bart2.R.

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```
# load packages
library(tidyverse)
## -- Attaching packages ----- tidyverse 1.3.2 --
                  v purrr 1.0.1
## v ggplot2 3.4.0
## v tibble 3.1.8 v dplyr 1.0.10
## v tidyr 1.2.1
                    v stringr 1.5.0
                  v forcats 0.5.2
## v readr 2.1.3
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(BART)
## Warning: package 'BART' was built under R version 4.2.3
## Loading required package: nlme
## Attaching package: 'nlme'
## The following object is masked from 'package:dplyr':
##
##
      collapse
##
## Loading required package: nnet
## Loading required package: survival
library(here)
## here() starts at C:/Users/Eliyahu/OneDrive - Johns Hopkins/theory_stat2/project
library(tictoc)
## Warning: package 'tictoc' was built under R version 4.2.3
# set seed
set.seed(1)
# load data
diabetes <- read_csv(here("data", "diabetes_binary_health_indicators_BRFSS2015.csv"))
```

```
## Rows: 253680 Columns: 22
## -- Column specification ------
## Delimiter: ","
## dbl (22): Diabetes_binary, HighBP, HighChol, CholCheck, BMI, Smoker, Stroke,...
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
# subset men over 30 for type 2 diabetes
diabetes2 <- diabetes %>%
   filter((Sex == 1) & (Age >= 3)) %>% #age category 3 is 30-34
    select(-c(Sex))%>% # sex is no longer variable
   slice_sample(n= 5000) # subset data for computational efficiency
# set the feature and target sets
x2 <- diabetes2 %>%
    select(-c(Diabetes_binary)) %>%
   as.data.frame()
y2 <- diabetes2 %>%
   select(Diabetes_binary)%>%
   deframe()
# set the smoker variable
x2_treat1 <- x2 %>%
   mutate(Smoker = 1)
x2_treat0 <- x2 %>%
   mutate(Smoker = 0)
tic('BART')
# BART fit
bartfit <- gbart(x.train = x2, y.train = y2, type = 'pbart', ntree = 50, printevery = 1000)
## *****Calling gbart: type=2
## ****Data:
## data:n,p,np: 5000, 20, 0
## y1,yn: 0.000000, 0.000000
## x1,x[n*p]: 0.000000, 7.000000
## *****Number of Trees: 50
## *****Number of Cut Points: 1 ... 7
## *****burn,nd,thin: 100,10000,10
## ****Prior:beta,alpha,tau,nu,lambda,offset: 2,0.95,0.212132,3,1,-1.0052
## *****Dirichlet:sparse,theta,omega,a,b,rho,augment: 0,0,1,0.5,1,20,0
## ****printevery: 1000
##
## MCMC
## done 0 (out of 10100)
## done 1000 (out of 10100)
## done 2000 (out of 10100)
## done 3000 (out of 10100)
## done 4000 (out of 10100)
## done 5000 (out of 10100)
## done 6000 (out of 10100)
## done 7000 (out of 10100)
## done 8000 (out of 10100)
```

```
## done 9000 (out of 10100)
## done 10000 (out of 10100)
## time: 104s
## trcnt, tecnt: 1000,0
toc()
## BART: 106.22 sec elapsed
# predictions
predict1 <- predict(bartfit, x2_treat1)</pre>
## *****In main of C++ for bart prediction
## tc (threadcount): 1
## number of bart draws: 1000
## number of trees in bart sum: 50
## number of x columns: 20
## from x,np,p: 20, 5000
## ***using serial code
predict0 <- predict(bartfit, x2_treat0)</pre>
## *****In main of C++ for bart prediction
## tc (threadcount): 1
## number of bart draws: 1000
## number of trees in bart sum: 50
## number of x columns: 20
## from x,np,p: 20, 5000
## ***using serial code
# ate estimation
e_ate <- mean(predict1$prob.test.mean - predict0$prob.test.mean)</pre>
print(e_ate)
```

[1] 0.0148356