

combination

library package

Warning: package 'pROC' was built under R version 4.3.3

Warning: package 'bartMachine' was built under R version 4.3.3

Warning: package 'randomForest' was built under R version 4.3.3

Warning: package 'missForest' was built under R version 4.3.3

Warning: package 'ggplot2' was built under R version 4.3.3

Warning: package 'dbarts' was built under R version 4.3.3

Warning: package 'BART' was built under R version 4.3.3

Combine all package

```
bart_package <- function(train_x,train_y,test_x,test_y,package,repeat_time,
                          num_trees,alpha,beta) {
  # define value
  results <- data.frame()
  time_BART <- c()
  mse_BART <- c()
  time_dbarts <- c()
  mse_dbarts <- c()
  time_BM <- c()
  mse_BM <- c()
}
```

```

time_SB <- c()
mse_SB <- c()
packageName <- ""
for(i in 1:repeat_time){
  for(package_name in package){
    if (package_name == "BART"){
      t <- system.time({fit <- pbart(x.train = train_x,
                                   y.train = train_y,
                                   x.test = test_x,
                                   ntree = num_trees,
                                   base = alpha,
                                   power = beta)})

      e_time <- as.numeric(t["elapsed"])
      predictions <- colMeans(fit$yhat.test)
      mse_score <- mean((test_y - predictions)^2)
      time_BART[i] <- e_time
      mse_BART[i] <- mse_score

    }

    if(package_name == "dbarts"){
      t <- system.time({bart_model <- bart(
        x.train = train_x,
        y.train = train_y,
        x.test = test_x,
        ntree = num_trees,
        power = beta,
        base = alpha
      )})

      e_time <- as.numeric(t["elapsed"])
      predictions <- colMeans(bart_model$yhat.test)
      mse_score <- mean((test_y - predictions)^2)
      time_dbarts[i] <- e_time
      mse_dbarts[i] <- mse_score

    }

    if(package_name == "bartMachine"){
      bart_model <- bartMachine(
        X = train_x,

```

```

        y = train_y,
        num_trees = num_trees,
        beta = beta,
        alpha = alpha

    )
    # The value of calculating the time required for modeling
    e_time <- bart_model$time_to_build
    predictions <- predict(bart_model,test_x,type = "prob")
    mse_score <- mean((test_y - predictions)^2)
    time_BM[i] <- e_time
    mse_BM[i] <- mse_score

}

if(package_name=="SoftBart"){
    t <- system.time({bart_model <- softbart(X = train_x, Y = train_y, X_test = test_x,
                                             hypers = Hypers(train_x, train_y, num_tree = num_trees, gamma = 0.001),
                                             opts = Opts(num_burn = 200, num_save = 1000, update_tau = TRUE)))

    #print(t)
    e_time <- as.numeric(t["elapsed"])
    #print(e_time)
    predictions <- bart_model$y_hat_test_mean
    mse_score <- mean((test_y - predictions)^2)
    time_SB[i] <- e_time
    mse_SB[i] <- mse_score
    #print(1)
}
}
}

for (package_name in package) {
    if (package_name == "BART"){
        time_mu <- mean(time_BART)
        time_sd <- sd(time_BART)
        mse_mu <- mean(mse_BART)
        mse_sd <- sd(mse_BART)
        new_row <- data.frame(MSE_mean = mse_mu, MSE_se = mse_sd,
                             running_time_mean = time_mu,running_time_sd = time_sd,
                             package_name = package_name)
        results <- rbind(results, new_row)
    }
    if (package_name == "dbarts"){

```

```

time_mu <- mean(time_dbarts)
time_sd <- sd(time_dbarts)
mse_mu <- mean(mse_dbarts)
mse_sd <- sd(mse_dbarts)
new_row <- data.frame(MSE_mean = mse_mu, MSE_se = mse_sd,
                      running_time_mean = time_mu, running_time_sd = time_sd,
                      package_name = package_name)
results <- rbind(results, new_row)
}
if (package_name == "bartMachine"){
  time_mu <- mean(time_BM)
  time_sd <- sd(time_BM)
  mse_mu <- mean(mse_BM)
  mse_sd <- sd(mse_BM)
  new_row <- data.frame(MSE_mean = mse_mu, MSE_se = mse_sd,
                        running_time_mean = time_mu, running_time_sd = time_sd,
                        package_name = package_name)
  results <- rbind(results, new_row)
}
if (package_name == "SoftBart"){
  time_mu <- mean(time_SB)
  time_sd <- sd(time_SB)
  mse_mu <- mean(mse_SB)
  mse_sd <- sd(mse_SB)
  new_row <- data.frame(MSE_mean = mse_mu, MSE_se = mse_sd,
                        running_time_mean = time_mu, running_time_sd = time_sd,
                        package_name = package_name)
  results <- rbind(results, new_row)
}
}
return(results)
}

```

```

set.seed(316)
cancer <- read.csv("C:/Users/pyk/Desktop/nus/RA/project/imodels-data-master/data_cleaned/bre
x <- cancer[, -18]
y <- cancer[, 18]

train_indices <- createDataPartition(y, p = 0.8, list = FALSE)
train_x <- x[train_indices, ]
test_x <- x[-train_indices, ]
train_y <- y[train_indices]

```

```
test_y <- y[-train_indices]

p <- c("BART","dbarts","bartMachine","SoftBart")

fit <- bart_package(train_x,train_y,test_x,test_y,package = p,repeat_time = 3,
                    num_trees = 20,alpha = 0.95,beta = 2)
```

```
*****Into main of pbart
*****Data:
data:n,p,np: 222, 17, 55
y1,yn: 1, 0
x1,x[n*p]: 3.000000, 1.000000
xp1,xp[np*p]: 4.000000, 0.000000
*****Number of Trees: 20
*****Number of Cut Points: 5 ... 1
*****burn and ndpost: 100, 1000
*****Prior:mybeta,alpha,tau: 2.000000,0.950000,0.335410
*****binaryOffset: -0.558392
*****Dirichlet:sparse,theta,omega,a,b,rho,augment: 0,0,1,0.5,1,17,0
*****nkeeptrain,nkeeptest,nkeeptreedraws: 1000,1000,1000
*****printevery: 100
*****skiptr,skipte,skiptreedraws: 1,1,1
```

```
MCMC
done 0 (out of 1100)
done 100 (out of 1100)
done 200 (out of 1100)
done 300 (out of 1100)
done 400 (out of 1100)
done 500 (out of 1100)
done 600 (out of 1100)
done 700 (out of 1100)
done 800 (out of 1100)
done 900 (out of 1100)
done 1000 (out of 1100)
time: 0s
check counts
trcnt,tecnt: 1000,1000
```

```
Running BART with binary y
```

```

number of trees: 20
number of chains: 1, number of threads 1
tree thinning rate: 1
Prior:
  k prior fixed to 2.000000
  power and base for tree prior: 2.000000 0.950000
  use quantiles for rule cut points: false
  proposal probabilities: birth/death 0.50, swap 0.10, change 0.40; birth 0.50
data:
  number of training observations: 222
  number of test observations: 55
  number of explanatory variables: 17

Cutoff rules c in  $x \leq c$  vs  $x > c$ 
Number of cutoffs: (var: number of possible c):
(1: 100) (2: 100) (3: 100) (4: 100) (5: 100)
(6: 100) (7: 100) (8: 100) (9: 100) (10: 100)
(11: 100) (12: 100) (13: 100) (14: 100) (15: 100)
(16: 100) (17: 100)
offsets:
  reg : 0.00 0.00 0.00 0.00 0.00
  test: 0.00 0.00 0.00 0.00 0.00
Running mcmc loop:
iteration: 100 (of 1000)
iteration: 200 (of 1000)
iteration: 300 (of 1000)
iteration: 400 (of 1000)
iteration: 500 (of 1000)
iteration: 600 (of 1000)
iteration: 700 (of 1000)
iteration: 800 (of 1000)
iteration: 900 (of 1000)
iteration: 1000 (of 1000)
total seconds in loop: 0.058036

Tree sizes, last iteration:
[1] 3 1 3 3 3 3 2 2 2 2 2 3 2 2 3 1 2 2
2 2

Variable Usage, last iteration (var:count):
(1: 0) (2: 0) (3: 3) (4: 2) (5: 3)
(6: 3) (7: 1) (8: 0) (9: 1) (10: 2)
(11: 2) (12: 0) (13: 3) (14: 0) (15: 2)

```

(16: 1) (17: 2)

DONE BART

bartMachine initializing with 20 trees...

bartMachine vars checked...

Warning: The response y is integer, bartMachine will run regression.

bartMachine java init...

bartMachine factors created...

bartMachine before preprocess...

bartMachine after preprocess... 17 total features...

bartMachine sigsq estimated...

bartMachine training data finalized...

Now building bartMachine for regression...

evaluating in sample data...done

[1] "Preprocessing data frame"

[1] "Using default grouping; if this is not desired, preprocess data frame manually using pr

Finishing warmup 100

Finishing warmup 200

Finishing save 100

Finishing save 200

Finishing save 300

Finishing save 400

Finishing save 500

Finishing save 600

Finishing save 700

Finishing save 800

Finishing save 900

Finishing save 1000

*****Into main of pbart

*****Data:

data:n,p,np: 222, 17, 55

y1,yn: 1, 0

x1,x[n*p]: 3.000000, 1.000000

xp1,xp[np*p]: 4.000000, 0.000000

*****Number of Trees: 20

*****Number of Cut Points: 5 ... 1

*****burn and ndpost: 100, 1000

*****Prior:mybeta,alpha,tau: 2.000000,0.950000,0.335410

*****binaryOffset: -0.558392

*****Dirichlet:sparse,theta,omega,a,b,rho,augment: 0,0,1,0.5,1,17,0

*****nkeeptrain,nkeeptest,nkeepreedraws: 1000,1000,1000

*****printevery: 100

*****skiptr,skipte,skiptreedraws: 1,1,1

MCMC

done 0 (out of 1100)
done 100 (out of 1100)
done 200 (out of 1100)
done 300 (out of 1100)
done 400 (out of 1100)
done 500 (out of 1100)
done 600 (out of 1100)
done 700 (out of 1100)
done 800 (out of 1100)
done 900 (out of 1100)
done 1000 (out of 1100)
time: 0s
check counts
trcnt,tecnt: 1000,1000

Running BART with binary y

number of trees: 20
number of chains: 1, number of threads 1
tree thinning rate: 1
Prior:
 k prior fixed to 2.000000
 power and base for tree prior: 2.000000 0.950000
 use quantiles for rule cut points: false
 proposal probabilities: birth/death 0.50, swap 0.10, change 0.40; birth 0.50
data:
 number of training observations: 222
 number of test observations: 55
 number of explanatory variables: 17

Cutoff rules c in $x \leq c$ vs $x > c$

Number of cutoffs: (var: number of possible c):
(1: 100) (2: 100) (3: 100) (4: 100) (5: 100)
(6: 100) (7: 100) (8: 100) (9: 100) (10: 100)
(11: 100) (12: 100) (13: 100) (14: 100) (15: 100)
(16: 100) (17: 100)

offsets:

reg : 0.00 0.00 0.00 0.00 0.00
test: 0.00 0.00 0.00 0.00 0.00

Running mcmc loop:

iteration: 100 (of 1000)


```
iteration: 200 (of 1000)
iteration: 300 (of 1000)
iteration: 400 (of 1000)
iteration: 500 (of 1000)
iteration: 600 (of 1000)
iteration: 700 (of 1000)
iteration: 800 (of 1000)
iteration: 900 (of 1000)
iteration: 1000 (of 1000)
total seconds in loop: 0.058565
```

```
Tree sizes, last iteration:
[1] 2 3 2 1 3 2 3 3 2 2 2 3 3 2 3 3 2 3
2 2
```

```
Variable Usage, last iteration (var:count):
(1: 1) (2: 2) (3: 6) (4: 1) (5: 2)
(6: 1) (7: 2) (8: 1) (9: 1) (10: 3)
(11: 0) (12: 0) (13: 3) (14: 1) (15: 0)
(16: 1) (17: 3)
DONE BART
```

```
bartMachine initializing with 20 trees...
bartMachine vars checked...
Warning: The response y is integer, bartMachine will run regression.
bartMachine java init...
bartMachine factors created...
bartMachine before preprocess...
bartMachine after preprocess... 17 total features...
bartMachine sigsq estimated...
bartMachine training data finalized...
Now building bartMachine for regression...
evaluating in sample data...done
[1] "Preprocessing data frame"
[1] "Using default grouping; if this is not desired, preprocess data frame manually using pr
Finishing warmup 100
Finishing warmup 200
Finishing save 100
Finishing save 200
Finishing save 300
Finishing save 400
Finishing save 500
Finishing save 600
```

```

Finishing save 700
Finishing save 800
Finishing save 900
Finishing save 1000
*****Into main of pbart
*****Data:
data:n,p,np: 222, 17, 55
y1,yn: 1, 0
x1,x[n*p]: 3.000000, 1.000000
xp1,xp[np*p]: 4.000000, 0.000000
*****Number of Trees: 20
*****Number of Cut Points: 5 ... 1
*****burn and ndpost: 100, 1000
*****Prior:mybeta,alpha,tau: 2.000000,0.950000,0.335410
*****binaryOffset: -0.558392
*****Dirichlet:sparse,theta,omega,a,b,rho,augment: 0,0,1,0.5,1,17,0
*****nkeeptrain,nkeptest,nkeepreedraws: 1000,1000,1000
*****printevery: 100
*****skiptr,skipte,skiptreedraws: 1,1,1

```

MCMC

```

done 0 (out of 1100)
done 100 (out of 1100)
done 200 (out of 1100)
done 300 (out of 1100)
done 400 (out of 1100)
done 500 (out of 1100)
done 600 (out of 1100)
done 700 (out of 1100)
done 800 (out of 1100)
done 900 (out of 1100)
done 1000 (out of 1100)
time: 0s
check counts
trcnt,tecnt: 1000,1000

```

Running BART with binary y

```

number of trees: 20
number of chains: 1, number of threads 1
tree thinning rate: 1
Prior:
    k prior fixed to 2.000000

```

```

    power and base for tree prior: 2.000000 0.950000
    use quantiles for rule cut points: false
    proposal probabilities: birth/death 0.50, swap 0.10, change 0.40; birth 0.50
data:
    number of training observations: 222
    number of test observations: 55
    number of explanatory variables: 17

Cutoff rules c in  $x \leq c$  vs  $x > c$ 
Number of cutoffs: (var: number of possible c):
(1: 100) (2: 100) (3: 100) (4: 100) (5: 100)
(6: 100) (7: 100) (8: 100) (9: 100) (10: 100)
(11: 100) (12: 100) (13: 100) (14: 100) (15: 100)
(16: 100) (17: 100)
offsets:
    reg : 0.00 0.00 0.00 0.00 0.00
    test: 0.00 0.00 0.00 0.00 0.00
Running mcmc loop:
iteration: 100 (of 1000)
iteration: 200 (of 1000)
iteration: 300 (of 1000)
iteration: 400 (of 1000)
iteration: 500 (of 1000)
iteration: 600 (of 1000)
iteration: 700 (of 1000)
iteration: 800 (of 1000)
iteration: 900 (of 1000)
iteration: 1000 (of 1000)
total seconds in loop: 0.057012

Tree sizes, last iteration:
[1] 3 2 2 2 2 2 2 3 2 2 2 2 3 2 2 2 2 2
2 1

Variable Usage, last iteration (var:count):
(1: 1) (2: 4) (3: 3) (4: 3) (5: 0)
(6: 1) (7: 1) (8: 1) (9: 1) (10: 0)
(11: 0) (12: 1) (13: 1) (14: 2) (15: 0)
(16: 2) (17: 1)
DONE BART

bartMachine initializing with 20 trees...
bartMachine vars checked...

```

```

Warning: The response y is integer, bartMachine will run regression.
bartMachine java init...
bartMachine factors created...
bartMachine before preprocess...
bartMachine after preprocess... 17 total features...
bartMachine sigsq estimated...
bartMachine training data finalized...
Now building bartMachine for regression...
evaluating in sample data...done
[1] "Preprocessing data frame"
[1] "Using default grouping; if this is not desired, preprocess data frame manually using pr
Finishing warmup 100
Finishing warmup 200
Finishing save 100
Finishing save 200
Finishing save 300
Finishing save 400
Finishing save 500
Finishing save 600
Finishing save 700
Finishing save 800
Finishing save 900
Finishing save 1000

```

```
print(fit)
```

	MSE_mean	MSE_se	running_time_mean	running_time_sd	package_name
1	1.3951936	0.0222138705	0.25000000	0.00000000	BART
2	1.3727706	0.0097309230	0.07666667	0.02081666	dbarts
3	0.1845267	0.0002833019	0.31948725	0.12082123	bartMachine
4	0.1931440	0.0011923028	4.32333333	0.08504901	SoftBart