

# Package ‘modelr’

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**Title** Modelling Functions that Work with the Pipe

**Version** 0.1.8

**Description** Functions for modelling that help you seamlessly integrate modelling into a pipeline of data manipulation and visualisation.

**License** GPL-3

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**BugReports** <https://github.com/tidyverse/modelr/issues>

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add_predictions	<i>Add predictions to a data frame</i>
-----------------	--

---

## Description

Add predictions to a data frame

## Usage

```
add_predictions(data, model, var = "pred", type = NULL)
```

```
spread_predictions(data, ..., type = NULL)
```

```
gather_predictions(data, ..., .pred = "pred", .model = "model", type = NULL)
```

## Arguments

data	A data frame used to generate the predictions.
model	add_predictions takes a single model;
var	The name of the output column, default value is pred
type	Prediction type, passed on to stats::predict(). Consult predict() documentation for given model to determine valid values.
...	gather_predictions and spread_predictions take multiple models. The name will be taken from either the argument name or the name of the model.
.pred, .model	The variable names used by gather_predictions.

**Value**

A data frame. `add_prediction` adds a single new column, with default name `pred`, to the input data. `spread_predictions` adds one column for each model. `gather_predictions` adds two columns `.model` and `.pred`, and repeats the input rows for each model.

**Examples**

```
df <- tibble::tibble(
  x = sort(runif(100)),
  y = 5 * x + 0.5 * x ^ 2 + 3 + rnorm(length(x))
)
plot(df)

m1 <- lm(y ~ x, data = df)
grid <- data.frame(x = seq(0, 1, length = 10))
grid %>% add_predictions(m1)

m2 <- lm(y ~ poly(x, 2), data = df)
grid %>% spread_predictions(m1, m2)
grid %>% gather_predictions(m1, m2)
```

---

add_predictors	<i>Add predictors to a formula</i>
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---

**Description**

This merges a one- or two-sided formula `f` with the right-hand sides of all formulas supplied in `...`

**Usage**

```
add_predictors(f, ..., fun = "+")
```

**Arguments**

<code>f</code>	A formula.
<code>...</code>	Formulas whose right-hand sides will be merged to <code>f</code> .
<code>fun</code>	A function name indicating how to merge the right-hand sides.

**Examples**

```
f <- lhs ~ rhs
add_predictors(f, ~var1, ~var2)

# Left-hand sides are ignored:
add_predictors(f, lhs1 ~ var1, lhs2 ~ var2)

# fun can also be set to a function like "*":
add_predictors(f, ~var1, ~var2, fun = "*")
```

---

add\_residuals                    *Add residuals to a data frame*

---

## Description

Add residuals to a data frame

## Usage

```
add_residuals(data, model, var = "resid")
spread_residuals(data, ...)
gather_residuals(data, ..., .resid = "resid", .model = "model")
```

## Arguments

data	A data frame used to generate the residuals
model, var	add_residuals takes a single model; the output column will be called resid
...	gather_residuals and spread_residuals take multiple models. The name will be taken from either the argument name or the name of the model.
.resid, .model	The variable names used by gather_residuals.

## Value

A data frame. add\_residuals adds a single new column, .resid, to the input data. spread\_residuals adds one column for each model. gather\_predictions adds two columns .model and .resid, and repeats the input rows for each model.

## Examples

```
df <- tibble::tibble(
  x = sort(runif(100)),
  y = 5 * x + 0.5 * x ^ 2 + 3 + rnorm(length(x))
)
plot(df)

m1 <- lm(y ~ x, data = df)
df %>% add_residuals(m1)

m2 <- lm(y ~ poly(x, 2), data = df)
df %>% spread_residuals(m1, m2)
df %>% gather_residuals(m1, m2)
```

---

bootstrap	<i>Generate n bootstrap replicates.</i>
-----------	---

---

**Description**

Generate n bootstrap replicates.

**Usage**

```
bootstrap(data, n, id = ".id")
```

**Arguments**

data	A data frame
n	Number of bootstrap replicates to generate
id	Name of variable that gives each model a unique integer id.

**Value**

A data frame with n rows and one column: strap

**See Also**

Other resampling techniques: [resample\\_bootstrap\(\)](#), [resample\\_partition\(\)](#), [resample\(\)](#)

**Examples**

```
library(purrr)
boot <- bootstrap(mtcars, 100)

models <- map(boot$strap, ~ lm(mpg ~ wt, data = .))
tidied <- map_df(models, broom::tidy, .id = "id")

hist(subset(tidied, term == "wt")$estimate)
hist(subset(tidied, term == "(Intercept)")$estimate)
```

---

crossv_mc	<i>Generate test-training pairs for cross-validation</i>
-----------	--

---

**Description**

crossv\_kfold splits the data into k exclusive partitions, and uses each partition for a test-training split. crossv\_mc generates n random partitions, holding out test of the data for training. crossv\_loo performs leave-one-out cross-validation, i.e., n = nrow(data) training partitions containing n - 1 rows each.

**Usage**

```
crossv_mc(data, n, test = 0.2, id = ".id")  
  
crossv_kfold(data, k = 5, id = ".id")  
  
crossv_loo(data, id = ".id")
```

**Arguments**

data	A data frame
n	Number of test-training pairs to generate (an integer).
test	Proportion of observations that should be held out for testing (a double).
id	Name of variable that gives each model a unique integer id.
k	Number of folds (an integer).

**Value**

A data frame with columns `test`, `train`, and `.id`. `test` and `train` are list-columns containing [resample\(\)](#) objects. The number of rows is `n` for `crossv_mc()`, `k` for `crossv_kfold()` and `nrow(data)` for `crossv_loo()`.

**Examples**

```
cv1 <- crossv_kfold(mtcars, 5)  
cv1  
  
library(purrr)  
cv2 <- crossv_mc(mtcars, 100)  
models <- map(cv2$train, ~ lm(mpg ~ wt, data = .))  
errs <- map2_dbl(models, cv2$test, rmse)  
hist(errs)
```

---

data_grid	<i>Generate a data grid.</i>
-----------	------------------------------

---

**Description**

To visualise a model, it is very useful to be able to generate an evenly spaced grid of points from the data. `data_grid` helps you do this by wrapping around [tidyr::expand\(\)](#).

**Usage**

```
data_grid(data, ..., .model = NULL)
```

**Arguments**

data	A data frame
...	Variables passed on to <code>tidyr::expand()</code>
.model	A model. If supplied, any predictors needed for the model not present in ... will be filled in with "typical" values.

**See Also**

[seq\\_range\(\)](#) for generating ranges from continuous variables.

**Examples**

```
data_grid(mtcars, vs, am)

# For continuous variables, seq_range is useful
data_grid(mtcars, mpg = mpg)
data_grid(mtcars, mpg = seq_range(mpg, 10))

# If you supply a model, missing predictors will be filled in with
# typical values
mod <- lm(mpg ~ wt + cyl + vs, data = mtcars)
data_grid(mtcars, .model = mod)
data_grid(mtcars, cyl = seq_range(cyl, 9), .model = mod)
```

---

fit\_with

*Fit a list of formulas*


---

**Description**

`fit_with()` is a pipe-friendly tool that applies a list of formulas to a fitting function such as `stats::lm()`. The list of formulas is typically created with `formulas()`.

**Usage**

```
fit_with(data, .f, .formulas, ...)
```

**Arguments**

data	A dataset used to fit the models.
.f	A fitting function such as <code>stats::lm()</code> , <code>lme4::lmer()</code> or <code>rstanarm::stan_glmer()</code> .
.formulas	A list of formulas specifying a model.
...	Additional arguments passed on to .f

**Details**

Assumes that .f takes the formula either as first argument or as second argument if the first argument is data. Most fitting functions should fit these requirements.

**See Also**

[formulas\(\)](#)

**Examples**

```
# fit_with() is typically used with formulas().
disp_fits <- mtcars %>% fit_with(lm, formulas(~disp,
  additive = ~drat + cyl,
  interaction = ~drat * cyl,
  full = add_predictors(interaction, ~am, ~vs)
))

# The list of fitted models is named after the names of the list of
# formulas:
disp_fits$full

# Additional arguments are passed on to .f
mtcars %>% fit_with(glm, list(am ~ disp), family = binomial)
```

---

formulas	<i>Create a list of formulas</i>
----------	----------------------------------

---

**Description**

`formulas()` creates a list of two-sided formulas by merging a unique left-hand side to a list of right-hand sides.

**Usage**

```
formulas(.response, ...)

formulae(.response, ...)
```

**Arguments**

`.response` A one-sided formula used as the left-hand side of all resulting formulas.  
`...` List of formulas whose right-hand sides will be merged to `.response`.

**Examples**

```
# Provide named arguments to create a named list of formulas:
models <- formulas(~lhs,
  additive = ~var1 + var2,
  interaction = ~var1 * var2
)
models$additive

# The formulas are created sequentially, so that you can refer to
# previously created formulas:
```



```

formulas(~lhs,
  linear = ~var1 + var2,
  hierarchical = add_predictors(linear, ~(1 | group))
)

```

---

geom\_ref\_line      *Add a reference line (ggplot2).*

---

### Description

Add a reference line (ggplot2).

### Usage

```
geom_ref_line(h, v, size = 2, colour = "white")
```

### Arguments

h, v	Position of horizontal or vertical reference line
size	Line size
colour	Line colour

---

heights      *Height and income data.*

---

### Description

You might have heard that taller people earn more. Is it true? You can try and answer the question by exploring this dataset extracted from the [National Longitudinal Study](#), which is sponsored by the U.S. Bureau of Labor Statistics.

### Usage

```
heights
```

### Format

**income** Yearly income. The top two percent of values were averaged and that average was used to replace all values in the top range.

**height** Height, in inches

**weight** Weight, in pounds

**age** Age, in years, between 47 and 56.

**marital** Marital status

**sex** Sex

**education** Years of education

**afqt** Percentile score on Armed Forces Qualification Test.

## Details

This contains data as at 2012.

---

model-quality

*Compute model quality for a given dataset*

---

## Description

Three summaries are immediately interpretable on the scale of the response variable:

- `rmse()` is the root-mean-squared-error
- `mae()` is the mean absolute error
- `qae()` is quantiles of absolute error.

Other summaries have varying scales and interpretations:

- `mape()` mean absolute percentage error.
- `rsae()` is the relative sum of absolute errors.
- `mse()` is the mean-squared-error.
- `rsquare()` is the variance of the predictions divided by the variance of the response.

## Usage

```
mse(model, data)
```

```
rmse(model, data)
```

```
mae(model, data)
```

```
rsquare(model, data)
```

```
qae(model, data, probs = c(0.05, 0.25, 0.5, 0.75, 0.95))
```

```
mape(model, data)
```

```
rsae(model, data)
```

## Arguments

<code>model</code>	A model
<code>data</code>	The dataset
<code>probs</code>	Numeric vector of probabilities

## Examples

```
mod <- lm(mpg ~ wt, data = mtcars)
mse(mod, mtcars)
rmse(mod, mtcars)
rsquare(mod, mtcars)
mae(mod, mtcars)
qae(mod, mtcars)
mape(mod, mtcars)
rsae(mod, mtcars)
```

---

model_matrix	<i>Construct a design matrix</i>
--------------	----------------------------------

---

## Description

This is a thin wrapper around `stats::model.matrix()` which returns a tibble. Use it to determine how your modelling formula is translated into a matrix, and thence into an equation.

## Usage

```
model_matrix(data, formula, ...)
```

## Arguments

data	A data frame
formula	A modelling formula
...	Other arguments passed onto <code>stats::model.matrix()</code>

## Value

A tibble.

## Examples

```
model_matrix(mtcars, mpg ~ cyl)
model_matrix(iris, Sepal.Length ~ Species)
model_matrix(iris, Sepal.Length ~ Species - 1)
```

---

na.warn	<i>Handle missing values with a warning</i>
---------	---

---

### Description

This NA handler ensures that those models that support the `na.action` parameter do not silently drop missing values. It wraps around `stats::na.exclude()` so that there is one prediction/residual for input row. To apply it globally, run `options(na.action = na.warn)`.

### Usage

```
na.warn(object)
```

### Arguments

object	A data frame
--------	--------------

### Examples

```
df <- tibble::tibble(  
  x = 1:10,  
  y = c(5.1, 9.7, NA, 17.4, 21.2, 26.6, 27.9, NA, 36.3, 40.4)  
)  
# Default behaviour is to silently drop  
m1 <- lm(y ~ x, data = df)  
resid(m1)  
  
# Use na.action = na.warn to warn  
m2 <- lm(y ~ x, data = df, na.action = na.warn)  
resid(m2)
```

---

permute	<i>Generate n permutation replicates.</i>
---------	---

---

### Description

A permutation test involves permuting one or more variables in a data set before performing the test, in order to break any existing relationships and simulate the null hypothesis. One can then compare the true statistic to the generated distribution of null statistics.

### Usage

```
permute(data, n, ..., .id = ".id")  
  
permute_(data, n, columns, .id = ".id")
```

**Arguments**

data	A data frame
n	Number of permutations to generate.
...	Columns to permute. This supports bare column names or dplyr <a href="#">dplyr::select_helpers</a>
.id	Name of variable that gives each model a unique integer id.
columns	In permute_, vector of column names to permute.

**Value**

A data frame with n rows and one column: perm

**Examples**

```
library(purrr)
perms <- permute(mtcars, 1000, mpg)

models <- map(perms$perm, ~ lm(mpg ~ wt, data = .))
glanced <- map_df(models, broom::glance, .id = "id")

# distribution of null permutation statistics
hist(glanced$statistic)
# confirm these are roughly uniform p-values
hist(glanced$p.value)

# test against the unpermuted model to get a permutation p-value
mod <- lm(mpg ~ wt, mtcars)
mean(glanced$statistic > broom::glance(mod)$statistic)
```

---

resample	A "lazy" resample.
----------	--------------------

---

**Description**

Often you will resample a dataset hundreds or thousands of times. Storing the complete resample each time would be very inefficient so this class instead stores a "pointer" to the original dataset, and a vector of row indexes. To turn this into a regular data frame, call `as.data.frame`, to extract the indices, use `as.integer`.

**Usage**

```
resample(data, idx)
```

**Arguments**

<code>data</code>	The data frame
<code>idx</code>	A vector of integer indexes indicating which rows have been selected. These values should lie between 1 and <code>nrow(data)</code> but they are not checked by this function in the interests of performance.

**See Also**

Other resampling techniques: [bootstrap\(\)](#), [resample\\_bootstrap\(\)](#), [resample\\_partition\(\)](#)

**Examples**

```
resample(mtcars, 1:10)

b <- resample_bootstrap(mtcars)
b
as.integer(b)
as.data.frame(b)

# Many modelling functions will do the coercion for you, so you can
# use a resample object directly in the data argument
lm(mpg ~ wt, data = b)
```

---

<code>resample_bootstrap</code>	<i>Generate a bootstrap replicate</i>
---------------------------------	---------------------------------------

---

**Description**

Generate a bootstrap replicate

**Usage**

```
resample_bootstrap(data)
```

**Arguments**

<code>data</code>	A data frame
-------------------	--------------

**See Also**

Other resampling techniques: [bootstrap\(\)](#), [resample\\_partition\(\)](#), [resample\(\)](#)

**Examples**

```
coef(lm(mpg ~ wt, data = resample_bootstrap(mtcars)))
coef(lm(mpg ~ wt, data = resample_bootstrap(mtcars)))
coef(lm(mpg ~ wt, data = resample_bootstrap(mtcars)))
```

---

resample\_partition     *Generate an exclusive partitioning of a data frame*

---

**Description**

Generate an exclusive partitioning of a data frame

**Usage**

```
resample_partition(data, p)
```

**Arguments**

data	A data frame
p	A named numeric vector giving where the value is the probability that an observation will be assigned to that group.

**See Also**

Other resampling techniques: [bootstrap\(\)](#), [resample\\_bootstrap\(\)](#), [resample\(\)](#)

**Examples**

```
ex <- resample_partition(mtcars, c(test = 0.3, train = 0.7))
mod <- lm(mpg ~ wt, data = ex$train)
rmse(mod, ex$test)
rmse(mod, ex$train)
```

---

resample\_permutation     *Create a resampled permutation of a data frame*

---

**Description**

Create a resampled permutation of a data frame

**Usage**

```
resample_permutation(data, columns, idx = NULL)
```

**Arguments**

data	A data frame
columns	Columns to be permuted
idx	Indices to permute by. If not given, generates them randomly

**Value**

A permutation object; use `as.data.frame` to convert to a permuted data frame

seq\_range

*Generate a sequence over the range of a vector***Description**

Generate a sequence over the range of a vector

**Usage**

```
seq_range(x, n, by, trim = NULL, expand = NULL, pretty = FALSE)
```

**Arguments**

x	A numeric vector
n, by	Specify the output sequence either by supplying the length of the sequence with n, or the spacing between value with by. Specifying both is an error. I recommend that you name these arguments in order to make it clear to the reader.
trim	Optionally, trim values off the tails. $\text{trim} / 2 * \text{length}(x)$ values are removed from each tail.
expand	Optionally, expand the range by $\text{expand} * (1 + \text{range}(x))$ (computed after trimming).
pretty	If TRUE, will generate a pretty sequence. If n is supplied, this will use <code>pretty()</code> instead of <code>seq()</code> . If by is supplied, it will round the first value to a multiple of by.

**Examples**

```
x <- rcauchy(100)
seq_range(x, n = 10)
seq_range(x, n = 10, trim = 0.1)
seq_range(x, by = 1, trim = 0.1)

# Make pretty sequences
y <- runif(100)
seq_range(y, n = 10)
seq_range(y, n = 10, pretty = TRUE)
seq_range(y, n = 10, expand = 0.5, pretty = TRUE)

seq_range(y, by = 0.1)
seq_range(y, by = 0.1, pretty = TRUE)
```



---

sim	<i>Simple simulated datasets</i>
-----	----------------------------------

---

**Description**

These simple simulated datasets are useful for teaching modelling basics.

**Usage**

```
sim1
sim2
sim3
sim4
```

---

typical	<i>Find the typical value</i>
---------	-------------------------------

---

**Description**

For numeric, integer, and ordered factor vectors, it returns the median. For factors, characters, and logical vectors, it returns the most frequent value. If multiple values are tied for most frequent, it returns them all. NA missing values are always silently dropped.

**Usage**

```
typical(x, ...)
```

**Arguments**

x	A vector
...	Arguments used by methods

**Examples**

```
# median of numeric vector
typical(rpois(100, lambda = 10))

# most frequent value of character or factor
x <- sample(c("a", "b", "c"), 100, prob = c(0.6, 0.2, 0.2), replace = TRUE)
typical(x)
typical(factor(x))

# if tied, returns them all
```

```
x <- c("a", "a", "b", "b", "c")
typical(x)

# median of an ordered factor
typical(ordered(c("a", "a", "b", "c", "d")))
```

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