## How to use Fast Step Graph

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To install the last version of this package directly from GitHub uncomment and run:

```
# library(devtools)
# use "quiet = FALSE" if you want to see the outputs of this command
# devtools::install_github("juancolonna/FastStepGraph", quiet = TRUE, force = TRUE)
# Then, load it:
library(FastStepGraph)
```

Simulate Gaussian Data with an Autoregressive (AR) Model:

```
set.seed(1234567)
phi <- 0.4
p <- 50 # number of variables (dimension)
n <- 30 # number of samples
# Generate Data from a Gaussian distribution
data <- FastStepGraph::SigmaAR(n, p, phi)
X <- scale(data$X) # standardizing variables</pre>
```

To fit the Omega matrix with FastStepGraph() function you have to know the optimal values of  $\alpha_{\mathbf{f}}$  and  $\alpha_{\mathbf{b}}$ . If you don't know these values, try to find them using cross-validation as follows:

```
t0 <- Sys.time() # INITIAL TIME
res <- FastStepGraph::cv.FastStepGraph(X, data_shuffle = TRUE)
difftime(Sys.time(), t0, units = "secs")
#> Time difference of 3.542877 secs
# print(res$alpha_f_opt)
# print(res$alpha_b_opt)
```

If your input variables are non-standardized (with zero mean and unit variance), we recommend that you set data\_scale=TURE. Subsequently, calculate the Omega matrix by calling the FastStepGraph() function passing the optimal parameters  $\alpha_{f}$  and  $\alpha_{b}$  found by cross-validation to fit the final model:

```
t0 <- Sys.time() # INITIAL TIME
G <- FastStepGraph::FastStepGraph(X, alpha_f = res$alpha_f_opt, alpha_b = res$alpha_b_opt)
difftime(Sys.time(), t0, units = "secs")
#> Time difference of 0.002274036 secs
# print(G$Omega)
```

You can also perform these two steps, the cross-validation to obtain the ideal parameters and return the fitted model, in a single step by setting the return\_model=TRUE option as follows:

```
t0 <- Sys.time() # INITIAL TIME
res <- FastStepGraph::cv.FastStepGraph(X, return_model=TRUE, data_shuffle = TRUE)</pre>
```

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```
difftime(Sys.time(), t0, units = "secs")
#> Time difference of 3.470215 secs
# print(res$alpha_f_opt)
# print(res$alpha_b_opt)
# print(res$Omega)
```

The arguments n\_folds = 5, alpha\_f\_min = 0.1, alpha\_f\_max = 0.9, n\_alpha = 32 (size of the grid search) and nei.max = 5, have defaults values and can be omitted. Note that, cv.FastStepGraph(X) is not an exhaustive grid search over  $\alpha_f$  and  $\alpha_b$ . This is a heuristic that tests only a few  $\alpha_b$  values starting with the rule  $\alpha_b = \frac{\alpha_f}{2}$ . It is recommended to shuffle the rows of X before running cross-validation. The default value is data\_shuffle = TRUE, but if you want to disable row shuffle, set it to data\_shuffle = FALSE.

To increase time performance, you can run cv.FastStepGraph(X, parallel = TRUE) in parallel. Before, you'll need to install and register a parallel backend. To run on a Linux system the **doParallel** dependency must be installed install.packages("doParallel"). These parallel packages will also require the following dependencies: foreach, iterators and parallel. Make sure you satisfy them. Then, call the method setting the parameter parallel = TRUE, as follows:

```
t0 <- Sys.time() # INITIAL TIME
# use 'n_cores = NULL' to set the maximum number of cores minus one on your machine
res <- FastStepGraph::cv.FastStepGraph(X, return_model=TRUE, parallel = TRUE, n_cores = 2)
difftime(Sys.time(), t0, units = "secs")
# print(res$alpha_f_opt)
# print(res$alpha_b_opt)
# print(res$Omega)</pre>
```

Remember, you can set the n\_cores parameter to a value equal to the number of cores you have, but be careful as this may overload your system. Setting it to 1 disables parallel processing, and setting it to a number greater than the number of available cores does not improve efficiency.