

Package ‘svines’

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Title Stationary Vine Copula Models

Version 0.3.0

Description Provides functionality to fit and simulate from stationary vine copula models for time series, see Nagler et al. (2022)
<[doi:10.1016/j.jeconom.2021.11.015](https://doi.org/10.1016/j.jeconom.2021.11.015)>.

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LazyData true

URL <https://tnagler.github.io/svines/>,
<https://github.com/tnagler/svines>

BugReports <https://github.com/tnagler/svines/issues>

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Author Thomas Nagler [aut, cre]

Maintainer Thomas Nagler <mail@tnagler.com>

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returns	<i>Stock returns of 20 companies</i>
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Description

Daily log returns of 20 companies from the insurance, automotive, technology, energy, and aviation sectors. The observation period is from 2015-01-01 to 2019-12-31. Trading dates are not stored in the dataset.

Usage

returns

Format

A data frame with 1,296 observations and 20 numeric variables, one for each company: Allianz, AXA, Generali, MetLife, Prudential, Ping An, BMW, General Motors, Toyota, Hyundai, Microsoft, Apple, Amazon, Alphabet, Alibaba, Exxon, Shell, PetroChina, Airbus, and Boeing.

Source

Historical prices obtained from [Yahoo Finance](#).

svine	<i>Stationary vine distribution models</i>
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Description

Automated fitting or creation of custom S-vine distribution models

Usage

```
svine(
  data,
  p,
  var_types,
  margin_families = univariateML::univariateML_models,
  selcrit = "aic",
  ...
)
```

Arguments

<code>data</code>	a matrix or data.frame of data.
<code>p</code>	the Markov order.
<code>var_types</code>	variable types; a character vector with one entry per variable: "c" for continuous, "d" for discrete. Defaults to all-continuous when missing.
<code>margin_families</code>	either a character vector of univariateML::univariateML_models to select from for every margin, or a list with one entry for every variable. For a discrete variable, the corresponding entry must contain only suitable discrete families. Use "empirical" for empirical CDFs.
<code>selcrit</code>	criterion for family selection, either "loglik", "aic", "bic", "mbicv".
<code>...</code>	arguments passed to <code>svinecop()</code> .

Value

Returns the fitted model as an object with classes `svine` and [svine_dist](#). A list with entries

- `$margins`: list of marginal models from [univariateML::univariateML_models](#),
- `$copula`: an object of `svinecop_dist`.

References

Nagler, T., Krüger, D., and Min, A. (2022). Stationary vine copula models for multivariate time series. *Journal of Econometrics*, 227(2), 305–324. doi:[10.1016/j.jeconom.2021.11.015](https://doi.org/10.1016/j.jeconom.2021.11.015).

See Also

[svine_dist](#), [svine_loglik](#), [svine_sim](#), [svine_bootstrap_models](#)

Examples

```
# load data set
data(returns)

# fit parametric S-vine model with Markov order 1
fit <- svine(returns[1:100, 1:3], p = 1, family_set = "parametric")
fit
summary(fit)
plot(fit$copula)
contour(fit$copula)
logLik(fit)

pairs(svine_sim(500, rep = 1, fit))
```

svinecop

Stationary vine copula models

Description

Automated fitting or creation of custom S-vine copula models

Usage

```
svinecop(
  data,
  p,
  var_types,
  family_set = "all",
  cs_structure = NA,
  out_vertices = NA,
  in_vertices = NA,
  type = "S",
  par_method = "mle",
  nonpar_method = "constant",
  mult = 1,
  selcrit = "aic",
  weights = numeric(),
  psi0 = 0.9,
  presel = TRUE,
  trunc_lvl = Inf,
  tree_crit = "tau",
  threshold = 0,
  keep_data = FALSE,
  show_trace = FALSE,
  cores = 1
)
```

Arguments

<code>data</code>	a matrix or data.frame of pseudo-observations (approximately uniform margins). For discrete variables declared in <code>var_types</code> , append one left-limit column $F(x-)$ per discrete variable after all regular columns. Use <code>svine()</code> to handle this transformation automatically.
<code>p</code>	the Markov order.
<code>var_types</code>	variable types; a character vector with one entry per variable: "c" for continuous, "d" for discrete. Omitting <code>var_types</code> when <code>cs_structure</code> is not provided will silently fit a continuous model.
<code>family_set</code>	a character vector of families; see <code>rvinecopulib::bicop()</code> for additional options.
<code>cs_structure</code>	the cross-sectional vine structure (see <code>rvinecopulib::rvine_structure()</code>); <code>cs_structure = NA</code> performs automatic structure selection.
<code>out_vertices</code>	a permutation of 1, ..., d specifying the out-vertices. It is selected automatically when no structure is provided. A fixed <code>cs_structure</code> requires explicit <code>out_vertices</code> and <code>in_vertices</code> .
<code>in_vertices</code>	a permutation of 1, ..., d specifying the in-vertices. It is selected automatically when no structure is provided. A fixed <code>cs_structure</code> requires explicit <code>out_vertices</code> and <code>in_vertices</code> .
<code>type</code>	type of stationary vine; "S" (default) for general S-vines, "D" for Smith's long D-vine, "M" for Beare and Seo's M-vine.
<code>par_method</code>	the estimation method for parametric models, either "mle" for sequential maximum likelihood, "itau" for inversion of Kendall's tau (only available for one-parameter families and "t").
<code>nonpar_method</code>	the estimation method for nonparametric models, either "constant" for the standard transformation estimator, or "linear"/"quadratic" for the local-likelihood approximations of order one/two.
<code>mult</code>	multiplier for the smoothing parameters of nonparametric families. Values larger than 1 make the estimate more smooth, values less than 1 less smooth.
<code>selcrit</code>	criterion for family selection, either "loglik", "aic", "bic", "mbic", or "mbicv".
<code>weights</code>	optional vector of weights for each observation.
<code>psi0</code>	prior probability of a non-independence copula (only used for <code>selcrit = "mbic"</code> and <code>selcrit = "mbicv"</code>).
<code>presel</code>	whether the family set should be thinned out according to symmetry characteristics of the data.
<code>trunc_lvl</code>	the truncation level. Only Inf, corresponding to no truncation, is currently supported.
<code>tree_crit</code>	the criterion for tree selection, one of "tau", "rho", "hoeffd", or "mcor" for Kendall's τ , Spearman's ρ , Hoeffding's D , and maximum correlation, respectively.
<code>threshold</code>	for thresholded vine copulas; NA indicates that the threshold should be selected automatically by <code>rvinecopulib::mBICV()</code> .
<code>keep_data</code>	whether the data should be stored (necessary for using <code>fitted()</code>).

show_trace	logical; whether a trace of the fitting progress should be printed.
cores	number of cores to use; if more than 1, estimation of pair copulas within a tree is done in parallel.

Value

Returns the fitted model as an object with classes `svinecop` and `svinecop_dist`. Also inherits from `vinecop`, `vinecop_dist` such that many functions from `rvinecopulib` can be called.

Examples

```
# load data set
data(returns)

# convert to pseudo observations with empirical cdf for marginal distributions
u <- pseudo_obs(returns[1:100, 1:3])

# fit parametric S-vine copula model with Markov order 1
fit <- svinecop(u, p = 1, family_set = "parametric")
fit
summary(fit)
plot(fit)
contour(fit)
logLik(fit)

pairs(svinecop_sim(500, rep = 1, fit))
```

svinecop_dist	<i>Custom S-vine models</i>
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Description

Custom S-vine models

Usage

```
svinecop_dist(
  pair_copulas,
  cs_structure,
  p,
  out_vertices,
  in_vertices,
  var_types = rep("c", dim(cs_structure)[1])
)
```

Arguments

<code>pair_copulas</code>	A nested list of 'bicop_dist' objects, where <code>pair_copulas[[t]][[e]]</code> corresponds to the pair-copula at edge <code>e</code> in tree <code>t</code> . Only the most-left unique pair copulas are used, others can be omitted.
<code>cs_structure</code>	The cross-sectional structure. Either a matrix, or an <code>rvine_structure</code> object; see <code>rvinecopulib::rvine_structure()</code>
<code>p</code>	the Markov order.
<code>out_vertices</code>	a permutation of <code>1, ..., d</code> specifying the out-vertices.
<code>in_vertices</code>	a permutation of <code>1, ..., d</code> specifying the in-vertices.
<code>var_types</code>	variable types; a character vector with one entry per variable: "c" for continuous, "d" for discrete. For discrete variables, methods on the returned object expect data in the compact layout: all regular CDF $F(x)$ columns first, then one left-limit column $F(x-)$ per discrete variable.

Value

Returns the model as an object with classes `svinecop_dist`. Also inherits from `vinecop_dist` such that many functions from `rvinecopulib` can be called.

See Also

[svinecop_loglik](#), [svinecop_sim](#), [svinecop_hessian](#), [svinecop_scores](#)

Examples

```
cs_struct <- cvine_structure(1:2)
pcs <- list(
  list( # first tree
    bicop_dist("clayton", 0, 3), # cross sectional copula
    bicop_dist("gaussian", 0, -0.1) # serial copula
  ),
  list( # second tree
    bicop_dist("gaussian", 0, 0.2), bicop_dist("indep")
  ),
  list( # third tree
    bicop_dist("indep")
  )
)

cop <- svinecop_dist(
  pcs, cs_struct, p = 1, out_vertices = 1:2, in_vertices = 1:2)
```

svinecop_hessian	<i>Expected Hessian for S-vine copula models</i>
------------------	--

Description

Expected Hessian for S-vine copula models

Usage

```
svinecop_hessian(u, model, cores = 1)
```

Arguments

u	the data; should have approximately uniform margins.
model	model inheriting from class svinecop_dist .
cores	number of cores to use; if larger than one, computations are performed in parallel on cores batches.

Value

A `model$nparams`-by-`model$nparams` estimate of the expected Hessian, obtained by averaging the per-observation Hessian contributions. Rows and columns correspond to model parameters in the order: copula parameters of the first tree, copula parameters of the second tree, etc. Duplicated parameters in the copula model are omitted.

See Also

[svinecop_scores](#)

Examples

```
# load data set
data(returns)

# convert to uniform margins
u <- pseudo_obs(returns[1:100, 1:3])

# fit parametric S-vine copula model with Markov order 1
fit <- svinecop(u, p = 1, family_set = "parametric")

svinecop_loglik(u, fit)
svinecop_scores(u, fit)
svinecop_hessian(u, fit)
```

svinecop_loglik	<i>Log-likelihood for S-vine copula models</i>
-----------------	--

Description

Log-likelihood for S-vine copula models

Usage

```
svinecop_loglik(u, model, cores = 1)
```

Arguments

u	the data; should have approximately uniform margins.
model	model inheriting from class svinecop_dist .
cores	number of cores to use; if larger than one, computations are performed in parallel on cores batches.

Value

The log-likelihood of the data under the model.

Examples

```
# load data set
data(returns)

# convert to uniform margins
u <- pseudo_obs(returns[1:100, 1:3])

# fit parametric S-vine copula model with Markov order 1
fit <- svinecop(u, p = 1, family_set = "parametric")

svinecop_loglik(u, fit)
svinecop_scores(u, fit)
svinecop_hessian(u, fit)
```

svinecop_pseudo_residuals	<i>Pseudo-residuals of S-vine copula models</i>
---------------------------	---

Description

Pseudo-residuals are defined as the Rosenblatt transform of the data, conditional on the past. Under a correctly specified model, they are approximately *iid* uniform on $[0, 1]^d$.

Usage

```
svinecop_pseudo_residuals(u, model, cores = 1)
```

Arguments

u the data; should have approximately uniform margins.
model model inheriting from class [svinecop_dist](#).
cores number of cores to use; if larger than one, computations are performed in parallel on cores batches.

Value

An n-by-d matrix of pseudo-residuals, where $n = \text{NROW}(u) - \text{model}\p and d is the cross-sectional dimension.

Examples

```
# load data set
data(returns)

# convert to pseudo observations with empirical cdf for marginal distributions
u <- pseudo_obs(returns[1:100, 1:3])

# fit parametric S-vine copula model with Markov order 1
fit <- svinecop(u, p = 1, family_set = "parametric")

# compute pseudo-residuals
# (should be independent uniform across variables and time)
v <- svinecop_pseudo_residuals(u, fit)
pairs(cbind(v[-1, ], v[-nrow(v), ]))
```

svinecop_scores	<i>Log-likelihood scores for S-vine copula models</i>
-----------------	---

Description

Log-likelihood scores for S-vine copula models

Usage

```
svinecop_scores(u, model, cores = 1)
```

Arguments

u the data; should have approximately uniform margins.
model model inheriting from class [svinecop_dist](#).
cores number of cores to use; if larger than one, computations are performed in parallel on cores batches.

Value

An n-by-k matrix containing the score vectors in its rows, where $n = \text{NROW}(u) - \text{model}\p and $k = \text{model}\$npars$. The rows correspond to the effective time points after the first p observations. The columns correspond to model parameters in the order: copula parameters of first tree, copula parameters of second tree, etc. Duplicated parameters in the copula model are omitted.

See Also

[svinecop_hessian](#)

Examples

```
# load data set
data(returns)

# convert to uniform margins
u <- pseudo_obs(returns[1:100, 1:3])

# fit parametric S-vine copula model with Markov order 1
fit <- svinecop(u, p = 1, family_set = "parametric")

svinecop_loglik(u, fit)
svinecop_scores(u, fit)
svinecop_hessian(u, fit)
```

svinecop_sim

Simulate from a S-vine copula model

Description

Simulate from a S-vine copula model

Usage

```
svinecop_sim(n, rep, model, past = NULL, qrng = FALSE, cores = 1)
```

Arguments

n	how many steps of the time series to simulate.
rep	number of replications; rep time series of length n are generated.
model	a S-vine copula model object (inheriting from svinecop_dist).
past	(optional) matrix of past observations. If provided, time series are simulated conditional on the past.
qrng	if TRUE, generates quasi-random numbers using the multivariate Generalized Halton sequence up to dimension 300 and the Generalized Sobol sequence in higher dimensions (default qrng = FALSE).
cores	number of cores to use; if larger than one, computations are performed in parallel over replications.

Value

An n-by-d-by-rep array, where d is the cross-sectional dimension of the model. This reduces to an n-by-d matrix if rep == 1.

Examples

```
# load data set
data(returns)

# convert to uniform margins
u <- pseudo_obs(returns[1:100, 1:3])

# fit parametric S-vine copula model with Markov order 1
fit <- svinecop(u, p = 1, family_set = "parametric")

pairs(u) # original data
pairs(svinecop_sim(100, rep = 1, model = fit)) # simulated data

# simulate the next day conditionally on the past 500 times
pairs(t(svinecop_sim(1, rep = 100, model = fit, past = u)[1, , ]))
```

svine_bootstrap_models

Bootstrap S-vine models

Description

Computes bootstrap replicates of a given model using the one-step block multiplier bootstrap of Nagler et al. (2022).

Usage

```
svine_bootstrap_models(n_models, model)
```

Arguments

n_models	number of bootstrap replicates.
model	the initial fitted model

Value

A list of length n_models, with each entry representing one bootstrapped model as object of class [svine](#).

References

Nagler, T., Krüger, D., and Min, A. (2022). Stationary vine copula models for multivariate time series. *Journal of Econometrics*, 227(2), 305–324. doi:[10.1016/j.jeconom.2021.11.015](https://doi.org/10.1016/j.jeconom.2021.11.015).

Examples

```

data(returns)
dat <- returns[1:100, 1:2]

# fit parametric S-vine model with Markov order 1
model <- svine(dat, p = 1, family_set = "parametric")

# compute 10 bootstrap replicates of the model
boot_models <- svine_bootstrap_models(10, model)

# compute bootstrap replicates of 90%-quantile of X_1 + X_2.
mu_boot <- sapply(
  boot_models,
  function(m) {
    xx <- rowSums(t(svine_sim(1, 10^2, m, past = dat)[1, ,]))
    quantile(xx, 0.9)
  }
)

```

svine_dist

*Custom S-vine distribution models***Description**

Custom S-vine distribution models

Usage

```
svine_dist(margins, copula)
```

Arguments

margins	A list with one marginal distribution per variable. Each element is either a univariateML object or a list with callables \$p (CDF), \$q (quantile function), and for discrete margins \$p_sub (left-limit CDF, i.e. $F(x-)$). Discrete list-form margins must also carry attr(, "type") = "discrete" and class including "svine_margin".
copula	the copula model; an object of class svinecop_dist with cross-sectional dimension d.

Value

Returns the model as an object with class svine_dist. A list with entries

- \$margins: list of marginal models from [univariateML::univariateML_models](#),
- \$copula: an object of svinecop_dist.

See Also

[svine_dist](#), [svine_loglik](#), [svine_sim](#), [svine_bootstrap_models](#)

Examples

```
## marginal objects
# create dummy univariateML models
univ1 <- univ2 <- univariateML::mlnorm(rnorm(10))

# modify the parameters to N(5, 10) and N(0, 2) distributions
univ1[] <- c(5, 10)
univ2[] <- c(0, 2)

## copula object
cs_struct <- cvine_structure(1:2)
pcs <- list(
  list( # first tree
    bicop_dist("clayton", 0, 3), # cross sectional copula
    bicop_dist("gaussian", 0, -0.1) # serial copula
  ),
  list( # second tree
    bicop_dist("gaussian", 0, 0.2), bicop_dist("indep")
  ),
  list( # third tree
    bicop_dist("indep")
  )
)

cop <- svinecop_dist(
  pcs, cs_struct, p = 1, out_vertices = 1:2, in_vertices = 1:2)

model <- svine_dist(margins = list(univ1, univ2), copula = cop)
summary(model)
```

svine_hessian

Expected Hessian of a parametric S-vine model

Description

Expected Hessian of a parametric S-vine model

Usage

```
svine_hessian(x, model, cores = 1)
```

Arguments

x	the data.
model	S-vine model (inheriting from svine_dist).
cores	number of cores to use.

Value

A k-by-k estimate of the expected Hessian, where k is the total number of model parameters. The estimate averages the per-observation Hessian contributions. Parameters are ordered as follows: marginal parameters, copula parameters of first tree, copula parameters of second tree, etc. Duplicated parameters in the copula model are omitted.

Examples

```
data(returns)
dat <- returns[1:100, 1:2]

# fit parametric S-vine model with Markov order 1
model <- svine(dat, p = 1, family_set = "parametric")

# Implementation of asymptotic variances
I <- cov(svine_scores(dat, model))
H <- svine_hessian(dat, model)
Hi <- solve(H)
n_eff <- nrow(dat) - model$copula$p
Hi %*% I %*% t(Hi) / n_eff
```

svine_loglik	<i>Log-likelihood for S-vine models</i>
--------------	---

Description

Log-likelihood for S-vine models

Usage

```
svine_loglik(x, model, cores = 1)
```

Arguments

x	the data.
model	model inheriting from class svine_dist .
cores	number of cores to use; if larger than one, computations are performed in parallel on cores batches.

Value

The log-likelihood of the data under the model.

Examples

```
# load data set
data(returns)

# fit parametric S-vine model with Markov order 1
fit <- svine(returns[1:100, 1:3], p = 1, family_set = "parametric")

svine_loglik(returns[1:100, 1:3], fit)
```

svine_pseudo_residuals

Pseudo-residuals of S-vine models

Description

Pseudo-residuals are defined as the Rosenblatt transform of the data, conditional on the past. Under a correctly specified model, they are approximately *iid* uniform on $[0, 1]^d$.

Usage

```
svine_pseudo_residuals(x, model, cores = 1)
```

Arguments

x	the data.
model	model inheriting from class svine_dist .
cores	number of cores to use; if larger than one, computations are performed in parallel on cores batches.

Value

An n-by-d matrix of pseudo-residuals, where $n = \text{NROW}(x) - \text{model\$copula\$p}$ and d is the cross-sectional dimension.

Examples

```
# load data set
data(returns)

dat <- returns[1:100, 1:3]

# fit parametric S-vine model with Markov order 1
fit <- svine(dat, p = 1, family_set = "parametric")

# compute pseudo-residuals
# (should be independent uniform across variables and time)
v <- svine_pseudo_residuals(dat, fit)
pairs(cbind(v[-1, ], v[-nrow(v), ]))
```

svine_scores	<i>Score function of parametric S-vine models</i>
--------------	---

Description

Score function of parametric S-vine models

Usage

```
svine_scores(x, model, cores = 1)
```

Arguments

x	the data.
model	S-vine model (inheriting from svine_dist).
cores	number of cores to use.

Value

An n-by-k matrix, where $n = \text{NROW}(x) - \text{model}\$copula\$p$ and k is the total number of model parameters. The rows correspond to the effective time points after the first p observations. Parameters are ordered as follows: marginal parameters, copula parameters of first tree, copula parameters of second tree, etc. Duplicated parameters in the copula model are omitted.

Examples

```
data(returns)
dat <- returns[1:100, 1:2]

# fit parametric S-vine model with Markov order 1
model <- svine(dat, p = 1, family_set = "parametric")

# Implementation of asymptotic variances
I <- cov(svine_scores(dat, model))
H <- svine_hessian(dat, model)
Hi <- solve(H)
n_eff <- nrow(dat) - model$copula$p
Hi %*% I %*% t(Hi) / n_eff
```

svine_sim	<i>Simulate from a S-vine model</i>
-----------	-------------------------------------

Description

Simulate from a S-vine model

Usage

```
svine_sim(n, rep, model, past = NULL, qrng = FALSE, cores = 1)
```

Arguments

n	how many steps of the time series to simulate.
rep	number of replications; rep time series of length n are generated.
model	a S-vine copula model object (inheriting from svinecop_dist).
past	(optional) matrix of past observations. If provided, time series are simulated conditional on the past.
qrng	if TRUE, generates quasi-random numbers using the multivariate Generalized Halton sequence up to dimension 300 and the Generalized Sobol sequence in higher dimensions (default qrng = FALSE).
cores	number of cores to use; if larger than one, computations are performed in parallel over replications.

Value

An n-by-d-byrep array, where d is the cross-sectional dimension of the model. This reduces to an n-by-d matrix if rep == 1.

Examples

```
# load data set
data(returns)
returns <- returns[1:100, 1:3]

# fit parametric S-vine model with Markov order 1
fit <- svine(returns, p = 1, family_set = "parametric")

pairs(returns) # original data
pairs(svine_sim(100, rep = 1, model = fit)) # simulated data

# simulate the next day conditionally on the past 500 times
pairs(t(svine_sim(1, rep = 100, model = fit, past = returns)[1, , ]))
```

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