BARS Package Test Output

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Introduction and Installation

Hello! Thank you for taking the time to test the bars package in R. This package lays out the foundation for Bayesian Adaptive Regression Splines, a common tool used to fit nonlinear data from a Bayesian perspective. This algorithm utilizes various MCMC techniques, such as Reversible-Jump algorithms and Gibbs sampling to adaptively predict the optimal number of knots and their locations without overfitting the data. I am still writing the Vignette and will upload it to the repository when it is completed.

Please run the code below and ensure all outputs appear exactly as the file TEST_FILE.pdf which can be found on the GitHub repository for this project: https://github.com/aashen12/BARS. In its entirety, the file should take around 1 minute to run completely, depending on your processor speed.

If you encounter any errors that you cannot fix after multiple attempts, please send me a screenshot of the full error message as well as the entire pdf/Rmd file with the error. You can email me (Andy) at aashen@ucla.edu. Please do not hesitate to email me if you have any questions as well.

Thanks again, and enjoy!

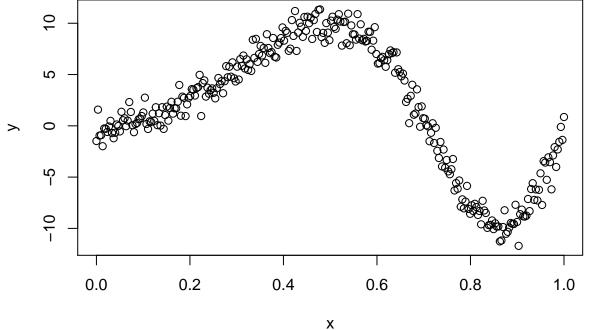
To install the bars package, please run the following command in your RStudio console:

devtools::install_github("aashen12/BARS") # only run this command once

Your name here

Data Generation

```
rm(list = ls())
set.seed(12) # feel free to modify once you get the code working
n <- 300
x <- seq(0,1,length.out=n)
y <- sin(2*pi*x^2)*10+rnorm(n)
plot(x, y)</pre>
```



Feel free to play around with the data and modify however you like once you get this code working.

Bayesian Spline with Random Knots

```
library(bars) #crucial step in calling the package
library(mvtnorm) #this line is necessary...for now
nmcmc <- 5000
res <- bars(its = nmcmc, verbose = TRUE)

## Iteration number 1000 sigma^2 = 0.8929665
## Iteration number 2000 sigma^2 = 0.8660429
## Iteration number 3000 sigma^2 = 0.9946919
## Iteration number 4000 sigma^2 = 1.086504
## Iteration number 5000 sigma^2 = 1.050121
```

Results

Beta

```
beta <- res$beta
beta <- beta[,colSums(is.na(beta)) != nrow(beta)]
round(beta[nrow(beta),], 2)

## [1] 11.86 61.17 -48.31 51.35 64.75 -23.57 -34.99 NA NA NA
## [11] NA NA</pre>
```

Sigma

```
sig <- res$sig
tail(round(sig, 2))
## [1] 0.99 0.96 1.07 1.05 1.10 1.05</pre>
```

Knots

```
knots <- res$knots
knots <- knots[,colSums(is.na(knots)) != nrow(knots)]
round(knots[nrow(knots),], 2)

## [1] 0.82 0.63 0.92 0.87 0.57 0.47 NA NA NA NA NA
res$knots_total

## [1] 6</pre>
```

Signs

```
signs <- res$signs
signs <- signs[,colSums(is.na(signs)) != nrow(signs)]
round(signs[nrow(signs),], 2)
## [1] 1 1 1 1 -1 1 NA NA NA NA</pre>
```

Basis Functions

0.18

0.18

0.18

0.36

0.37

0.37

[298,] 1

[299,] 1

[300,] 1

```
X <- res$X
tail(round(X, 2))
           basis_vec basis_vec basis_vec basis_vec basis_vec
## [295,] 1
                0.17
                          0.35
                                   0.06
                                             0.11
                                                                 0.51
## [296,] 1
                0.17
                          0.36
                                    0.06
                                             0.12
                                                          0
                                                                 0.51
## [297,] 1
                0.17
                          0.36
                                    0.07
                                             0.12
                                                          0
                                                                 0.52
```

0.12

0.13

0.13

0

0

0

0.52

0.52

0.53

0.07

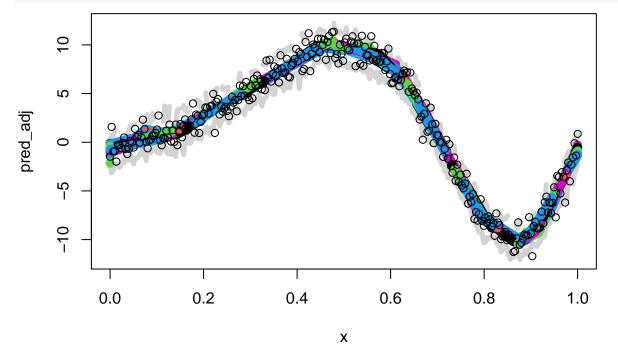
0.07

0.08

Plotting Predicted Values

```
knotnum <- ncol(knots)
mean.pred <- matrix(NA, nrow = nmcmc, ncol = length(x))
pred <- mean.pred
for(p in 1:nmcmc) {
   splb <- spline.basis(nknot = knotnum, knots = knots[p,], signs = signs[p,])
   mean.pred[p,] <- splb %*% beta[p,]
   pred[p,] <- mean.pred[p,] + rnorm(length(x), sd = sqrt(sig[p]))
}
mean.pred <- t(mean.pred)

pred_adj <- t(apply(pred, 2, quantile, probs = c(0.025, 0.975), na.rm = TRUE))
matplot(x, pred_adj, col = "lightgrey", lwd = 5, type = "l")
matplot(x, mean.pred, type = "l", lwd = 9, add = TRUE)
points(x, y)</pre>
```



Plot of Averages

```
plot(x, rowMeans(mean.pred, na.rm = TRUE), lwd = 8, type = "l", col = "royalblue")
points(x, y)
```

