances,
   MSS = 10,
   MAC = 9,
   FD = 0.25,
   TC = 0
)
# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
fitted(simple_model)</pre>
```

format_splits

Description

Function formats split data and calculates split distances, split times and average split velocity

Usage

format_splits(distance, time)

Arguments

distance	Numeric vector
time	Numeric vector

Value

Data frame with the following columns:

split Split number

split_distance_start Distance at which split starts

split_distance_stop Distance at which split ends

split_distance Split distance

split_time_start Time at which distance starts

split_time_stop Time at which distance ends

split_time Split time

split_mean_velocity Mean velocity over split distance

split_mean_acceleration Mean acceleration over split distance

```
data("split_times")
john_data <- split_times[split_times$athlete == "John", ]
format_splits(john_data$distance, john_data$time)</pre>
```

get_air_resistance Get Air Resistance

Description

get_air_resistance estimates air resistance in Newtons

Usage

```
get_air_resistance(
  velocity,
  bodymass = 75,
  bodyheight = 1.75,
  barometric_pressure = 760,
  air_temperature = 25,
  wind_velocity = 0
)
```

Arguments

velocity	Instantaneous running velocity in meters per second (m/s)	
bodymass	In kilograms (kg). Default is 75kg	
bodyheight	In meters (m). Default is 1.75m	
barometric_pressure		
	In Torrs. Default is 760Torrs	
air_temperature		
	In Celzius (C). Default is 25C	
wind_velocity	In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is $0m/s$ (no wind)	

Value

Air resistance in Newtons (N)

References

Arsac LM, Locatelli E. 2002. Modeling the energetics of 100-m running by using speed curves of world champions. Journal of Applied Physiology 92:1781–1788. DOI: 10.1152/japplphys-iol.00754.2001.

Samozino P, Rabita G, Dorel S, Slawinski J, Peyrot N, Saez de Villarreal E, Morin J-B. 2016. A simple method for measuring power, force, velocity properties, and mechanical effectiveness in sprint running: Simple method to compute sprint mechanics. Scandinavian Journal of Medicine & Science in Sports 26:648–658. DOI: 10.1111/sms.12490.

van Ingen Schenau GJ, Jacobs R, de Koning JJ. 1991. Can cycle power predict sprint running performance? European Journal of Applied Physiology and Occupational Physiology 63:255–260. DOI: 10.1007/BF00233857.

Examples

```
get_air_resistance(
 velocity = 5,
 bodymass = 80,
 bodyheight = 1.90,
 barometric_pressure = 760,
 air_temperature = 16,
 wind_velocity = -0.5
)
```

jb_morin

JB Morin Sample Dataset

Description

Sample radar gun data provided by Jean-Benoît Morin on his website. See https://jbmorin.net/2017/12/13/a-spreadsheet-for-sprint-acceleration-force-velocity-power-profiling/for more details.

Usage

data(jb_morin)

Format

Data frame with 2 variables and 232 observations:

time Time in seconds

velocity Velocity in m/s

Details

This dataset represents a sample data provided by Jean-Benoît Morin on a single individual running approximately 35m from a stand still position that is measured with the radar gun. Individual's body mass is 75kg, height is 1.72m. Conditions of the run are the following: air temperature 25C, barometric pressure 760mmHg, wind velocity 0m/s.

The purpose of including this dataset in the package is to check the agreement of the model estimates with Jean-Benoît Morin Microsoft Excel spreadsheet.

Author(s)

Jean-Benoît Morin Inter-university Laboratory of Human Movement Biology Saint-Étienne, France https://jbmorin.net/

laser_gun_data

References

Morin JB. 2017.A spreadsheet for Sprint acceleration Force-Velocity-Power profiling. Available at https://jbmorin.net/2017/12/13/a-spreadsheet-for-sprint-acceleration-force-velocity-power-profiling/ (accessed October 27, 2020).

laser_gun_data Laser Gun Data

Description

Performance of 35m sprint by a youth basketball player done using standing start. Sample was collected by laser gun (CMP3 Distance Sensor, Noptel Oy, Oulu, Finland) and was sampled at a rate of 2.56 KHz. A polynomial function modeling the relationship between distance and time was employed and subsequently resampled at a frequency of 1,000 Hz using MusclelabTM v10.232.107.5298, a software developed by Ergotest Technology AS located in Langesund, Norway. Data was further modified by calculating raw acceleration using dv/dt (using smoothed velocity provided by the system), and then smoothed out using 4th-order Butterworth filter with a cutoff frequency of 1 Hz.

Usage

data(laser_gun_data)

Format

Data frame with 6 variables and 4805 observations:

time Time vector in seconds

distance Distance vector in meters

velocity Smoothed velocity vector in m/s; this represent step-averaged velocity

raw_velocity Raw velocity vector in m/s

- **raw_acceleration** Raw acceleration vector in m/s/s; calculated using difference in smoothed velocity divided by time difference (i.e., dv/dt method of derivation)
- **butter_acceleration** Smoothed acceleration vector in m/s/s; smoothed out using 4th-order Butterworth filter with a cutoff frequency of 1 Hz

LPS_session

Description

LPS Basketball Session Dataset

Usage

data(LPS_session)

Format

Data frame with 5 variables and 91,099 observations:

time Time in seconds from the start of the session

x x-coordinate in meters provided by the LPS

y y-coordinate in meters provided by the LPS

velocity Velocity provided by LPS in m/s

acceleration Acceleration provided by LPS in m/s

Details

This dataset represents a sample data provided by Local Positioning System (LPS) on a single individual performing a single basketball practice session (aprox. 90min). Sampling frequency is 20Hz.

model_functions Model functions

Description

Family of functions that serve a purpose of estimating short sprint parameters

model_in_situ estimates short sprint parameters using velocity-acceleration trace, provided by the monitoring systems like GPS or LPS. See references for the information

model_radar_gun estimates short sprint parameters using time-velocity trace, with additional parameter TC serving as intercept

model_laser_gun alias for model_radar_gun

model_tether estimates short sprint parameters using distance-velocity trace (e.g., tether devices).

model_tether_DC estimates short sprint parameters using distance-velocity trace (e.g., tether devices) with additional distance correction DC parameter

model_time_distance estimates short sprint parameters using time distance trace

model_functions

model_time_distance_FD estimates short sprint parameters using time-distance trace with additional estimated flying distance correction parameter FD

model_time_distance_FD_fixed estimates short sprint parameters using time-distance trace with additional flying distance correction parameter FD which is fixed by the user

model_time_distance estimates short sprint parameters using time distance trace with additional time correction parameter TC

model_time_distance estimates short sprint parameters using time distance trace with additional distance correction parameter DC

model_time_distance estimates short sprint parameters using time distance trace with additional time correction TC and distance correction TC parameters

model_timing_gates estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells)

model_timing_gates_TC estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional time correction parameter TC

model_timing_gates_FD estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional estimated flying distance correction parameter FD

model_timing_gates_FD_fixed estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional flying distance correction parameter FD which is fixed by the user

model_timing_gates_DC estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional distance correction parameter DC

model_timing_gates_TC_DC estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional time correction TC and distance correction DC parameters

Usage

```
model_in_situ(
  velocity,
  acceleration,
  weights = 1,
  velocity_threshold = NULL,
  velocity_step = 0.2,
  n_observations = 2,
  CV = NULL,
  na.rm = FALSE,
)
model_radar_gun(
  time,
  velocity.
  weights = 1,
  CV = NULL,
  use_observed_MSS = FALSE,
  na.rm = FALSE,
  . . .
```

```
)
model_laser_gun(
  time,
  velocity,
 weights = 1,
 CV = NULL,
 use_observed_MSS = FALSE,
 na.rm = FALSE,
  . . .
)
model_tether(
 distance,
 velocity,
 weights = 1,
 CV = NULL,
 use_observed_MSS = FALSE,
 na.rm = FALSE,
  . . .
)
model_tether_DC(
 distance,
 velocity,
 weights = 1,
 CV = NULL,
 use_observed_MSS = FALSE,
 na.rm = FALSE,
  . . .
)
model_time_distance(time, distance, weights = 1, CV = NULL, na.rm = FALSE, ...)
model_time_distance_FD(
  time,
 distance,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
  . . .
)
model_time_distance_FD_fixed(
  time,
  distance,
 weights = 1,
 FD = 0,
```

```
CV = NULL,
 na.rm = FALSE,
  . . .
)
model_time_distance_TC(
 time,
 distance,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
  . . .
)
model_time_distance_DC(
  time,
 distance,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
  . . .
)
model_time_distance_TC_DC(
 time,
 distance,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
  . . .
)
model_timing_gates(distance, time, weights = 1, CV = NULL, na.rm = FALSE, ...)
model_timing_gates_TC(
 distance,
  time,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
  . . .
)
model_timing_gates_FD(
 distance,
  time,
 weights = 1,
 CV = NULL,
```

```
na.rm = FALSE,
  . . .
)
model_timing_gates_FD_fixed(
  distance,
  time,
  weights = 1,
  FD = 0,
  CV = NULL,
  na.rm = FALSE,
  . . .
)
model_timing_gates_DC(
  distance,
  time,
  weights = 1,
  CV = NULL,
  na.rm = FALSE,
  . . .
)
model_timing_gates_TC_DC(
  distance,
  time,
  weights = 1,
  CV = NULL,
  na.rm = FALSE,
  . . .
```

)

Arguments

weights	Numeric vector. Default is 1	
velocity_threshold		
	Velocity cutoff. If NULL (default), velocity of the observation with the fastest acceleration is taken as the cutoff value	
velocity_step	Velocity increment size for finding max acceleration. Default is 0.2 m/s	
n_observations	Number of top acceleration observations to keep in velocity bracket. Default is 2	
CV	Should cross-validation be used to estimate model fit? Default is NULL. Otherwise use integer indicating number of folds	
na.rm	Logical. Default is FALSE	
	Forwarded to nlsLM function	
time, velocity, c	listance, acceleration	
	Numeric vector	

model_functions

use_observed_MSS Should observed peak velocity be used as MSS parameter? Default is FALSE FD Flying distance parameter. Default is 0

Value

List object with the following elements:

data Data frame used to estimate the sprint parameters

model_info Extra information regarding model used

model Model returned by the nlsLM function

parameters List with the following estimated parameters: MSS, MAC, TAU, and PMAX

correction List with additional model correcitons

predictions Data frame with .predictor, .observed, .predicted, and .residual columns

model_fit List with multiple model fit estimators

CV If cross-validation is performed, this will included the data as above, but for each fold

References

Samozino P. 2018. A Simple Method for Measuring Force, Velocity and Power Capabilities and Mechanical Effectiveness During Sprint Running. In: Morin J-B, Samozino P eds. Biomechanics of Training and Testing. Cham: Springer International Publishing, 237–267. DOI: 10.1007/978-3-319-05633-3_11.

Clavel, P., Leduc, C., Morin, J.-B., Buchheit, M., & Lacome, M. (2023). Reliability of individual acceleration-speed profile in-situ in elite youth soccer players. Journal of Biomechanics, 153, 111602. https://doi.org/10.1016/j.jbiomech.2023.111602

Morin, J.-B. (2021). The "in-situ" acceleration-speed profile for team sports: testing players without testing them. JB Morin, PhD – Sport Science website. Accessed 31. Dec. 2023. https://jbmorin.net/2021/07/29/the-in-situ-sprint-profile-for-team-sports-testing-players-without-testing-them/

```
# Model In-Situ (Embedded profiling)
data("LPS_session")
m1 <- model_in_situ(
    velocity = LPS_session$velocity,
    acceleration = LPS_session$acceleration,
    # Use specific cutoff value
    velocity_threshold = 4)
m1
plot(m1)
# Model Radar Gun (includes Time Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 6, 0.1))
# Add some noise
df$velocity <- df$velocity + rnorm(n = nrow(df), 0, 10^-2)</pre>
```

```
m1 <- model_radar_gun(time = df$time, velocity = df$velocity)</pre>
m1
plot(m1)
# Model Laser Gun (includes Time Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 6, 0.1))
# Add some noise
df$velocity <- df$velocity + rnorm(n = nrow(df), 0, 10^-2)
m1 <- model_laser_gun(time = df$time, velocity = df$velocity)</pre>
m1
plot(m1)
# Model Tether
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 6, 0.5))</pre>
m1 <- model_tether(distance = df$distance, velocity = df$velocity)</pre>
m1
plot(m1)
# Model Tether with Distance Correction (DC)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0.001, 6, 0.5), DC = 5)
m1 <- model_tether_DC(distance = df$distance, velocity = df$velocity)</pre>
m1
plot(m1)
# Model Time-Distance trace (simple, without corrections)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5))
m1 <- model_time_distance(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Time-Distance trace (with Flying Distance Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), FD = 0.5)
m1 <- model_time_distance_FD(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Time-Distance trace (with Flying Distance Correction fixed)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), FD = 0.5)
m1 <- model_time_distance_FD_fixed(time = df$time, distance = df$distance, FD = 0.5)</pre>
m1
plot(m1)
```

Model Time-Distance trace (with Time Correction)

```
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), TC = 1.5)
m1 <- model_time_distance_TC(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Time-Distance trace (with Distance Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), DC = -5)
m1 <- model_time_distance_DC(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Time-Distance trace (with Time and Distance Corrections)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), TC = -1.3, DC = 5)
m1 <- model_time_distance_TC_DC(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Timing Gates (simple, without corrections)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40))
m1 <- model_timing_gates(distance = df$distance, time = df$time)</pre>
m1
plot(m1)
# Model Timing Gates (with Time Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), TC = 0.2)
m1 <- model_timing_gates_TC(distance = df$distance, time = df$time)</pre>
m1
plot(m1)
# Model Timing Gates (with Flying Distance Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), FD = 0.5)
m1 <- model_timing_gates_FD(distance = df$distance, time = df$time)</pre>
m1
plot(m1)
# Model Timing Gates (with Flying Distance Correction fixed)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), FD = 0.5)
m1 <- model_timing_gates_FD_fixed(distance = df$distance, time = df$time)</pre>
m1
plot(m1)
# Model Timing Gates (with Distance Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), DC = 1.5)
m1 <- model_timing_gates_DC(distance = df$distance, time = df$time)</pre>
m1
plot(m1)
```

```
# Model Timing Gates (with Time and Distance Corrections)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), TC = 0.25, DC = 1.5)
m1 <- model_timing_gates_TC_DC(distance = df$distance, time = df$time)
m1
plot(m1)</pre>
```

optimal_functions Optimal profile functions

Description

Family of functions that serve a purpose of finding optimal sprint or force-velocity profile

<code>optimal_FV</code> finds "optimal" F0 and V0 where time at distance is minimized, while keeping the power the same

optimal_MSS_MAC finds "optimal" MSS and MAS where time at distance is minimized, while keeping the Pmax the same

Usage

```
optimal_FV(
   distance,
   F0,
   V0,
   bodymass = 75,
   inertia = 0,
   resistance = 0,
   method = "max",
   ...
)
```

optimal_MSS_MAC(distance, MSS, MAC)

Arguments

distance	Numeric vector
F0, V0	Numeric vectors. FV profile parameters
bodymass	Body mass in kg
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
method	Method to be utilized. Options are "peak" and "max" (default)

	Arguments passed on to get_air_resistance	
	velocity Instantaneous running velocity in meters per second (m/s)	
	bodyheight In meters (m). Default is 1.75m	
barometric_pressure In Torrs. Default is 760Torrs air_temperature In Celzius (C). Default is 25C		
MSS, MAC	Numeric vectors. Model parameters	

Value

optimal_FV returns s data frame with the following columns

F0 Original F0 V0 Original F0 **bodymass** Bodymass inertia Inertia resistance Resistance **Pmax** Maximal power estimated using F0 * V0 / 4 Pmax_rel Relative maximal power slope FV profile slope distance Distance time Time to cover distance Ppeak Peak power estimated quantitatively Ppeak_rel Relative peak power Ppeak_dist Distance at which peak power is manifested Ppeak_time Time at which peak power is manifested F0_optim Optimal F0 F0_coef Ratio between F0_optim an F0 V0_optim Optimal V0 **V0_coef** Ratio between V0_optim an V0 **Pmax_optim** Optimal maximal power estimated F0_optim * V0_optim / 4 **Pmax_rel_optim** Optimal relative maximal power slope_optim Optimal FV profile slope profile_imb Percent ratio between slope and optimal slope time_optim Time to cover distance when profile is optimal time_gain Difference in time to cover distance between time_optimal and time **Ppeak_optim** Optimal peak power estimated quantitatively Ppeak_rel_optim Optimal relative peak power Ppeak_dist_optim Distance at which optimal peak power is manifested

Ppeak_time_optim Time at which optimal peak power is manifested optimal_MSS_MAC returns a data frame with the following columns MSS Original MSS MAC Original MAC Pmax_rel Relative maximal power estimated using MSS * MAC / 4 slope Sprint profile slope distance Distance time Time to cover distance MSS_optim Optimal MSS MSS_coef Ratio between MSS_optim an MSS MAC_optim Optimal MAC MAC_coef Ratio between MAC_optim an MAC Pmax_rel_optim Optimal relative maximal power estimated using MSS_optim * MAC_optim / 4 slope_optim Optimal sprint profile slope profile_imb Percent ratio between slope and optimal slope time_optim Time to cover distance when profile is optimal time_gain Difference in time to cover distance between time_optimal and time

References

Samozino P, Peyrot N, Edouard P, Nagahara R, Jimenez-Reyes P, Vanwanseele B, Morin J. 2022. Optimal mechanical force-velocity profile for sprint acceleration performance. Scandinavian Journal of Medicine & Science in Sports 32:559–575. DOI: 10.1111/sms.14097.

```
MSS <- 10
MAC <- 8
bodymass <- 75
fv <- create_FVP(MSS, MAC, bodymass)
dist <- seq(5, 40, by = 5)
opt_MSS_MAC_profile <- optimal_MSS_MAC(
    distance = dist,
    MSS,
    MAC
)[["profile_imb"]]
opt_FV_profile <- optimal_FV(
    distance = dist,
    fv$F0,
    fv$V0,
    fv$bodymass
```

plot.shorts_model

```
)[["profile_imb"]]
opt_FV_profile_peak <- optimal_FV(
    distance = dist,
    fv$F0,
    fv$V0,
    fv$bodymass,
    method = "peak"
)[["profile_imb"]]
plot(x = dist, y = opt_MSS_MAC_profile, type = "1", ylab = "Profile imbalance")
lines(x = dist, y = opt_FV_profile, type = "1", col = "blue")
lines(x = dist, y = opt_FV_profile_peak, type = "1", col = "red")
abline(h = 100, col = "gray", lty = 2)
```

plot.shorts_model S3 method for plotting shorts_model object

Description

S3 method for plotting shorts_model object

Usage

S3 method for class 'shorts_model'
plot(x, type = "model", ...)

Arguments

x	shorts_model object
type	Type of plot. Can be "model" (default), "kinematics-time", "kinematics-distance", or "residuals"
	Not used

Value

ggplot object

```
# Simple model with radar gun data
instant_velocity <- data.frame(
   time = c(0, 1, 2, 3, 4, 5, 6),
   velocity = c(0.00, 4.99, 6.43, 6.84, 6.95, 6.99, 7.00)
)
radar_model <- with(
   instant_velocity,
   model_radar_gun(time, velocity)</pre>
```

```
)
plot(radar_model)
plot(radar_model, "kinematics-time")
plot(radar_model, "kinematics-distance")
plot(radar_model, "residuals")
```

predict.shorts_model S3 method for making predictions using shorts_model

Description

S3 method for making predictions using shorts_model

Usage

```
## S3 method for class 'shorts_model'
predict(object, ...)
```

Arguments

object	shorts_model object
	Forwarded to generic predict() function

Examples

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
   gates = split_distances,
   MSS = 10,
   MAC = 9,
   FD = 0.25,
   TC = 0
)
# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
predict(simple_model)</pre>
```

Description

Predicts kinematic from known MSS and MAC parameters

Usage

```
predict_velocity_at_time(time, MSS, MAC)
predict_distance_at_time(time, MSS, MAC)
predict_acceleration_at_time(time, MSS, MAC)
predict_time_at_distance(distance, MSS, MAC)
predict_time_at_distance_FV(
  distance,
  F0,
  V0,
  bodymass = 75,
  inertia = 0,
  resistance = 0,
  . . .
)
predict_velocity_at_distance(distance, MSS, MAC)
predict_acceleration_at_distance(distance, MSS, MAC)
predict_acceleration_at_velocity(velocity, MSS, MAC)
predict_air_resistance_at_time(time, MSS, MAC, ...)
predict_air_resistance_at_distance(distance, MSS, MAC, ...)
predict_force_at_velocity(
  velocity,
  MSS,
  MAC,
  bodymass = 75,
  inertia = 0,
  resistance = 0,
)
```

```
predict_force_at_time(
  time,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
 resistance = 0,
  . . .
)
predict_force_at_distance(
  distance,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
  resistance = 0,
  . . .
)
predict_power_at_distance(
 distance,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
  resistance = 0,
  . . .
)
predict_power_at_time(
  time,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
  resistance = 0,
  . . .
)
predict_relative_power_at_distance(
 distance,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
  resistance = 0,
  . . .
```

```
)
predict_relative_power_at_time(
  time,
 MSS,
 MAC,
 bodymass = 75,
 inertia = 0,
 resistance = 0,
  . . .
)
predict_work_till_time(time, ...)
predict_work_till_distance(distance, ...)
predict_kinematics(
 object = NULL,
 MSS,
 MAC,
 max_time = 6,
 frequency = 100,
 bodymass = 75,
 inertia = 0,
 resistance = 0,
 add_inertia_to_vertical = TRUE,
  . . .
)
```

Arguments

time, distance, velocity

	Numeric vectors	
MSS, MAC	Numeric vectors. Model parameters	
F0, V0	Numeric vectors. FV profile parameters	
bodymass	Body mass in kg. Used to calculate relative power and forwarded to get_air_resistance	
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation	
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)	
	Arguments passed on to get_air_resistance	
	bodyheight In meters (m). Default is 1.75m	
	barometric_pressure In Torrs. Default is 760Torrs	
	air_temperature In Celzius (C). Default is 25C	
	wind_velocity In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)	

object	If <code>shorts_model</code> object is provided, estimated parameters will be used. Otherwise provide MSS and MAC parameters	
max_time	Predict from 0 to max_time. Default is 6seconds	
frequency	Number of samples within one second. Default is 100Hz	
add_inertia_to_	vertical	
	Should inertia be added to bodymass when calculating vertical force? Use TRUE	
	(Default) when using weight vest, and FALSE when dragging sled	

Value

Numeric vector

Data frame with kinetic and kinematic variables

References

Haugen TA, Tønnessen E, Seiler SK. 2012. The Difference Is in the Start: Impact of Timing and Start Procedure on Sprint Running Performance: Journal of Strength and Conditioning Research 26:473–479. DOI: 10.1519/JSC.0b013e318226030b.

Jovanović, M., Vescovi, J.D. (2020). shorts: An R Package for Modeling Short Sprints. Preprint available at SportRxiv. https://doi.org/10.31236/osf.io/4jw62

Samozino P. 2018. A Simple Method for Measuring Force, Velocity and Power Capabilities and Mechanical Effectiveness During Sprint Running. In: Morin J-B, Samozino P eds. Biomechanics of Training and Testing. Cham: Springer International Publishing, 237–267. DOI: 10.1007/978-3-319-05633-3_11.

```
MSS <- 8
MAC <- 9
time_seq <- seq(0, 6, length.out = 10)
df <- data.frame(
   time = time_seq,
   distance_at_time = predict_distance_at_time(time_seq, MSS, MAC),
   velocity_at_time = predict_velocity_at_time(time_seq, MSS, MAC),
   acceleration_at_time = predict_acceleration_at_time(time_seq, MSS, MAC))
)
df$time_at_distance <- predict_time_at_distance(df$distance_at_time, MSS, MAC)
df$velocity_at_distance <- predict_velocity_at_distance(df$distance_at_time, MSS, MAC)
df$velocity_at_distance <- predict_velocity_at_distance(df$distance_at_time, MSS, MAC)
df$acceleration_at_distance <- predict_acceleration_at_distance(df$distance_at_time, MSS, MAC)
df$acceleration_at_velocity <- predict_acceleration_at_velocity(df$velocity_at_time, MSS, MAC)
# Power calculation uses shorts::get_air_resistance function and its defaults
# values to calculate power. Use the ... to setup your own parameters for power
```

```
# values to calculate power. Use the ... to setup your own parameter
# calculations
df$power_at_time <- predict_power_at_time(
   time = df$time, MSS = MSS, MAC = MAC,
```

```
# Check shorts::get_air_resistance for available params
bodymass = 100, bodyheight = 1.85
)
df
# Example for predict_kinematics
split_times <- data.frame(
    distance = c(5, 10, 20, 30, 35),
    time = c(1.20, 1.96, 3.36, 4.71, 5.35)
)
# Simple model
simple_model <- with(
    split_times,
    model_timing_gates(distance, time)
)
predict_kinematics(simple_model)
```

print.shorts_model S3 method for printing shorts_model object

Description

S3 method for printing shorts_model object

Usage

```
## S3 method for class 'shorts_model'
print(x, ...)
```

Arguments

х	<pre>shorts_model object</pre>
	Not used

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
  gates = split_distances,
  MSS = 10,
  MAC = 9,
  FD = 0.25,
  TC = 0
)</pre>
```

```
# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
simple_model</pre>
```

probe_functions Probe profile functions

Description

Family of functions that serve a purpose of probing sprint or force-velocity profile. This is done by increasing individual sprint parameter for a percentage and calculating which parameter improvement yield biggest deduction in sprint tim

probe_FV "probes" F0 and V0 and calculates which one improves sprint time for a defined distance

probe_MSS_MAC "probes" MSS and MAC and calculates which one improves sprint time for a defined distance

Usage

```
probe_FV(
   distance,
   F0,
   V0,
   bodymass = 75,
   inertia = 0,
   resistance = 0,
   perc = 2.5,
   ...
)
```

probe_MSS_MAC(distance, MSS, MAC, perc = 2.5)

Arguments

distance	Numeric vector
F0, V0	Numeric vectors. FV profile parameters
bodymass	Body mass in kg
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
perc	Numeric vector. Probing percentage. Default is 2.5 percent
	Arguments passed on to get_air_resistance
	velocity Instantaneous running velocity in meters per second (m/s) bodyheight In meters (m). Default is 1.75m

	barometric_pressure In Torrs. Default is 760Torrs		
	air_temperature In Celzius (C). Default is 25C		
	wind_velocity In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)		
MSS, MAC	Numeric vectors. Model parameters		

Value

probe_FV returns a data frame with the following columns

F0 Original F0 V0 Original F0 bodymass Bodymass inertia Inertia resistance Resistance **Pmax** Maximal power estimated using F0 * V0 / 4 Pmax_rel Relative maximal power slope FV profile slope distance Distance time Time to cover distance probe_perc Probe percentage F0_probe Probing F0 F0 probe time Predicted time for distance when F0 is probed F0_probe_time_gain Difference in time to cover distance between time_optimal and time V0_probe Probing V0 **V0 probe time** Predicted time for distance when V0 is probed V0_probe_time_gain Difference in time to cover distance between time_optimal and time profile_imb Percent ratio between V0_probe_time_gain and F0_probe_time_gain probe_MSS_MAC returns a data frame with the following columns MSS Original MSS MAC Original MAC Pmax_rel Relative maximal power estimated using MSS * MAC / 4 slope Sprint profile slope distance Distance time Time to cover distance probe_perc Probe percentage MSS_probe Probing MSS MSS_probe_time Predicted time for distance when MSS is probed MSS_probe_time_gain Difference in time to cover distance between probe time and time MAC probe Probing MAC MAC_probe_time Predicted time for distance when MAC is probed MAC_probe_time_gain Difference in time to cover distance between probing time and time profile_imb Percent ratio between MSS_probe_time_gain and MAC_probe_time_gain

Examples

```
MSS <- 10
MAC <- 8
bodymass <- 75
fv <- create_FVP(MSS, MAC, bodymass)</pre>
dist <- seq(5, 40, by = 5)
probe_MSS_MAC_profile <- probe_MSS_MAC(</pre>
  distance = dist,
  MSS.
  MAC
)[["profile_imb"]]
probe_FV_profile <- probe_FV(</pre>
  distance = dist,
  fv$F0,
  fv$V0,
  fv$bodymass
)[["profile_imb"]]
plot(x = dist, y = probe_MSS_MAC_profile, type = "l", ylab = "Profile imbalance")
lines(x = dist, y = probe_FV_profile, type = "1", col = "blue")
abline(h = 100, col = "gray", lty = 2)
```

radar_gun_data Radar Gun Data

Description

Data generated from known MSS and TAU and measurement error for N=5 athletes using radar gun with sampling frequency of 100Hz over 6 seconds.

Usage

```
data(radar_gun_data)
```

Format

Data frame with 4 variables and 3000 observations:

athlete Character string

bodyweight Bodyweight in kilograms

time Time reported by the radar gun in seconds

velocity Velocity reported by the radar gun in m/s

residuals.shorts_model

S3 method for returning residuals of shorts_model

Description

S3 method for returning residuals of shorts_model

Usage

```
## S3 method for class 'shorts_model'
residuals(object, ...)
```

Arguments

object	shorts_model object
	Extra arguments. Not used

Examples

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
   gates = split_distances,
   MSS = 10,
   MAC = 9,
   FD = 0.25,
   TC = 0
)
# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
residuals(simple_model)</pre>
```

split_times Split Testing Data

Description

Data generated from known MSS and TAU and measurement error for N=5 athletes using 6 timing gates: 5m, 10m, 15m, 20m, 30m, 40m

Usage

data(split_times)

Format

Data frame with 4 variables and 30 observations:

athlete Character stringbodyweight Bodyweight in kilogramsdistance Distance of the timing gates from the sprint start in meterstime Time reported by the timing gate

summary.shorts_model S3 method for providing summary for the shorts_model object

Description

S3 method for providing summary for the shorts_model object

Usage

S3 method for class 'shorts_model'
summary(object, ...)

Arguments

object	shorts_model object
	Not used

Examples

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
   gates = split_distances,
   MSS = 10,
   MAC = 9,
   FD = 0.25,
   TC = 0
)
# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
summary(simple_model)</pre>
```

vescovi

Description

Timing gates sprint times involving 52 female athletes. Timing gates were located at 5m, 10m, 20m, 30m, and 35m. See **Details** for more information.

Usage

data(vescovi)

Format

Data frame with 17 variables and 52 observations:

Team Team or sport. Contains the following levels: 'W Soccer' (Women Soccer), 'FH Sr' (Field Hockey Seniors), 'FH U21' (Field Hockey Under 21), and 'FH U17' (Field Hockey Under 17)

Surface Type of testing surface. Contains the following levels: 'Hard Cours' and 'Natural Grass'

Athlete ID

Age Athlete age in years

Height Body height in cm

Bodyweight Body weight in kg

BMI Body Mass Index

BSA Body Surface Area. Calculated using Mosteller equation sqrt((height/weight)/3600)

5m Time in seconds at 5m gate

10m Time in seconds at 10m gate

20m Time in seconds at 20m gate

30m Time in seconds at 30m gate

35m Time in seconds at 35m gate

10m-5m split Split time in seconds between 10m and 5m gate

20m-10m split Split time in seconds between 20m and 10m gate

30m-20m split Split time in seconds between 30m and 20m gate

35m-30m split Split time in seconds between 35m and 30m gate

Details

This data-set represents sub-set of data from a total of 220 high-level female athletes (151 soccer players and 69 field hockey players). Using a random number generator, a total of 52 players (35 soccer and 17 field hockey) were selected for this data-set. Soccer players were older (24.6±3.6 vs. 18.9±2.7 yr, p < 0.001), however there were no differences for height (167.3±5.9 vs. 167.0±5.7 cm, p = 0.886), body mass (62.5±5.9 vs. 64.0±9.4 kg, p = 0.500) or any sprint interval time (p > 0.650).

The protocol for assessing linear sprint speed has been described previously (Vescovi 2014, 2016, 2012) and was identical for each cohort. Briefly, all athletes performed a standardized warm-up that included general exercises such as jogging, shuffling, multi-directional movements, and dynamic stretching exercises. Infrared timing gates (Brower Timing, Utah) were positioned at the start line and at 5, 10, 20, and 35 meters at a height of approximately 1.0 meter. Participants stood with their lead foot positioned approximately 5 cm behind the initial infrared beam (i.e., start line). Only forward movement was permitted (no leaning or rocking backwards) and timing started when the laser of the starting gate was triggered. The best 35 m time, and all associated split times were kept for analysis. The assessment of linear sprints using infrared timing gates does not require familiarization (Moir, Button, Glaister, and Stone 2004).

Author(s)

Jason D. Vescovi University of Toronto Faculty of Kinesiology and Physical Education Graduate School of Exercise Science Toronto, ON Canada <vescovij@gmail.com>

References

Moir G, Button C, Glaister M, Stone MH (2004). "Influence of Familiarization on the Reliability of Vertical Jump and Acceleration Sprinting Performance in Physically Active Men." The Journal of Strength and Conditioning Research, 18(2), 276. ISSN 1064-8011, 1533-4287. doi:10.1519/R-13093.1.

Vescovi JD (2012). "Sprint Speed Characteristics of High-Level American Female Soccer Players: Female Athletes in Motion (FAiM) Study." Journal of Science and Medicine in Sport, 15(5), 474-478. ISSN 14402440. doi:10.1016/j.jsams.2012.03.006.

Vescovi JD (2014). "Impact of Maximum Speed on Sprint Performance During High-Level Youth Female Field Hockey Matches: Female Athletes in Motion (FAiM) Study." International Journal of Sports Physiology and Performance, 9(4), 621-626. ISSN 1555-0265, 1555-0273. doi:10.1123/ijspp.2013-0263.

Vescovi JD (2016). "Locomotor, Heart-Rate, and Metabolic Power Characteristics of Youth Women's Field Hockey: Female Athletes in Motion (FAiM) Study." Research Quarterly for Exercise and Sport, 87(1), 68-77. ISSN 0270-1367, 2168-3824. doi:10.1080/02701367.2015.1124972.

Index

* datasets dynaspeed, 8 jb_morin, 16 laser_gun_data, 17 LPS_session, 18 radar_gun_data, 38 split_times, 39 vescovi, 41 coef.shorts_model, 2 confint.shorts_model, 3 convert_FVP, 4 create_FVP, 5 create_sprint_trace, 6, 8 create_timing_gates_splits, 7, 7 dynaspeed, 8 find_acceleration_critical_distance (find_functions), 9 find_acceleration_critical_time (find_functions), 9 find_functions, 9 find_optimal_distance, 12 find_peak_power_distance (find_functions), 9 find_peak_power_time (find_functions), 9 find_power_critical_distance (find_functions), 9 find_power_critical_time (find_functions), 9 find_velocity_critical_distance (find_functions), 9 find_velocity_critical_time (find_functions), 9 fitted.shorts_model, 13 format_splits, 14 get_air_resistance, 4, 5, 10, 15, 27, 33, 36 ggplot, 29

jb_morin, 16 laser_gun_data, 17 LPS_session, 18 model_functions, 18 model_in_situ (model_functions), 18 model_laser_gun (model_functions), 18 model_radar_gun, 18 model_radar_gun (model_functions), 18 model_tether (model_functions), 18 model_tether_DC (model_functions), 18 model_time_distance (model_functions), 18 model_time_distance_DC (model_functions), 18 model_time_distance_FD (model_functions), 18 model_time_distance_FD_fixed (model_functions), 18 model_time_distance_TC (model_functions), 18 model_time_distance_TC_DC (model_functions), 18 model_timing_gates (model_functions), 18 model_timing_gates_DC (model_functions), 18 model_timing_gates_FD (model_functions), 18 model_timing_gates_FD_fixed (model_functions), 18 model_timing_gates_TC (model_functions), 18 model_timing_gates_TC_DC (model_functions), 18

nlsLM, 22, 23

optimal_functions, 26
optimal_FV, 12

INDEX

optimal_FV (optimal_functions), 26 optimal_MSS_MAC (optimal_functions), 26 plot.shorts_model, 29 predict.shorts_model, 30 predict_acceleration_at_distance (predict_kinematics), 31 predict_acceleration_at_time (predict_kinematics), 31 predict_acceleration_at_velocity (predict_kinematics), 31 predict_air_resistance_at_distance (predict_kinematics), 31 predict_air_resistance_at_time (predict_kinematics), 31 predict_distance_at_time (predict_kinematics), 31 predict_force_at_distance (predict_kinematics), 31 predict_force_at_time (predict_kinematics), 31 predict_force_at_velocity (predict_kinematics), 31 predict_kinematics, 31 predict_power_at_distance, 4, 5 predict_power_at_distance (predict_kinematics), 31 predict_power_at_time (predict_kinematics), 31 predict_relative_power_at_distance (predict_kinematics), 31 predict_relative_power_at_time (predict_kinematics), 31 predict_time_at_distance (predict_kinematics), 31 predict_time_at_distance_FV (predict_kinematics), 31 predict_velocity_at_distance (predict_kinematics), 31 predict_velocity_at_time (predict_kinematics), 31 predict_work_till_distance (predict_kinematics), 31 predict_work_till_time (predict_kinematics), 31 print.shorts_model, 35 probe_functions, 36 probe_FV (probe_functions), 36 probe_MSS_MAC (probe_functions), 36

radar_gun_data, 38
residuals.shorts_model, 39

split_times, 39
summary.shorts_model, 40

vescovi, 41