

# Package: mmcc (via r-universe)

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**Type** Package

**Title** tidy mcmc.list using data.table

**Version** 0.0.9.9000

**Description** Tidy up, diagnose, and visualise your mcmc samples quickly and easily so you can get on with your analysis.

**License** MIT + file LICENSE

**Encoding** UTF-8

**LazyData** true

**Imports** data.table, magrittr, purrr, coda, generics, graphics, stats, glue

**RoxygenNote** 7.1.2

**Suggests** knitr, rmarkdown, ggplot2, covr, greta, bayesplot, reticulate, rjags, testthat, rstan (>= 2.17.0), spelling

**VignetteBuilder** knitr

**Depends** R (>= 3.1.2)

**Roxygen** list(markdown = TRUE)

**Language** en-US

**Repository** <https://njtierney.r-universe.dev>

**RemoteUrl** <https://github.com/njtierney/mmcc>

**RemoteRef** HEAD

**RemoteSha** f1f271bf565edfaa21c2f4a1979dc478e450a9d9

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diag_autocorr	<i>Diagnostic: Calculate the autocorrelation for each chain and parameter</i>
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## Description

Diagnostic: Calculate the autocorrelation for each chain and parameter

## Usage

```
diag_autocorr(x, lags = NULL)
```

## Arguments

x	mcmc.list object or dataframe created by mcmc_to_dt
lags	integer the lag value that you want to use

## Value

a data.table data frame with the autocorrelation at each lag for each chain and parameter columns lag, acf, chain, and parameter

## Examples

```
library(coda)
data(line)
line_acf <- diag_autocorr(line)
line_acf
```

---

diag_ess	<i>Diagnostic: Calculate the effective sample size for each chain and parameter</i>
----------	---

---

**Description**

Diagnostic: Calculate the effective sample size for each chain and parameter

**Usage**

```
diag_ess(x)
```

**Arguments**

x mcmc.list object or dataframe created by mcmc\_to\_dt

**Value**

a data.table data frame with the effective sample size for each chain and parameter

**Examples**

```
library(coda)
data(line)
line_ess <- diag_ess(line)
line_ess
```

---

diag_mc_stderr	<i>Batch Means Standard Errors</i>
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**Description**

Calculate Monte Carlo Standard Errors Using Batch Means

**Usage**

```
diag_mc_stderr(x, b_size = "sqrt", warn = TRUE)
```

**Arguments**

x 'mcmc.list' or 'data.table' object  
b\_size character, method for determining size of batch (see Details)  
warn logical, give a warning if there are too few samples in the MCMC output

## Details

For the batch size, the options are 'sqrt', for using the square root of the MCMC output length, or 'cuberoot', for using the cube root of the output length. The default is 'sqrt'.

## References

Galin L. Jones, Murali Haran, Brian S. Caffo, and Ronald Neath (2006). "Fixed-Width Output Analysis for Markov Chain Monte Carlo," *Journal of the American Statistical Association*, 101, 1537–1547

## Examples

```
library(coda)
data(line)
r <- diag_mc_stderr(line)
r
```

---

example\_jags\_model      *Example JAGS Model*

---

## Description

This is an example JAGS model to use for examples in the `mmcc` package. The model is fit as a basic linear regression, with uniform priors on  $\beta_0$  and  $\beta_1$ , where  $y$  is assumed to be normal, with mean  $\mu$  and precision  $\tau$ . The model is drawn from the vignette, "Model summaries for a Bayesian linear regression", and the code to generate it can be found in the `data-raw` folder.

## Usage

```
data(example_jags_model)
```

## Format

An object of class `jags` of length 8.

## Examples

```
library(rjags)
example_jags_model$recompile()
model_dic <- dic.samples(example_jags_model, n.iter = 1000)
glance(model_dic)
```

---

example_stan_model	<i>Example STAN Model</i>
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**Description**

This is an example STAN model to use for examples in the mmcc package. It was created with the example code given below in examples.

**Usage**

```
data(example_stan_model)
```

**Format**

An object of class `stanfit` of dimension 5 x 4 x 3.

**Examples**

```
mcmc_to_dt(example_stan_model)
```

---

glance.dic	<i>Glance upon your dic samples to get the penalised deviance</i>
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**Description**

This provides a one-row dataframe (`data.table`) with information on the overall deviance, effective number of parameters (when type was `pD`) or optimism (when type was `popt`), and resulting penalised deviance.

**Usage**

```
## S3 method for class 'dic'  
glance(x, ...)
```

**Arguments**

<code>x</code>	a dic object generated by <code>rjags::dic.samples</code>
<code>...</code>	(optional) additional arguments to pass

**Value**

a one-row dataframe of summary information of the dic samples

**Author(s)**

Sam Clifford, <sj.clifford@gmail.com>

**Examples**

```
library(rjags)
example_jags_model$recompile()
model_dic <- dic.samples(example_jags_model, n.iter = 1000)
glance(model_dic)
```

---

`glance.mcmc.list`      *Glance upon your mcmc.list to get summary information*

---

**Description**

This provides a one-row dataframe with information on number of chains, the number of variables, the number of iterations, and the lower and upper values for effective sample size (ess), and rhat.

**Usage**

```
## S3 method for class 'mcmc.list'
glance(x, ...)
```

**Arguments**

`x`                    an mcmc.list object  
`...`                (optional) additional arguments to pass

**Value**

a one-row dataframe of summary information of the mcmc model

**Examples**

```
library(coda)
data(line)
glance(line)
```

---

`mcmc-dims`                    *Dimensions of MCMC objects*

---

**Description**

Retrieve dimensions of an MCMC object:

- `n_chain(x)` the number of chains
- `n_var(x)` the number of variables
- `n_iter(x)` the number of iterations

**Usage**

```
n_chain(x)
```

```
n_iter(x)
```

```
n_var(x)
```

**Arguments**

`x` an mcmc object, see "Details" for a list of supported mcmc objects.

**Details**

This is similar to `coda::nchain()` but more general, working for the following classes: `* mcmc *`  
`mcmc.list` `* data.table` (generated by `mcmc_to_dt`) `* stanfit` `* jags`

---

```
mcmc_to_dt
```

```
Convert mcmc.list to a tidy data.table object
```

---

**Description**

`mcmc_to_dt` use `data.table` to return a tidy dataframe from an "mcmc.list", or "stan" object.

**Usage**

```
mcmc_to_dt(mcmc_object, ...)
```

**Arguments**

`mcmc_object` an object of class "mcmc.list", as you would find with fitting a model using `jags.model()`, and `coda.samples`, or "stan", from fitting a stan model.

`...` additional arguments

**Value**

a tidy `data.table` dataframe of MCMC sample

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mcmc\_to\_dt.mcmc.list    *Convert mcmc.list to a tidy data.table object*

---

### Description

mcmc\_to\_dt use data.table to return a tidy dataframe from an "mcmc.list", or "stan" object.

### Usage

```
## S3 method for class 'mcmc.list'
mcmc_to_dt(mcmc_object, ..., colnames = NULL)
```

### Arguments

mcmc\_object    an object of class "mcmc.list", as you would find with fitting a model using jags.model(), and coda.samples, or "stan", from fitting a stan model.

...            additional arguments

colnames       which parameters we want from mcmc\_object, if NULL then all columns get selected

### Examples

```
library(coda)
data(line)
mcmc_to_dt(line)
```

---

mcmc\_to\_dt.stanfit    *MCMC tidiers (draft) for STAN*

---

### Description

MCMC tidiers (draft) for STAN

### Usage

```
## S3 method for class 'stanfit'
mcmc_to_dt(mcmc_object, ...)
```

### Arguments

mcmc\_object    an object of class "mcmc.list", as you would find with fitting a model using jags.model(), and coda.samples, or "stan", from fitting a stan model.

...            additional arguments



**Examples**

```
mcmc_to_dt(example_stan_model)
```

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<code>n_sims</code>	<i>Return the number of simulations</i>
---------------------	---

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**Description**

Return the number of simulations

**Usage**

```
n_sims(x)
```

**Arguments**

`x` mcmc list

**Value**

integer of number of simulations

**Author(s)**

Nicholas Tierney

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<code>thin_dt</code>	<i>thin_dt</i>
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**Description**

post-hoc thinning of MCMC chains which have been converted to a `data.table`

**Usage**

```
thin_dt(dt, thin = 1)
```

**Arguments**

`dt` an object of class "data.table" returned from `mcmc_to_dt`  
`thin` thinning interval

**Value**

a `data.table` dataframe

**Examples**

```
library(coda)
data(line)
mcmc_dt <- mcmc_to_dt(line)
thin_dt(mcmc_dt, thin = 10)
thin_dt(mcmc_dt, thin = 2)
thin_dt(mcmc_dt, thin = 20)
```

---

tidy.mcmc.list	<i>Return a tidy data summary of an MCMC object</i>
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**Description**

`tidy.mcmc.list` is a function that behaves like those from `broom`. It takes an `mcmc.list` object from `coda.samples` and return a data frame that summarises each parameters with its mean and quantiles and returns the output as a `data.table` object. This can be called as `tidy`. Currently summarises over all chains.

**Usage**

```
## S3 method for class 'mcmc.list'
tidy(x, conf_level = c(0.95), chain = FALSE, colnames = NULL, ...)
```

**Arguments**

<code>x</code>	object of class "mcmc.list", as you would find with fitting a model using <code>jags.model()</code> , and <code>coda.samples</code> .
<code>conf_level</code>	level of the credible interval to be calculated. Can be multiple values.
<code>chain</code>	whether or not to summarise each parameter for each chain
<code>colnames</code>	which parameters we want from <code>mcmc_object</code> , if <code>NULL</code> then all columns get selected
<code>...</code>	extra arguments

**Value**

a `data.table` containing parameter summaries

**Author(s)**

Sam Clifford, <sj.clifford@gmail.com>

**Examples**

```
library(coda)
data(line)
tidy(line)
# Optionally ask for a subset of parameters with a vector of `colnames`,
# and summarise for each chain:
tidy(line,
      chain = TRUE,
      colnames=c("alpha"))
# can provide two levels of confidence:
tidy(line, conf_level = c(0.95, 0.50))
tidy(line, conf_level = c(0.95))
tidy(line, conf_level = c(0.89, 0.25))
```

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