

# Package ‘metaSDTreg’

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**Title** Regression Models for Meta Signal Detection Theory

**Version** 0.2.1

**Description** Regression methods for the meta-SDT model. The package implements methods for cognitive experiments of metacognition as described in Kristensen, S. B., Sandberg, K., & Bibby, B. M. (2020). Regression methods for metacognitive sensitivity. *Journal of Mathematical Psychology*, 94. <doi:10.1016/j.jmp.2019.102297>.

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coef.metaSDTreg	<i>Coefficients method for metaSDTreg</i>
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---

### Description

Coefficients method for metaSDTreg

### Usage

```
## S3 method for class 'metaSDTreg'
coef(object, ...)
```

### Arguments

object	An object of class metaSDTreg.
...	For future methods.

### Value

A named vector of parameter estimates.

### Examples

```
## Declare simulated data as metaSDTdata
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

## Fit model to subset of data
fit <- metaSDTreg(A ~ signal,
                 data=metadata,
                 subset = m <= 20)
coef(fit)
```

---

lines.predict_roc	<i>Lines of predicted ROC curve</i>
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---

## Description

Plot lines method for objects of class 'predict\_roc'.

## Usage

```
## S3 method for class 'predict_roc'  
lines(x, thr = seq(-10, 10, length = 1000), ...)
```

## Arguments

x	A 'predict_roc' object to plot.
thr	The sequence of thresholds parametrising the ROC curve, if this is a function. Default to a length 1000 sequence from -10 to 10.
...	Additional arguments passed to lines.

## Value

Invisible.

## Examples

```
## Declare simulated data as metaSDTdata  
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')  
  
## Fit model to subset of data  
fit <- metaSDTreg(A ~ signal,  
                 data=metadata,  
                 subset = m <= 20)  
  
## Plot observed type 1 ROC points  
plot(predict_roc(metadata, type = '1'), xlim = 0:1, ylim = 0:1)  
  
## Add Model-predicted ROC curve (estimated from subset of data)  
lines(predict_roc(fit, type = '1'))
```

---

logLik.metaSDTreg      *Log-likelihood of metaSDTreg*

---

### Description

Extract the log-likelihood from a metaSDTreg object.

### Usage

```
## S3 method for class 'metaSDTreg'
logLik(object, ...)
```

### Arguments

object      Object of class 'metaSDTreg'.  
 ...      For further methods.

### Value

Numeric, the likelihood at the maximum found by the optimisation procedure.

### Examples

```
## Declare simulated data as metaSDTdata
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

## Fit model to subset of data
fit <- metaSDTreg(A ~ signal,
                 data=metadata,
                 subset = m <= 20)
logLik(fit)
```

---

metaSDTcontrol      *Control for metaSDTreg*

---

### Description

Control function for metaSDTreg. The function allows passing arguments to the maxLik and clm procedure used for optimisation and starting values, respectively.

**Usage**

```
metaSDTcontrol(
  meth = "BFGS",
  start.vals = NULL,
  hessianMeth = TRUE,
  returnMaxLik = FALSE,
  loglikUse = c("R", "C++"),
  scoreFun = FALSE,
  c.maxLik = list(),
  c.clm = list(maxIter = 1000, maxModIter = 20),
  ...
)
```

**Arguments**

meth	Method used by <code>maxLik</code> to perform maximisation, which must support constrained optimisation, see <code>maxLik</code> . Defaults to 'BFGS'.
start.vals	Starting values for the maximisation algorithm. Any NA entries will be set to zero. Defaults to NULL, in which case starting values will be obtained from the partial proportional odds model as implemented in the package <b>ordinal</b> using <code>clm</code> .
hessianMeth	Method to compute the Hessian of the optimisation. Passed to the argument <code>finalHessian</code> in <code>maxLik</code> .
returnMaxLik	Should the function return the fit object from <code>maxLik</code> rather than the <code>metaSDTreg</code> object? Defaults to FALSE.
loglikUse	For testing purposes. The C++ option is not presently included. Should the procedure use the log-likelihood implemented in R (default), or that in C++? Option will be dropped in future versions.
scoreFun	For testing purposes. Should the optimisation procedure use the analytic score function? If FALSE (and 'meth' relies on gradients) numerical methods are used. Presently, only numerical methods are implemented. Defaults to FALSE. Option default may change in future versions.
c.maxLik	A list that is passed to the 'MaxControl' object in <code>maxLik</code> . The list should contain control arguments to be passed to the maximisation method used by <code>maxLik</code> (the default being <code>maxBFGS</code> ). In particular, 'iterlim' is the number of iterations (default 200) and a larger number of 'printlevel' (default 0) prints more information from the maximisation procedure.
c.clm	A list that is passed to <code>clm.control</code> containing controls to be used by <code>clm</code> . The <code>clm</code> is used to determine starting values. By default, <code>maxIter</code> is set to 1000 and <code>maxModIter</code> is set to 20.
...	For future methods.

**Value**

A list of class 'metaSDTcontrol' containing the control arguments.

**Examples**

```
names(metaSDTcontrol())
```

---

metaSDTdata	<i>Construct metaSDTdata</i>
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---

**Description**

Constructor function for a metaSDTdata object.

**Usage**

```
metaSDTdata(data, type1, type2, signal)
```

**Arguments**

data	Data frame to be converted. Data should be in long format with one row corresponding to a single response.
type1	A string naming the variable containing the type 1 response, which should be an ordered factor with two levels where the first level corresponds to 'noise',
type2	A string naming the variable containing the ordinal type 2 response, which should be an ordered factor.
signal	A string naming the variable containing the signal.

**Details**

If type1 or type 2 is not an ordered factor, the function returns a warning. The function constructs a data frame containing variables named c('type1','type2','A','signal') along with any other variable in 'data' that is not given as an argument to the function. Because of this a warning is issued when the names c('type1','type2','A','signal') are present in 'data'.

**Value**

A data object of class 'metaSDTdata'. This has attributes 'L', the number of levels in the ordinal type 2 rating, and 'K' which is two times L (the number of levels of the ordinal variable 'A').

**Examples**

```
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

summary(metadata)
summary.data.frame(metadata)
```

---

metaSDTreg	<i>Meta-SDT regression</i>
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---

## Description

Fit the meta SDT regression model.

## Usage

```
metaSDTreg(
  formula,
  data,
  subset,
  na.action = na.fail,
  control = metaSDTcontrol()
)
```

## Arguments

formula	Formula specifying the regression model. Presently, the left-hand side should be the ordinal variable A, while the right-hand side must contain the signal variable from the metaSDTdata object. Note that the variable 'signal' has a special interpretation in the model, see 'Details' below.
data	Data frame to fit the model on. Should be declared as metaSDTdata using the <a href="#">metaSDTdata</a> function.
subset	Optional argument specifying a subset of the data to be used. A logical statement, which is evaluated in 'data'. Only rows with TRUE are used in the analysis.
na.action	Method for handling missing values, see na.action. Default is na.fail which stops the function with an error in the presence of missing values among variables entering 'formula'.
control	Optimisation control, see <a href="#">metaSDTcontrol</a> .

## Details

The input data frame should be of class 'metaSDTdata' as constructed by the function [metaSDTdata](#). This will contain a variable, 'signal' which presently must be included on the right-hand side of the formula. The left hand side must be the ordinal variable 'A' also contained in the metaSDTdata object. The 'signal' variable has a special interpretation as its coefficients are the signal sensitivities (type 1 d-prime and the two response specific type 2 sensitivities meta-d-prime). Presently all other variables, as determined by `all.vars(update(formula, . ~ . -signal))`, enter proportionally in the model, i.e. they share coefficients across response scale levels. See Kristensen & al. (2020) for details. This can be written formally as follows (in the notation of Kristensen & al. (2020)):

$$V_N = V_S = V,$$

and  $\beta_N$ ,  $\beta_S$  and  $\beta$  agree on all but the first entrance which is the signal sensitivities  $d_{mathcal{N}}$ ,  $d_{mathcal{S}}$  and  $d$ , respectively. Note that 'formula' specifies the mean of the latent variables and not the threshold model. Accordingly, attempts to remove the intercept in 'formula' will be ignored with a warning. The function fails when the left-hand side of 'formula' is not the ordinal variable 'A' and when 'data' is not a 'metaSDTdata' object. Future versions may be less defensive and simply issue a warning in these cases. Note that constrained optimisation is used to maximise the likelihood under ordinality of the thresholds. The variance-covariance matrix is not adjusted following the constrained estimation, cf. documentation of `maxLik`. When interpreting results from `summary.metaSDTreg()` care should be taken regarding the z-tester and associated p-value for threshold parameters, as the distribution of the statistic depends on the null hypothesis which will only be reasonable under special circumstances as in other types of ordinal models.

### Value

An object of class 'metaSDTreg'. This is a list object containing,

- `logLik`: The log likelihood after optimisation.
- `coefficients`: Estimated coefficients.
- `vcov`: Variance-covariance matrix of the `maxLik` object.
- `call`: The call issued to `metaSDTreg`.
- `na.act`: The NA action.

### References

Kristensen, S. B., Sandberg, K., & Bibby, B. M. (2020). Regression methods for metacognitive sensitivity. *Journal of Mathematical Psychology*, 94. <doi:10.1016/j.jmp.2019.102297>.

### Examples

```
## Declare simulated data as metaSDTdata
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

## Fit function to data of first 20 replications per subject
fit_sub <- metaSDTreg(A ~ signal,
                    data=metadata,
                    subset = m <= 20)
summary(fit_sub)

## Fit model to estimate thresholds and sensitivities
fit <- metaSDTreg(A ~ signal,
                 data=metadata)

## True values are
## c(tau.n.2=-0.5, tau.n.1=-0.2, theta=0,
##   tau.s.1=0.8, tau.s.2=1, d.n=0.5, d=1, d.s=0.75)
coef(fit)
```



---

plot.predict_roc	<i>Plot predicted ROC curve</i>
------------------	---------------------------------

---

### Description

Plot method for objects of class 'predict\_roc'.

### Usage

```
## S3 method for class 'predict_roc'  
plot(x, thr = seq(-10, 10, length = 1000), rline = TRUE, ...)
```

### Arguments

x	A 'predict_roc' object to plot.
thr	The sequence of thresholds parametrising the ROC curve, if this is a function. Default to a length 1000 sequence from -10 to 10.
rline	Logical, should the line of random discrimination be added to the plot? Defaults to TRUE.
...	Additional arguments passed to plot.

### Details

If neither 'xlab' nor 'ylab' is passed to plot the function supplies default x- and y-axis labels based on the type of ROC curve.

### Value

Invisible.

### Examples

```
## Declare simulated data as metaSDTdata  
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')  
  
## Fit model to subset of data  
fit <- metaSDTreg(A ~ signal,  
                 data=metadata,  
                 subset = m <= 20)  
  
## Model-predicted signal-specific ROC curve  
signalROC <- predict_roc(fit, type = 's')  
plot(signalROC, type = 'l')
```

---

points.predict\_roc      *Points from predicted ROC curve*

---

## Description

Plot points method for objects of class 'predict\_roc'.

## Usage

```
## S3 method for class 'predict_roc'  
points(x, thr = seq(-10, 10, length = 1000), ...)
```

## Arguments

x	A 'predict_roc' object to plot.
thr	The sequence of thresholds parametrising the ROC curve, if this is a function. Default to a length 1000 sequence from -10 to 10.
...	Additional arguments passed to lines.

## Value

Invisible.

## Examples

```
## Declare simulated data as metaSDTdata  
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')  
  
## Fit model to subset of data  
fit <- metaSDTreg(A ~ signal,  
                  data=metadata,  
                  subset = m <= 20)  
  
## Plot observed type 1 ROC points  
plot(predict_roc(metadata, type = '1'), xlim = 0:1, ylim = 0:1, pch = 'x')  
  
## Add Model-predicted ROC curve (estimated from subset of data)  
points(predict_roc(fit, type = '1'))
```

---

predict_roc	<i>Generic predict_roc method</i>
-------------	-----------------------------------

---

**Description**

Predict ROC curve from signal detection theory model.

**Usage**

```
predict_roc(object, ...)
```

**Arguments**

object	Signal detection theory model.
...	Further arguments passed to methods.

**Value**

An object of class 'predict\_roc'.

**See Also**

[predict\\_roc.metaSDTreg](#), [predict\\_roc.metaSDTdata](#).

---

predict_roc.metaSDTdata	<i>Observed ROC points</i>
-------------------------	----------------------------

---

**Description**

The observed points of the ROC curve from a 'metaSDTdata' object.

**Usage**

```
## S3 method for class 'metaSDTdata'
predict_roc(object, type = c("1", "n", "s"), s0 = 0, s1 = 1, ...)
```

**Arguments**

object	A 'metaSDTdata' object from which to calculate observed ROC points.
type	The type of ROC curve to predict. A character string, where '1' requests the type 1 ROC curve, 'n' requests the type 2 noise-specific and 's' the type 2 signal-specific ROC curve.
s0	Numeric, the value of object\$signal to regard as 'noise'. Defaults to 0.
s1	Numeric, the value of object\$signal to regard as 'signal'. Defaults to 1.
...	For future methods

**Details**

Note that the type 1 ROC points arise by using each criterion in turn to decide between 'signal' and 'noise'. Since this involves also the type 2 thresholds, such a curve is also sometimes referred to as a 'pseudo' ROC curve.

**Value**

A matrix two-column matrix of class 'predict\_roc' with one row of c(FA, HR) per threshold (FA: False Alarm rate, HR: Hit Rate).

**References**

Maniscalco, B., & Lau, H. (2014). Signal Detection Theory Analysis of Type 1 and Type 2 Data: Meta-d , Response-Specific Meta-d , and the Unequal Variance SDT Model. In S. M. Fleming, & C. D. Frith (Eds.), *The Cognitive Neuroscience of Metacognition* (pp. 25-66). : Springer Berlin Heidelberg.

**Examples**

```
## Declare simulated data as metaSDTdata
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

## Observed signal-specific ROC curve
signalROC <- predict_roc(metadata, type = 's')
```

---

predict\_roc.metaSDTreg

*Predicted ROC curve*

---

**Description**

Predict ROC curves from metaSDTreg object.

**Usage**

```
## S3 method for class 'metaSDTreg'
predict_roc(object, type = c("1", "n", "s"), s0 = 0, s1 = 1, ...)
```

**Arguments**

object	An object of class metaSDTreg.
type	The type of ROC curve to predict. A character string, where '1' requests the type 1 ROC curve (the default), 'n' requests the type 2 noise-specific and 's' the type 2 signal-specific ROC curve.
s0	Numeric, the value of 'signal' to regard as 'noise'. Defaults to 0.
s1	Numeric, the value of 'signal' to regard as 'signal'. Defaults to 1.
...	For future methods

**Details**

The 'metaSDTreg' object given to the function must have named coefficients with names as they would be if metaSDTreg is run without user-supplied starting values.

A ROC curve is a 2-D curve parametrised by some  $x$  given by  $c(\text{FA}(x), \text{HR}(x))$  where FA is the false alarm rate and HR is the hit rate. For example, for type 1 ROC,

$$\text{FA}(x) = 1 - \text{pnorm}(x - s0 * d),$$

$$\text{HR}(x) = 1 - \text{pnorm}(x - s1 * d),$$

where  $d$  is the signal sensitivity.

Note that the predicted ROC curve is for a reference individual in the regression, i.e. additional covariates are not entered into the ROC so that reparametrisation of the 'metaSDTreg' model is needed to change predictions.

**Value**

A function of class 'predict\_roc' containing the appropriate ROC curve. This is a function of  $x$  which returns  $c(\text{FA}, \text{HR})$ , where FA is the false alarm rate and HR is the hit rate.

**References**

Maniscalco, B., & Lau, H. (2014). Signal Detection Theory Analysis of Type 1 and Type 2 Data: Meta-d , Response-Specific Meta-d , and the Unequal Variance SDT Model. In S. M. Fleming, & C. D. Frith (Eds.), The Cognitive Neuroscience of Metacognition (pp. 25 66). : Springer Berlin Heidelberg.

**Examples**

```
## Declare simulated data as metaSDTdata
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

## Fit model to subset of data
fit <- metaSDTreg(A ~ signal,
                 data=metadata,
                 subset = m <= 20)

## Model-predicted signal-specific ROC curve
signalROC <- predict_roc(fit, type = 's')
```

---

print.metaSDTdata      *Print method for metaSDT data*

---

**Description**

Print method for metaSDT data

**Usage**

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

**Arguments**

x	The metaSDTdata object.
...	Additional arguments passed to print.data.frame

**Value**

Invisible.

**Examples**

```
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')  
print(metadata)
```

---

print.metaSDTreg	<i>Print method for metaSDTreg</i>
------------------	------------------------------------

---

**Description**

Print method for metaSDTreg

**Usage**

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

**Arguments**

x	An object of class metaSDTreg.
...	For future methods.

**Value**

Invisible.

**Examples**

```
## Declare simulated data as metaSDTdata  
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')  
  
## Fit model to subset of data  
(fit <- metaSDTreg(A ~ signal,  
  data=metadata,  
  subset = m <= 20))
```

---

```
print.summary.metaSDTdata
```

*Print method for a summary.metaSDTdata object*

---

**Description**

Print method for a summary.metaSDTdata object

**Usage**

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

**Arguments**

x	A metaSDTdata object.
...	For future methods.

**Value**

Invisible.

**Examples**

```
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')  
summary(metadata)
```

---

```
print.summary.metaSDTreg
```

*Print summary method for metaSDTreg*

---

**Description**

Print summary method for metaSDTreg

**Usage**

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

**Arguments**

x	An object of class summary.metaSDTreg.
...	For future methods.

**Value**

Invisible.

**Examples**

```
## Declare simulated data as metaSDTdata
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

## Fit model to subset of data
fit <- metaSDTreg(A ~ signal,
                 data=metadata,
                 subset = m <= 20)
summary(fit)
```

---

simMetaData

*Simulated metacognition experiment*

---

**Description**

Data is simulated from the meta-SDT model, formula (3)-(4) in Kristensen, Sandberg and Bibby (2020). The true values of the simulation parameters are  $c(\tau_{n.2}=-0.5, \tau_{n.1}=-0.2, \theta=0, \tau_{s.1}=0.8, \tau_{s.2}=1, d_{n.1}=0.5, d_{n.2}=1, d_{s.1}=0.75)$ .

**Usage**

```
simMetaData
```

**Format**

A data frame with 25000 obs. of 5 variables:

**id** Subject id.

**m** Observation id, within subject.

**S** Signal indicator (0 or 1).

**resp** Type 1 response (0 or 1).

**conf** Type 2 response, ordered (1, 2 or 3).

**References**

Kristensen, S. B., Sandberg, K., & Bibby, B. M. (2020). Regression methods for metacognitive sensitivity. *Journal of Mathematical Psychology*, 94. <doi:10.1016/j.jmp.2019.102297>.



---

summary.metaSDTdata     *Summarise a metaSDTdata object as a cognitive experiment.*

---

**Description**

Summarise a metaSDTdata object as a cognitive experiment.

**Usage**

```
## S3 method for class 'metaSDTdata'  
summary(object, ...)
```

**Arguments**

object             The metaSDTdata object.  
...                Additional arguments. Presently not used.

**Value**

A list with class 'summary.metaSDTdata' containing summaries.

**Examples**

```
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')  
summary(metadata)
```

---

summary.metaSDTreg     *Summary method for metaSDTreg*

---

**Description**

Summary method for metaSDTreg

**Usage**

```
## S3 method for class 'metaSDTreg'  
summary(object, ...)
```

**Arguments**

object             An object of class metaSDTreg.  
...                For future methods.

**Value**

A list with class 'summary.metaSDTreg' containing summaries.

**Examples**

```
## Declare simulated data as metaSDTdata
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

## Fit model to subset of data
fit <- metaSDTreg(A ~ signal,
                 data=metadata,
                 subset = m <= 20)
summary(fit)
```

---

vcov.metaSDTreg

*Variance-covariance method for metaSDTreg*

---

**Description**

Variance-covariance method for metaSDTreg

**Usage**

```
## S3 method for class 'metaSDTreg'
vcov(object, ...)
```

**Arguments**

object	An object of class metaSDTreg.
...	For future methods.

**Value**

A matrix of variances and covariances.

**Examples**

```
## Declare simulated data as metaSDTdata
metadata <- metaSDTdata(simMetaData, type1='resp', type2='conf', signal='S')

## Fit model to subset of data
fit <- metaSDTreg(A ~ signal,
                 data=metadata,
                 subset = m <= 20)

## Standard errors
sqrt(diag(vcov(fit)))
```

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