

# Package ‘MATA’

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**Title** Model-Averaged Tail Area (MATA) Confidence Interval and Distribution

**Version** 0.7.1

**Description** Calculates Model-Averaged Tail Area Wald (MATA-Wald) confidence intervals, and MATA-Wald confidence densities and distributions, which are constructed using single-model frequentist estimators and model weights. See Turek and Fletcher (2012) <[doi:10.1016/j.csda.2012.03.002](https://doi.org/10.1016/j.csda.2012.03.002)> and Fletcher et al (2019) <[doi:10.1007/s10651-019-00432-5](https://doi.org/10.1007/s10651-019-00432-5)> for details.

**License** GPL-2

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## Description

Functions for computing the Model-Averaged Tail Area Wald (MATA-Wald) confidence interval, density, and distribution. These are all constructed using single-model frequentist estimators and model weights. See details.

**Usage**

```

mata.wald(
  theta.hats,
  se.theta.hats,
  model.weights,
  mata.t,
  residual.dfs,
  alpha = 0.025
)

dmata.wald(
  theta,
  theta.hats,
  se.theta.hats,
  model.weights,
  mata.t,
  residual.dfs
)

pmata.wald(
  theta,
  theta.hats,
  se.theta.hats,
  model.weights,
  mata.t,
  residual.dfs
)

```

**Arguments**

<code>theta.hats</code>	A numeric vector containing the parameter estimates under each candidate model.
<code>se.theta.hats</code>	A numeric vector containing the estimated standard error of each value in <code>theta.hats</code> . If an element is zero, this corresponds to the parameter being fixed to the value give in <code>theta.hats</code> under a particular candidate model.
<code>model.weights</code>	A vector containing the weights for each candidate model (e.g. AIC weights or stacking weights). All model weights must be non-negative, and sum to one.
<code>mata.t</code>	Logical. TRUE for the normal linear model case, and FALSE otherwise. When TRUE, the argument <code>residual.dfs</code> must also be supplied.
<code>residual.dfs</code>	A vector containing the residual (error) degrees of freedom under each candidate model. This argument must be provided when <code>mata.t = TRUE</code> .
<code>alpha</code>	For <code>mata.wald</code> only, the desired lower and upper error rate. The value 0.025 corresponds to a 95% MATA-Wald confidence interval, and 0.05 to a 90% interval. Must be between 0 and 0.5. Default value is 0.025.
<code>theta</code>	For <code>dmata.wald</code> and <code>pmata.wald</code> only, a vector of <code>theta</code> values at which to evaluate the model-averaged confidence density or distribution function.

## Details

`mata.wald` may be used to construct model-averaged confidence intervals, using the Model-Averaged Tail Area (MATA) construction (see Turek and Fletcher (2012) for details). The idea underlying this construction is similar to that of a model-averaged Bayesian credible interval. This function returns the lower and upper confidence limits of a MATA-Wald interval.

Closely related, the `dmata.wald` and `pmata.wald` functions evaluate the MATA-Wald confidence density and confidence distribution functions, which were developed by Fletcher et al. (2019).

Two usages are supported. For the normal linear model, or any other model where a t-based interval is appropriate (e.g., quasi-poisson), using option `mata.t = TRUE` corresponds to a MATA-Wald confidence interval, density, or distribution corresponding to the solutions of equations (2) and (3) of Turek and Fletcher (2012). The argument `residual.dfs` is required for this usage.

When the sampling distribution for the estimator is asymptotically normal (e.g. MLEs), possibly after a transformation, use option `mata.t = FALSE`. This corresponds to solutions to the equations in Section 3.2 of Turek and Fletcher (2012).

If the parameter is fixed to a certain value under one or more candidate models, this value should be provided in the `theta.hats` argument, along with a corresponding value of zero in `se.theta.hats`. For the model-averaged confidence density, this results in a point mass equal to the sum of the weights for these models.

## Author(s)

Daniel Turek

## References

- Turek, D. and Fletcher, D. (2012). Model-Averaged Wald Confidence Intervals. *Computational Statistics and Data Analysis*, 56(9): 2809-2815.
- Fletcher, D. (2018). *Model Averaging*. Berlin, Heidelberg: Springer Briefs in Statistics.
- Fletcher, D., Dillingham, P. W., and Zeng, J. (2019). Model-averaged confidence distributions. *Environmental and Ecological Statistics*, 26: 367–384.

## See Also

`mata.wald`, `dmata.wald`, `pmata.wald`

## Examples

```
# Normal linear prediction:
# Generate single-model Wald and model-averaged MATA-Wald 95% confidence intervals
#
# Data 'y', covariates 'x1' and 'x2', all vectors of length 'n'.
# 'y' taken to have a normal distribution.
# 'x1' specifies treatment/group (factor).
# 'x2' a continuous covariate.
#
# Take the quantity of interest (theta) as the predicted response
# (expectation of y) when x1=1 (second group/treatment), and x2=15.
```

```

set.seed(0)
n = 20                                # 'n' is assumed to be even
x1 = c(rep(0,n/2), rep(1,n/2))      # two groups: x1=0, and x1=1
x2 = rnorm(n, mean=10, sd=3)
y = rnorm(n, mean = 3*x1 + 0.1*x2)  # data generation

x1 = factor(x1)
m1 = glm(y ~ x1)                     # using 'glm' provides AIC values.
m2 = glm(y ~ x1 + x2)                # using 'lm' doesn't.
aic = c(m1$aic, m2$aic)
delta.aic = aic - min(aic)
model.weights = exp(-0.5*delta.aic) / sum(exp(-0.5*delta.aic))
residual.dfs = c(m1$df.residual, m2$df.residual)

p1 = predict(m1, se=TRUE, newdata=list(x1=factor(1), x2=15))
p2 = predict(m2, se=TRUE, newdata=list(x1=factor(1), x2=15))
theta.hats = c(p1$fit, p2$fit)
se.theta.hats = c(p1$se.fit, p2$se.fit)

# AIC model weights
model.weights

# 95% Wald confidence interval for theta (under Model 1)
theta.hats[1] + c(-1,1)*qt(0.975, residual.dfs[1])*se.theta.hats[1]

# 95% Wald confidence interval for theta (under Model 2)
theta.hats[2] + c(-1,1)*qt(0.975, residual.dfs[2])*se.theta.hats[2]

# 95% MATA-Wald confidence interval for theta:
mata.wald(theta.hats, se.theta.hats, model.weights, mata.t = TRUE, residual.dfs = residual.dfs)

# Plot the model-averaged confidence density and distribution functions
# on the interval [2, 7]
thetas <- seq(2, 7, by = 0.1)

dens <- dmata.wald(thetas, theta.hats, se.theta.hats, model.weights,
                  mata.t = TRUE, residual.dfs = residual.dfs)

dists <- pmata.wald(thetas, theta.hats, se.theta.hats, model.weights,
                  mata.t = TRUE, residual.dfs = residual.dfs)

par(mfrow = c(2,1))
plot(thetas, dens, type = 'l', main = 'Model-Averaged Confidence Density')
plot(thetas, dists, type = 'l', main = 'Model-Averaged Confidence Distribution')

```

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