

# Package ‘mfbvar’

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**Type** Package

**Title** Mixed-Frequency Bayesian VAR Models

**Version** 0.5.1

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**Description** Estimation of mixed-frequency Bayesian vector autoregressive (VAR) models. The package implements a state space-based VAR model that handles mixed frequencies of the data. The model is estimated using Markov Chain Monte Carlo to numerically approximate the posterior distribution. Prior distributions that can be used include normal-inverse Wishart and normal-diffuse priors as well as steady-state priors. Stochastic volatility can be handled by common or factor stochastic volatility models.

**License** GPL-3

**LazyData** TRUE

**URL** <https://github.com/ankargren/mfbvar>

**BugReports** <https://github.com/ankargren/mfbvar/issues>

**Imports** Rcpp (>= 0.12.7), ggplot2 (>= 2.2.1), methods, lubridate, GIGrv, stochvol (>= 2.0.3), RcppParallel, dplyr, magrittr, tibble

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estimate_mfbvar	<i>Mixed-frequency Bayesian VAR</i>
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### Description

The main function for estimating a mixed-frequency BVAR.

### Usage

```
estimate_mfbvar(mfbvar_prior = NULL, prior, variance = "iw", ...)
```

### Arguments

mfbvar_prior	a mfbvar_prior object
prior	either "ss" (steady-state prior) or "minn" (Minnesota prior)
variance	form of the error variance-covariance matrix: "iw" for the inverse Wishart prior, "diffuse" for a diffuse prior, "csv" for common stochastic volatility or "fsv" for factor stochastic volatility
...	additional arguments to update_prior (if mfbvar_prior is NULL, the arguments are passed on to set_prior)

**Value**

An object of class `mfbvar`, `mfbvar_<prior>` and `mfbvar_<prior>_<variance>` containing posterior quantities as well as the prior object. For all choices of prior and variance, the returned object contains:

`Pi` Array of dynamic coefficient matrices; `Pi[, , r]` is the  $r$ th draw  
`Z` Array of monthly processes; `Z[, , r]` is the  $r$ th draw  
`Z_fcst` Array of monthly forecasts; `Z_fcst[, , r]` is the  $r$ th forecast. The first `n_lags` rows are taken from the data to offer a bridge between observations and forecasts and for computing nowcasts (i.e. with ragged edges).

**Steady-state priors:** If `prior = "ss"`, it also includes:

`psi` Matrix of steady-state parameter vectors; `psi[r, ]` is the  $r$ th draw  
`roots` The maximum eigenvalue of the lag polynomial (if `check_roots = TRUE`)

If `prior = "ssng"`, it also includes:

`psi` Matrix of steady-state parameter vectors; `psi[r, ]` is the  $r$ th draw  
`roots` The maximum eigenvalue of the lag polynomial (if `check_roots = TRUE`)  
`lambda_psi` Vector of draws of the global hyperparameter in the normal-Gamma prior  
`phi_psi` Vector of draws of the auxiliary hyperparameter in the normal-Gamma prior  
`omega_psi` Matrix of draws of the prior variances of `psi`; `omega_psi[r, ]` is the  $r$ th draw, where `diag(omega_psi[r, ])` is used as the prior covariance matrix for `psi`

**Constant error covariances:** If `variance = "iw"` or `variance = "diffuse"`, it also includes:

`Sigma` Array of error covariance matrices; `Sigma[, , r]` is the  $r$ th draw

**Time-varying error covariances:** If `variance = "csv"`, it also includes:

`Sigma` Array of error covariance matrices; `Sigma[, , r]` is the  $r$ th draw  
`phi` Vector of AR(1) parameters for the log-volatility regression; `phi[r]` is the  $r$ th draw  
`sigma` Vector of error standard deviations for the log-volatility regression; `sigma[r]` is the  $r$ th draw  
`f` Matrix of log-volatilities; `f[r, ]` is the  $r$ th draw

If `variance = "fsv"`, it also includes:

`facload` Array of factor loadings; `facload[, , r]` is the  $r$ th draw  
`latent` Array of latent log-volatilities; `latent[, , r]` is the  $r$ th draw  
`mu` Matrix of means of the log-volatilities; `mu[, r]` is the  $r$ th draw  
`phi` Matrix of AR(1) parameters for the log-volatilities; `phi[, r]` is the  $r$ th draw  
`sigma` Matrix of innovation variances for the log-volatilities; `sigma[, r]` is the  $r$ th draw

**References**

- Ankargren, S., Unosson, M., & Yang, Y. (2019) A Flexible Mixed-Frequency Bayesian Vector Autoregression with a Steady-State Prior. Manuscript.  
 Ankargren, S., & Jonéus, P. (2019) Simulation Smoothing for Nowcasting with Large Mixed-Frequency VARs. arXiv:1907.01075, <http://arxiv.org/abs/1907.01075>.

Ankargren, S., & Jonéus, P. (2019) Estimating Large Mixed-Frequency Bayesian VAR Models. Manuscript.  
Kastner, G., & Huber, F. (2018) Sparse Bayesian Vector Autoregressions in Huge Dimensions. arXiv:1704.03239, <http://arxiv.org/abs/1704.03239>.  
Schorfheide, F., & Song, D. (2015) Real-Time Forecasting With a Mixed-Frequency VAR. *Journal of Business & Economic Statistics*, 33(3), 366–380. <http://dx.doi.org/10.1080/07350015.2014.954707>

### See Also

[set\\_prior](#), [update\\_prior](#), [predict.mfbvar](#), [plot.mfbvar\\_minn](#), [plot.mfbvar\\_ss](#), [varplot](#), [summary.mfbvar](#)

### Examples

```
prior_obj <- set_prior(Y = mf_sweden, freq = c(rep("m", 4), "q"),
                      n_lags = 4, n_burnin = 20, n_reps = 20)
mod_minn <- estimate_mfbvar(prior_obj, prior = "minn")
```

---

mcmc\_sampler

*MCMC sampler*

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### Description

mcmc\_sampler is a generic function for deciding which specific MCMC algorithm to dispatch to. It is called internally.

### Usage

```
mcmc_sampler(x, ...)
```

### Arguments

x                    argument to dispatch on (of class prior\_obj)  
...                   additional named arguments passed on to the methods

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mdd	<i>Marginal data density estimation</i>
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**Description**

mdd estimates the (log) marginal data density.

**Usage**

```
mdd(x, ...)
```

**Arguments**

x	argument to dispatch on (of class <code>mfvar_ss</code> or <code>mfvar_minn</code> )
...	additional named arguments passed on to the methods

**Details**

This is a generic function. See the methods for more information.

**See Also**

[mdd.mfvar\\_ss\\_iw](#), [mdd.mfvar\\_minn\\_iw](#)

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<code>mdd.mfvar_minn_iw</code>	<i>Marginal data density method for class <code>mfvar_minn</code></i>
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---

**Description**

Estimate the marginal data density for the model with a Minnesota prior.

**Usage**

```
## S3 method for class 'mfvar_minn_iw'
mdd(x, ...)
```

**Arguments**

x	object of class <code>mfvar_minn</code>
...	additional arguments (currently only <code>p_trunc</code> for the degree of truncation is available)

**Details**

The method used for estimating the marginal data density is the proposal made by Schorfheide and Song (2015).

**Value**

The logarithm of the marginal data density.

**References**

Schorfheide, F., & Song, D. (2015) Real-Time Forecasting With a Mixed-Frequency VAR. *Journal of Business & Economic Statistics*, 33(3), 366–380. <http://dx.doi.org/10.1080/07350015.2014.954707>

**See Also**

[mdd](#), [mdd.mfbvar\\_ss\\_iw](#)

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mdd.mfbvar_ss_iw	<i>Marginal data density method for class mfbvar_ss</i>
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**Description**

Estimate the marginal data density for the model with a steady-state prior.

**Usage**

```
## S3 method for class 'mfbvar_ss_iw'
mdd(x, method = 1, ...)
```

**Arguments**

x	object of class mfbvar_ss
method	option for which method to choose for computing the mdd (1 or 2)
...	additional arguments (currently only p_trunc for the degree of truncation for method 2 is available)

**Details**

Two methods for estimating the marginal data density are implemented. Method 1 and 2 correspond to the two methods proposed by Fuentes-Albero and Melosi (2013) and Ankargren, Unosson and Yang (2018).

**Value**

The logarithm of the marginal data density.

## References

- Fuentes-Albero, C. and Melosi, L. (2013) Methods for Computing Marginal Data Densities from the Gibbs Output. *Journal of Econometrics*, 175(2), 132-141, <https://doi.org/10.1016/j.jeconom.2013.03.002>
- Ankargren, S., Unosson, M., & Yang, Y. (2018) A Mixed-Frequency Bayesian Vector Autoregression with a Steady-State Prior. Working Paper, Department of Statistics, Uppsala University No. 2018:3.

## See Also

[mdd](#), [mdd.mfbvar\\_minn\\_iw](#)

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mfbvar	<i>mfbvar: A package for mixed-frequency Bayesian vector autoregressive (VAR) models.</i>
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## Description

The mfbvar package makes estimation of Bayesian VARs with a mix of monthly and quarterly data simple. The prior for the regression parameters is normal with Minnesota-style prior moments. The package supports either an inverse Wishart prior for the error covariance matrix, yielding a standard normal-inverse Wishart prior, or a time-varying error covariance matrix by means of a factor stochastic volatility model through the [factorstochvol-package](#) package.

## Specifying the prior

The prior of the VAR model is specified using the function [set\\_prior](#). The function creates a prior object, which can be further updated using [update\\_prior](#). The model can be estimated using the steady-state prior, which requires the prior moments of the steady-state parameters. The function [interval\\_to\\_moments](#) is a helper function for obtaining these from prior intervals.

## Estimating the model

The model is estimated using the function [estimate\\_mfbvar](#). The error covariance matrix is given an inverse Wishart prior or modeled using factor stochastic volatility. If the former is used, [mdd](#) can be used to estimate to the marginal data density (marginal likelihood).

## Processing the output

Plots of the output can be obtained from calling the generic function [plot](#) (see [plot-mfbvar](#)). If factor stochastic volatility is used, the time-varying standard deviations can be plotted using [varplot](#). Predictions can be obtained from [predict.mfbvar](#).

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mf_sweden	<i>Real-time data set.</i>
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### Description

A dataset containing real-time data for mixed and quarterly frequencies.

### Usage

```
mf_sweden
```

### Format

A mixed-frequency data set of five Swedish macroeconomic variables.

**unemp** harmonized unemployment rate (source: OECD)

**infl** inflation rate (source: OECD)

**ip** industrial production (source: OECD)

**eti** economic tendency indicator (source: National Institute of Economic Research)

**gdp** GDP growth (source: Statistics Sweden)

### References

OECD (2016) MEI Archive: Revisions Analysis Dataset.

Billstam, M., Frändén, J., Samuelsson, J., Osterholm, P. (2016) Quasi-Real-Time Data of the Economic Tendency Survey. Working Paper No. 143, National Institute of Economic Research.

Statistics Sweden (2016) Revisions, expenditure approach and hours worked at each release.

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plot-mfbvar	<i>Plotting methods for posterior mfbvar objects</i>
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---

### Description

Methods for plotting posterior mfbvar objects (mfbvar\_minn and mfbvar\_ss).

### Usage

```
## S3 method for class 'mfbvar_ss'
plot(x, fcst_start = NULL, aggregate_fcst = TRUE,
     plot_start = NULL, pred_bands = 0.8, nrow_facet = NULL,
     ss_bands = 0.95, ...)
```

```
## S3 method for class 'mfbvar_ssng'
plot(x, fcst_start = NULL, aggregate_fcst = TRUE,
     plot_start = NULL, pred_bands = 0.8, nrow_facet = NULL,
```





```
mod_ss <- estimate_mfbvar(prior_obj, prior = "ss", variance = "fsv")
plot(mod_ss)
varplot(mod_ss)
```

---

plot.mfbvar\_prior      *Plot method for class mfbvar\_prior*

---

### Description

Method for plotting mfbvar\_prior objects.

### Usage

```
## S3 method for class 'mfbvar_prior'
plot(x, nrow_facet = NULL, ...)
```

### Arguments

x	object of class mfbvar_prior
nrow_facet	number of rows in facet
...	Currently not in use.

### Details

The function plots the data. If the prior moments for the steady-state parameters are available in x, these are included.

### Examples

```
prior_obj <- set_prior(Y = mf_sweden[, 4:5], freq = c("m", "q"),
                      n_lags = 4, n_burnin = 20, n_reps = 20, n_fcst = 4)
plot(prior_obj)
```

---

predict.mfbvar      *Predict method for class mfbvar*

---

### Description

Method for predicting mfbvar objects.

### Usage

```
## S3 method for class 'mfbvar'
predict(object, fcst_start = NULL,
        aggregate_fcst = TRUE, pred_bands = 0.8, ...)
```

**Arguments**

object	object of class mfbvar
fcst_start	The date (YYYY-MM-DD) of the first forecast. If not provided, dates from the original data is used if available.
aggregate_fcst	If forecasts of quarterly variables should be aggregated back to the quarterly frequency.
pred_bands	The level of the probability bands for the forecasts.
...	Currently not in use.

**Details**

Note that this requires that forecasts were made in the original mfbvar call.

**Examples**

```
prior_obj <- set_prior(Y = mf_sweden[, 4:5], freq = c("m", "q"),
                      n_lags = 4, n_burnin = 20, n_reps = 20, n_fcst = 4)
mod_minn <- estimate_mfbvar(prior_obj, prior = "minn")
predict(mod_minn)
```

---

print.mfbvar                      *Printing method for class mfbvar*

---

**Description**

Method for printing mfbvar objects.

**Usage**

```
## S3 method for class 'mfbvar'
print(x, ...)
```

**Arguments**

x	object of class mfbvar
...	Currently not in use.

**Examples**

```
prior_obj <- set_prior(Y = mf_sweden[, 4:5], d = "intercept",
                      freq = c("m", "q"), n_lags = 4, n_burnin = 20, n_reps = 20)
mod_minn <- estimate_mfbvar(prior_obj, prior = "minn")
mod_minn
```

---

```
print.mfbvar_prior      Print method for mfbvar_prior
```

---

### Description

Printing method for object of class `mfbvar_prior`, checking if information in the prior is sufficient for estimating models.

### Usage

```
## S3 method for class 'mfbvar_prior'
print(x, ...)
```

### Arguments

<code>x</code>	prior object (class <code>mfbvar_prior</code> )
<code>...</code>	additional arguments (currently unused)

### Details

The print method checks whether the steady-state and Minnesota priors can be used with the current specification. This check is minimal in the sense that it checks only prior elements with no defaults, and it only checks for estimation and not forecasting (for which the steady-state prior requires additional information).

### See Also

[set\\_prior](#), [update\\_prior](#), [estimate\\_mfbvar](#), [summary.mfbvar\\_prior](#)

### Examples

```
prior_obj <- set_prior(Y = mf_sweden, freq = c(rep("m", 4), "q"),
                      n_lags = 4, n_burnin = 100, n_reps = 100)
print(prior_obj)
```

---

```
set_prior      Set priors for mfbvar
```

---

### Description

The function creates an object storing all information needed for estimating a mixed-frequency BVAR. The object includes data as well as details for the model and its priors.

**Usage**

```
set_prior(Y, freq, aggregation = "average", prior_Pi_AR1 = rep(0,
  ncol(Y)), lambda1 = 0.2, lambda2 = 0.5, lambda3 = 1,
  lambda4 = 10000, block_exo = NULL, n_lags, n_fcst = 0,
  n_thin = 1, n_burnin, n_reps, d = NULL, d_fcst = NULL,
  prior_psi_mean = NULL, prior_psi_Omega = NULL, check_roots = FALSE,
  s = -1000, prior_ng = c(0.01, 0.01), prior_phi = c(0.9, 0.1),
  prior_sigma2 = c(0.01, 4), n_fac = NULL, n_cores = 1,
  a = 1/(ncol(Y)^2 * n_lags), verbose = FALSE, ...)
```

```
update_prior(prior_obj, ...)
```

```
check_prior(prior_obj)
```

**Arguments**

Y	matrix or data frame with data with quarterly variables stored in the final columns. For full compatibility with plot and predict functions, Y should contain monthly YYYY-MM-DD dates, either as a separate column or as row names. Data points for quarterly variables should be stored at the end of each quarter with NA otherwise.
freq	Character vector with elements 'm' (monthly) or 'q' (quarterly) for sampling frequency. Monthly variables must precede all quarterly variables.
aggregation	the aggregation scheme used for relating latent monthly series to their quarterly observations. The default is "average" for averaging over the monthly observations within each quarter. The alternative is "triangular" is to use the Mariano-Murasawa triangular set of weights. See details for more information.
prior_Pi_AR1	The prior means for the AR(1) coefficients.
lambda1	The overall tightness.
lambda2	(Only if variance is one of c("diffuse", "fsv")) The cross-variable tightness
lambda3	The tightness of the intercept prior variance.
lambda4	(Minnesota only) Prior variance of the intercept.
block_exo	(Only if variance is one of c("diffuse", "fsv")) Vector of indexes/names of variables to be treated as block exogenous
n_lags	The number of lags.
n_fcst	The number of periods to forecast.
n_thin	Store every n_thin draw
n_burnin	The number of burn-in replications.
n_reps	The number of replications.
d	(Steady state only) Either a matrix with same number of rows as Y and n_determ number of columns containing the deterministic terms or a string "intercept" for requesting an intercept as the only deterministic term.
d_fcst	(Steady state only) The deterministic terms for the forecasting period (not used if d = "intercept").

prior_psi_mean	(Steady state only) Vector of length $n\_determ \times n\_vars$ with the prior means of the steady-state parameters.
prior_psi_Omega	(Steady state only) Matrix of size $(n\_determ \times n\_vars) \times (n\_determ \times n\_vars)$ with the prior covariance of the steady-state parameters.#'
check_roots	Logical, if roots of the companion matrix are to be checked to ensure stationarity.
s	(Hierarchical steady state only) scalar giving the tuning parameter for the Metropolis-Hastings proposal for the kurtosis parameter. If $s < 0$ , then adaptive Metropolis-Hastings targeting an acceptance rate of 0.44 is used, where the scaling factor is restricted to the interval $[-abs(s), abs(s)]$
prior_ng	(Hierarchical steady state only) vector with two elements giving the parameters $c(c0, c1)$ of the hyperprior for the global shrinkage parameter
prior_phi	(Only used with common stochastic volatility) Vector with two elements $c(\text{mean}, \text{variance})$ for the AR(1) parameter in the log-volatility regression
prior_sigma2	(Only used with common stochastic volatility) Vector with two elements $c(\text{mean}, \text{df})$ for the innovation variance of the log-volatility regression
n_fac	(Only used with factor stochastic volatility) Number of factors to use for the factor stochastic volatility model
n_cores	(Only used with factor stochastic volatility) Number of cores to use for drawing regression parameters in parallel
a	(Only used with the Dirichlet-Laplace prior) Shrinkage hyperparameter $a$ (lower values impose more powerful shrinkage)
verbose	Logical, if progress should be printed to the console.
...	(Only used with factor stochastic volatility) Arguments to pass along to <a href="#">fsvsample</a> . See details.
prior_obj	an object of class <code>mfvar_prior</code>

## Details

Some support is provided for single-frequency data sets, where  $Y$  contains only monthly or only quarterly variables. The vector of frequencies given to `freq` should be set accordingly.

The aggregation weights that can be used for aggregation are intra-quarterly averages (`aggregation = "average"`), where the quarterly observations  $y_{q,t}$  are assumed to relate to the underlying monthly series  $z_{q,t}$  through:

$$y_{q,t} = \frac{1}{3}(z_{q,t} + z_{q,t-1} + z_{q,t-2})$$

If `aggregation = "triangular"`, then instead

$$y_{q,t} = \frac{1}{9}(z_{q,t} + 2z_{q,t-1} + 3z_{q,t-2} + 2z_{q,t-3} + z_{q,t-4})$$

The latter is typically used when modeling growth rates, and the former when working with log-levels.

If the steady-state prior is to be used, the deterministic matrix needs to be supplied, or a string indicating that the intercept should be the only deterministic term (`d = "intercept"`). If the latter,

d\_fcst is automatically set to be intercept only. Otherwise, if forecasts are requested ( $n\_fcst > 0$ ) also d\_fcst must be provided. Finally, the prior means of the steady-state parameters must (at the very minimum) also be provided in prior\_psi\_mean. The steady-state prior involves inverting the lag polynomial. For this reason, draws in which the largest eigenvalue (in absolute value) of the lag polynomial is greater than 1 are discarded and new draws are made if check\_roots = TRUE. The maximum number of attempts is 1,000.

For modeling stochastic volatility by the factor stochastic volatility model, the number of factors to use must be supplied. Further arguments can be passed along, but are not included as formal arguments. If the default settings are not overridden, the defaults used are as follows (see [fsvsample](#) for descriptions):

- priormu = c(0,10)
- priorphiidi = c(10,3)
- priorphifac = c(10,3)
- priorsigmaidi = 1
- priorsigmafac = 1
- priorfacload = 1
- restrict = "none"

The function update\_prior can be used to update an existing prior object. See the examples.

### See Also

[estimate\\_mfbvar](#), [update\\_prior](#), [interval\\_to\\_moments](#), [print.mfbvar\\_prior](#), [summary.mfbvar\\_prior](#), [fsvsample](#)

### Examples

```
prior_obj <- set_prior(Y = mf_sweden, freq = c(rep("m", 4), "q"),
                      n_lags = 4, n_burnin = 100, n_reps = 100)
prior_obj <- update_prior(prior_obj, n_fcst = 4)
```

---

summary.mfbvar

*Summary method for class mfbvar*

---

### Description

Method for summarizing mfbvar objects.

### Usage

```
## S3 method for class 'mfbvar'
summary(object, ...)
```

**Arguments**

object	object of class mfbvar
...	Currently not in use.

**Examples**

```
prior_obj <- set_prior(Y = mf_sweden[, 4:5], d = "intercept",  
                      freq = c("m", "q"), n_lags = 4, n_burnin = 20, n_reps = 20)  
mod_minn <- estimate_mfbvar(prior_obj, prior = "minn")  
summary(mod_minn)
```

---

summary.mfbvar\_prior *Summary method for mfbvar\_prior*

---

**Description**

summary method for object of class mfbvar\_prior, showing some basic information regarding the contents of the prior.

**Usage**

```
## S3 method for class 'mfbvar_prior'  
summary(object, ...)
```

**Arguments**

object	prior object (class mfbvar_prior)
...	additional arguments (currently unused)

**See Also**

[set\\_prior](#), [update\\_prior](#), [estimate\\_mfbvar](#), [print.mfbvar\\_prior](#)

**Examples**

```
prior_obj <- set_prior(Y = mf_sweden, freq = c(rep("m", 4), "q"),  
                      n_lags = 4, n_burnin = 100, n_reps = 100)  
summary(prior_obj)
```



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