Package: muse (via r-universe)

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Type Package

Title Multiple Unobserved Sources of Error State Space Models

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URL https://github.com/config-i1/muse

BugReports https://github.com/config-i1/muse/issues

Language en-GB

Description Implements the Power / Trend / Seasonal (PTS) model, a unified state-space framework based on the Multiple Source of Error (MSOE) model. It brings the trend, seasonal and irregular component models of Harvey (1989) <doi:10.1017/CBO9781107049994>, Durbin and Koopman (2012) <doi:10.1093/acprof:oso/9780199641178.001.0001>, Proietti (2000) <doi:10.1016/S0169-2070(00)00037-6>, Sbrana and Silvestrini (2023) <doi:10.1016/j.ijforecast.2022.03.003> and others together under a single estimation, selection and forecasting interface, with an optional Box-Cox power transformation. Models are estimated by maximum likelihood through the Kalman filter and smoother, with automatic component selection by information criteria.

License LGPL-2.1

Depends R (>= 3.0.2), greybox, smooth

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Contents

`initials` *Initial state values for a fitted* `pts` *object.*

Description

Returns the smoothed structural states at the first observation, with the `"Error"` and `"Fit"` columns dropped. For deterministic components – e.g. the slope under the global (`G`) trend, where the slope is fixed throughout the horizon – this equals the t = 0 initial value; for stochastic components it is the smoother's estimate at t = 1, a close proxy to the t = 0 initial.

Usage

```
initials(object, ...)
```

Arguments

`object`	A fitted object of class `"pts"`.
`...`	Unused.

Value

Named numeric vector of initial structural-state values.

`muse` *muse package*

Description

Package contains functions implementing Multiple Source of Error state space models for purposes of time series analysis and forecasting.

Details

Package:	muse
Type:	Package
Date:	2024-10-18 - Inf
License:	LGPL-2.1

The following functions are included in the package:

Author(s)

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See Also

`adam`

Examples

```
x <- rnorm(100,100,10)
```

`pts` *pts: Power / Trend / Seasonal state-space model*

Description

Estimates a PTS (Power / Trend / Seasonal) state-space model for a univariate time series. This is the user-facing entry point of the **muse** package and mirrors the calling convention used elsewhere in the **smooth** family: `pts()` estimates the model, and `forecast.pts` produces forecasts from the fitted object without re-estimating.

Usage

```
pts(data, model = "ZZZ", lags = stats::frequency(data), orders = list(ar
  = 0, ma = 0, select = FALSE), formula = NULL, regressors = c("use"),
  outliers = c("ignore", "use", "select"), level = 0.99, ic = c("AICc",
  "BICc", "BIC", "AIC"), lambda_estim = c("likelihood", "guerrero",
  "decomp-guerrero"), biasadj = FALSE, h = 0, holdout = FALSE,
  verbose = FALSE, ...)
```

Arguments

`data`	response series. Either a univariate `ts` / `zoo` / numeric vector, OR a matrix / `data.frame` whose first column is the response and whose remaining columns are external regressors (xregs).
`model`	3-letter PTS specification string. The three positions encode Power / Trend / Seasonal: • Power: `Z` to estimate Box-Cox λ, or a numeric value (e.g. `"0"`, `"0.5"`, `"1"`). • Trend: `Z` (auto), `N` (none / random walk), `L` (local linear), `D` (damped / smooth random walk), `G` (global / deterministic). • Seasonal: `Z` (auto), `N` (none), `D` (discrete / linear), `T` (trigonometric / equal).
`lags`	seasonal period (default `frequency(data)`). Scalar for the structural seasonal; may also be a vector `c(1, s)` mirroring `smooth::adam`'s convention – the last entry is the structural period, the full vector becomes the default lag set for the irregular's ARMA blocks (overridable per-fit via `orders$lags`).
`orders`	ARMA / SARMA spec for the irregular component. Three forms: • Full list `list(ar, ma, select)` for a non-seasonal $ARMA(p, q)$ – `ar`, `ma` non-negative scalars (default 0); `select = TRUE` asks the engine to search ARMA orders up to that cap. • Numeric shortcut `c(p, q)` – equivalent to `list(ar = p, ma = q, select = FALSE)`; `c(p)` is treated as `c(p, 0)`. • Seasonal `list(ar = c(p, P), ma = c(q, Q), lags = c(1, s))` for $SARMA(p, q)(P, Q)_s$ – `ar` / `ma` are length-L vectors paired position-wise with `lags`, with `lags[1] = 1` (non-seasonal block) and `lags[2] = s` (seasonal block). If `orders$lags` is omitted the default falls back to the top-level `lags` argument (or `c(1, frequency(data))`). The seasonal SARMA polynomial is multiplied internally, so the BFGS only optimises the free $\phi_i, \Phi_j, \theta_i, \Theta_j$ coefficients. `select = TRUE` runs a grid search over every (p', q', P', Q') tuple with $0 \leq p' \leq$ `ar[1]` and so on, and picks the candidate with the lowest `ic`. PTS has no differencing – any `orders$i` supplied is silently ignored.
`formula`	optional formula `response ~ x1 + x2 + ...`; only meaningful when `data` is a matrix or `data.frame`. Used to pick the response column + xreg columns explicitly.
`regressors`	handling of xregs. Currently only `"use"` (apply all supplied xregs as fixed-coefficient covariates). Adam's `"select"` and `"adapt"` modes are not yet implemented.

`outliers` what to do about outliers, mirroring `adam()`'s interface: `"ignore"` (default – fit ignores the possibility of outliers) or `"use"` (run the engine's outlier detector once after the structural fit, classify each event as AO / LS / SC, and refit with the detected events as fixed regressor dummies). Adam's `"select"` mode (IC-pruning of detected dummies) is not yet supported – passing it errors with a clear message. When `outliers = "use"` the detected events are returned on the fitted object as `$outliersDetected` (a data frame with `time` and `type` columns) and the corresponding dummy coefficients appear in `coef(m)` as `Beta(...)` entries.

`level` confidence level driving the outlier z-score threshold (default 0.99). Translated to a two-sided z via `qnorm((1 + level) / 2)`: 0.99 -> ~= 2.576, 0.95 -> ~= 1.960. The z drives the AO threshold; LS / SC scale proportionally to preserve the engine's relative stiffness (LS ~= 1.087*z, SC ~= 1.304*z). Ignored when `outliers = "ignore"`.

`ic` information criterion for automatic model selection; one of `"AICc"` (default), `"BICc"`, `"BIC"`, `"AIC"`. Matches the adam option set; AICc is the default, as in `adam`.

`lambda_estim` how the Box-Cox power λ is chosen when the power slot of `model` is `"Z"`; one of:

- `"likelihood"` (default) – estimate λ jointly with the structural parameters by maximum likelihood in the engine.
- `"guerrero"` – the classical Guerrero (1993) variance- stabilisation screen on raw season-length blocks.
- `"decomp-guerrero"` – Guerrero on an `msdecompose`-smoothed trend (the former default).

Ignored when a numeric power is supplied (e.g. `"0.5ZZ"`).

`biasadj` logical (default `FALSE`). Point forecasts are the back-transformed conditional *median* $g^{-1}(\mu)$. When `TRUE`, a second-order bias correction is applied so the point forecast approximates the conditional *mean* (as in `forecast::InvBoxCox(biasadj = TRUE)`). Prediction-interval quantiles are unaffected. Has no effect at $\lambda = 1$.

`h` forecast horizon. If `h > 0` a forecast is computed at fit time and cached on the object; `forecast(object, h)` can later recompute for a different horizon cheaply.

`holdout` logical. If `TRUE` and `h > 0`, the last `h` observations of `data` are withheld from estimation and returned in `$holdout` for later accuracy assessment.

`verbose` logical: print intermediate optimisation output.

`...` advanced / undocumented passthroughs. Supported keys:

- `B` - numeric vector of starting values for the optimiser (natural-scale variances, in the order returned in `$B` by a default fit). Mirrors the same hatch in `smooth::adam()`. The optimised vector is returned in the `$B` slot of the output regardless of whether the user supplied one.

Value

An object of class `c("pts", "smooth")`. Slot names mirror `smooth::adam()`'s return list where the concept is shared; pts-only extensions are flagged below.

- Inputs / spec: `y`, `model`, `modelUC*`, `lags`, `lambda*`

- Parameters: B (estimated parameter vector), covp* (parameter covariance), nParam – an adam-style 2 x 5 matrix (rows Estimated / Provided; columns nParamInternal, nParamXreg, nParamOccurrence, nParamScale, nParamAll). nparam() returns the [Estimated, nParamAll] cell. Estimated structural initials (level/slope, cycle, seasonal states) are folded into nParamInternal, exactly as smooth::adam does
- In-sample fit: fitted, residuals, states plus pts-specific comp* (additive BC-scale decomposition with Error/Fit columns)
- Cached forecast: forecast (original scale, if h > 0) and forecast_args* for cheap re-forecasting
- Likelihood + diagnostics: logLik, table* (C++ validation text)
- Scalars read by plot.smooth / diagnostics: distribution = "dnorm", loss = "likelihood", occurrence = NULL, holdout
- Bookkeeping: call, timeElapsed

AIC / AICc / BIC / BICc are derived on demand via the methods, not stored on the object. (* = pts-specific extension.)

Author(s)

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See Also

forecast.pts

Examples

```
# Automatic model selection (Power / Trend / Seasonal) on monthly data
model <- pts(AirPassengers, model = "ZZZ", h = 12, holdout = TRUE)
model

# A fixed specification: no transform, local-linear trend, trigonometric
# seasonality, with a 12-step forecast
fixedModel <- pts(AirPassengers, model = "1LT", h = 12)
forecast(fixedModel, h = 12)
```

---

pts-methods *S3 methods for objects of class* pts

Description

Standard accessors and printing for fitted `pts` objects. `forecast(object, h)` generates forecasts from the fitted parameters without re-running the optimiser; `predict(object)` returns the in-sample fitted values.

`print.pts` mirrors `smooth::print.adam` (`smooth/R/adam.R:5862`) section by section, substituting the PTS-specific MSOE concepts where ETS-specific ones do not apply: the "initialisation" line becomes the Box-Cox λ block, and the "Persistence vector g" block becomes the MSOE innovation variances (the same variances are otherwise the structural pieces of the B parameter vector). The C++ validation diagnostics live on `$cppOutput` should the user want them.

`forecast.pts` uses the C++ `forecastOnly` entry point: it skips re-estimation and just propagates the Kalman filter `h` steps forward from the fitted state, so changing `h` is cheap. `interval` selects the variance source:

"prediction" (default) state propagation + future shocks (the engine's `yForV`). Bands the next observation.

"confidence" conditional-mean variance. In a state-space model $var(E[y_{t+h} \mid obs]) = var(y_{t+h} \mid obs) - \sigma^2$, where σ^2 is the BC-scale residual variance: we read it straight off the fitted scale, no re-forecast needed.

"simulated" empirical quantiles of `nsim` forward paths from `simulate.pts()`. Returns the path matrix in `$scenarios` when `scenarios = TRUE`.

"none" no bands; `lower = upper = mean`.

`level` accepts a vector; `lower` / `upper` are then $h \times nLevels$ matrices (or $1 \times nLevels$ when `cumulative = TRUE`). `side` produces two-sided / upper-only / lower-only intervals – the absent side is set to $\mp\infty$ on the BC scale and back-transformed to the original-scale support boundary (0 for $\lambda > 0$, $-\infty$ for the identity transform). `cumulative = TRUE` collapses the horizon into one total – exact for `interval = "simulated"` (sum within each path); the engine does not expose cross-step state covariance, so the other intervals fall back to simulation totals.

Usage

```
## S3 method for class 'pts'
print(x, digits = 4, ...)

## S3 method for class 'pts'
plot(x, which = c(1, 2, 4, 6), ...)

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

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

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

## S3 method for class 'pts'
```

```
vcov(object, ...)

## S3 method for class 'pts'
nobs(object, all = FALSE, ...)

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

## S3 method for class 'pts'
predict(object, newdata = NULL, ...)

## S3 method for class 'pts'
forecast(object, h = 10, newdata = NULL,
  interval = c("prediction", "confidence", "simulated", "none"),
  level = 0.95, side = c("both", "upper", "lower"), cumulative = FALSE,
  nsim = NULL, scenarios = FALSE, biasadj = NULL, ...)

## S3 method for class 'pts'
initials(object, ...)
```

Arguments

x, object	A fitted object of class "pts".
digits	number of significant digits used in printed output.
...	further arguments passed to underlying generics.
which	integer vector of plot panels to draw; passed through to plot.smooth (see ?smooth::plot.adam). Defaults to c(1, 2, 4, 6).
all	logical; if TRUE the holdout sample is included in the observation count. Default FALSE.
newdata	unused (reserved for forecast paths with new regressors).
h	forecast horizon (steps ahead).
interval	interval type; one of "prediction", "confidence", "simulated", or "none".
level	confidence level for prediction intervals (default 0.95).
side	one of "both", "upper", "lower" – selects two-sided vs one-sided intervals.
cumulative	if TRUE, return the cumulative h-step total.
nsim	number of simulated paths when interval = "simulated" or under the cumulative fallback (default 10000).
scenarios	if TRUE and interval = "simulated", return the simulated path matrix as $scenarios.
biasadj	point-forecast back-transform: FALSE (median, the default) or TRUE (bias-corrected mean). Defaults to the value the model was fitted with (object$biasadj).

Value

The methods return the following:

- `fitted`, `predict` – the in-sample fitted values as a `ts` / numeric vector on the original (back-transformed) scale.
- `residuals` – the model residuals as a `ts` / numeric vector on the Box-Cox scale.
- `coef` – a named numeric vector of the estimated parameters (structural variances plus any ARMA / damping coefficients).
- `vcov` – the parameter variance-covariance `matrix`.
- `nobs` – an integer, the number of in-sample observations (plus the holdout when `all = TRUE`).
- `nparam` – an integer, the number of estimated parameters (the degrees of freedom used by the information criteria).
- `logLik` – an object of class `"logLik"`: the maximised log-likelihood, with `df` and `nobs` attributes.
- `sigma`, `extractSigma` – a numeric scalar, the residual standard deviation; `extractScale` – the maximum-likelihood scale of the assumed distribution.
- `actuals` – the original response series.
- `modelType` – a character string with the PTS model code (e.g. `"PTS(0,N,T)"`); `modelName` – a human-readable description; `errorType` – the character error type (`"A"`, additive on the Box-Cox scale).
- `lags` – the integer seasonal period; `orders` – a `list` of the irregular ARMA orders (`ar`, `ma`, `lags`); `initials` – a named numeric vector of the estimated initial structural-state values.
- `forecast` – an object of class `"pts.forecast"`: a `list` with the point forecast `$mean` and the interval bounds `$lower` / `$upper` (original scale), together with the forecast variance, level, and interval metadata.
- `print` and `plot` are called for their side effects (console output and a diagnostic plot, respectively) and invisibly return their first argument.

Author(s)

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References

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- Pankratz, A., & Dudley, U. (1987). Forecasts of power-transformed series. *Journal of Forecasting*, 6(4), 239-248. doi:10.1002/for.3980060403
- Guerrero, V. M. (1993). Time-series analysis supported by power transformations. *Journal of Forecasting*, 12(1), 37-48. doi:10.1002/for.3980120104

Examples

```
model <- pts(AirPassengers, model = "1LT", h = 12, holdout = TRUE)
print(model)
fitted(model)
residuals(model)
# forecast 12 steps ahead with 95% prediction intervals
forecast(model, h = 12, interval = "prediction", level = 0.95)
```

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