Vignette to fit the main model

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R Markdown

This is a vignette document on how to run the code associated with the article Maity, A. K., Pati, D., and Mallick, B. K. (2024). Bayesian Time-to-event Density Regression with Application to High Dimensional Data

```
# for multuivariate normal distribution
require(mvtnorm)
## Loading required package: mvtnorm
                    \# for truncated normal distribution
library(msm)
require(tmvtnorm)
                    # for truncated multivariate normal distribution
## Loading required package: tmvtnorm
## Loading required package: Matrix
## Loading required package: stats4
## Loading required package: gmm
## Loading required package: sandwich
require(truncdist) # for random sample from truncated Gamma distribution
## Loading required package: truncdist
## Loading required package: evd
require(MHadaptive) # for makePositivedefinite function
## Loading required package: MHadaptive
## Loading required package: MASS
require(foreach)
                 # for foreach
## Loading required package: foreach
require(doParallel) # for cluster creation
## Loading required package: doParallel
## Loading required package: iterators
## Loading required package: parallel
require(SuppDists) # for rinvGauss function
## Loading required package: SuppDists
require(ipred)
                    # for integrated Brier Score
## Loading required package: ipred
```

```
require(survival) # for Surv function
## Loading required package: survival
require(ggplot2)
                  # for confidence band
## Loading required package: ggplot2
n <- 150 # sample size
p <- 200 # number of covariates
x <- rmvnorm(n, mean = rep(0, p)) # design matrix
k <- 5 # number of true variables
beta.t \leftarrow c(sample(c(1, -1), size = k, replace = TRUE), rep(0, p - k))
linear.part <- as.vector(x %*% beta.t)</pre>
sigma.square <- 1
c < -0.1
d < -2
# Set the parameters
nburnin <- 1000
       <- 1000
nmc
thin
        <- 1
N
        <- 61
        <- 2.5
sig
        <- 0.1
nu
       <- seq(from = c, to = d, length.out = N)</pre>
knots
CM
        <- covariance_matrix(knots, sig, nu) # covariance matrix of g</pre>
# Data generation
xi <- rnorm(n, mean = linear.part, sd = sqrt(sigma.square)) # response</pre>
a <- 1
b \leftarrow min(xi) - 1
y \leftarrow (1/a)^{(1/5)} * (xi - b)^{(1/5)}
T <- exp(y) # AFT model
C <- rtrunc(n, spec = "gamma", shape = 1, scale = 10, a = 2) # 20% censor
time <- pmin(T, C) # observed time is min of censored and true
status = time == T  # set to 1 if event is observed
1 - sum(status)/length(y) # censoring rate
## [1] 0.2066667
censor.rate <- 1 - sum(status)/length(y) # censoring rate</pre>
censor.rate
## [1] 0.2066667
ct <- as.matrix(cbind(time = time, status = status)) # censored time
X <- scale(x)</pre>
        <- min(log(time)) - 0.1
        \leftarrow \max(\log(\text{time})) + 0.1
knots <- seq(from = c, to = d, length.out = N)</pre>
        <- covariance_matrix(knots, sig, nu) # covariance matrix of g</pre>
```

```
# Training set and Test set
          <- 100
ntrain
ntest
           <- n - ntrain
cttrain
          <- ct[1:ntrain, ]</pre>
cttest
          <- ct[(ntrain + 1):n, ]
test.index1 <- which(cttest[, 2] == 1) # which are NOT censored</pre>
          <- cttest[test.index1, ]</pre>
cttest1
Xtrain
           <- X[1:ntrain, ]
         <- X[(ntrain + 1):n, ]
Xtest
```

The following will fit the model and provide the significant variables. However, the function gph() and the file helper.R has to be loaded in the R console before running the following.

```
gph.fit <- gph(ct = cttrain, X = Xtrain, nburnin = nburnin, nmc = nmc, thin = thin, method.mcmc = "MH",
## [1] 100
## [1] 200
## [1] 300
## [1] 400
## [1] 500
## [1] 600
## [1] 700
## [1] 800
## [1] 900
## [1] 1000
## [1] 1100
## [1] 1200
## [1] 1300
## [1] 1400
## [1] 1500
## [1] 1600
## [1] 1700
## [1] 1800
## [1] 1900
## [1] 2000
beta.post.gp <- gph.fit$BetaHat</pre>
cluster
           <- kmeans(abs(beta.post.gp), centers = 2)$cluster # return cluster indices</pre>
cluster1
            <- which(cluster == 1)</pre>
cluster2 <- which(cluster == 2)</pre>
min.cluster <- ifelse(length(cluster1) < length(cluster2), 1, 2)
discovery <- which(cluster == min.cluster)</pre>
discovery # significant variables
```

[1] 1 2 3 4 5