

Package ‘jafar’

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bsfp.predict.oos	<i>Out-of-sample prediction for BSFP</i>
------------------	--

Description

Modified version of the function `bsfp.predict` from the GitHub repo BSFP for out-of-sample predictions.

Usage

```
bsfp.predict.oos(
  bsfp.fit,
  test_data,
  response_type = "continuous",
  model_params = NULL,
  nsample,
  progress = TRUE,
  starting_values = NULL
)
```

Arguments

<code>bsfp.fit</code>	Results from fitting <code>bsfp</code> on training data.
<code>test_data</code>	Matrix-list dataset of held-out test data.
<code>response_type</code>	Continuous or binary response. Must be one of 'continuous' (default) or 'binary'.
<code>model_params</code>	May be NULL if <code>model_params=NULL</code> in <code>bsfp</code> fit. Otherwise, specify as (<code>error_vars</code> , <code>joint_vars</code> , <code>indiv_vars</code> , <code>beta_vars</code> , <code>response_vars</code>).
<code>nsample</code>	Integer specifying number of Gibbs sampling iterations
<code>progress</code>	Boolean determining if progress of the sampler be displayed
<code>starting_values</code>	List of starting values for \mathbf{V} , \mathbf{U}_s , \mathbf{W}_s , \mathbf{V}_s for $s = 1, \dots, q$. If NULL, initialize from prior.

Details

Generate new scores for held-out test data based on a training fit of BSFP. Uses the estimated ranks and joint and individual loadings. Cannot be used if missing values are present in test data.

Value

Returns a list with the following parameters:

test_data	Test data provided by user
EY.draw	List of posterior samples for the $E(Y X)$, i.e. $\beta_0 + \mathbf{V}\beta_{joint} + \sum_{s=1}^q \mathbf{V}_s\beta_s$ for each Gibbs sampling iteration.
V.draw	List of posterior samples for joint scores, \mathbf{V}
U.train	List of posterior samples for joint loadings for each source, \mathbf{U}_s for $s = 1, \dots, q$ given by the training BSFP fit
W.train	List of posterior samples for individual loadings for each source, \mathbf{W}_s for $s = 1, \dots, q$ given by the training BSFP fit
Vs.draw	List of posterior samples for individual scores for each source, \mathbf{V}_s for $s = 1, \dots, q$
ranks	Vector with the estimated joint and individual ranks. ranks[1] is the estimated joint rank. ranks[2:(q+1)] correspond to the individual ranks for each source.
tau2.train	List of posterior samples for the response variance if the response was continuous given by training BSFP fit
beta.train	List of posterior samples for the regression coefficients used in the predictive model given by training BSFP fit
Xm.draw	List of posterior samples for missing predictors imputations

features_reorder_HC	<i>predictors preprocess: reorder features via hierarchical clustering for better visualization</i>
---------------------	---

Description

predictors preprocess: reorder features via hierarchical clustering for better visualization

Usage

```
features_reorder_HC(X_m, X_m_test = NULL, K0_HC = 15)
```

Arguments

X_m	Train set predictors
X_m_test	Test set predictors
K0_HC	Reference number of clusters for hierarchical clustering (default: 15)

Value

List of preprocessed features and rescaling factors

Description

Fits a Joint Additive FActor Regression (JAFAR) model using Gibbs sampling. Variation across multiple data-views is explained via shared and study-specific latent factors. Default and optional outputs include posterior means of the induced covariances, posterior samples of residual variances, latent factors, and factor loadings. Supports parallel computation and tempered updates to limit rank estimation in extreme large-p-small-n settings.

Usage

```
gibbs_jafar(
  X_m,
  y = NULL,
  yBinary = F,
  K0 = NULL,
  K0_m = NULL,
  tMCMC = 20000,
  tBurnIn = 15000,
  tThin = 10,
  hyperparams = list(),
  get_latent_vars = TRUE,
  get_last_sample = FALSE,
  parallel = TRUE,
  tempered = FALSE,
  rescale_pred = FALSE
)
```

Arguments

<code>X_m</code>	Multi-view input data. Rows should correspond to samples, columns to features. (list of length M ; m -th element: matrix $n \times p_{mm}$).
<code>y</code>	Vector responses (of length n). Set to <code>NULL</code> for unsupervised mode (default: <code>NULL</code>).
<code>yBinary</code>	Logical, indicating if the response(s) are binary (default: <code>FALSE</code>).
<code>K0</code>	Upper bound to numbers of shared latent factors (optional) If <code>NULL</code> , <code>K0</code> is set to $3 \cdot \log(\max(p_m))$
<code>K0_m</code>	Upper bounds to numbers of view-specific latent factors (optional) Length should equal <code>length(X_m)</code> . If <code>NULL</code> , <code>K0[m]</code> is set to $3 \cdot \log(\max(p_m[m]))$
<code>tMCMC</code>	Total number of MCMC iterations (default: 20000).
<code>tBurnIn</code>	Number of burn-in iterations (default: 15000).
<code>tThin</code>	Thinning interval for saving samples (default: 10).
<code>hyperparams</code>	List of hyperparameters for the D-CUSP prior distributions. Missing hyperparameters are replaced by defaults encoded in jafar_set_hyperparameters .
<code>get_latent_vars</code>	Logical, whether to return latent factors and loading matrices (default: <code>TRUE</code>).

<code>get_last_sample</code>	Logical, whether to return only the last sample of the MCMC chain (default: FALSE).
<code>parallel</code>	Logical, whether to use parallel computation for the loadings' update (default: TRUE).
<code>tempered</code>	Logical, temperature parameter for tempered sampling (default: FALSE, no tempering).
<code>rescale_pred</code>	Logical, whether to rescale loadings when computing response predictions (default: FALSE).

Details

The number of samples in output is $t_{\text{Eff}} = (t_{\text{MCMC}} - t_{\text{BurnIn}}) \% t_{\text{Thin}}$. The output list contains:

- `KNumber` shared latent factors (vector of length t_{Eff}).
- `K_GmNumber` view-specific latent factors (matrix $t_{\text{Eff}} \times M$).
- `K_Lm_eff` Numbers of shared factors active in each view (matrix $t_{\text{Eff}} \times M$).
- `K_Gm_eff` Numbers of specific factors active in each view (matrix $t_{\text{Eff}} \times M$).
- `active_Lm` Binary indicators of shared factors activity across views (binary array $t_{\text{Eff}} \times K \times M$ s).
- `Cov_m_mean` Posterior mean of the covariance matrix for each dataset (list of length M ; m-th element: matrix $p_m[m] \times p_m[m]$).
- `Marg_Var_m` Marginal variances of features (list of length M ; m-th element: matrix $t_{\text{Eff}} \times p_m[m]$).
- `s2_inv_m` Inverse residual variances across views (list of length M ; m-th element: matrix $t_{\text{Eff}} \times p_m[m]$).
- `mu_m` Features intercepts across views (list of length M ; m-th element: matrix $t_{\text{Eff}} \times p_m[m]$).
- `hyper_param` List of hyperparameters used for the model, including user-specified values and defaults ones were missing.

If `is_supervised = TRUE`:

- `K_T_eff` Numbers of shared factors active in the response (vector of length t_{Eff}).
- `K_Tm_eff` Numbers of specific factors active in the response (matrix $t_{\text{Eff}} \times M$).
- `active_TB` Binary indicators of shared factors activity in the response (binary matrix $t_{\text{Eff}} \times K$).
- `active_Tm` Binary indicators of specific factors activity in the response (list of length M ; m-th element: matrix $t_{\text{Eff}} \times K_Gm[m]$).
- `s2_inv` Response inverse residual variances (vector of length t_{Eff}).
- `mu_y` Response intercept (vector of length t_{Eff}).
- `ThetaResponse` loadings on shared factors (matrix $t_{\text{Eff}} \times K$).
- `Theta_m` Response loadings on specific factors (list of length M ; m-th element: matrix $t_{\text{Eff}} \times K_Gm[m]$).
- `y_M` Latent probit utilities (matrix $t_{\text{Eff}} \times n$). (only if `yBinary = TRUE`).

If `get_latent_vars = TRUE`:

- `Lambda_m` Loadings matrices on shared factors (list of length M ; m-th element: array $t_{\text{Eff}} \times p_m[m] \times K$).

- `Gamma_mLoadings` matrices on view-specific factors (list of length `M`; `m`-th element: array $t_{\text{Eff}} \times p_m[m] \times K_{\text{Gm}}[m]$).
- `etaShared` latent factors (array $t_{\text{Eff}} \times n \times K$).
- `phi_mView`-specific latent factors (list of length `M`; `m`-th element: array $t_{\text{Eff}} \times n \times K_{\text{Gm}}[m]$).

If the input matrices `X_m` contain missing values:

- `Xm_MCPPosterior` samples of imputed values for missing entries. A list of length `M`; the `m`-th element is itself a list (one per feature with missingness), each containing an $t_{\text{Eff}} \times n_{\text{miss}}$ matrix of imputed values across MCMC iterations.
- `na_idxList` of length `M`; the `m`-th element gives the column indices of missing entries in `X_m[[m]]`.
- `na_row_idxList` of length `M`; the `m`-th element gives the corresponding row indices of missing entries in `X_m[[m]]`.

If `get_last_sample = TRUE`:

- `last_sampleList` of posterior values of all parameters at the last MCMC iteration, including latent factors, loadings, residual variances, and hyperparameters.

Value

A list containing posterior samples, latent variables (if requested), and other relevant model outputs.

Note

- Ensure that all matrices in `X_m` have the same number of rows (subjects).
- Missing data in `X_m` are allowed as NA and imputed in the MCMC.

gibbs_jfr

Gibbs Sampler for JFR

Description

Fits a Joint Factor Regression (JFR) model using Gibbs sampling. The model can be fitted in both unsupervised (no response) and supervised (with response `y`) settings. Default and optional outputs include posterior means of the induced covariances, posterior samples of residual variances, latent factors, and factor loadings. Supports parallel computation and tempered updates to limit rank estimation in extreme large- p -small- n settings.

Usage

```
gibbs_jfr(
  X_m,
  y = NULL,
  yBinary = F,
  K0 = NULL,
  tMCMC = 20000,
  tBurnIn = 15000,
  tThin = 10,
  hyperparams = list(),
```

```

    get_latent_vars = TRUE,
    get_last_sample = FALSE,
    parallel = TRUE,
    tempered = FALSE,
    rescale_pred = FALSE
  )

```

Arguments

<code>X_m</code>	Multi-view input data. Rows should correspond to samples, columns to features. (list of length M ; m -th element: matrix $n \times p_m$).
<code>y</code>	Vector responses (of length n). Set to NULL for unsupervised mode (default: NULL).
<code>yBinary</code>	Logical, indicating if the response(s) are binary (default: FALSE).
<code>K0</code>	Upper bound to numbers of latent factors (optional) If NULL, $K0$ is set to $3 \cdot \log(\max(p_m))$
<code>tMCMC</code>	Total number of MCMC iterations (default: 20000).
<code>tBurnIn</code>	Number of burn-in iterations (default: 15000).
<code>tThin</code>	Thinning interval for saving samples (default: 10).
<code>hyperparams</code>	List of hyperparameters for the D-CUSP prior distributions. Missing hyperparameters are replaced by defaults encoded in jafar_set_hyperparameters .
<code>get_latent_vars</code>	Logical, whether to return latent factors and loading matrices (default: TRUE).
<code>get_last_sample</code>	Logical, whether to return only the last sample of the MCMC chain (default: FALSE).
<code>parallel</code>	Logical, whether to use parallel computation for the loadings' update (default: TRUE).
<code>tempered</code>	Logical, temperature parameter for tempered sampling (default: FALSE, no tempering).
<code>rescale_pred</code>	Logical, whether to rescale loadings when computing response predictions (default: FALSE).

Details

The number of samples in output is $tEff = (tMCMC - tBurnIn) \% \% tThin$. The output list contains:

- `KNumber` latent factors (vector of length $tEff$).
- `K_Lm_eff` Numbers of latent factors active in each view (matrix $tEff \times M$).
- `active_Lm` Binary indicators of latent factors activity across views (binary array $tEff \times K \times M$).
- `Cov_m_mean` Posterior mean of the covariance matrix for each dataset (list of length M ; m -th element: matrix $p_m[m] \times p_m[m]$).
- `Marg_Var_m` Marginal variances of features (list of length M ; m -th element: matrix $tEff \times p_m[m]$).
- `s2_inv_m` Inverse residual variances across views (list of length M ; m -th element: matrix $tEff \times p_m[m]$).
- `mu_m` Features intercepts across views (list of length M ; m -th element: matrix $tEff \times p_m[m]$).

- `hyper_paramList` of hyperparameters used for the model, including user-specified values and defaults ones were missing.

If `is_supervised = TRUE`:

- `K_T_eff` Numbers of latent factors active in the response (vector of length `tEff`).
- `active_TB` Binary indicators of latent factors activity in the response (binary matrix `tEff x K`).
- `s2_invResponse` inverse residual variances (vector of length `tEff`).
- `mu_yResponse` intercept (vector of length `tEff`).
- `ThetaResponse` loadings on latent factors (matrix `tEff x K`).
- `y_MCLatent` probit utilities (matrix `tEff x n`). (only if `yBinary = TRUE`).

If `get_latent_vars = TRUE`:

- `Lambda_mLoadings` matrices on latent factors (list of length `M`; m-th element: array `tEff x p_m[m] x K`).
- `etaLatent` factors (array `tEff x n x K`).

If the input matrices `X_m` contain missing values:

- `Xm_MCP` Posterior samples of imputed values for missing entries. A list of length `M`; the m-th element is itself a list (one per feature with missingness), each containing an `tEff x n_miss` matrix of imputed values across MCMC iterations.
- `na_idxList` of length `M`; the m-th element gives the column indices of missing entries in `X_m[[m]]`.
- `na_row_idxList` of length `M`; the m-th element gives the corresponding row indices of missing entries in `X_m[[m]]`.

If `get_last_sample = TRUE`:

- `last_sampleList` of posterior values of all parameters at the last MCMC iteration, including latent factors, loadings, residual variances, and hyperparameters.

Value

A list containing posterior samples, latent variables (if requested), and other relevant model outputs.

Note

- Ensure that all matrices in `X_m` have the same number of rows (subjects).
- Missing data in `X_m` are allowed as NA and imputed in the MCMC.

jafar_set_hyperparameters

Set Hyperparameters for JAFAR and JFR models

Description

Helper function to set hyperparameters for [gibbs_jfr](#) and [gibbs_jafar](#). Missing hyperparameters are assigned default values. Supports both unsupervised and supervised (response-guided) settings.

Usage

```
jafar_set_hyperparameters(hyperparams_list, M, is_supervised = FALSE)
```

Arguments

`hyperparams_list` A named list of hyperparameters to be used in the model.

`M` Integer, number of data-views.

`is_supervised` Logical, whether the model is supervised (default: FALSE).

Details

Default hyperparameters include:

- `seed`: random seed for reproducibility (default: 123).
- `t0`, `t1`, `t0_adapt`: adaptation parameters for MCMC (default: `t0=-1`, `t1=-5e-4`, `t0_adapt=200`).
- `a_m`, `b_m`: shape and rate of inverse-gamma prior for idiosyncratic noise in each view. Scalars of vectors of length `M` (default: `a_m[m]=3`, `b_m[m]=1`).
- `prec0m`: precision of normal prior on intercepts. Scalar of vector of length `M` (default: `prec0m[m]=2`).
- `var_spike`: variance of normal spike in cusps. Scalar of vector of length `M` (default: `var_spike[m]=0.005`).
- `a_chi`, `b_chi`: hyperparameters for slab inverse-gamma prior in cusps. Scalars of vectors of length `M` (default: `a_chi[m]=0.5`, `b_chi[m]=0.1`).
- `alpha_L`, `alpha_G`: DP concentration parameters giving the expected number of factors, shared and local. Scalars of vectors of length `M` (default: `alpha_L[m]=5`, `alpha_G[m]=5`).

If `is_supervised = TRUE`, additional hyperparameters for the response model are

- `a_sig`, `b_sig`: shape and rate of inverse-gamma prior for idiosyncratic noise (default: `a_sig=3`, `b_sig=1`).
- `prec0`: precision of normal prior on intercept (default: `prec0=2`).
- `var_spike_y`: variance of normal spike (default: `var_spike_y=0.005`).
- `a_theta`, `b_theta`: hyperparameters for slab inverse-gamma prior in the slab (default: `a_theta=0.5`, `b_theta=0.1`).
- `a_xi`, `b_xi`: shape parameters for beta prior on mixture weight in response loadings (default: `a_xi=3`, `b_xi=2`).

Value

A named list of hyperparameters with defaults filled in where missing. Scalar values are replicated `M` times where necessary.

multiviewMatchAlign	<i>Perform rotational alignment using multi-view MatchAlign.</i>
---------------------	--

Description

Perform rotational alignment using multi-view MatchAlign.

Usage

```
multiviewMatchAlign(ris_MCMC)
```

Arguments

risMCMC	Posterior samples, as returned by gibbs_jafar or gibbs_jfr.
---------	---

Value

A modified version of the input risMCMC, with latent factors, loading matrices, and (if supervised) response loadings rotated according to multi-view MatchAlign.

plot_coefficients	<i>Plot induced regression coefficients for $y X=x$</i>
-------------------	--

Description

Plot induced regression coefficients for $y|X=x$

Usage

```
plot_coefficients(yPred, out_path = "~/Desktop/", out_name = "coefficients")
```

Arguments

yPred	Response predictions, output of predict_y or predict_y_raw
out_path	Output path where the generated plot will be saved (default: "~/Desktop/")
out_name	Output file name (default: "coefficients")

plot_correlations	<i>Plot the empirical and inferred within-view correlation matrices</i>
-------------------	---

Description

Plot the empirical and inferred within-view correlation matrices

Usage

```
plot_correlations(
  risMCMC,
  X_m = NULL,
  out_path = "~/Desktop/",
  out_name = "cor_matrices"
)
```

Arguments

risMCMC	Posterior samples, output of gibbs_jafar or gibbs_jfr
X_m	Training set multi-view predictors (optional, default: NULL). If NULL, only inferred correlation matrices are visualized. If not NULL, the empirical correlation matrices are displayed besides the inferred ones
out_path	Output path where the generated plot will be saved (default: "~/Desktop/")
out_name	Output file name (default: "cor_matrices")

plot_loadings	<i>Plot posterior means of factor loadings.</i>
---------------	---

Description

Rotational alignment must be performed in advanced through the function multiviewMatchAlign

Usage

```
plot_loadings(
  risMCMC,
  out_path = "~/Desktop/",
  out_name_shared = "shared_loadings",
  out_name_specific = "specific_loadings"
)
```

Arguments

risMCMC	Posterior samples, output of gibbs_jafar or gibbs_jfr
out_path	Output path where the generated plot will be saved
out_name_shared	Output file name for the shared component plot (default: "n_factors_shared")
out_name_specific	Output file name for the specific components plot (default: "n_factor_specific")

plot_n_factors	<i>Plot MCMC samples of the inferred number of factors</i>
----------------	--

Description

Plot MCMC samples of the inferred number of factors

Usage

```
plot_n_factors(
  risMCMC,
  out_path = "~/Desktop",
  out_name_shared = "n_factors_shared",
  out_name_specific = "n_factor_specific"
)
```

Arguments

risMCMC	Posterior samples, output of gibbs_jafar or gibbs_jfr
out_path	Output path where the generated plot will be saved
out_name_shared	Output file name for the shared component plot (default: "n_factors_shared")
out_name_specific	Output file name for the specific components plot (default: "n_factor_specific")

plot_predictions	<i>Plot response predictions against true values</i>
------------------	--

Description

Plot response predictions against true values

Usage

```
plot_predictions(
  yPred,
  yTrue,
  risMCMC,
  out_path = "~/Desktop/",
  out_name = "predictions"
)
```

Arguments

yPred	Response predictions, output of predict_y or predict_y_raw
yTrue	True values of the responses
risMCMC	Posterior samples, output of gibbs_jafar or gibbs_jfr
out_path	Output path where the generated plot will be saved (default: "~/Desktop/")
out_name	Output file name (default: "predictions")

predict_y	<i>Response predictions and induced regression coefficients for JAFAR and JFR</i>
-----------	---

Description

Response predictions and induced regression coefficients for JAFAR and JFR

Usage

```
predict_y(Xpred, risMCMC, rescale_pred = FALSE)
```

Arguments

Xpred	A list of M features' matrices, the m-th of dimension nPred x p_m[m] or possibly with missing (X_m[[m]]=NULL)).
risMCMC	Output of gibbs_jafar or gibbs_jfr containing posterior samples.
rescale_pred	Logical, whether to rescale loadings when computing response predictions (default: FALSE).

Value

A list containing posterior samples of the predicted responses (matrix tEff x nPred), and of the induced regression coefficients for each view (list of length M; m-th element: tEff x p_m[m]).

predict_y_raw	<i>Response predictions for JAFAR and JFR</i>
---------------	---

Description

Response predictions for JAFAR and JFR

Usage

```
predict_y_raw(Xpred, risMCMC, rescale_pred = FALSE)
```

Arguments

Xpred	A list of M features' matrices, the m-th of dimension nPred x p_m[m] or possibly with missing (X_m[[m]]=NULL)).
risMCMC	Output of gibbs_jafar or gibbs_jfr containing posterior samples.
rescale_pred	Logical, whether to rescale loadings when computing response predictions (default: FALSE).

Value

A list containing posterior samples of the predicted responses (matrix tEff x nPred).

preprocess_X	<i>predictors preprocess: center & rescale + cdf transform (optional)</i>
--------------	---

Description

predictors preprocess: center & rescale + cdf transform (optional)

Usage

```
preprocess_X(X_m, X_m_test = NULL, copula = F)
```

Arguments

X_m	Train set predictors
X_m_test	Test set predictors
copula	Apply cdf transformation

Value

List of preprocessed features and rescaling factors

preprocess_y	<i>response preprocess: center & rescale</i>
--------------	--

Description

response preprocess: center & rescale

Usage

```
preprocess_y(yTrain, yTest = NULL)
```

Arguments

yTrain	Train set responses
yTest	Test set responses

Value

List of preprocessed responses and rescaling factors

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