

# Package ‘fable’

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**Title** Forecasting Models for Tidy Time Series

**Version** 0.3.2

**Description** Provides a collection of commonly used univariate and multivariate time series forecasting models including automatically selected exponential smoothing (ETS) and autoregressive integrated moving average (ARIMA) models. These models work within the 'fable' framework provided by the 'fabletools' package, which provides the tools to evaluate, visualise, and combine models in a workflow consistent with the tidyverse.

**License** GPL-3

**URL** <https://fable.tidyverts.org>, <https://github.com/tidyverts/fable>

**BugReports** <https://github.com/tidyverts/fable/issues>

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---

 AR *Estimate a AR model*


---

**Description**

Searches through the vector of lag orders to find the best AR model which has lowest AIC, AICc or BIC value. It is implemented using OLS, and behaves comparably to `stats::ar.ols()`.

**Usage**

```
AR(formula, ic = c("aicc", "aic", "bic"), ...)
```

**Arguments**

formula	Model specification (see "Specials" section).
ic	The information criterion used in selecting the model.
...	Further arguments for arima

**Details**

Exogenous regressors and `common_xregs` can be specified in the model formula.

**Value**

A model specification.

**Specials**

**pdq:** The order special is used to specify the lag order for the auto-regression.

```
order(p = 0:15, fixed = list())
```

**p** The order of the auto-regressive (AR) terms. If multiple values are provided, the one which minimises `ic` will be chosen.  
**fixed** A named list of fixed parameters for coefficients. The names identify the coefficient, beginning with `ar`, and then follow the order of the terms.

**xreg:** Exogenous regressors can be included in an AR model without explicitly using the `xreg()` special. Common exogenous regressor specials as specified in `common_xregs` can also be used. These regressors are handled using `stats::model.frame()`, and so interactions and other functionality behaves similarly to `stats::lm()`.

The inclusion of a constant in the model follows the similar rules to `stats::lm()`, where including `1` will add a constant and `0` or `-1` will remove the constant. If left out, the inclusion of a constant will be determined by minimising `ic`.

```
xreg(..., fixed = list())
```

... Bare expressions for the exogenous regressors (such as `log(x)`)  
**fixed** A named list of fixed parameters for coefficients. The names identify the coefficient, and should match the name of the corresponding term in the formula.

**See Also**

[Forecasting: Principles and Practices, Vector autoregressions \(section 11.2\)](#)

**Examples**

```
luteinizing_hormones <- as_tsibble(lh)
fit <- luteinizing_hormones %>%
  model(AR(value ~ order(3)))

report(fit)

fit %>%
  forecast() %>%
  autoplot(luteinizing_hormones)
```

---

 ARIMA

*Estimate an ARIMA model*


---

**Description**

Searches through the model space specified in the specials to identify the best ARIMA model, with the lowest AIC, AICc or BIC value. It is implemented using `stats::arima()` and allows ARIMA models to be used in the fable framework.

**Usage**

```
ARIMA(
  formula,
  ic = c("aicc", "aic", "bic"),
  selection_metric = function(x) x[[ic]],
  stepwise = TRUE,
  greedy = TRUE,
  approximation = NULL,
  order_constraint = p + q + P + Q <= 6 & (constant + d + D <= 2),
  unitroot_spec = unitroot_options(),
  trace = FALSE,
  ...
)
```

**Arguments**

<code>formula</code>	Model specification (see "Specials" section).
<code>ic</code>	The information criterion used in selecting the model.
<code>selection_metric</code>	A function used to compute a metric from an Arima object which is minimised to select the best model.
<code>stepwise</code>	Should stepwise be used? (Stepwise can be much faster)

greedy	Should the stepwise search move to the next best option immediately?
approximation	Should CSS (conditional sum of squares) be used during model selection? The default (NULL) will use the approximation if there are more than 150 observations or if the seasonal period is greater than 12.
order_constraint	A logical predicate on the orders of p, d, q, P, D, Q and constant to consider in the search. See "Specials" for the meaning of these terms.
unitroot_spec	A specification of unit root tests to use in the selection of d and D. See <code>unitroot_options()</code> for more details.
trace	If TRUE, the selection_metric of estimated models in the selection procedure will be outputted to the console.
...	Further arguments for <code>stats::arima()</code>

### Value

A model specification.

### Parameterisation

The fable `ARIMA()` function uses an alternate parameterisation of constants to `stats::arima()` and `forecast::Arima()`. While the parameterisations are equivalent, the coefficients for the constant/mean will differ.

In fable, the parameterisation used is:

$$(1 - \phi_1 B - \dots - \phi_p B^p)(1 - B)^d y_t = c + (1 + \theta_1 B + \dots + \theta_q B^q) \varepsilon_t$$

In stats and forecast, an ARIMA model is parameterised as:

$$(1 - \phi_1 B - \dots - \phi_p B^p)(y'_t - \mu) = (1 + \theta_1 B + \dots + \theta_q B^q) \varepsilon_t$$

where  $\mu$  is the mean of  $(1 - B)^d y_t$  and  $c = \mu(1 - \phi_1 - \dots - \phi_p)$ .

### Specials

The *specials* define the space over which ARIMA will search for the model that best fits the data. If the RHS of formula is left blank, the default search space is given by `pdq() + PDQ()`: that is, a model with candidate seasonal and nonseasonal terms, but no exogenous regressors. Note that a seasonal model requires at least 2 full seasons of data; if this is not available, ARIMA will revert to a nonseasonal model with a warning.

To specify a model fully (avoid automatic selection), the intercept and `pdq()/PDQ()` values must be specified. For example, `formula = response ~ 1 + pdq(1, 1, 1) + PDQ(1, 0, 0)`.

**pdq:** The `pdq` special is used to specify non-seasonal components of the model.

```
pdq(p = 0:5, d = 0:2, q = 0:5,
    p_init = 2, q_init = 2, fixed = list())
```

- p The order of the non-seasonal auto-regressive (AR) terms. If multiple values are provided, the one which minimises ic.
- d The order of integration for non-seasonal differencing. If multiple values are provided, one of the values will be selected.
- q The order of the non-seasonal moving average (MA) terms. If multiple values are provided, the one which minimises ic.
- p\_init If stepwise = TRUE, p\_init provides the initial value for p for the stepwise search procedure.
- q\_init If stepwise = TRUE, q\_init provides the initial value for q for the stepwise search procedure.
- fixed A named list of fixed parameters for coefficients. The names identify the coefficient, beginning with either ar or ma,

**PDQ:** The PDQ special is used to specify seasonal components of the model. To force a non-seasonal fit, specify PDQ(0, 0, 0) in the RHS of the model formula. Note that simply omitting PDQ from the formula will *not* result in a non-seasonal fit.

```
PDQ(P = 0:2, D = 0:1, Q = 0:2, period = NULL,
     P_init = 1, Q_init = 1, fixed = list())
```

- P The order of the seasonal auto-regressive (SAR) terms. If multiple values are provided, the one which minimises ic.
- D The order of integration for seasonal differencing. If multiple values are provided, one of the values will be selected.
- Q The order of the seasonal moving average (SMA) terms. If multiple values are provided, the one which minimises ic.
- period The periodic nature of the seasonality. This can be either a number indicating the number of observations in each season or a vector of 0s and 1s.
- P\_init If stepwise = TRUE, P\_init provides the initial value for P for the stepwise search procedure.
- Q\_init If stepwise = TRUE, Q\_init provides the initial value for Q for the stepwise search procedure.
- fixed A named list of fixed parameters for coefficients. The names identify the coefficient, beginning with either sar or sma,

**xreg:** Exogenous regressors can be included in an ARIMA model without explicitly using the xreg() special. Common exogenous regressor specials as specified in [common\\_xregs](#) can also be used. These regressors are handled using `stats::model.frame()`, and so interactions and other functionality behaves similarly to `stats::lm()`.

The inclusion of a constant in the model follows the similar rules to `stats::lm()`, where including 1 will add a constant and 0 or -1 will remove the constant. If left out, the inclusion of a constant will be determined by minimising ic.

```
xreg(..., fixed = list())
```

- ... Bare expressions for the exogenous regressors (such as log(x))
- fixed A named list of fixed parameters for coefficients. The names identify the coefficient, and should match the name of the regressor.

## See Also

[Forecasting: Principles and Practices, ARIMA models \(chapter 9\)](#) [Forecasting: Principles and Practices, Dynamic regression models \(chapter 10\)](#)

## Examples

```
# Manual ARIMA specification
USAccDeaths %>%
  as_tsibble() %>%
  model(arima = ARIMA(log(value) ~ 0 + pdq(0, 1, 1) + PDQ(0, 1, 1)) %>%
  report())

# Automatic ARIMA specification
library(tsibble)
```

```
library(dplyr)
tsibbledata::global_economy %>%
  filter(Country == "Australia") %>%
  model(ARIMA(log(GDP) ~ Population))
```

---

breusch_godfrey	<i>Breusch-Godfrey Test</i>
-----------------	-----------------------------

---

### Description

Breusch-Godfrey test for higher-order serial correlation.

### Usage

```
breusch_godfrey(x, ...)

## S3 method for class 'TSLM'
breusch_godfrey(x, order = 1, type = c("Chisq", "F"), ...)
```

### Arguments

x	A model object to be tested.
...	Further arguments for methods.
order	The maximum order of serial correlation to test for.
type	The type of test statistic to use.

### See Also

[lmtest::bgtest\(\)](#)

---

components.ETS	<i>Extract estimated states from an ETS model.</i>
----------------	--

---

### Description

Extract estimated states from an ETS model.

### Usage

```
## S3 method for class 'ETS'
components(object, ...)
```

### Arguments

object	An estimated model.
...	Unused.



**Value**

A `fabletools::dable()` containing estimated states.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(ets = ETS(log(value) ~ season("A"))) %>%
  components()
```

---

CROSTON

*Croston's method*


---

**Description**

Based on Croston's (1972) method for intermittent demand forecasting, also described in Shenstone and Hyndman (2005). Croston's method involves using simple exponential smoothing (SES) on the non-zero elements of the time series and a separate application of SES to the times between non-zero elements of the time series.

**Usage**

```
CROSTON(
  formula,
  opt_crit = c("mse", "mae"),
  type = c("croston", "sba", "sbj"),
  ...
)
```

**Arguments**

<code>formula</code>	Model specification (see "Specials" section).
<code>opt_crit</code>	The optimisation criterion used to optimise the parameters.
<code>type</code>	Which variant of Croston's method to use. Defaults to "croston" for Croston's method, but can also be set to "sba" for the Syntetos-Boylan approximation, and "sbj" for the Shale-Boylan-Johnston method.
<code>...</code>	Not used.

**Details**

Note that forecast distributions are not computed as Croston's method has no underlying stochastic model. In a later update, we plan to support distributions via the equivalent stochastic models that underly Croston's method (Shenstone and Hyndman, 2005)

There are two variant methods available which apply multiplicative correction factors to the forecasts that result from the original Croston's method. For the Syntetos-Boylan approximation (`type = "sba"`), this factor is  $1 - \alpha/2$ , and for the Shale-Boylan-Johnston method (`type = "sbj"`), this factor is  $1 - \alpha/(2 - \alpha)$ , where  $\alpha$  is the smoothing parameter for the interval SES application.

**Value**

A model specification.

**Specials**

**demand:** The demand special specifies parameters for the demand SES application.

```
demand(initial = NULL, param = NULL, param_range = c(0, 1))
```

**initial**            The initial value for the demand application of SES.

**param**             The smoothing parameter for the demand application of SES.

**param\_range**    If param = NULL, the range of values over which to search for the smoothing parameter.

**interval:** The interval special specifies parameters for the interval SES application.

```
interval(initial = NULL, param = NULL, param_range = c(0, 1))
```

**initial**            The initial value for the interval application of SES.

**param**             The smoothing parameter for the interval application of SES.

**param\_range**    If param = NULL, the range of values over which to search for the smoothing parameter.

**References**

Croston, J. (1972) "Forecasting and stock control for intermittent demands", *Operational Research Quarterly*, **23**(3), 289-303.

Shenstone, L., and Hyndman, R.J. (2005) "Stochastic models underlying Croston's method for intermittent demand forecasting". *Journal of Forecasting*, **24**, 389-402.

Kourentzes, N. (2014) "On intermittent demand model optimisation and selection". *International Journal of Production Economics*, **156**, 180-190. doi: [10.1016/j.ijpe.2014.06.007](https://doi.org/10.1016/j.ijpe.2014.06.007).

**Examples**

```
library(tsibble)
sim_poisson <- tsibble(
  time = yearmonth("2012 Dec") + seq_len(24),
  count = rpois(24, lambda = 0.3),
  index = time
)

sim_poisson %>%
  autoplot(count)

sim_poisson %>%
  model(CROSTON(count)) %>%
  forecast(h = "2 years") %>%
  autoplot(sim_poisson)
```

---

ETS *Exponential smoothing state space model*

---

### Description

Returns ETS model specified by the formula.

### Usage

```
ETS(
  formula,
  opt_crit = c("lik", "amse", "mse", "sigma", "mae"),
  nmse = 3,
  bounds = c("both", "usual", "admissible"),
  ic = c("aicc", "aic", "bic"),
  restrict = TRUE,
  ...
)
```

### Arguments

<code>formula</code>	Model specification (see "Specials" section).
<code>opt_crit</code>	The optimization criterion. Defaults to the log-likelihood "lik", but can also be set to "mse" (Mean Square Error), "amse" (Average MSE over first nmse forecast horizons), "sigma" (Standard deviation of residuals), or "mae" (Mean Absolute Error).
<code>nmse</code>	If <code>opt_crit == "amse"</code> , nmse provides the number of steps for average multi-step MSE ( $1 \leq \text{nmse} \leq 30$ ).
<code>bounds</code>	Type of parameter space to impose: "usual" indicates all parameters must lie between specified lower and upper bounds; "admissible" indicates parameters must lie in the admissible space; "both" (default) takes the intersection of these regions.
<code>ic</code>	The information criterion used in selecting the model.
<code>restrict</code>	If TRUE (default), the models with infinite variance will not be allowed. These restricted model components are AMM, AAM, AMA, and MMA.
<code>...</code>	Other arguments

### Details

Based on the classification of methods as described in Hyndman et al (2008).

The methodology is fully automatic. The model is chosen automatically if not specified. This methodology performed extremely well on the M3-competition data. (See Hyndman, et al, 2002, below.)

**Value**

A model specification.

**Specials**

The *specials* define the methods and parameters for the components (error, trend, and seasonality) of an ETS model. If more than one method is specified, ETS will consider all combinations of the specified models and select the model which best fits the data (minimising `aic`). The method argument for each specials have reasonable defaults, so if a component is not specified an appropriate method will be chosen automatically.

There are a couple of limitations to note about ETS models:

- It does not support exogenous regressors.
- It does not support missing values. You can complete missing values in the data with imputed values (e.g. with `tidyr::fill()`), or by fitting a different model type and then calling `fabletools::interpolate()` before fitting the model.

**error:** The error special is used to specify the form of the error term.

```
error(method = c("A", "M"))
```

`method` The form of the error term: either additive ("A") or multiplicative ("M"). If the error is multiplicative, the data must

**trend:** The trend special is used to specify the form of the trend term and associated parameters.

```
trend(method = c("N", "A", "Ad"),
      alpha = NULL, alpha_range = c(1e-04, 0.9999),
      beta = NULL, beta_range = c(1e-04, 0.9999),
      phi = NULL, phi_range = c(0.8, 0.98))
```

`method` The form of the trend term: either none ("N"), additive ("A"), multiplicative ("M") or damped variants ("Ad",

`alpha` The value of the smoothing parameter for the level. If `alpha = 0`, the level will not change over time. Convers

`alpha_range` If `alpha=NULL`, `alpha_range` provides bounds for the optimised value of `alpha`.

`beta` The value of the smoothing parameter for the slope. If `beta = 0`, the slope will not change over time. Convers

`beta_range` If `beta=NULL`, `beta_range` provides bounds for the optimised value of `beta`.

`phi` The value of the dampening parameter for the slope. If `phi = 0`, the slope will be dampened immediately (no

`phi_range` If `phi=NULL`, `phi_range` provides bounds for the optimised value of `phi`.

**season:** The season special is used to specify the form of the seasonal term and associated parameters. To specify a nonseasonal model you would include `season(method = "N")`.

```
season(method = c("N", "A", "M"), period = NULL,
      gamma = NULL, gamma_range = c(1e-04, 0.9999))
```

`method` The form of the seasonal term: either none ("N"), additive ("A") or multiplicative ("M"). All specified method

`period` The periodic nature of the seasonality. This can be either a number indicating the number of observations in e

`gamma` The value of the smoothing parameter for the seasonal pattern. If `gamma = 0`, the seasonal pattern will not chan

`gamma_range` If `gamma=NULL`, `gamma_range` provides bounds for the optimised value of `gamma`.

## References

Hyndman, R.J., Koehler, A.B., Snyder, R.D., and Grose, S. (2002) "A state space framework for automatic forecasting using exponential smoothing methods", *International J. Forecasting*, **18**(3), 439–454.

Hyndman, R.J., Akram, Md., and Archibald, B. (2008) "The admissible parameter space for exponential smoothing models". *Annals of Statistical Mathematics*, **60**(2), 407–426.

Hyndman, R.J., Koehler, A.B., Ord, J.K., and Snyder, R.D. (2008) *Forecasting with exponential smoothing: the state space approach*, Springer-Verlag. <http://www.exponentialsMOOTHING.net>.

## See Also

[Forecasting: Principles and Practices, Exponential smoothing \(chapter 8\)](#)

## Examples

```
as_tsibble(USAccDeaths) %>%
  model(ETS(log(value) ~ season("A")))
```

---

fitted.AR

*Extract fitted values from a fable model*

---

## Description

Extracts the fitted values.

## Usage

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

## Arguments

`object`            The time series model used to produce the forecasts  
`...`                Additional arguments for forecast model methods.

## Value

A vector of fitted values.

## Examples

```
as_tsibble(lh) %>%
  model(AR(value ~ order(3))) %>%
  fitted()
```

---

fitted.ARIMA	<i>Extract fitted values from a fable model</i>
--------------	---

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted values.

**Examples**

```
USAccDeaths %>%
  as_tsibble() %>%
  model(arima = ARIMA(log(value) ~ pdq(0, 1, 1) + PDQ(0, 1, 1))) %>%
  fitted()
```

---

fitted.croston	<i>Extract fitted values from a fable model</i>
----------------	---

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted values.

**Examples**

```
library(tsibble)
sim_poisson <- tsibble(
  time = yearmonth("2012 Dec") + seq_len(24),
  count = rpois(24, lambda = 0.3),
  index = time
)

sim_poisson %>%
  model(CROSTON(count)) %>%
  tidy()
```

---

fitted.ETS	<i>Extract fitted values from a fable model</i>
------------	---

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted values.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(ets = ETS(log(value) ~ season("A"))) %>%
  fitted()
```

---

fitted.fable\_theta      *Extract fitted values from a fable model*

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object            The time series model used to produce the forecasts  
...                Additional arguments for forecast model methods.

**Value**

A vector of fitted values.

**Examples**

```
library(tsibbledata)  
vic_elec %>%  
  model(avg = MEAN(Demand)) %>%  
  fitted()
```

---

fitted.model\_mean      *Extract fitted values from a fable model*

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object            The time series model used to produce the forecasts  
...                Additional arguments for forecast model methods.



**Value**

A vector of fitted values.

**Examples**

```
library(tsibbledata)
vic_elec %>%
  model(avg = MEAN(Demand)) %>%
  fitted()
```

---

fitted.NNETAR	<i>Extract fitted values from a fable model</i>
---------------	---

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted values.

**Examples**

```
as_tsibble(airmiles) %>%
  model(nn = NNETAR(box_cox(value, 0.15))) %>%
  fitted()
```

---

fitted.RW	<i>Extract fitted values from a fable model</i>
-----------	---

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted values.

**Examples**

```
as_tsibble(Nile) %>%
  model(NAIVE(value)) %>%
  fitted()

library(tsibbledata)
aus_production %>%
  model(snaive = SNAIVE(Beer ~ lag("year"))) %>%
  fitted()
```

---

fitted.TSLM	<i>Extract fitted values from a fable model</i>
-------------	---

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object            The time series model used to produce the forecasts  
 ...                Additional arguments for forecast model methods.

**Value**

A vector of fitted values.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(lm = TSLM(log(value) ~ trend() + season())) %>%
  fitted()
```

---

<code>fitted.VAR</code>	<i>Extract fitted values from a fable model</i>
-------------------------	---

---

**Description**

Extracts the fitted values.

**Usage**

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

**Arguments**

object            The time series model used to produce the forecasts  
 ...                Additional arguments for forecast model methods.

**Value**

A vector of fitted values.

**Examples**

```
lung_deaths <- cbind(mdeaths, fdeaths) %>%
  as_tsibble(pivot_longer = FALSE)

lung_deaths %>%
  model(VAR(vars(mdeaths, fdeaths) ~ AR(3))) %>%
  fitted()
```

---

`forecast.AR`*Forecast a model from the fable package*

---

## Description

Produces forecasts from a trained model.

## Usage

```
## S3 method for class 'AR'
forecast(
  object,
  new_data = NULL,
  specials = NULL,
  bootstrap = FALSE,
  times = 5000,
  ...
)
```

## Arguments

<code>object</code>	The time series model used to produce the forecasts
<code>new_data</code>	A <code>tsibble</code> containing future information used to forecast.
<code>specials</code>	(passed by <code>fabletools::forecast.mdl_df()</code> ).
<code>bootstrap</code>	If TRUE, then forecast distributions are computed using simulation with resampled errors.
<code>times</code>	The number of sample paths to use in estimating the forecast distribution when <code>bootstrap = TRUE</code> .
<code>...</code>	Additional arguments for forecast model methods.

## Value

A list of forecasts.

## Examples

```
as_tsibble(lh) %>%
  model(AR(value ~ order(3))) %>%
  forecast()
```

---

forecast.ARIMA	<i>Forecast a model from the fable package</i>
----------------	--

---

## Description

Produces forecasts from a trained model.

## Usage

```
## S3 method for class 'ARIMA'
forecast(
  object,
  new_data = NULL,
  specials = NULL,
  bootstrap = FALSE,
  times = 5000,
  ...
)
```

## Arguments

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
times	The number of sample paths to use in estimating the forecast distribution when <code>bootstrap = TRUE</code> .
...	Additional arguments for forecast model methods.

## Value

A list of forecasts.

## Examples

```
USAccDeaths %>%
  as_tsibble() %>%
  model(arima = ARIMA(log(value) ~ pdq(0, 1, 1) + PDQ(0, 1, 1))) %>%
  forecast()
```

---

forecast.croston	<i>Forecast a model from the fable package</i>
------------------	--

---

**Description**

Produces forecasts from a trained model.

**Usage**

```
## S3 method for class 'croston'
forecast(object, new_data, specials = NULL, ...)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
...	Additional arguments for forecast model methods.

**Value**

A list of forecasts.

**Examples**

```
library(tsibble)
sim_poisson <- tsibble(
  time = yearmonth("2012 Dec") + seq_len(24),
  count = rpois(24, lambda = 0.3),
  index = time
)

sim_poisson %>%
  model(CROSTON(count)) %>%
  forecast()
```

---

forecast.ETS	<i>Forecast a model from the fable package</i>
--------------	--

---

**Description**

Produces forecasts from a trained model.

**Usage**

```
## S3 method for class 'ETS'
forecast(
  object,
  new_data,
  specials = NULL,
  simulate = FALSE,
  bootstrap = FALSE,
  times = 5000,
  ...
)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
simulate	If TRUE, prediction intervals are produced by simulation rather than using analytic formulae.
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
times	The number of sample paths to use in estimating the forecast distribution if simulated intervals are used.
...	Additional arguments for forecast model methods.

**Value**

A list of forecasts.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(ets = ETS(log(value) ~ season("A"))) %>%
  forecast()
```

---

forecast.fable\_theta *Forecast a model from the fable package*

---

**Description**

Produces forecasts from a trained model.

**Usage**

```
## S3 method for class 'fable_theta'
forecast(
  object,
  new_data,
  specials = NULL,
  bootstrap = FALSE,
  times = 5000,
  ...
)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
times	The number of sample paths to use in estimating the forecast distribution when <code>bootstrap = TRUE</code> .
...	Additional arguments for forecast model methods.

**Value**

A list of forecasts.

**Examples**

```
USAccDeaths %>%
  as_tsibble() %>%
  model(arima = ARIMA(log(value) ~ pdq(0, 1, 1) + PDQ(0, 1, 1))) %>%
  forecast()
```

---

forecast.model\_mean     *Forecast a model from the fable package*

---

**Description**

Produces forecasts from a trained model.



**Usage**

```
## S3 method for class 'model_mean'  
forecast(  
  object,  
  new_data,  
  specials = NULL,  
  bootstrap = FALSE,  
  times = 5000,  
  ...  
)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
times	The number of sample paths to use in estimating the forecast distribution when <code>bootstrap = TRUE</code> .
...	Additional arguments for forecast model methods.

**Value**

A list of forecasts.

**Examples**

```
library(tsibbledata)  
vic_elec %>%  
  model(avg = MEAN(Demand)) %>%  
  forecast()
```

---

`forecast.NNETAR`*Forecast a model from the fable package*

---

**Description**

Produces forecasts from a trained model.

**Usage**

```
## S3 method for class 'NNETAR'
forecast(
  object,
  new_data,
  specials = NULL,
  simulate = TRUE,
  bootstrap = FALSE,
  times = 1000,
  ...
)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
simulate	If TRUE, prediction intervals are produced by simulation rather than using analytic formulae.
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
times	The number of sample paths to use in estimating the forecast distribution if simulated intervals are used.
...	Additional arguments for forecast model methods.

**Value**

A list of forecasts.

**Examples**

```
as_tsibble(airmiles) %>%
  model(nn = NNETAR(box_cox(value, 0.15))) %>%
  forecast(times = 10)
```

---

forecast.RW

*Forecast a model from the fable package*

---

**Description**

Produces forecasts from a trained model.

**Usage**

```
## S3 method for class 'RW'
forecast(
  object,
  new_data,
  specials = NULL,
  simulate = FALSE,
  bootstrap = FALSE,
  times = 5000,
  ...
)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
simulate	If TRUE, prediction intervals are produced by simulation rather than using analytic formulae.
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
times	The number of sample paths to use in estimating the forecast distribution when <code>bootstrap = TRUE</code> .
...	Additional arguments for forecast model methods.

**Value**

A list of forecasts.

**Examples**

```
as_tsibble(Nile) %>%
  model(NAIVE(value)) %>%
  forecast()

library(tsibbledata)
aus_production %>%
  model(snaive = SNAIVE(Beer ~ lag("year"))) %>%
  forecast()
```

forecast.TSLM

*Forecast a model from the fable package***Description**

Produces forecasts from a trained model.

**Usage**

```
## S3 method for class 'TSLM'
forecast(
  object,
  new_data,
  specials = NULL,
  bootstrap = FALSE,
  approx_normal = TRUE,
  times = 5000,
  ...
)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
approx_normal	Should the resulting forecast distributions be approximated as a Normal distribution instead of a Student's T distribution. Returning Normal distributions (the default) is a useful approximation to make it easier for using TSLM models in model combinations or reconciliation processes.
times	The number of sample paths to use in estimating the forecast distribution when <code>bootstrap = TRUE</code> .
...	Additional arguments for forecast model methods.

**Value**

A list of forecasts.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(lm = TSLM(log(value) ~ trend() + season())) %>%
  forecast()
```

---

forecast.VAR	<i>Forecast a model from the fable package</i>
--------------	--

---

### Description

Produces forecasts from a trained model.

### Usage

```
## S3 method for class 'VAR'
forecast(
  object,
  new_data = NULL,
  specials = NULL,
  bootstrap = FALSE,
  times = 5000,
  ...
)
```

### Arguments

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
times	The number of sample paths to use in estimating the forecast distribution when <code>bootstrap = TRUE</code> .
...	Additional arguments for forecast model methods.

### Value

A list of forecasts.

### Examples

```
lung_deaths <- cbind(mdeaths, fdeaths) %>%
  as_tsibble(pivot_longer = FALSE)

lung_deaths %>%
  model(VAR(vars(mdeaths, fdeaths) ~ AR(3))) %>%
  forecast()
```

---

generate.AR	<i>Generate new data from a fable model</i>
-------------	---

---

### Description

Simulates future paths from a dataset using a fitted model. Innovations are sampled by the model's assumed error distribution. If `bootstrap` is `TRUE`, innovations will be sampled from the model's residuals. If `new_data` contains the `.innov` column, those values will be treated as innovations.

### Usage

```
## S3 method for class 'AR'
generate(x, new_data = NULL, specials = NULL, bootstrap = FALSE, ...)
```

### Arguments

<code>x</code>	A fitted model.
<code>new_data</code>	A <code>tsibble</code> containing future information used to forecast.
<code>specials</code>	(passed by <code>fabletools::forecast.mdl_df()</code> ).
<code>bootstrap</code>	If <code>TRUE</code> , then forecast distributions are computed using simulation with resampled errors.
<code>...</code>	Additional arguments for forecast model methods.

### See Also

[fabletools::generate.mdl\\_df](#)

### Examples

```
as_tsibble(1h) %>%
  model(AR(value ~ order(3))) %>%
  generate()
```

---

generate.ARIMA	<i>Generate new data from a fable model</i>
----------------	---

---

### Description

Simulates future paths from a dataset using a fitted model. Innovations are sampled by the model's assumed error distribution. If `bootstrap` is `TRUE`, innovations will be sampled from the model's residuals. If `new_data` contains the `.innov` column, those values will be treated as innovations.

### Usage

```
## S3 method for class 'ARIMA'
generate(x, new_data, specials, bootstrap = FALSE, ...)
```

**Arguments**

x	A fitted model.
new_data	A tsibble containing future information used to forecast.
specials	(passed by <a href="#">fabletools::forecast.mdl_df()</a> ).
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
...	Additional arguments for forecast model methods.

**See Also**

[fabletools::generate.mdl\\_df](#)

**Examples**

```
fable_fit <- as_tsibble(USAccDeaths) %>%
  model(model = ARIMA(value ~ 0 + pdq(0,1,1) + PDQ(0,1,1)))
fable_fit %>% generate(times = 10)
```

---

generate.ETS

*Generate new data from a fable model*

---

**Description**

Simulates future paths from a dataset using a fitted model. Innovations are sampled by the model's assumed error distribution. If bootstrap is TRUE, innovations will be sampled from the model's residuals. If new\_data contains the .innov column, those values will be treated as innovations.

**Usage**

```
## S3 method for class 'ETS'
generate(x, new_data, specials, bootstrap = FALSE, ...)
```

**Arguments**

x	A fitted model.
new_data	A tsibble containing future information used to forecast.
specials	(passed by <a href="#">fabletools::forecast.mdl_df()</a> ).
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
...	Additional arguments for forecast model methods.

**See Also**

[fabletools::generate.mdl\\_df](#)

## Examples

```
as_tsibble(USAccDeaths) %>%
  model(ETS(log(value) ~ season("A"))) %>%
  generate(times = 100)
```

---

generate.model\_mean     *Generate new data from a fable model*

---

## Description

Simulates future paths from a dataset using a fitted model. Innovations are sampled by the model's assumed error distribution. If `bootstrap` is `TRUE`, innovations will be sampled from the model's residuals. If `new_data` contains the `.innov` column, those values will be treated as innovations.

## Usage

```
## S3 method for class 'model_mean'
generate(x, new_data, bootstrap = FALSE, ...)
```

## Arguments

<code>x</code>	A fitted model.
<code>new_data</code>	A tsibble containing future information used to forecast.
<code>bootstrap</code>	If <code>TRUE</code> , then forecast distributions are computed using simulation with resampled errors.
<code>...</code>	Additional arguments for forecast model methods.

## See Also

[fabletools::generate.mdl\\_df](#)

## Examples

```
library(tsibbledata)
vic_elec %>%
  model(avg = MEAN(Demand)) %>%
  generate()
```



---

generate.NNETAR	<i>Generate new data from a fable model</i>
-----------------	---

---

**Description**

Simulates future paths from a dataset using a fitted model. Innovations are sampled by the model's assumed error distribution. If `bootstrap` is `TRUE`, innovations will be sampled from the model's residuals. If `new_data` contains the `.innov` column, those values will be treated as innovations.

**Usage**

```
## S3 method for class 'NNETAR'
generate(x, new_data, specials = NULL, bootstrap = FALSE, ...)
```

**Arguments**

<code>x</code>	A fitted model.
<code>new_data</code>	A <code>tsibble</code> containing future information used to forecast.
<code>specials</code>	(passed by <code>fabletools::forecast.mdl_df()</code> ).
<code>bootstrap</code>	If <code>TRUE</code> , then forecast distributions are computed using simulation with resampled errors.
<code>...</code>	Additional arguments for forecast model methods.

**See Also**

[fabletools::generate.mdl\\_df](#)

**Examples**

```
as_tsibble(airmiles) %>%
  model(nn = NNETAR(box_cox(value, 0.15))) %>%
  generate()
```

---

generate.RW	<i>Generate new data from a fable model</i>
-------------	---

---

**Description**

Simulates future paths from a dataset using a fitted model. Innovations are sampled by the model's assumed error distribution. If `bootstrap` is `TRUE`, innovations will be sampled from the model's residuals. If `new_data` contains the `.innov` column, those values will be treated as innovations.

**Usage**

```
## S3 method for class 'RW'
generate(x, new_data, bootstrap = FALSE, ...)
```

**Arguments**

x	A fitted model.
new_data	A tsibble containing future information used to forecast.
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
...	Additional arguments for forecast model methods.

**See Also**

[fabletools::generate.mdl\\_df](#)

**Examples**

```
as_tsibble(Nile) %>%
  model(NAIVE(value)) %>%
  generate()

library(tsibbledata)
aus_production %>%
  model(snaive = SNAIVE(Beer ~ lag("year"))) %>%
  generate()
```

---

generate.TSLM

*Generate new data from a fable model*

---

**Description**

Simulates future paths from a dataset using a fitted model. Innovations are sampled by the model's assumed error distribution. If `bootstrap` is TRUE, innovations will be sampled from the model's residuals. If `new_data` contains the `.innov` column, those values will be treated as innovations.

**Usage**

```
## S3 method for class 'TSLM'
generate(x, new_data, specials, bootstrap = FALSE, ...)
```

**Arguments**

x	A fitted model.
new_data	A tsibble containing future information used to forecast.
specials	(passed by <a href="#">fabletools::forecast.mdl_df()</a> ).
bootstrap	If TRUE, then forecast distributions are computed using simulation with resampled errors.
...	Additional arguments for forecast model methods.

**See Also**

[fabletools::generate.mdl\\_df](#)

**Examples**

```
as_tsibble(USAccDeaths) %>%  
  model(lm = TSLM(log(value) ~ trend() + season())) %>%  
  generate()
```

---

glance.AR

*Glance a AR*

---

**Description**

Construct a single row summary of the AR model.

**Usage**

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

**Arguments**

x                    model or other R object to convert to single-row data frame  
...                   other arguments passed to methods

**Details**

Contains the variance of residuals (`sigma2`), the log-likelihood (`log_lik`), and information criterion (AIC, AICc, BIC).

**Value**

A one row tibble summarising the model's fit.

**Examples**

```
as_tsibble(lh) %>%  
  model(AR(value ~ order(3))) %>%  
  glance()
```

---

glance.ARIMA

*Glance an ARIMA model*


---

### Description

Construct a single row summary of the ARIMA model.

### Usage

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

### Arguments

`x` model or other R object to convert to single-row data frame  
`...` other arguments passed to methods

### Format

A data frame with 1 row, with columns:

**sigma2** The unbiased variance of residuals. Calculated as  $\text{sum}(\text{residuals}^2) / (\text{num\_observations} - \text{num\_parameters} + 1)$

**log\_lik** The log-likelihood

**AIC** Akaike information criterion

**AICc** Akaike information criterion, corrected for small sample sizes

**BIC** Bayesian information criterion

**ar\_roots, ma\_roots** The model's characteristic roots

### Value

A one row tibble summarising the model's fit.

### Examples

```
USAccDeaths %>%
  as_tsibble() %>%
  model(arima = ARIMA(log(value) ~ pdq(0, 1, 1) + PDQ(0, 1, 1))) %>%
  glance()
```

---

glance.ETS	<i>Glance an ETS model</i>
------------	----------------------------

---

**Description**

Construct a single row summary of the ETS model.

**Usage**

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

**Arguments**

x	model or other R object to convert to single-row data frame
...	other arguments passed to methods

**Details**

Contains the variance of residuals (`sigma2`), the log-likelihood (`log_lik`), and information criterion (AIC, AICc, BIC).

**Value**

A one row tibble summarising the model's fit.

**Examples**

```
as_tsibble(USAccDeaths) %>%  
  model(ets = ETS(log(value) ~ season("A"))) %>%  
  glance()
```

---

glance.fable_theta	<i>Glance a theta method</i>
--------------------	------------------------------

---

**Description**

Construct a single row summary of the average method model.

**Usage**

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

**Arguments**

x                    model or other R object to convert to single-row data frame  
...                   other arguments passed to methods

**Details**

Contains the variance of residuals (sigma2).

**Value**

A one row tibble summarising the model's fit.

---

glance.model_mean	<i>Glance a average method model</i>
-------------------	--------------------------------------

---

**Description**

Construct a single row summary of the average method model.

**Usage**

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

**Arguments**

x                    model or other R object to convert to single-row data frame  
...                   other arguments passed to methods

**Details**

Contains the variance of residuals (sigma2).

**Value**

A one row tibble summarising the model's fit.

**Examples**

```
library(tsibbledata)  
vic_elec %>%  
  model(avg = MEAN(Demand)) %>%  
  glance()
```

---

glance.NNETAR	<i>Glance a NNETAR model</i>
---------------	------------------------------

---

**Description**

Construct a single row summary of the NNETAR model. Contains the variance of residuals ( $\sigma^2$ ).

**Usage**

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

**Arguments**

x	model or other R object to convert to single-row data frame
...	other arguments passed to methods

**Value**

A one row tibble summarising the model's fit.

**Examples**

```
as_tibble(airmiles) %>%
  model(nn = NNETAR(box_cox(value, 0.15))) %>%
  glance()
```

---

glance.RW	<i>Glance a lag walk model</i>
-----------	--------------------------------

---

**Description**

Construct a single row summary of the lag walk model. Contains the variance of residuals ( $\sigma^2$ ).

**Usage**

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

**Arguments**

x	model or other R object to convert to single-row data frame
...	other arguments passed to methods

**Value**

A one row tibble summarising the model's fit.

**Examples**

```
as_tsibble(Nile) %>%
  model(NAIVE(value)) %>%
  glance()

library(tsibbledata)
aus_production %>%
  model(snaive = SNAIVE(Beer ~ lag("year"))) %>%
  glance()
```

---

glance.TSLM

*Glance a TSLM*


---

**Description**

Construct a single row summary of the TSLM model.

**Usage**

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

**Arguments**

```
x          model or other R object to convert to single-row data frame
...        other arguments passed to methods
```

**Details**

Contains the R squared (`r_squared`), variance of residuals (`sigma2`), the log-likelihood (`log_lik`), and information criterion (AIC, AICc, BIC).

**Value**

A one row tibble summarising the model's fit.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(lm = TSLM(log(value) ~ trend() + season())) %>%
  glance()
```



---

glance.VAR	<i>Glance a VAR</i>
------------	---------------------

---

**Description**

Construct a single row summary of the VAR model.

**Usage**

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

**Arguments**

x	model or other R object to convert to single-row data frame
...	other arguments passed to methods

**Details**

Contains the variance of residuals (`sigma2`), the log-likelihood (`log_lik`), and information criterion (AIC, AICc, BIC).

**Value**

A one row tibble summarising the model's fit.

**Examples**

```
lung_deaths <- cbind(mdeaths, fdeaths) %>%
  as_tsibble(pivot_longer = FALSE)

lung_deaths %>%
  model(VAR(vars(mdeaths, fdeaths) ~ AR(3))) %>%
  glance()
```

---

interpolate.ARIMA	<i>Interpolate missing values from a fable model</i>
-------------------	--

---

**Description**

Applies a model-specific estimation technique to predict the values of missing values in a tsibble, and replace them.

**Usage**

```
## S3 method for class 'ARIMA'
interpolate(object, new_data, specials, ...)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
...	Additional arguments for forecast model methods.

**Value**

A tibble of the same dimension of `new_data` with missing values interpolated.

**Examples**

```
library(tsibbledata)

olympic_running %>%
  model(arima = ARIMA(Time ~ trend())) %>%
  interpolate(olympic_running)
```

---

interpolate.model\_mean

*Interpolate missing values from a fable model*

---

**Description**

Applies a model-specific estimation technique to predict the values of missing values in a tibble, and replace them.

**Usage**

```
## S3 method for class 'model_mean'
interpolate(object, new_data, specials, ...)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
...	Additional arguments for forecast model methods.

**Value**

A tibble of the same dimension of `new_data` with missing values interpolated.

## Examples

```
library(tsibbledata)

olympic_running %>%
  model(mean = MEAN(Time)) %>%
  interpolate(olympic_running)
```

---

interpolate.TSLM      *Interpolate missing values from a fable model*

---

## Description

Applies a model-specific estimation technique to predict the values of missing values in a tsibble, and replace them.

## Usage

```
## S3 method for class 'TSLM'
interpolate(object, new_data, specials, ...)
```

## Arguments

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
...	Additional arguments for forecast model methods.

## Value

A tibble of the same dimension of `new_data` with missing values interpolated.

## Examples

```
library(tsibbledata)

olympic_running %>%
  model(lm = TSLM(Time ~ trend())) %>%
  interpolate(olympic_running)
```

---

MEAN

*Mean models*

---

### Description

MEAN() returns an iid model applied to the formula's response variable.

### Usage

```
MEAN(formula, ...)
```

### Arguments

formula	Model specification.
...	Not used.

### Details

The model does not use any specials, and so everything on the formula's right-hand-side will be ignored.

### Value

A model specification.

### Specials

**window:** The window special is used to specify a rolling window for the mean.

```
window(size = NULL)
```

size The size (number of observations) for the rolling window. If NULL (default), a rolling window will not be used.

### See Also

[Forecasting: Principles and Practices, Some simple forecasting methods \(section 3.2\)](#)

### Examples

```
library(tsibbledata)
vic_elec %>%
  model(avg = MEAN(Demand))
```

---

NNETAR

*Neural Network Time Series Forecasts*

---

**Description**

Feed-forward neural networks with a single hidden layer and lagged inputs for forecasting univariate time series.

**Usage**

```
NNETAR(formula, n_nodes = NULL, n_networks = 20, scale_inputs = TRUE, ...)
```

**Arguments**

formula	Model specification (see "Specials" section).
n_nodes	Number of nodes in the hidden layer. Default is half of the number of input nodes (including external regressors, if given) plus 1.
n_networks	Number of networks to fit with different random starting weights. These are then averaged when producing forecasts.
scale_inputs	If TRUE, inputs are scaled by subtracting the column means and dividing by their respective standard deviations. Scaling is applied after transformations.
...	Other arguments passed to <code>\link[nnet]{nnet}</code> .

**Details**

A feed-forward neural network is fitted with lagged values of the response as inputs and a single hidden layer with `size` nodes. The inputs are for lags 1 to `p`, and lags `m` to `mP` where `m` is the seasonal period specified.

If exogenous regressors are provided, its columns are also used as inputs. Missing values are currently not supported by this model. A total of `repeats` networks are fitted, each with random starting weights. These are then averaged when computing forecasts. The network is trained for one-step forecasting. Multi-step forecasts are computed recursively.

For non-seasonal data, the fitted model is denoted as an NNAR( $p,k$ ) model, where  $k$  is the number of hidden nodes. This is analogous to an AR( $p$ ) model but with non-linear functions. For seasonal data, the fitted model is called an NNAR( $p,P,k$ )[ $m$ ] model, which is analogous to an ARIMA( $p,0,0$ )( $P,0,0$ )[ $m$ ] model but with non-linear functions.

**Value**

A model specification.

**Specials**

**AR:** The AR special is used to specify auto-regressive components in each of the nodes of the neural network.

```
AR(p = NULL, P = 1, period = NULL)
```

<code>p</code>	The order of the non-seasonal auto-regressive (AR) terms. If <code>p = NULL</code> , an optimal number of lags will be selected for
<code>P</code>	The order of the seasonal auto-regressive (SAR) terms.
<code>period</code>	The periodic nature of the seasonality. This can be either a number indicating the number of observations in each se

**xreg:** Exogenous regressors can be included in an NNETAR model without explicitly using the `xreg()` special. Common exogenous regressor specials as specified in `common_xregs` can also be used. These regressors are handled using `stats::model.frame()`, and so interactions and other functionality behaves similarly to `stats::lm()`.

```
xreg(...)
```

```
... Bare expressions for the exogenous regressors (such as log(x))
```

### See Also

[Forecasting: Principles and Practices, Neural network models \(section 11.3\)](#)

### Examples

```
as_tsibble(airmiles) %>%
  model(nn = NNETAR(box_cox(value, 0.15)))
```

---

refit.AR

*Refit an AR model*

---

### Description

Applies a fitted AR model to a new dataset.

### Usage

```
## S3 method for class 'AR'
refit(object, new_data, specials = NULL, reestimate = FALSE, ...)
```

### Arguments

<code>object</code>	The time series model used to produce the forecasts
<code>new_data</code>	A <code>tsibble</code> containing future information used to forecast.
<code>specials</code>	(passed by <code>fabletools::forecast.mdl_df()</code> ).
<code>reestimate</code>	If TRUE, the coefficients for the fitted model will be re-estimated to suit the new data.
<code>...</code>	Additional arguments for forecast model methods.

### Value

A refitted model.

**Examples**

```
lung_deaths_male <- as_tsibble(mdeaths)
lung_deaths_female <- as_tsibble(fdeaths)

fit <- lung_deaths_male %>%
  model(AR(value ~ 1 + order(10)))

report(fit)

fit %>%
  refit(lung_deaths_female) %>%
  report()
```

---

refit.ARIMA

*Refit an ARIMA model*


---

**Description**

Applies a fitted ARIMA model to a new dataset.

**Usage**

```
## S3 method for class 'ARIMA'
refit(object, new_data, specials = NULL, reestimate = FALSE, ...)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
reestimate	If TRUE, the coefficients for the fitted model will be re-estimated to suit the new data.
...	Additional arguments for forecast model methods.

**Value**

A refitted model.

**Examples**

```
lung_deaths_male <- as_tsibble(mdeaths)
lung_deaths_female <- as_tsibble(fdeaths)

fit <- lung_deaths_male %>%
  model(ARIMA(value ~ 1 + pdq(2, 0, 0) + PDQ(2, 1, 0)))

report(fit)
```

```
fit %>%
  refit(lung_deaths_female) %>%
  report()
```

---

refit.ETS

*Refit an ETS model*


---

## Description

Applies a fitted ETS model to a new dataset.

## Usage

```
## S3 method for class 'ETS'
refit(
  object,
  new_data,
  specials = NULL,
  reestimate = FALSE,
  reinitialise = TRUE,
  ...
)
```

## Arguments

object	The time series model used to produce the forecasts
new_data	A <code>tsibble</code> containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
reestimate	If TRUE, the coefficients for the fitted model will be re-estimated to suit the new data.
reinitialise	If TRUE, the initial parameters will be re-estimated to suit the new data.
...	Additional arguments for forecast model methods.

## Examples

```
lung_deaths_male <- as_tsibble(mdeaths)
lung_deaths_female <- as_tsibble(fdeaths)

fit <- lung_deaths_male %>%
  model(ETS(value))

report(fit)

fit %>%
  refit(lung_deaths_female, reinitialise = TRUE) %>%
  report()
```



---

refit.model_mean	<i>Refit a MEAN model</i>
------------------	---------------------------

---

**Description**

Applies a fitted average method model to a new dataset.

**Usage**

```
## S3 method for class 'model_mean'
refit(object, new_data, specials = NULL, reestimate = FALSE, ...)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
reestimate	If TRUE, the mean for the fitted model will be re-estimated to suit the new data.
...	Additional arguments for forecast model methods.

**Examples**

```
lung_deaths_male <- as_tsibble(mdeaths)
lung_deaths_female <- as_tsibble(fdeaths)

fit <- lung_deaths_male %>%
  model(MEAN(value))

report(fit)

fit %>%
  refit(lung_deaths_female) %>%
  report()
```

---

refit.NNETAR	<i>Refit a NNETAR model</i>
--------------	-----------------------------

---

**Description**

Applies a fitted NNETAR model to a new dataset.

**Usage**

```
## S3 method for class 'NNETAR'
refit(object, new_data, specials = NULL, reestimate = FALSE, ...)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
reestimate	If TRUE, the networks will be initialized with random starting weights to suit the new data. If FALSE, for every network the best individual set of weights found in the pre-estimation process is used as the starting weight vector.
...	Additional arguments for forecast model methods.

**Value**

A refitted model.

**Examples**

```
lung_deaths_male <- as_tsibble(mdeaths)
lung_deaths_female <- as_tsibble(fdeaths)

fit <- lung_deaths_male %>%
  model(NNETAR(value))

report(fit)

fit %>%
  refit(new_data = lung_deaths_female, reestimate = FALSE) %>%
  report()
```

---

 refit.RW

*Refit a lag walk model*


---

**Description**

Applies a fitted random walk model to a new dataset.

**Usage**

```
## S3 method for class 'RW'
refit(object, new_data, specials = NULL, reestimate = FALSE, ...)
```

**Arguments**

object	The time series model used to produce the forecasts
new_data	A tsibble containing future information used to forecast.
specials	(passed by <code>fabletools::forecast.mdl_df()</code> ).
reestimate	If TRUE, the lag walk model will be re-estimated to suit the new data.
...	Additional arguments for forecast model methods.

**Details**

The models NAIVE and SNAIVE have no specific model parameters. Using `refit` for one of these models will provide the same estimation results as one would use `fabletools::model(NAIVE(...))` (or `fabletools::model(SNAIVE(...))`).

**Examples**

```
lung_deaths_male <- as_tsibble(mdeaths)
lung_deaths_female <- as_tsibble(fdeaths)

fit <- lung_deaths_male %>%
  model(RW(value ~ drift()))

report(fit)

fit %>%
  refit(lung_deaths_female) %>%
  report()
```

---

refit.TSLM

*Refit a TSLM*


---

**Description**

Applies a fitted TSLM to a new dataset.

**Usage**

```
## S3 method for class 'TSLM'
refit(object, new_data, specials = NULL, reestimate = FALSE, ...)
```

**Arguments**

<code>object</code>	The time series model used to produce the forecasts
<code>new_data</code>	A <code>tsibble</code> containing future information used to forecast.
<code>specials</code>	(passed by <code>fabletools::forecast.mdl_df()</code> ).
<code>reestimate</code>	If TRUE, the coefficients for the fitted model will be re-estimated to suit the new data.
<code>...</code>	Additional arguments for forecast model methods.

**Examples**

```
lung_deaths_male <- as_tsibble(mdeaths)
lung_deaths_female <- as_tsibble(fdeaths)

fit <- lung_deaths_male %>%
  model(TSLM(value ~ trend() + season()))
```

```

report(fit)

fit %>%
  refit(lung_deaths_female) %>%
  report()

```

---

residuals.AR                    *Extract residuals from a fable model*

---

### Description

Extracts the residuals.

### Usage

```

## S3 method for class 'AR'
residuals(object, type = c("innovation", "regression"), ...)

```

### Arguments

object                    The time series model used to produce the forecasts  
type                        The type of residuals to extract.  
...                         Additional arguments for forecast model methods.

### Value

A vector of fitted residuals.

### Examples

```

as_tsibble(lh) %>%
  model(AR(value ~ order(3))) %>%
  residuals()

```

---

residuals.ARIMA                *Extract residuals from a fable model*

---

### Description

Extracts the residuals.

### Usage

```

## S3 method for class 'ARIMA'
residuals(object, type = c("innovation", "regression"), ...)

```

**Arguments**

object           The time series model used to produce the forecasts  
 type            The type of residuals to extract.  
 ...             Additional arguments for forecast model methods.

**Value**

A vector of fitted residuals.

**Examples**

```
USAccDeaths %>%
  as_tsibble() %>%
  model(arima = ARIMA(log(value) ~ pdq(0, 1, 1) + PDQ(0, 1, 1))) %>%
  residuals()
```

---

residuals.croston       *Extract residuals from a fable model*

---

**Description**

Extracts the residuals.

**Usage**

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

**Arguments**

object           The time series model used to produce the forecasts  
 ...             Additional arguments for forecast model methods.

**Value**

A vector of fitted residuals.

**Examples**

```
library(tsibble)
sim_poisson <- tsibble(
  time = yearmonth("2012 Dec") + seq_len(24),
  count = rpois(24, lambda = 0.3),
  index = time
)

sim_poisson %>%
  model(CROSTON(count)) %>%
  residuals()
```

---

residuals.ETS	<i>Extract residuals from a fable model</i>
---------------	---

---

**Description**

Extracts the residuals.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted residuals.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(ets = ETS(log(value) ~ season("A"))) %>%
  residuals()
```

---

residuals.fable_theta	<i>Extract residuals from a fable model</i>
-----------------------	---

---

**Description**

Extracts the residuals.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted residuals.

**Examples**

```
library(tsibbledata)
vic_elec %>%
  model(avg = MEAN(Demand)) %>%
  residuals()
```

---

residuals.model\_mean *Extract residuals from a fable model*

---

**Description**

Extracts the residuals.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted residuals.

**Examples**

```
library(tsibbledata)
vic_elec %>%
  model(avg = MEAN(Demand)) %>%
  residuals()
```

---

residuals.NNETAR	<i>Extract residuals from a fable model</i>
------------------	---

---

**Description**

Extracts the residuals.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted residuals.

**Examples**

```
as_tsibble(airmiles) %>%
  model(nn = NNETAR(box_cox(value, 0.15))) %>%
  residuals()
```

---

residuals.RW	<i>Extract residuals from a fable model</i>
--------------	---

---

**Description**

Extracts the residuals.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.



**Value**

A vector of fitted residuals.

**Examples**

```
as_tsibble(Nile) %>%
  model(NAIVE(value)) %>%
  residuals()

library(tsibbledata)
aus_production %>%
  model(snaive = SNAIVE(Beer ~ lag("year"))) %>%
  residuals()
```

---

residuals.TSLM	<i>Extract residuals from a fable model</i>
----------------	---

---

**Description**

Extracts the residuals.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted residuals.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(lm = TSLM(log(value) ~ trend() + season())) %>%
  residuals()
```

---

residuals.VAR	<i>Extract residuals from a fable model</i>
---------------	---

---

**Description**

Extracts the residuals.

**Usage**

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

**Arguments**

object	The time series model used to produce the forecasts
...	Additional arguments for forecast model methods.

**Value**

A vector of fitted residuals.

**Examples**

```
lung_deaths <- cbind(mdeaths, fdeaths) %>%
  as_tsibble(pivot_longer = FALSE)

lung_deaths %>%
  model(VAR(vars(mdeaths, fdeaths) ~ AR(3))) %>%
  residuals()
```

---

RW	<i>Random walk models</i>
----	---------------------------

---

**Description**

`RW()` returns a random walk model, which is equivalent to an `ARIMA(0,1,0)` model with an optional drift coefficient included using `drift()`. `naive()` is simply a wrapper to `rwf()` for simplicity. `snaive()` returns forecasts and prediction intervals from an `ARIMA(0,0,0)(0,1,0)m` model where `m` is the seasonal period.

**Usage**

```
RW(formula, ...)

NAIVE(formula, ...)

SNAIVE(formula, ...)
```

**Arguments**

formula      Model specification (see "Specials" section).  
 ...          Not used.

**Details**

The random walk with drift model is

$$Y_t = c + Y_{t-1} + Z_t$$

where  $Z_t$  is a normal iid error. Forecasts are given by

$$Y_n(h) = ch + Y_n$$

. If there is no drift (as in naive), the drift parameter  $c=0$ . Forecast standard errors allow for uncertainty in estimating the drift parameter (unlike the corresponding forecasts obtained by fitting an ARIMA model directly).

The seasonal naive model is

$$Y_t = Y_{t-m} + Z_t$$

where  $Z_t$  is a normal iid error.

**Value**

A model specification.

**Specials**

**lag:** The lag special is used to specify the lag order for the random walk process. If left out, this special will automatically be included.

lag(lag = NULL)

lag      The lag order for the random walk process. If lag = m, forecasts will return the observation from m time periods ago. This

**drift:** The drift special can be used to include a drift/trend component into the model. By default, drift is not included unless drift() is included in the formula.

drift(drift = TRUE)

drift      If drift = TRUE, a drift term will be included in the model.

**See Also**

[Forecasting: Principles and Practices, Some simple forecasting methods \(section 3.2\)](#)

**Examples**

```
library(tsibbledata)
```

```

aus_production %>%
  model(rw = RW(Beer ~ drift()))

as_tsibble(Nile) %>%
  model(NAIVE(value))
library(tsibbledata)
aus_production %>%
  model(snaive = SNAIVE(Beer ~ lag("year")))

```

---

 THETA

*Theta method*


---

### Description

The theta method of Assimakopoulos and Nikolopoulos (2000) is equivalent to simple exponential smoothing with drift. This is demonstrated in Hyndman and Billah (2003).

### Usage

```
THETA(formula, ...)
```

### Arguments

formula	Model specification.
...	Not used.

### Details

The series is tested for seasonality using the test outlined in A&N. If deemed seasonal, the series is seasonally adjusted using a classical multiplicative decomposition before applying the theta method. The resulting forecasts are then reseasonalized.

More general theta methods are available in the `forecTheta` package.

### Value

A model specification.

### Specials

**season:** The season special is used to specify the parameters of the seasonal adjustment via classical decomposition.

```
season(period = NULL, method = c("multiplicative", "additive"))
```

period	The periodic nature of the seasonality. This can be either a number indicating the number of observations in each season.
method	The type of classical decomposition to apply. The original Theta method always used multiplicative seasonal decomposition.

**Author(s)**

Rob J Hyndman, Mitchell O'Hara-Wild

**References**

Assimakopoulos, V. and Nikolopoulos, K. (2000). The theta model: a decomposition approach to forecasting. *International Journal of Forecasting* **16**, 521-530.

Hyndman, R.J., and Billah, B. (2003) Unmasking the Theta method. *International J. Forecasting*, **19**, 287-290.

**Examples**

```
# Theta method with transform
deaths <- as_tsibble(USAccDeaths)
deaths %>%
  model(theta = THETA(log(value))) %>%
  forecast(h = "4 years") %>%
  autoplot(deaths)

# Compare seasonal specifications
library(tsibbledata)
library(dplyr)
aus_retail %>%
  filter(Industry == "Clothing retailing") %>%
  model(theta_multiplicative = THETA(Turnover ~ season(method = "multiplicative")),
        theta_additive = THETA(Turnover ~ season(method = "additive"))) %>%
  accuracy()
```

---

tidy.AR

*Tidy a fable model*

---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'AR'
tidy(x)
```

**Arguments**

x                    An object to be converted into a tidy `tibble::tibble()`.

**Value**

The model's coefficients in a tibble.

**Examples**

```
as_tsibble(lh) %>%
  model(AR(value ~ order(3))) %>%
  tidy()
```

---

tidy.ARIMA	<i>Tidy a fable model</i>
------------	---------------------------

---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'ARIMA'
tidy(x, ...)
```

**Arguments**

`x` An object to be converted into a tidy `tibble::tibble()`.  
`...` Additional arguments to tidying method.

**Value**

The model's coefficients in a tibble.

**Examples**

```
USAccDeaths %>%
  as_tsibble() %>%
  model(arima = ARIMA(log(value) ~ pdq(0, 1, 1) + PDQ(0, 1, 1))) %>%
  tidy()
```

---

tidy.croston	<i>Tidy a fable model</i>
--------------	---------------------------

---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'croston'
tidy(x, ...)
```

**Arguments**

`x` An object to be converted into a tidy `tibble::tibble()`.  
`...` Additional arguments to tidying method.

**Value**

The model's coefficients in a tibble.

**Examples**

```
library(tsibble)
sim_poisson <- tsibble(
  time = yearmonth("2012 Dec") + seq_len(24),
  count = rpois(24, lambda = 0.3),
  index = time
)

sim_poisson %>%
  model(CROSTON(count)) %>%
  tidy()
```

---

tidy.ETS

*Tidy a fable model*


---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'ETS'
tidy(x, ...)
```

**Arguments**

`x` An object to be converted into a tidy `tibble::tibble()`.  
`...` Additional arguments to tidying method.

**Value**

The model's coefficients in a tibble.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(ets = ETS(log(value) ~ season("A"))) %>%
  tidy()
```

---

tidy.fable_theta	<i>Tidy a fable model</i>
------------------	---------------------------

---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'fable_theta'
tidy(x, ...)
```

**Arguments**

x                    An object to be converted into a tidy `tibble::tibble()`.  
 ...                 Additional arguments to tidying method.

**Value**

The model's coefficients in a tibble.

**Examples**

```
USAccDeaths %>%
  as_tsibble() %>%
  model(arima = ARIMA(log(value) ~ pdq(0, 1, 1) + PDQ(0, 1, 1))) %>%
  tidy()
```

---

tidy.model_mean	<i>Tidy a fable model</i>
-----------------	---------------------------

---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'model_mean'
tidy(x, ...)
```

**Arguments**

x                    An object to be converted into a tidy `tibble::tibble()`.  
 ...                 Additional arguments to tidying method.



**Value**

The model's coefficients in a tibble.

**Examples**

```
library(tsibbledata)
vic_elec %>%
  model(avg = MEAN(Demand)) %>%
  tidy()
```

---

tidy.NNETAR

*Tidy a fable model*


---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'NNETAR'
tidy(x, ...)
```

**Arguments**

`x` An object to be converted into a tidy `tibble::tibble()`.  
`...` Additional arguments to tidying method.

**Value**

The model's coefficients in a tibble.

**Examples**

```
as_tsibble(airmiles) %>%
  model(nn = NNETAR(box_cox(value, 0.15))) %>%
  tidy()
```

---

tidy.RW

*Tidy a fable model*


---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'RW'
tidy(x, ...)
```

**Arguments**

`x` An object to be converted into a tidy `tibble::tibble()`.  
`...` Additional arguments to tidying method.

**Value**

The model's coefficients in a tibble.

**Examples**

```
as_tsibble(Nile) %>%
  model(NAIVE(value)) %>%
  tidy()

library(tsibbledata)
aus_production %>%
  model(snaive = SNAIVE(Beer ~ lag("year"))) %>%
  tidy()
```

---

tidy.TSLM

*Tidy a fable model*


---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'TSLM'
tidy(x, ...)
```

**Arguments**

x                    An object to be converted into a tidy `tibble::tibble()`.  
 ...                  Additional arguments to tidying method.

**Value**

The model's coefficients in a tibble.

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(lm = TSLM(log(value) ~ trend() + season())) %>%
  tidy()
```

---

tidy.VAR	<i>Tidy a fable model</i>
----------	---------------------------

---

**Description**

Returns the coefficients from the model in a tibble format.

**Usage**

```
## S3 method for class 'VAR'
tidy(x)
```

**Arguments**

x                    An object to be converted into a tidy `tibble::tibble()`.

**Value**

The model's coefficients in a tibble.

**Examples**

```
lung_deaths <- cbind(mdeaths, fdeaths) %>%
  as_tsibble(pivot_longer = FALSE)

lung_deaths %>%
  model(VAR(vars(mdeaths, fdeaths) ~ AR(3))) %>%
  tidy()
```

TSLM

*Fit a linear model with time series components***Description**

The model formula will be handled using `stats::model.matrix()`, and so the the same approach to include interactions in `stats::lm()` applies when specifying the formula. In addition to `stats::lm()`, it is possible to include `common_xregs` in the model formula, such as `trend()`, `season()`, and `fourier()`.

**Usage**

```
TSLM(formula)
```

**Arguments**

```
formula      Model specification.
```

**Value**

A model specification.

**Specials**

**xreg:** Exogenous regressors can be included in a TSLM model without explicitly using the `xreg()` special. Common exogenous regressor specials as specified in `common_xregs` can also be used. These regressors are handled using `stats::model.frame()`, and so interactions and other functionality behaves similarly to `stats::lm()`.

```
xreg(...)
```

```
... Bare expressions for the exogenous regressors (such as log(x))
```

**See Also**

[stats::lm\(\)](#), [stats::model.matrix\(\)](#) [Forecasting: Principles and Practices, Time series regression models \(chapter 6\)](#)

**Examples**

```
as_tsibble(USAccDeaths) %>%
  model(lm = TSLM(log(value) ~ trend() + season()))
```

```
library(tsibbledata)
olympic_running %>%
  model(TSLM(Time ~ trend())) %>%
  interpolate(olympic_running)
```

---

unitroot_options	<i>Options for the unit root tests for order of integration</i>
------------------	---

---

### Description

By default, a kpss test (via `feasts::unitroot_kpss()`) will be performed for testing the required first order differences, and a test of the seasonal strength (via `feasts::feat_stl()` seasonal\_strength) being above the 0.64 threshold is used for determining seasonal required differences.

### Usage

```
unitroot_options(
  ndiffs_alpha = 0.05,
  nsdiffs_alpha = 0.05,
  ndiffs_pvalue = ~feasts::unitroot_kpss(.)["kpss_pvalue"],
  nsdiffs_pvalue = ur_seasonal_strength(0.64)
)
```

### Arguments

`ndiffs_alpha`, `nsdiffs_alpha`

The level for the test specified in the pval functions. As long as  $pval < \alpha$ , differences will be added.

`ndiffs_pvalue`, `nsdiffs_pvalue`

A function (or lambda expression) that provides a p-value for the unit root test. As long as  $pval < \alpha$ , differences will be added.

For the function for the seasonal p-value, the seasonal period will be provided as the `.period` argument to this function. A vector of data to test is available as `.` or `.x`.

### Value

A list of parameters

---

VAR	<i>Estimate a VAR model</i>
-----	-----------------------------

---

### Description

Searches through the vector of lag orders to find the best VAR model which has lowest AIC, AICc or BIC value. It is implemented using OLS per equation.

### Usage

```
VAR(formula, ic = c("aicc", "aic", "bic"), ...)
```

**Arguments**

<code>formula</code>	Model specification (see "Specials" section).
<code>ic</code>	The information criterion used in selecting the model.
<code>...</code>	Further arguments for <code>arma</code>

**Details**

Exogenous regressors and `common_xregs` can be specified in the model formula.

**Value**

A model specification.

**Specials**

**pdq:** The AR special is used to specify the lag order for the auto-regression.

`AR(p = 0:5)`

`p` The order of the auto-regressive (AR) terms. If multiple values are provided, the one which minimises `ic` will be chosen.

**xreg:** Exogenous regressors can be included in an VAR model without explicitly using the `xreg()` special. Common exogenous regressor specials as specified in `common_xregs` can also be used. These regressors are handled using `stats::model.frame()`, and so interactions and other functionality behaves similarly to `stats::lm()`.

The inclusion of a constant in the model follows the similar rules to `stats::lm()`, where including `1` will add a constant and `0` or `-1` will remove the constant. If left out, the inclusion of a constant will be determined by minimising `ic`.

`xreg(...)`

`...` Bare expressions for the exogenous regressors (such as `log(x)`)

**See Also**

[Forecasting: Principles and Practices, Vector autoregressions \(section 11.2\)](#)

**Examples**

```
lung_deaths <- cbind(mdeaths, fdeaths) %>%
  as_tsibble(pivot_longer = FALSE)

fit <- lung_deaths %>%
  model(VAR(vars(mdeaths, fdeaths) ~ AR(3)))

report(fit)
```

```
fit %>%  
  forecast() %>%  
  autoplot(lung_deaths)
```

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