

Lab 9

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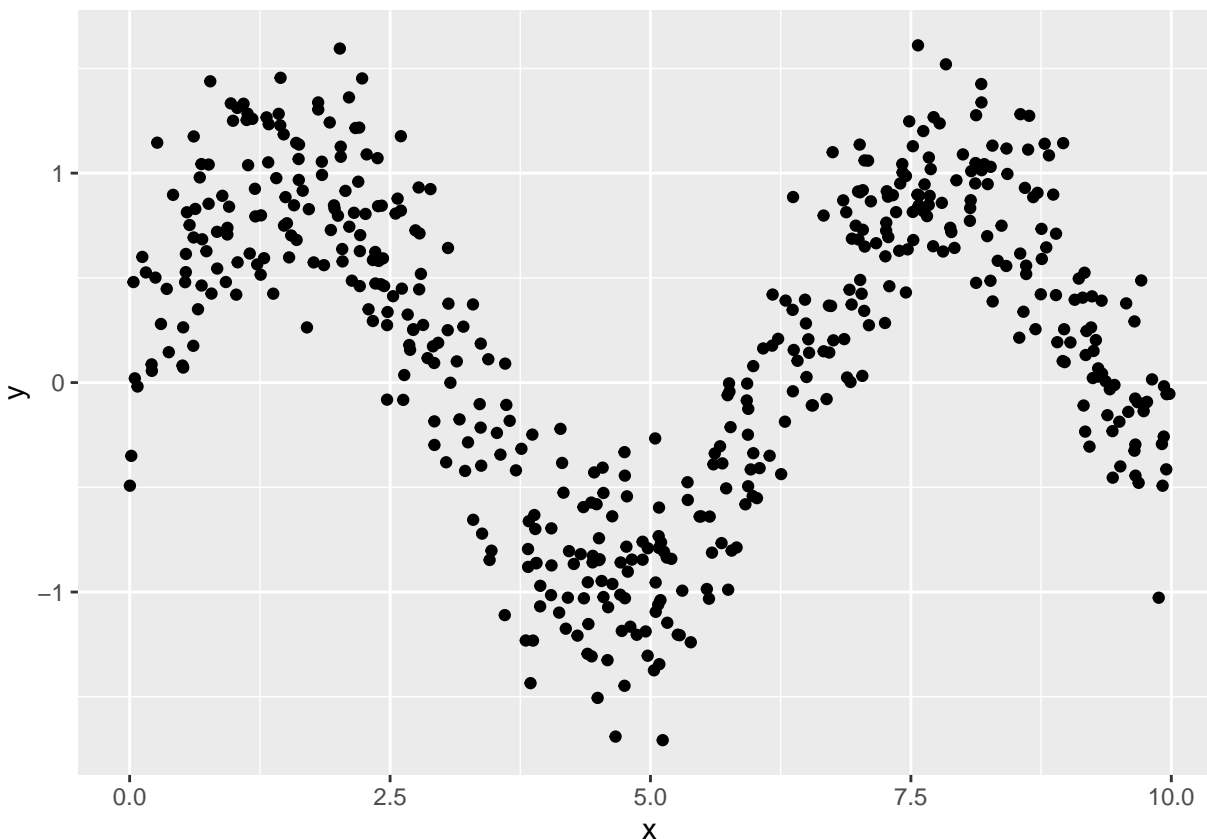
Here we will learn about trees, bagged trees and random forests. You can use the **YARF** package if it works, otherwise, use the **randomForest** package (the standard).

Let's take a look at the simulated sine curve data from practice lecture 12. Below is the code for the data generating process:

```
rm(list = ls())
n = 500
sigma = 0.3
x_min = 0
x_max = 10
f_x = function(x){sin(x)}
y_x = function(x, sigma){f_x(x) + rnorm(n, 0, sigma)}
x_train = runif(n, x_min, x_max)
y_train = y_x(x_train, sigma)
```

Plot an example dataset of size 500:

```
pacman::p_load(ggplot2)
ggplot(data.frame(x = x_train, y = y_train)) +
  geom_point(aes(x = x, y = y))
```



Create a test set of size 500 as well

```
x_test = runif(n, x_min, x_max)
y_test = y_x(x_test, sigma)
```

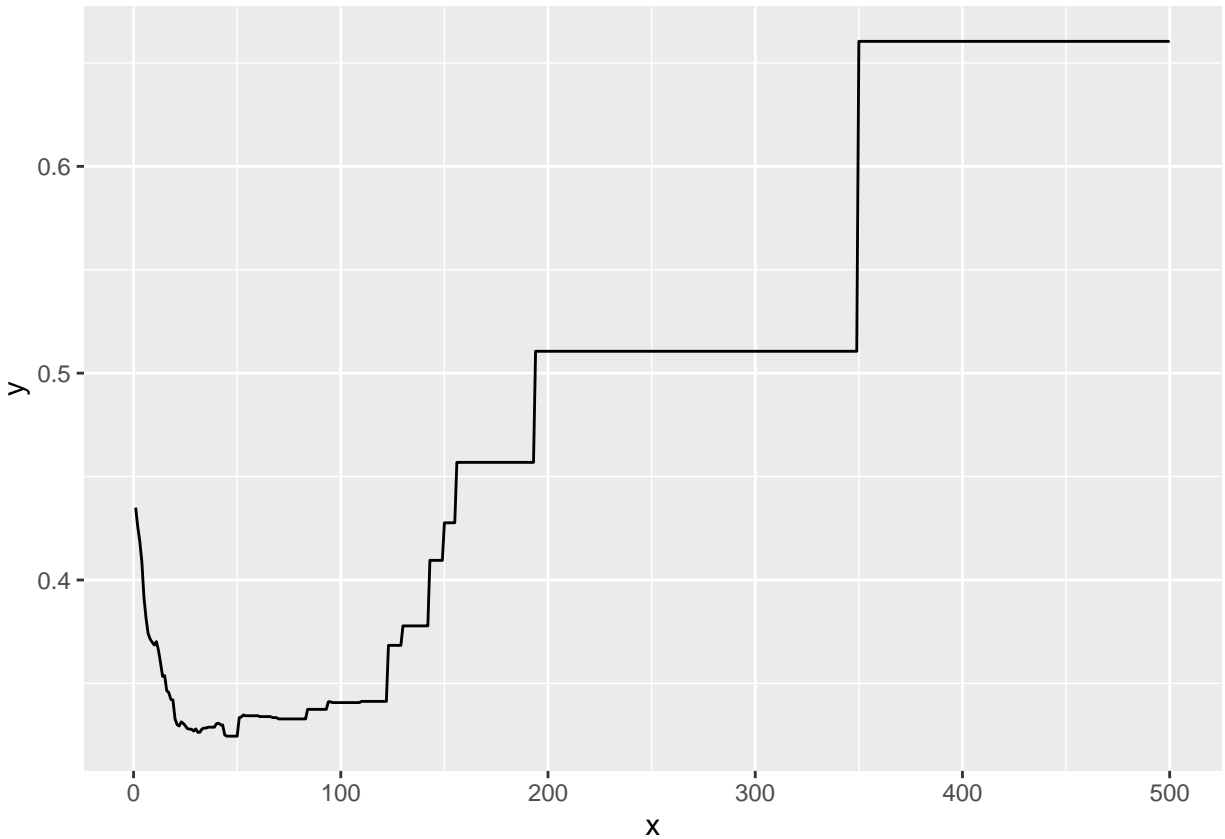
Locate the optimal node size hyperparameter for the regression tree model. I believe you can use `randomForest` here by setting `ntree = 1`, `replace = FALSE`, `sampsize = n` (`mtry` is already set to be 1 because there is only one feature) and then you can set `nodesize`. Plot node size by out of sample SE.

```
pacman::p_load(randomForest)

node_sizes = 1:n

se_by_node_sizes = array(NA, length(node_sizes))
for(i in 1:length(node_sizes)){
  rf_mod = randomForest(x = data.frame(x = x_train), y = y_train, ntree = 1, replace = FALSE, sampsize = n)
  y_hat_test = predict(rf_mod, data.frame(x = x_test))
  se_by_node_sizes[i] = sd(y_test - y_hat_test)
}

ggplot(data.frame(x = node_sizes, y = se_by_node_sizes)) +
  geom_line(aes(x = x, y = y))
```



```
scale_x_reverse()
```

```
## <ScaleContinuousPosition>
## Range:
## Limits: 0 -- 1
```

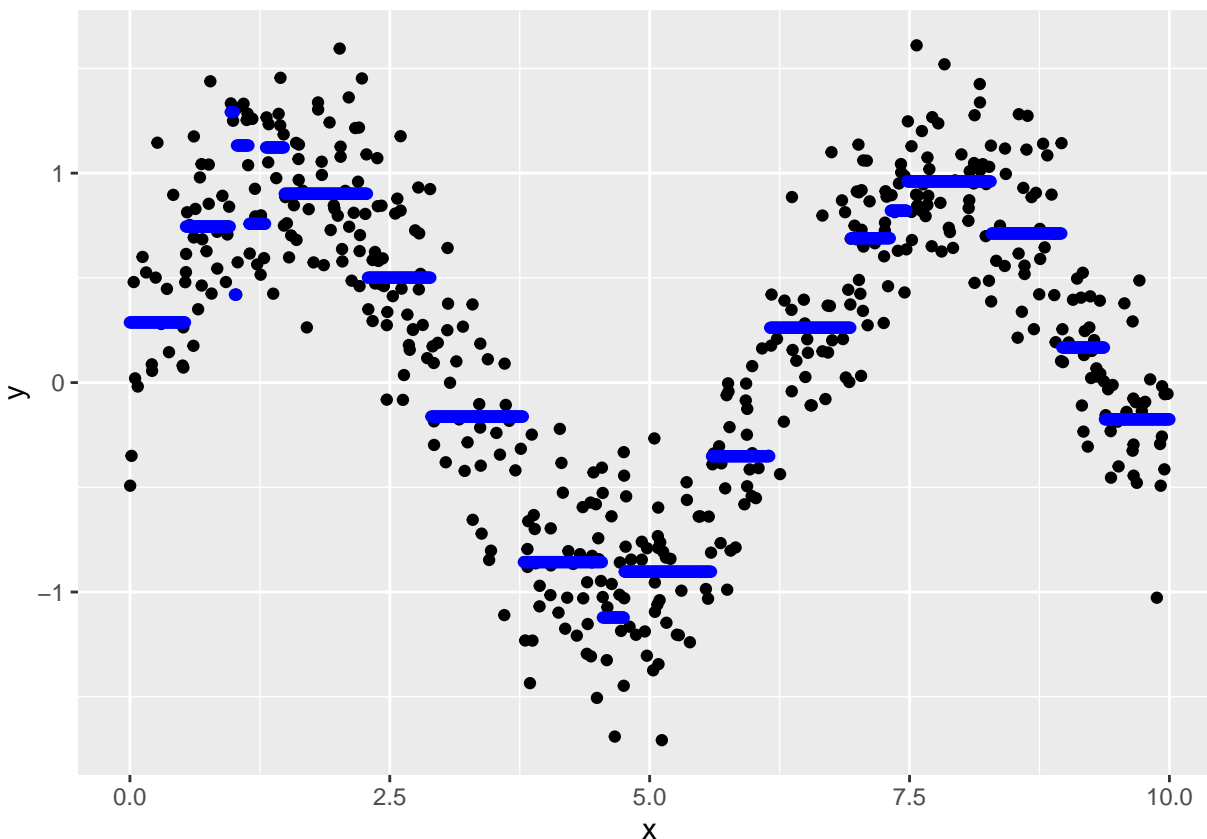
```
which.min(se_by_node_sizes)
```

```
## [1] 45
```

Plot the regression tree model with the optimal node size.

```
rf_mod = randomForest(x = data.frame(x = x_train), y = y_train, ntree = 1, replace = FALSE, sampsize = n,
resolution = 0.01
x_grid = seq(from = x_min, to = x_max, by = resolution)
g_x = predict(rf_mod, data.frame(x = x_grid))

ggplot(data.frame(x = x_grid, y = g_x)) +
  aes(x = x, y = y) +
  geom_point(data = data.frame(x = x_train, y = y_train)) +
  geom_point(color = "blue")
```



Provide the bias-variance decomposition of this DGP fit with this model. It is a lot of code, but it is in the practice lectures. If your three numbers don't add up within two significant digits, increase your resolution.

```
n_train = 20
n_test = 1000
Nsim = 1000

training_gs = matrix(NA, nrow = Nsim, ncol = 2)
x_training = matrix(NA, nrow = Nsim, ncol = n_train)
y_training = matrix(NA, nrow = Nsim, ncol = n_train)
all_oos_residuals = matrix(NA, nrow = Nsim, ncol = n_test)

for (nsim in 1 : Nsim){
  x_train = runif(n_train, x_min, x_max)
  delta_train = rnorm(n_train, 0, sigma)
  y_train = f_x(x_train) + delta_train
  x_training[nsim, ] = x_train
  y_training[nsim, ] = y_train

  g_model = lm(y_train ~ ., data.frame(x = x_train))
  training_gs[nsim, ] = coef(g_model)

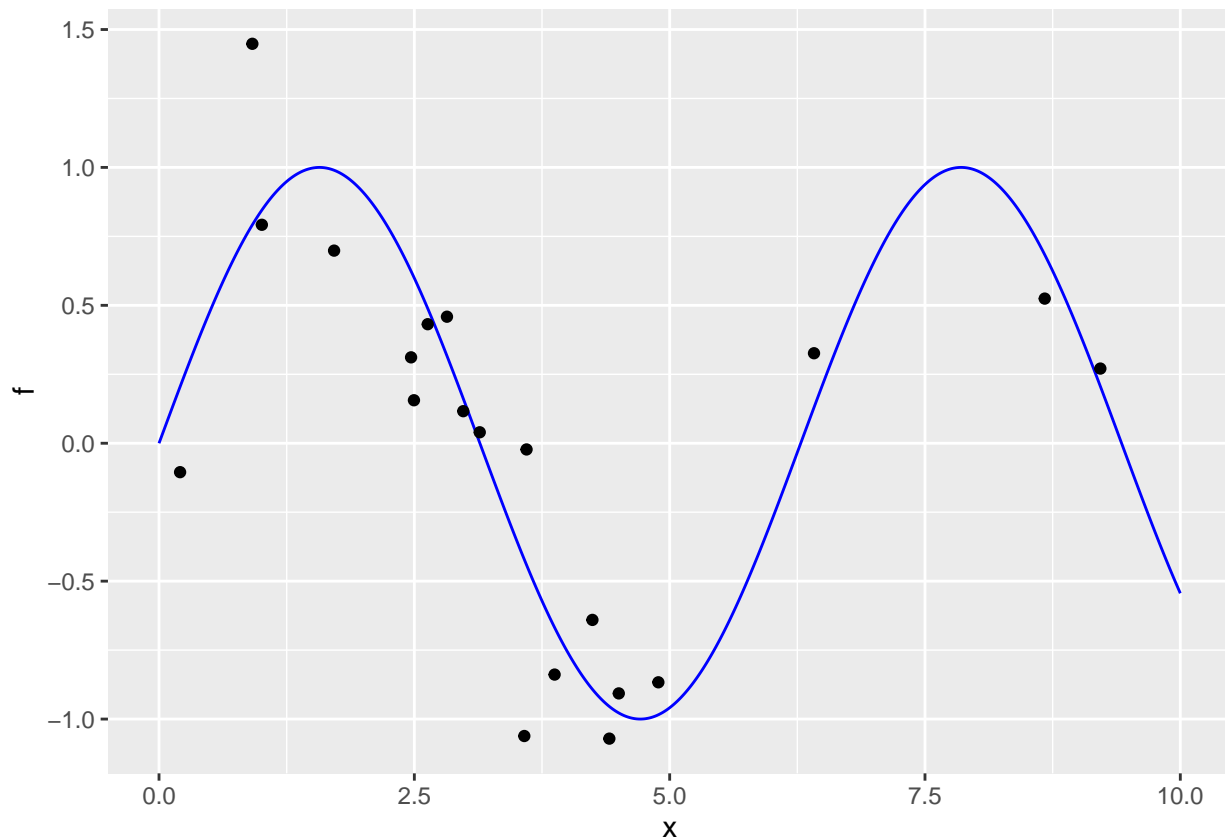
  x_test = runif(n_test, x_min, x_max)
  delta_test = rnorm(n_test, 0, sigma)
```

```

y_test = f_x(x_test) + delta_test
y_hat_test = predict(g_model, data.frame(x = x_test))
all_oos_residuals[nsim, ] = y_test - y_hat_test
}

pacman::p_load(ggplot2)
resolution = 10000
x = seq(x_min, x_max, length.out = resolution)
f_x_2 = data.frame(x = x, f = f_x(x))
ggplot(f_x_2, aes(x, f)) +
  geom_line(col = "blue") +
  geom_point(aes(x, y), data = data.frame(x = x_training[1, ], y = y_training[1, ]))

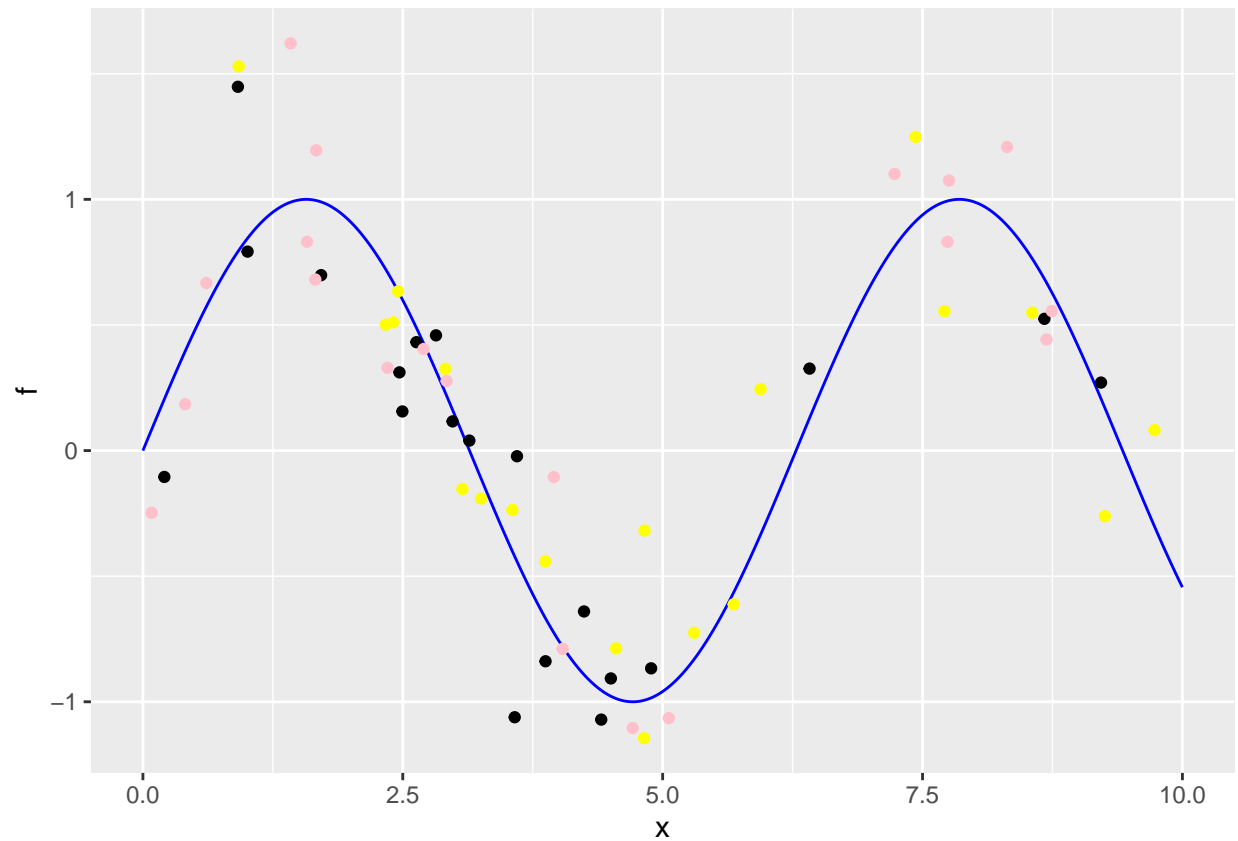
```



```

ggplot(f_x_2, aes(x, f)) +
  geom_line(col = "blue") +
  geom_point(aes(x, y), data = data.frame(x = x_training[1, ], y = y_training[1, ]), col = "black") +
  geom_point(aes(x, y), data = data.frame(x = x_training[2, ], y = y_training[2, ]), col = "yellow") +
  geom_point(aes(x, y), data = data.frame(x = x_training[3, ], y = y_training[3, ]), col = "pink")

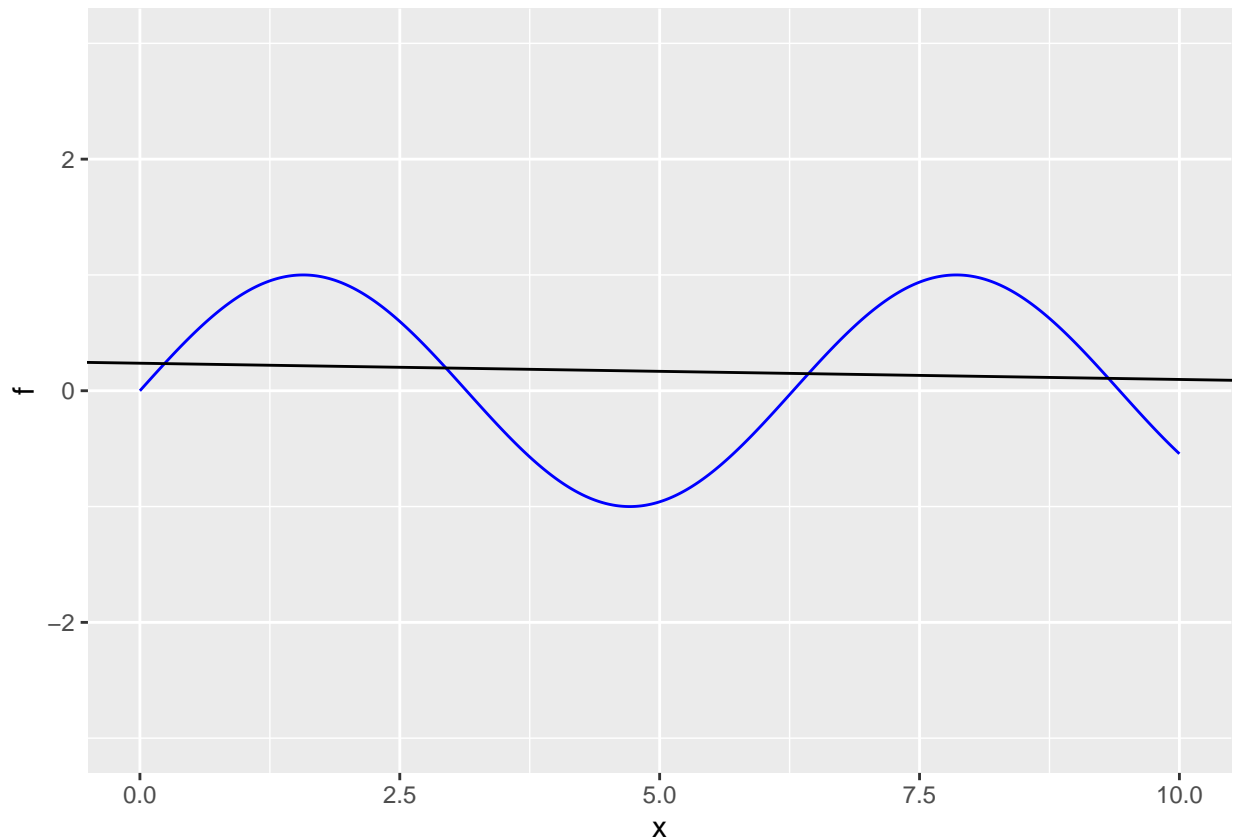
```



```
MSE = mean(c(all_oos_residuals)^2)
MSE
```

```
## [1] 0.5843797
```

```
g_average = colMeans(training_gs)
ggplot(f_x_2, aes(x, f)) +
  geom_line(col = "blue") +
  geom_abline(intercept = g_average[1], slope = g_average[2], col = "black") +
  ylim(-3, 3)
```



```
x = seq(x_min, x_max, length.out = resolution)
g_avg_x = g_average[1] + g_average[2] * x
f = sin(x)
biases = f - g_avg_x
expe_bias_g_sq = mean(biases^2)
expe_bias_g_sq
```

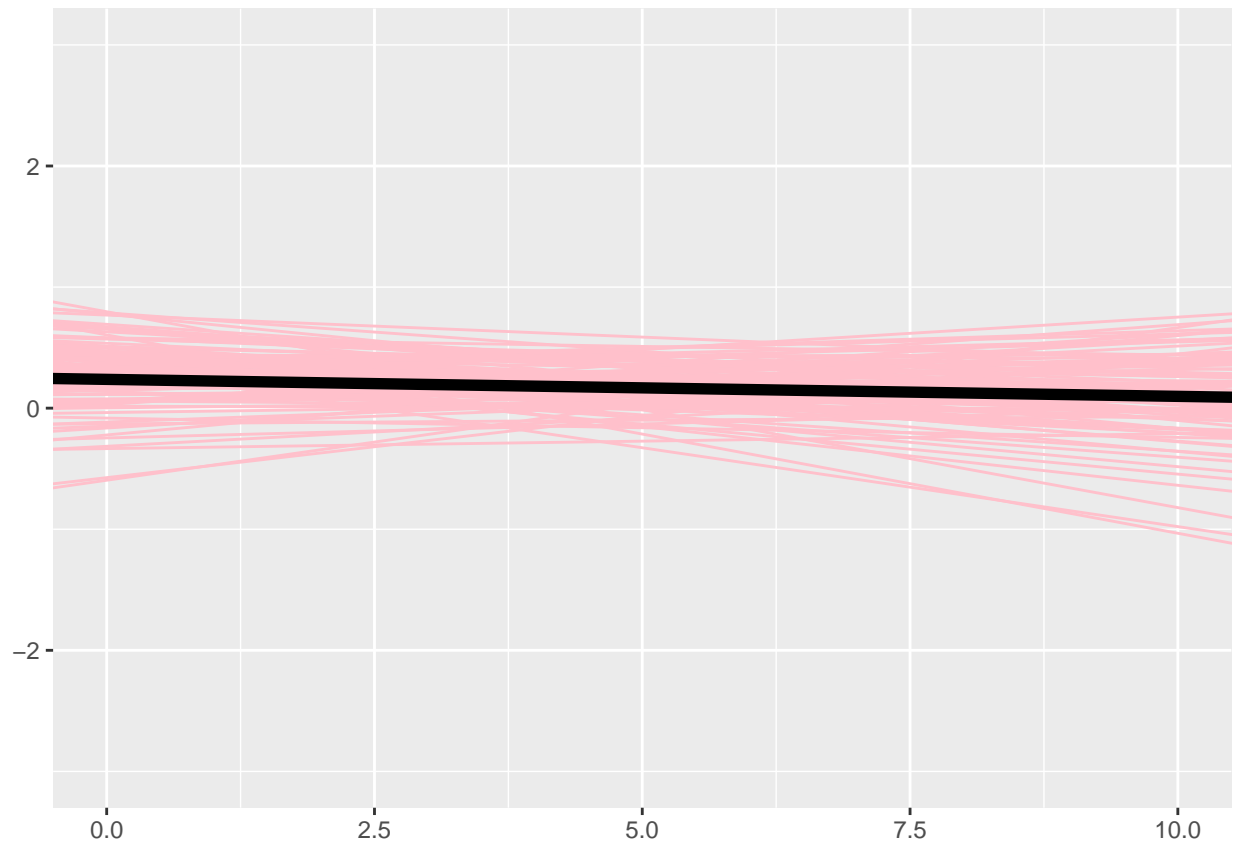
```
## [1] 0.4414531
```

```
plot_obj = ggplot() +
  xlim(x_min, x_max) + ylim(x_min^2, x_max^2)

for (nsim in 1 : min(Nsim, 100)){ #otherwise takes too long
  plot_obj = plot_obj + geom_abline(intercept = training_gs[nsim, 1], slope = training_gs[nsim, 2], col
}

plot_obj +
  geom_abline(intercept = g_average[1], slope = g_average[2], col = "black", lwd = 2) +
  ylim(-3,3)
```

```
## Scale for 'y' is already present. Adding another scale for 'y', which will
## replace the existing scale.
```



```
# geom_line(data = f_x_df, aes(x, f), col = "green", size = 1)

x = seq(x_min, x_max, length.out = resolution)
expe_g_x = g_average[1] + g_average[2] * x
var_x_s = array(NA, Nsim)
for (nsim in 1 : Nsim){
  g_x = training_gs[nsim, 1] + training_gs[nsim, 2] * x
  var_x_s[nsim] = mean((g_x - expe_g_x)^2)
}
expe_var_g = mean(var_x_s)
expe_var_g
```

```
## [1] 0.05249671
```

```
MSE
```

```
## [1] 0.5843797
```

```
sigma^2
```

```
## [1] 0.09
```



```
expe_bias_g_sq
```

```
## [1] 0.4414531
```

```
expe_var_g
```

```
## [1] 0.05249671
```

```
sigma^2 + expe_bias_g_sq + expe_var_g
```

```
## [1] 0.5839498
```

```
rm(list = ls())
```

Take a sample of $n = 2000$ observations from the diamonds data.

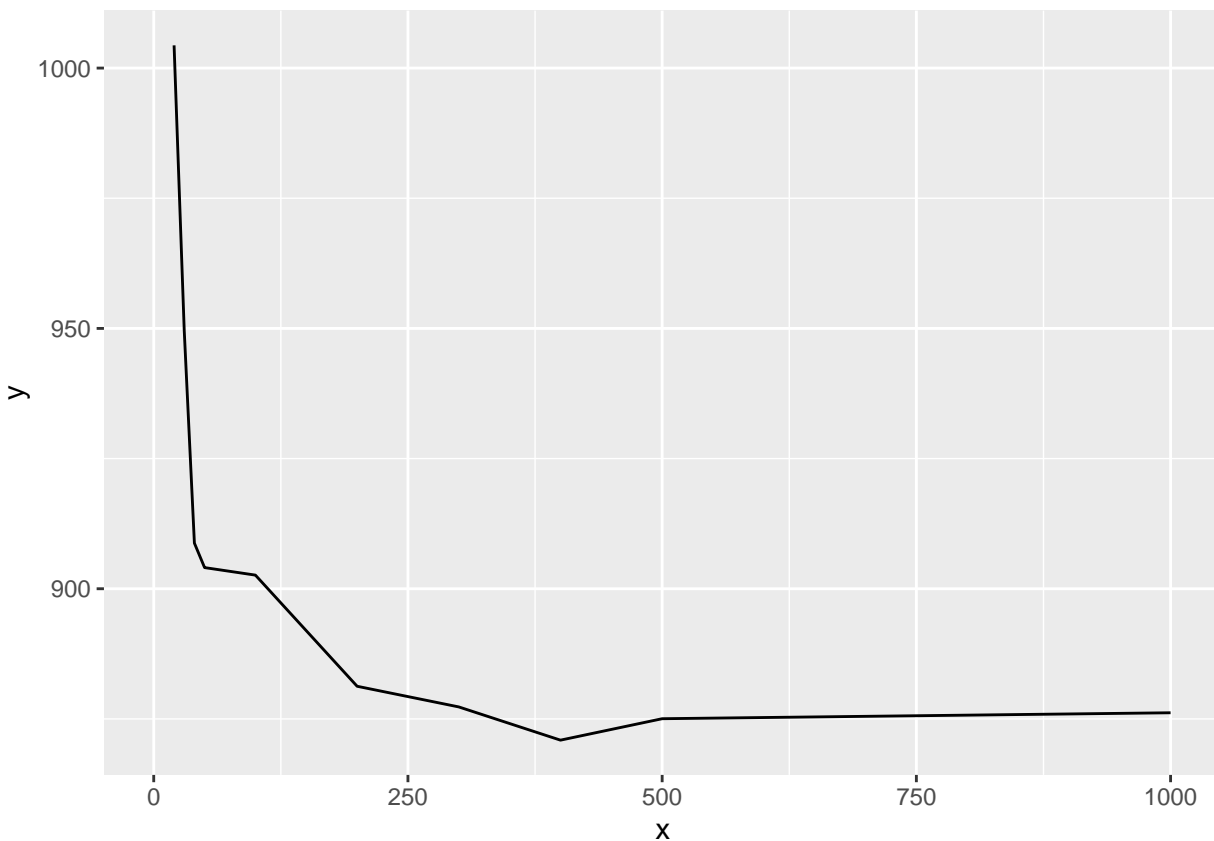
```
pacman::p_load(dplyr)
diamond_samp = diamonds %>%
  sample_n(2000)
```

find the bootstrap s_e for a RF model using 1, 2, 5, 10, 20, 30, 40, 50, 100, 200, 300, 400, 500, 1000 trees. If you are using the `randomForest` package, you can calculate oob residuals via $e_{oob} = y_{train} - rf_mod\$predicted$.

```
num_trees = c(1, 2, 5, 10, 20, 30, 40, 50, 100, 200, 300, 400, 500, 1000)
oob_se_by_num_trees = array(NA, length(num_trees))
for(i in 1:length(num_trees)){
  rf_mod = randomForest(price ~., data = diamond_samp, ntree = num_trees[i])
  oob_se_by_num_trees[i] = sd(diamond_samp$price - rf_mod$predicted)
}

ggplot(data.frame(x = num_trees, y = oob_se_by_num_trees)) +
  geom_line(aes(x = x, y = y))
```

```
## Warning: Removed 4 row(s) containing missing values (geom_path).
```



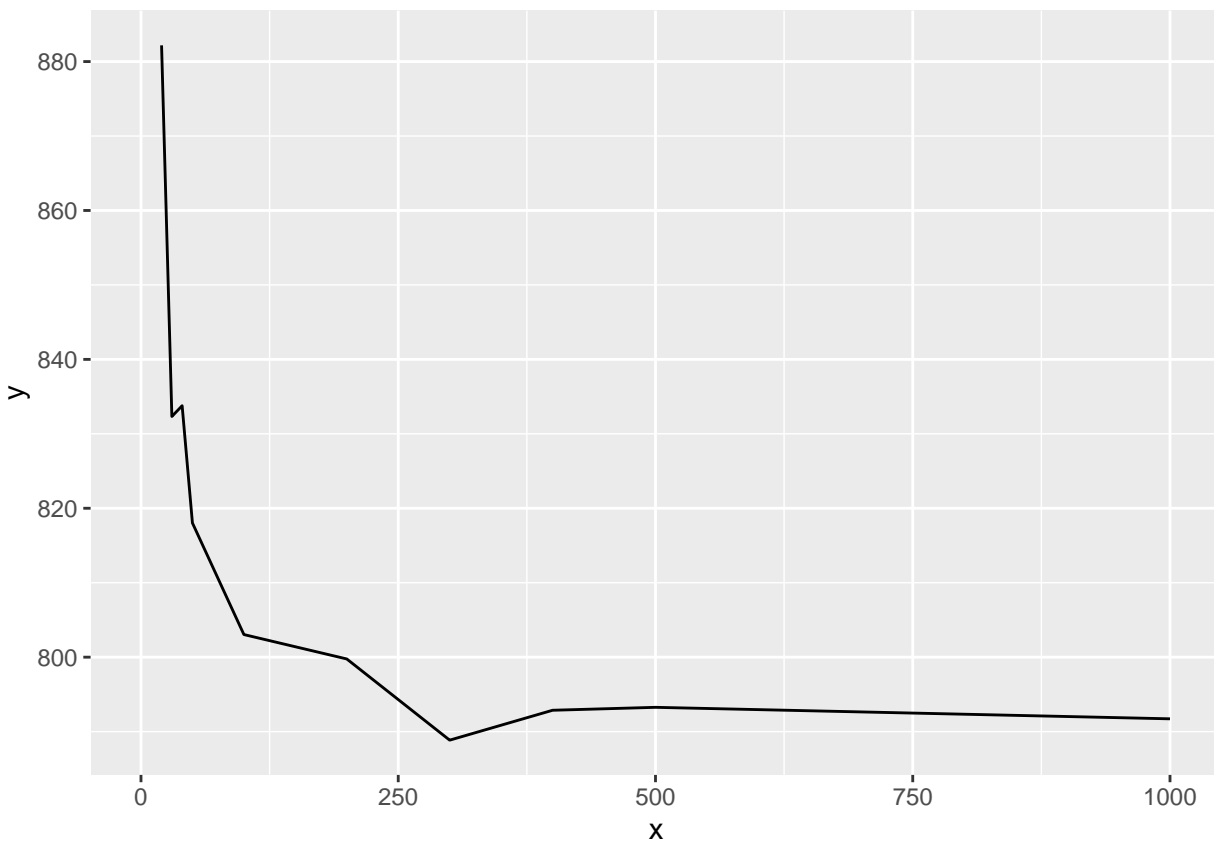
Using the diamonds data, find the bootstrap oob s_e for a bagged-tree model using 1, 2, 5, 10, 20, 30, 40, 50, 100, 200, 300, 400, 500, 1000 trees. If you are using the `randomForest` package, you can create the bagged tree model via setting an argument within the RF constructor function.

```
num_trees = c(1, 2, 5, 10, 20, 30, 40, 50, 100, 200, 300, 400, 500, 1000)
oob_se_by_num_trees_bag = array(NA, length(num_trees))

for(i in 1:length(num_trees)){
  rf_mod = randomForest(price ~., data = diamond_samp, ntree = num_trees[i], mtry = ncol(diamond_samp)
  oob_se_by_num_trees_bag[i] = sd(diamond_samp$price - rf_mod$predicted)
}

ggplot(data.frame(x = num_trees, y = oob_se_by_num_trees_bag)) +
  geom_line(aes(x = x, y = y))
```

```
## Warning: Removed 4 row(s) containing missing values (geom_path).
```



What is the percentage gain / loss in performance of the RF model vs bagged trees model?

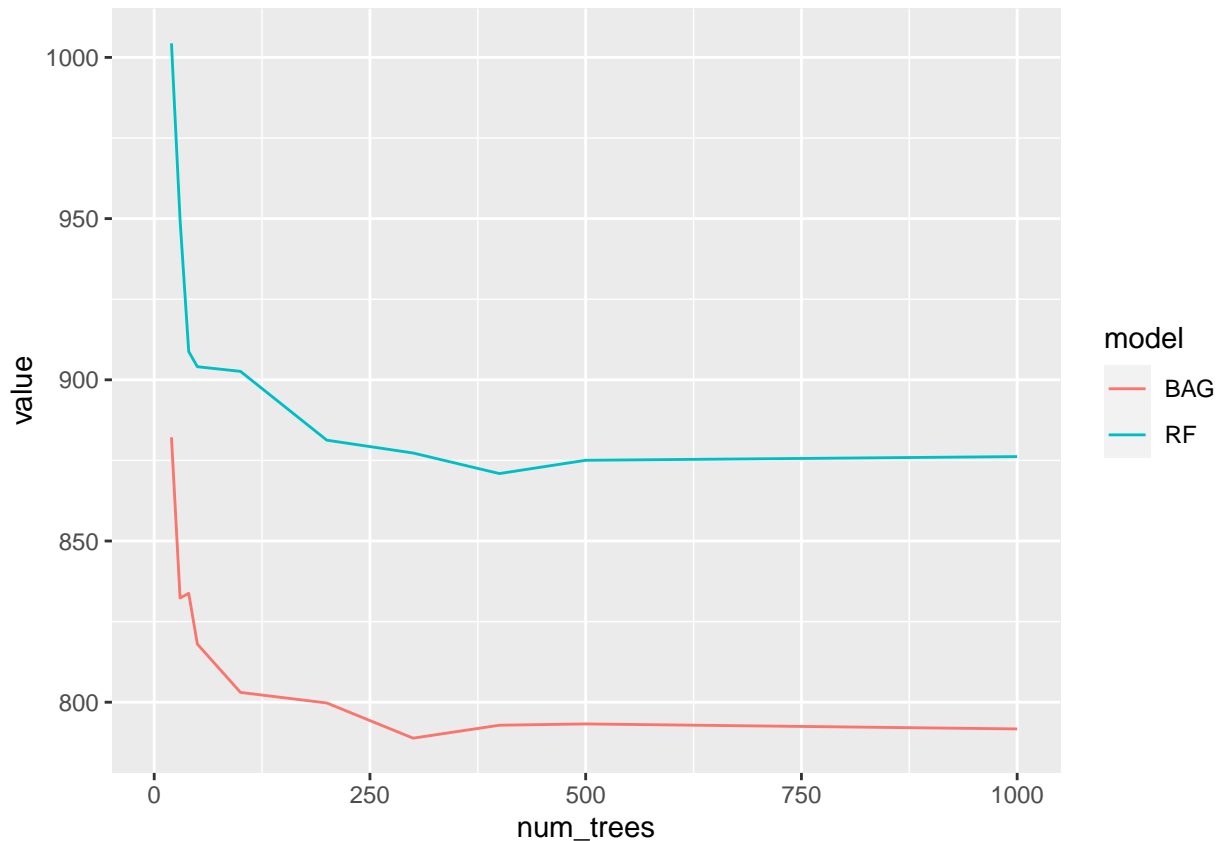
```
(oob_se_by_num_trees - oob_se_by_num_trees_bag) / oob_se_by_num_trees_bag * 100
```

```
## [1]      NA      NA      NA      NA 13.850269 14.085703  8.994056
## [8] 10.517395 12.400595 10.192240 11.212516  9.844600 10.309096 10.668566
```

Plot bootstrap s_e by number of trees for both RF and bagged trees.

```
ggplot(rbind(data.frame(num_trees = num_trees, value = oob_se_by_num_trees, model = "RF"), data.frame(n
  geom_line(aes(x = num_trees, y = value, color = model))
```

```
## Warning: Removed 8 row(s) containing missing values (geom_path).
```

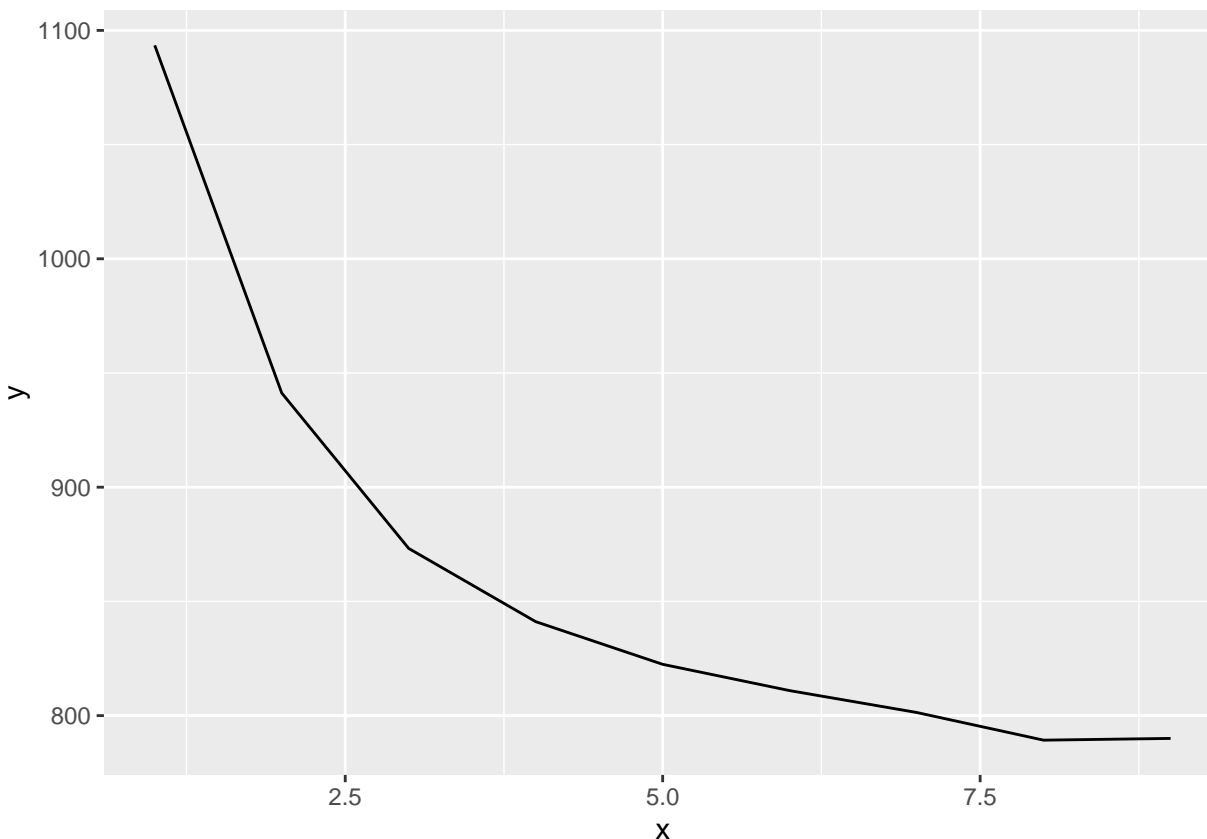


Build RF models for 500 trees using different `mtry` values: 1, 2, ... the maximum. That maximum will be the number of features assuming that we do not binarize categorical features if you are using `randomForest` or the number of features assuming binarization of the categorical features if you are using `YARF`. Calculate bootstrap `s_e` for all `mtry` values.

```
mtrys = 1:(ncol(diamond_samp) - 1)
oob_se_by_mtrys = array(NA, length(mtrys))

for(i in 1:length(mtrys)){
  rf_mod = randomForest(price ~., data = diamond_samp, mtry = mtrys[i])
  oob_se_by_mtrys[i] = sd(diamond_samp$price - rf_mod$predicted)
}

ggplot(data.frame(x = mtrys, y = oob_se_by_mtrys)) +
  geom_line(aes(x = x, y = y))
```



```
rm(list = ls())
```

Take a sample of $n = 2000$ observations from the adult data.

```
pacman::p_load_gh("coatless/ucidata")
data(adult)
adult = na.omit(adult) #kill any observations with missingness
adult_samp = adult %>%
  sample_n(2000)
```

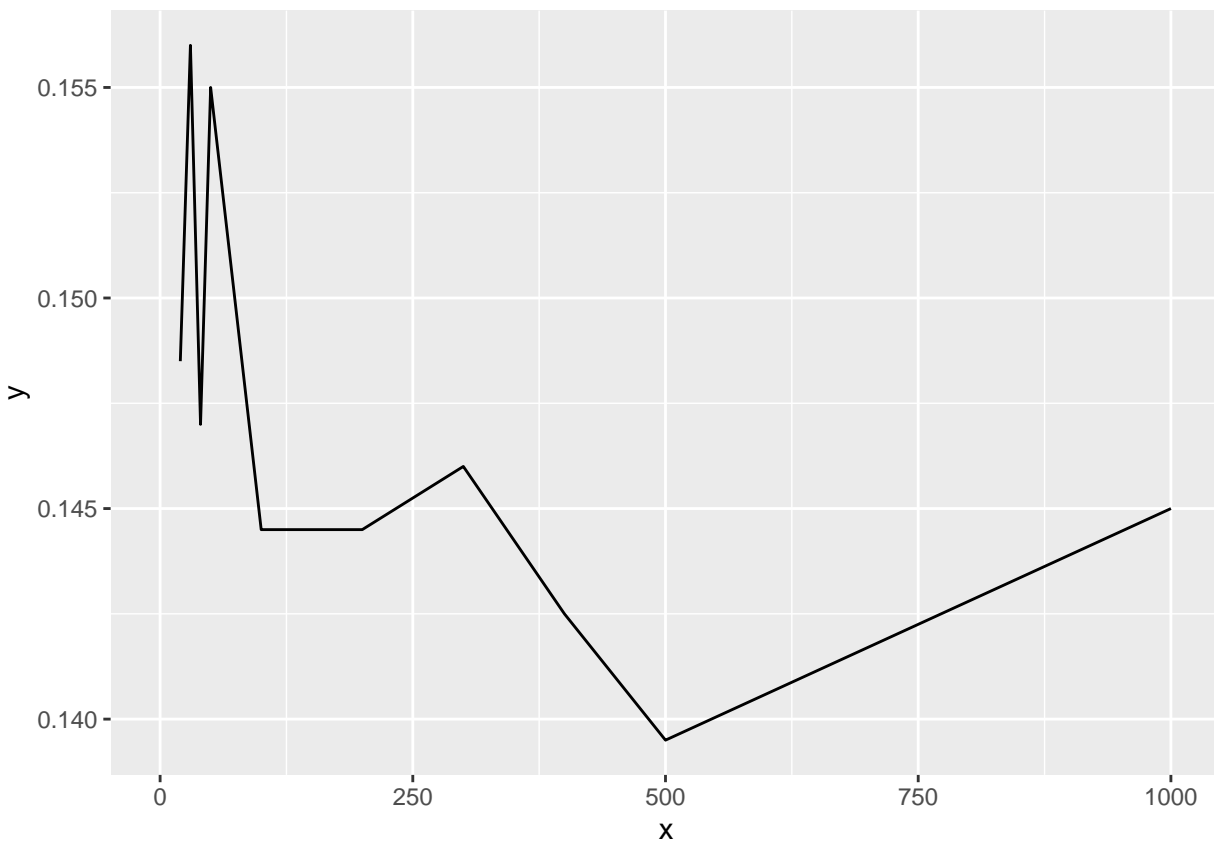
Using the adult data, find the bootstrap misclassification error for an RF model using 1, 2, 5, 10, 20, 30, 40, 50, 100, 200, 300, 400, 500, 1000 trees.

```
num_trees = c(1, 2, 5, 10, 20, 30, 40, 50, 100, 200, 300, 400, 500, 1000)
oob_me_by_num_trees = array(NA, length(num_trees))

for(i in 1:length(num_trees)){
  rf_mod = randomForest(income ~., data = adult_samp, ntree = num_trees[i])
  oob_me_by_num_trees[i] = mean(adult_samp$income != rf_mod$predicted)
}

ggplot(data.frame(x = num_trees, y = oob_me_by_num_trees)) +
  geom_line(aes(x = x, y = y))
```

```
## Warning: Removed 4 row(s) containing missing values (geom_path).
```



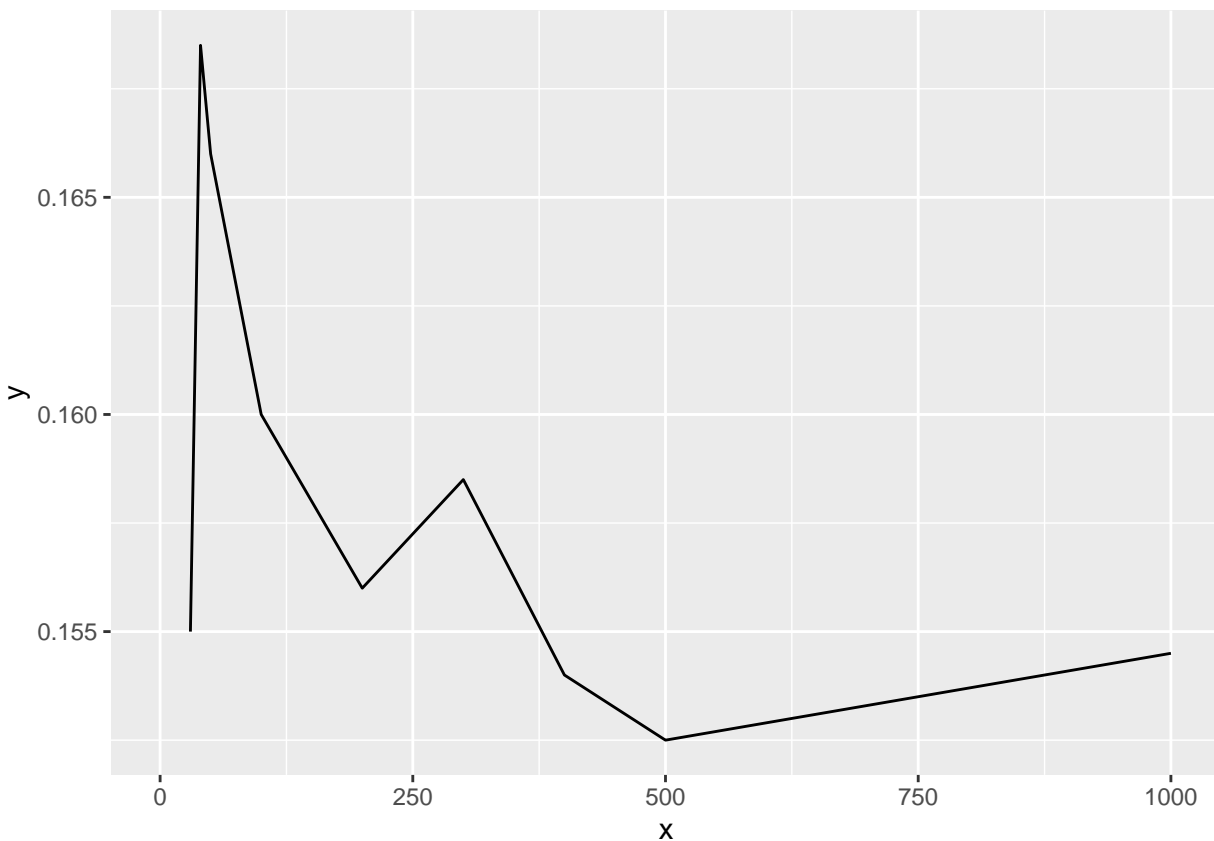
Using the adult data, find the bootstrap misclassification error for a bagged-tree model using 1, 2, 5, 10, 20, 30, 40, 50, 100, 200, 300, 400, 500, 1000 trees.

```
oob_me_by_num_trees_bag = array(NA, length(num_trees))

for(i in 1:length(num_trees)){
  rf_mod = randomForest(income ~., data = adult_samp, ntree = num_trees[i], mtry = ncol(adult) - 1)
  oob_me_by_num_trees_bag[i] = mean(adult_samp$income != rf_mod$predicted)
}

ggplot(data.frame(x = num_trees, y = oob_me_by_num_trees_bag)) +
  geom_line(aes(x = x, y = y))
```

```
## Warning: Removed 5 row(s) containing missing values (geom_path).
```



What is the percentage gain / loss in performance of the RF model vs bagged trees model?

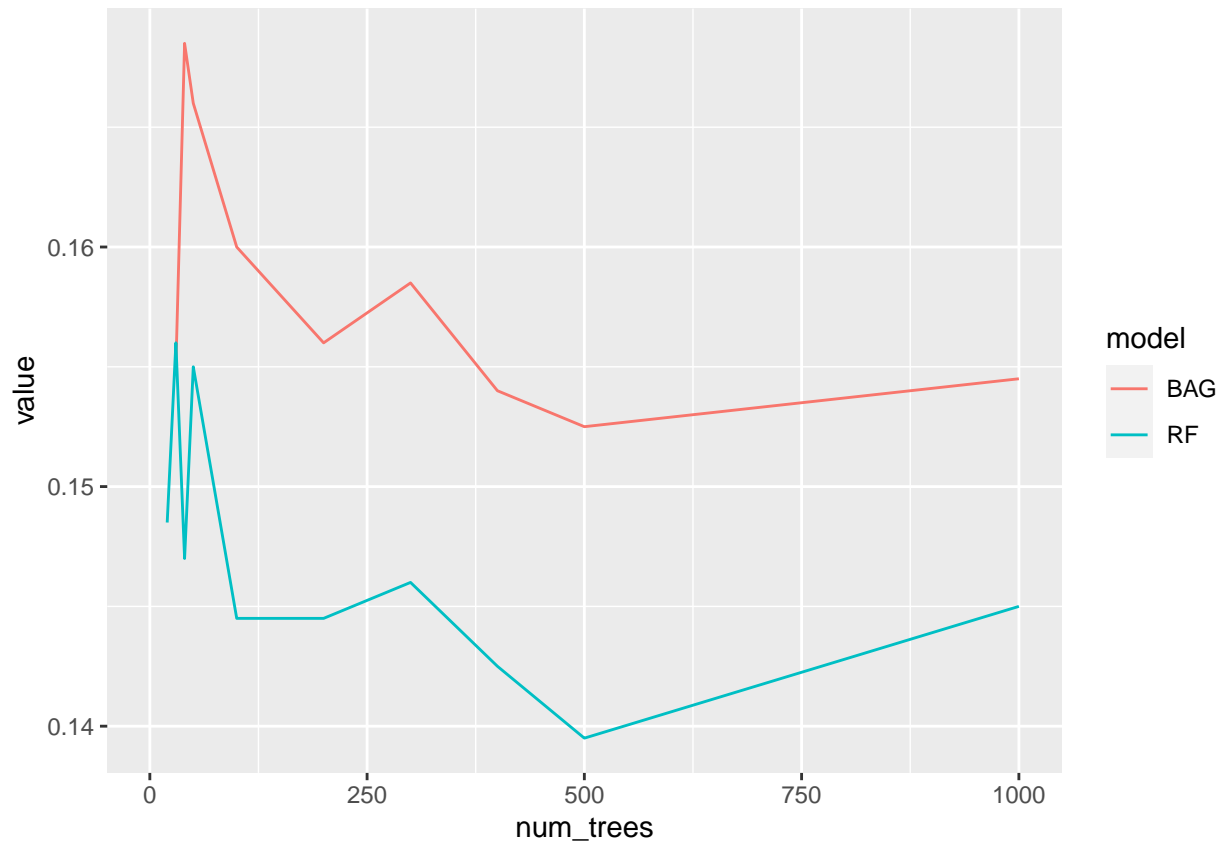
```
(oob_me_by_num_trees - oob_me_by_num_trees_bag) / (oob_me_by_num_trees_bag * 100)
```

```
## [1] NA NA NA NA NA
## [6] 6.451613e-05 -1.275964e-03 -6.626506e-04 -9.687500e-04 -7.371795e-04
## [11] -7.886435e-04 -7.467532e-04 -8.524590e-04 -6.148867e-04
```

Plot bootstrap misclassification error by number of trees for both RF and bagged trees.

```
ggplot(rbind(data.frame(num_trees = num_trees, value = oob_me_by_num_trees, model = "RF"), data.frame(num_trees = num_trees, value = oob_me_by_num_trees_bag, model = "bagged")),
  geom_line(aes(x = num_trees, y = value, color = model)))
```

```
## Warning: Removed 9 row(s) containing missing values (geom_path).
```

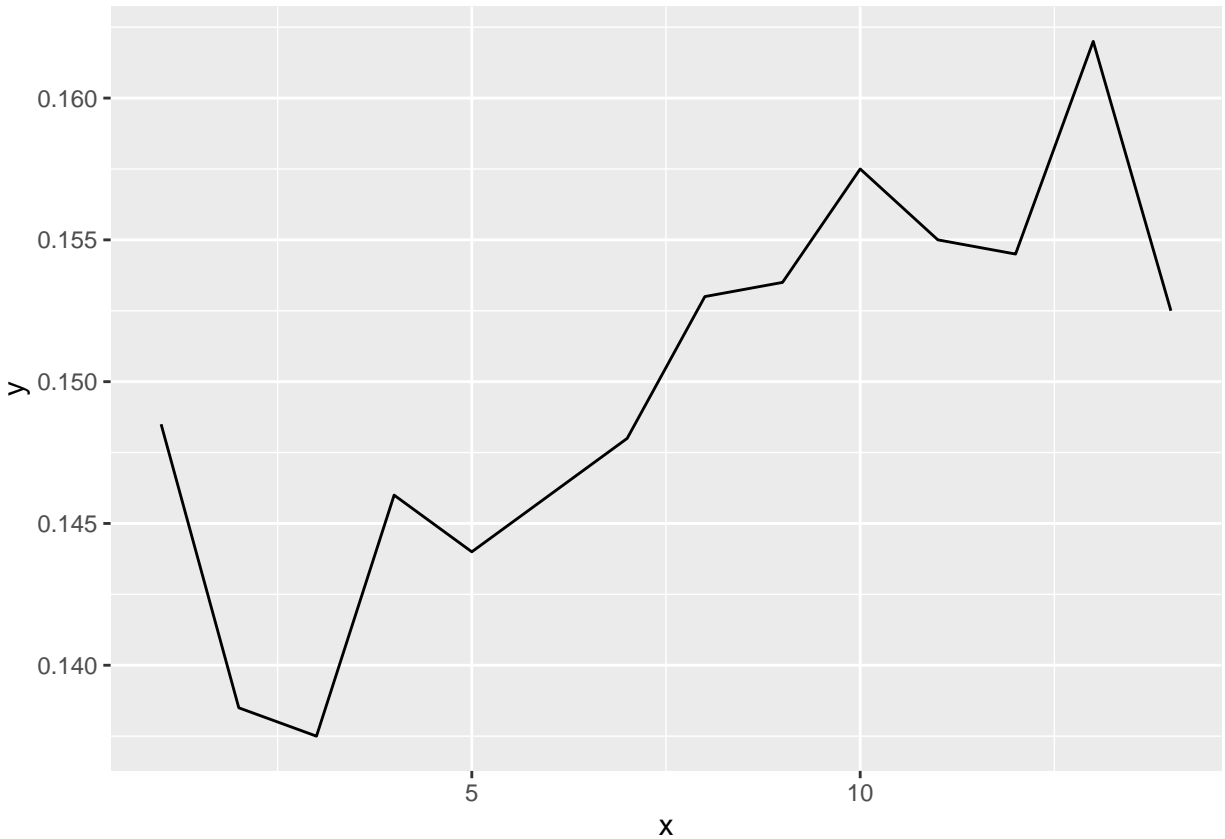


Build RF models for 500 trees using different `mtry` values: 1, 2, ... the maximum (see above as maximum is defined by the specific RF algorithm implementation).

```
mtrys = 1:(ncol(adult_samp) - 1)
oob_me_by_mtrys = array(NA, length(mtrys))

for(i in 1:length(mtrys)){
  rf_mod = randomForest(income ~., data = adult_samp, mtry = mtrys[i])
  oob_me_by_mtrys[i] = mean(adult_samp$income != rf_mod$predicted)
}

ggplot(data.frame(x = mtrys, y = oob_me_by_mtrys)) +
  geom_line(aes(x = x, y = y))
```

```
rm(list = ls())
```

Write a function `random_bagged_ols` which takes as its arguments `X` and `y` with further arguments `num_ols_models` defaulted to 100 and `mtry` defaulted to `NULL` which then gets set within the function to be 50% of available features. This argument builds an OLS on a bootstrap sample of the data and uses only `mtry < p` of the available features. The function then returns all the `lm` models as a list with size `num_ols_models`.

```
random_bagged_ols = function(X, y, num_ols_model = 100, mtry = NULL){
  lm_models = array(NA, num_ols_model)

  for(i in 1:num_ols_model){
    columnsNum = round(runif(1, min = 1, max = ncol(X)))

    X_training = X[, sample(ncol(X), columnsNum)]

    n_not = round(runif(1, min = 1, max = nrow(X)))
    n = round(runif(1, min = 1, max = nrow(X)))

    matrix_X = X_training[0:n_not]

    for(i in 1:n_not){
      matrix_X[i,] = X_training[n[i],]
    }
  }
}
```

```

vector_y = array(NA, n_not)

for(i in 1:n_not){
  vector_y[i] = y[n[i]]
}

models = lm(vector ~. + 0, data.frame(matrix_X))
lm_models[i] =c(models$coefficients)
}
lm_models
}

```

Load up the Boston Housing Data and separate into X and y.

```

#From lab 10
pacman::p_load(MASS)
y = Boston$medv
X = Boston[, 1:13]

```

Similar to lab 1, write a function that takes a matrix and punches holes (i.e. sets entries equal to NA) randomly with an argument `prob_missing`.

```

#From lab 10
punching_holes = function(X, prob_missing){
  for(i in 1:nrow(X)){
    for(j in 1:ncol(X)) {
      if(runif(1) < prob_missing) {
        X[i,j] = NA
      }
    }
  }
  X
}

```

Create a matrix `Xmiss` which is `X` but has missingness with probability of 10%.

```

Xmiss = punching_holes(X, .10)
Xmiss

```

```

##      crim    zn indus chas    nox    rm   age      dis rad tax ptratio  black
## 1  0.00632 18.0  2.31    0 0.5380 6.575 65.2  4.0900  NA  296    15.3 396.90
## 2      NA   0.0  7.07    0    NA 6.421   NA  4.9671   2 242      NA    NA
## 3  0.02729  0.0  7.07    0    NA 7.185 61.1  4.9671   2 242    17.8 392.83
## 4  0.03237  0.0  2.18    0 0.4580   NA   NA  6.0622   3 222    18.7    NA
## 5  0.06905  0.0  2.18    0 0.4580 7.147 54.2  6.0622   3 222    18.7 396.90
## 6  0.02985  0.0  2.18    0 0.4580 6.430 58.7  6.0622  NA 222    18.7 394.12
## 7  0.08829 12.5  7.87    0 0.5240 6.012 66.6  5.5605   5 311    15.2 395.60
## 8  0.14455 12.5  7.87   NA 0.5240 6.172 96.1  5.9505  NA 311    15.2 396.90
## 9  0.21124 12.5  7.87   NA 0.5240 5.631 100.0 6.0821   5 311    15.2 386.63
## 10 0.17004 12.5  7.87    0 0.5240 6.004 85.9  6.5921   5 311    15.2 386.71
## 11 0.22489 12.5  7.87    0 0.5240 6.377 94.3  6.3467  NA 311    15.2 392.52
## 12      NA 12.5  7.87    0 0.5240 6.009 82.9  6.2267   5 311    15.2 396.90

```

## 13	0.09378	12.5	7.87	0	0.5240	5.889	39.0	NA	5	311	15.2	390.50
## 14	NA	0.0	8.14	0	0.5380	5.949	61.8	4.7075	4	307	21.0	396.90
## 15	0.63796	0.0	8.14	0	NA	6.096	84.5	4.4619	NA	307	21.0	380.02
## 16	NA	0.0	8.14	0	NA	5.834	56.5	4.4986	NA	307	21.0	395.62
## 17	1.05393	0.0	NA	0	0.5380	5.935	29.3	4.4986	4	307	21.0	386.85
## 18	NA	0.0	NA	0	0.5380	NA	81.7	4.2579	4	307	21.0	386.75
## 19	0.80271	0.0	8.14	NA	0.5380	5.456	36.6	3.7965	4	307	21.0	288.99
## 20	NA	0.0	8.14	0	0.5380	5.727	69.5	3.7965	NA	NA	21.0	390.95
## 21	1.25179	NA	8.14	0	0.5380	5.570	98.1	3.7979	4	307	21.0	376.57
## 22	0.85204	0.0	8.14	0	0.5380	5.965	89.2	4.0123	4	307	21.0	392.53
## 23	1.23247	0.0	8.14	0	0.5380	6.142	91.7	3.9769	4	307	21.0	396.90
## 24	0.98843	0.0	8.14	0	0.5380	5.813	100.0	NA	NA	NA	21.0	394.54
## 25	0.75026	0.0	8.14	0	0.5380	NA	NA	4.3996	4	307	21.0	394.33
## 26	0.84054	0.0	8.14	NA	NA	5.599	85.7	4.4546	4	307	21.0	NA
## 27	NA	NA	8.14	0	0.5380	5.813	90.3	4.6820	4	307	21.0	NA
## 28	0.95577	0.0	8.14	0	0.5380	6.047	88.8	4.4534	4	307	21.0	306.38
## 29	0.77299	NA	NA	0	0.5380	6.495	94.4	4.4547	NA	307	21.0	387.94
## 30	1.00245	0.0	8.14	0	0.5380	6.674	87.3	4.2390	NA	307	21.0	380.23
## 31	1.13081	0.0	8.14	0	0.5380	5.713	94.1	4.2330	4	307	NA	360.17
## 32	1.35472	0.0	8.14	NA	0.5380	6.072	100.0	4.1750	4	307	21.0	376.73
## 33	1.38799	NA	8.14	0	0.5380	5.950	82.0	3.9900	4	NA	21.0	232.60
## 34	1.15172	0.0	8.14	0	0.5380	5.701	95.0	3.7872	4	307	21.0	358.77
## 35	1.61282	0.0	NA	0	0.5380	6.096	96.9	3.7598	4	307	21.0	248.31
## 36	0.06417	0.0	5.96	0	0.4990	5.933	68.2	3.3603	5	279	NA	396.90
## 37	0.09744	0.0	NA	0	0.4990	5.841	61.4	3.3779	5	279	19.2	NA
## 38	0.08014	0.0	5.96	0	0.4990	5.850	41.5	3.9342	5	279	19.2	396.90
## 39	0.17505	0.0	5.96	0	0.4990	5.966	30.2	3.8473	5	279	NA	393.43
## 40	0.02763	NA	2.95	0	NA	NA	21.8	NA	3	252	18.3	395.63
## 41	0.03359	75.0	2.95	0	0.4280	7.024	15.8	5.4011	3	252	18.3	395.62
## 42	NA	0.0	6.91	0	0.4480	6.770	2.9	5.7209	3	NA	17.9	385.41
## 43	0.14150	0.0	6.91	0	0.4480	6.169	6.6	5.7209	3	233	17.9	383.37
## 44	0.15936	0.0	6.91	0	0.4480	6.211	6.5	5.7209	3	233	17.9	394.46
## 45	0.12269	0.0	6.91	0	0.4480	6.069	40.0	5.7209	NA	233	17.9	389.39
## 46	0.17142	0.0	6.91	0	0.4480	5.682	33.8	NA	3	233	17.9	396.90
## 47	NA	0.0	6.91	0	0.4480	5.786	33.3	5.1004	3	233	17.9	396.90
## 48	0.22927	0.0	6.91	0	0.4480	6.030	85.5	NA	3	233	NA	392.74
## 49	NA	0.0	6.91	0	0.4480	5.399	NA	5.8700	3	233	17.9	396.90
## 50	0.21977	0.0	NA	0	0.4480	5.602	62.0	6.0877	3	NA	17.9	396.90
## 51	0.08873	21.0	5.64	0	0.4390	5.963	45.7	6.8147	4	243	NA	395.56
## 52	0.04337	NA	5.64	NA	NA	6.115	63.0	6.8147	NA	NA	16.8	393.97
## 53	0.05360	21.0	5.64	0	0.4390	6.511	21.1	6.8147	4	243	NA	396.90
## 54	0.04981	21.0	NA	0	NA	5.998	21.4	6.8147	4	243	16.8	396.90
## 55	0.01360	75.0	4.00	0	NA	5.888	47.6	7.3197	3	469	NA	396.90
## 56	0.01311	90.0	1.22	0	0.4030	7.249	21.9	8.6966	5	226	17.9	NA
## 57	0.02055	85.0	0.74	0	NA	6.383	35.7	9.1876	2	313	17.3	NA
## 58	NA	100.0	1.32	0	NA	6.816	40.5	8.3248	5	256	15.1	392.90
## 59	0.15445	25.0	5.13	0	0.4530	6.145	29.2	NA	8	284	NA	390.68
## 60	0.10328	25.0	5.13	0	NA	5.927	47.2	6.9320	8	284	19.7	396.90
## 61	0.14932	25.0	NA	0	NA	5.741	66.2	7.2254	8	284	19.7	395.11
## 62	0.17171	25.0	5.13	0	0.4530	5.966	93.4	6.8185	8	284	19.7	378.08
## 63	0.11027	NA	5.13	0	0.4530	6.456	67.8	7.2255	8	284	19.7	396.90
## 64	0.12650	25.0	5.13	0	0.4530	6.762	43.4	7.9809	8	284	NA	NA
## 65	0.01951	17.5	1.38	NA	0.4161	7.104	59.5	9.2229	3	216	18.6	393.24
## 66	0.03584	80.0	NA	0	0.3980	NA	NA	6.6115	4	337	16.1	396.90

## 67	0.04379	80.0	3.37	0	0.3980	5.787	31.1	6.6115	4	337	16.1	396.90
## 68	0.05789	12.5	6.07	0	0.4090	5.878	NA	6.4980	4	345	18.9	396.21
## 69	0.13554	12.5	6.07	0	0.4090	5.594	36.8	6.4980	NA	345	18.9	396.90
## 70	0.12816	12.5	6.07	0	NA	5.885	NA	6.4980	4	345	18.9	NA
## 71	0.08826	0.0	10.81	0	0.4130	6.417	NA	5.2873	4	305	19.2	383.73
## 72	0.15876	0.0	10.81	0	0.4130	NA	17.5	5.2873	4	305	19.2	376.94
## 73	0.09164	0.0	10.81	0	NA	6.065	7.8	NA	4	305	19.2	390.91
## 74	0.19539	0.0	10.81	0	0.4130	6.245	6.2	5.2873	4	305	19.2	377.17
## 75	0.07896	0.0	12.83	0	0.4370	6.273	6.0	4.2515	5	398	18.7	394.92
## 76	0.09512	0.0	12.83	0	0.4370	6.286	45.0	4.5026	5	398	18.7	383.23
## 77	NA	0.0	12.83	0	NA	6.279	74.5	4.0522	5	398	18.7	373.66
## 78	0.08707	0.0	12.83	0	0.4370	6.140	45.8	4.0905	5	398	18.7	386.96
## 79	0.05646	0.0	NA	0	NA	6.232	53.7	5.0141	5	398	18.7	386.40
## 80	0.08387	0.0	12.83	0	0.4370	5.874	36.6	4.5026	5	398	18.7	NA
## 81	NA	25.0	4.86	0	0.4260	6.727	NA	5.4007	NA	281	NA	396.90
## 82	0.04462	25.0	NA	0	0.4260	6.619	70.4	NA	NA	281	19.0	395.63
## 83	0.03659	25.0	4.86	0	0.4260	6.302	32.2	5.4007	4	281	19.0	396.90
## 84	0.03551	25.0	4.86	0	0.4260	6.167	46.7	5.4007	4	281	19.0	390.64
## 85	0.05059	0.0	4.49	0	0.4490	6.389	48.0	4.7794	3	247	NA	396.90
## 86	0.05735	0.0	4.49	0	0.4490	6.630	56.1	4.4377	3	247	18.5	392.30
## 87	0.05188	NA	4.49	0	0.4490	6.015	45.1	NA	3	NA	18.5	NA
## 88	0.07151	0.0	4.49	0	0.4490	6.121	56.8	3.7476	3	247	18.5	395.15
## 89	0.05660	0.0	3.41	0	0.4890	7.007	86.3	3.4217	2	270	17.8	396.90
## 90	0.05302	0.0	3.41	0	0.4890	7.079	63.1	3.4145	2	270	17.8	396.06
## 91	0.04684	0.0	3.41	0	0.4890	6.417	66.1	3.0923	2	NA	17.8	392.18
## 92	0.03932	0.0	NA	0	NA	6.405	73.9	3.0921	2	270	17.8	393.55
## 93	0.04203	28.0	15.04	0	0.4640	6.442	53.6	3.6659	4	270	18.2	395.01
## 94	0.02875	28.0	15.04	0	0.4640	NA	NA	3.6659	NA	270	NA	396.33
## 95	0.04294	NA	15.04	0	0.4640	NA	77.3	3.6150	4	270	18.2	396.90
## 96	0.12204	0.0	2.89	0	0.4450	6.625	57.8	3.4952	NA	276	18.0	NA
## 97	0.11504	0.0	2.89	0	0.4450	6.163	69.6	3.4952	2	276	18.0	391.83
## 98	0.12083	0.0	NA	0	0.4450	8.069	76.0	NA	2	276	18.0	396.90
## 99	0.08187	0.0	2.89	0	NA	7.820	36.9	3.4952	2	276	18.0	393.53
## 100	0.06860	0.0	2.89	0	NA	7.416	62.5	3.4952	2	276	18.0	396.90
## 101	NA	0.0	8.56	0	0.5200	NA	79.9	NA	5	384	20.9	394.76
## 102	0.11432	0.0	8.56	0	0.5200	6.781	NA	2.8561	5	384	NA	395.58
## 103	0.22876	0.0	8.56	0	0.5200	6.405	85.4	2.7147	5	384	20.9	70.80
## 104	0.21161	0.0	8.56	0	0.5200	NA	87.4	2.7147	5	384	20.9	394.47
## 105	0.13960	0.0	8.56	0	0.5200	6.167	90.0	2.4210	5	NA	20.9	392.69
## 106	0.13262	NA	8.56	0	0.5200	5.851	96.7	2.1069	5	384	20.9	394.05
## 107	0.17120	0.0	8.56	0	0.5200	5.836	91.9	2.2110	5	384	20.9	395.67
## 108	0.13117	0.0	8.56	0	0.5200	6.127	85.2	2.1224	5	384	20.9	387.69
## 109	0.12802	0.0	8.56	0	0.5200	6.474	97.1	2.4329	5	NA	20.9	395.24
## 110	0.26363	0.0	8.56	0	0.5200	6.229	91.2	2.5451	5	384	20.9	391.23
## 111	0.10793	NA	8.56	0	NA	6.195	54.4	2.7778	5	384	20.9	393.49
## 112	0.10084	0.0	10.01	0	0.5470	6.715	81.6	2.6775	6	432	17.8	395.59
## 113	0.12329	0.0	10.01	0	NA	5.913	92.9	NA	6	432	17.8	394.95
## 114	0.22212	0.0	10.01	NA	0.5470	6.092	95.4	NA	6	NA	17.8	396.90
## 115	0.14231	0.0	NA	0	0.5470	6.254	84.2	2.2565	6	432	NA	388.74
## 116	0.17134	0.0	10.01	0	0.5470	5.928	88.2	NA	NA	432	17.8	344.91
## 117	0.13158	0.0	10.01	0	0.5470	6.176	72.5	2.7301	6	432	17.8	393.30
## 118	0.15098	0.0	10.01	0	0.5470	6.021	82.6	2.7474	6	432	17.8	394.51
## 119	0.13058	0.0	10.01	0	NA	5.872	73.1	2.4775	6	432	17.8	338.63
## 120	NA	0.0	10.01	0	0.5470	5.731	65.2	2.7592	6	432	17.8	391.50

## 121	0.06899	0.0	25.65	0	0.5810	5.870	69.7	2.2577	2	188	19.1	389.15
## 122	0.07165	0.0	25.65	0	0.5810	6.004	84.1	2.1974	2	188	NA	377.67
## 123	0.09299	0.0	25.65	0	0.5810	5.961	92.9	2.0869	2	188	19.1	378.09
## 124	NA	0.0	25.65	0	0.5810	5.856	97.0	1.9444	2	188	NA	370.31
## 125	0.09849	0.0	25.65	0	0.5810	5.879	95.8	2.0063	2	188	19.1	379.38
## 126	0.16902	0.0	25.65	0	0.5810	NA	88.4	1.9929	2	188	19.1	385.02
## 127	0.38735	0.0	25.65	0	0.5810	5.613	95.6	NA	2	188	19.1	359.29
## 128	0.25915	0.0	21.89	0	0.6240	5.693	96.0	1.7883	4	437	21.2	392.11
## 129	0.32543	0.0	21.89	0	0.6240	6.431	98.8	1.8125	4	437	21.2	396.90
## 130	0.88125	0.0	21.89	0	0.6240	5.637	94.7	1.9799	NA	437	21.2	396.90
## 131	0.34006	NA	21.89	0	0.6240	NA	98.9	2.1185	4	437	21.2	395.04
## 132	1.19294	NA	21.89	0	0.6240	NA	97.7	2.2710	4	437	21.2	396.90
## 133	0.59005	0.0	21.89	0	0.6240	6.372	97.9	2.3274	4	437	21.2	385.76
## 134	0.32982	NA	21.89	0	NA	5.822	95.4	2.4699	4	437	21.2	388.69
## 135	0.97617	0.0	21.89	0	0.6240	5.757	98.4	2.3460	4	437	21.2	262.76
## 136	0.55778	0.0	21.89	NA	0.6240	6.335	98.2	2.1107	4	437	21.2	394.67
## 137	0.32264	0.0	21.89	0	0.6240	5.942	93.5	1.9669	4	437	21.2	378.25
## 138	0.35233	0.0	21.89	0	0.6240	6.454	NA	1.8498	4	437	21.2	394.08
## 139	0.24980	0.0	21.89	0	0.6240	5.857	98.2	1.6686	4	437	21.2	392.04
## 140	0.54452	0.0	21.89	0	0.6240	6.151	97.9	1.6687	4	437	21.2	396.90
## 141	NA	0.0	21.89	0	0.6240	6.174	NA	NA	4	437	NA	388.08
## 142	1.62864	0.0	21.89	0	0.6240	5.019	100.0	1.4394	4	437	NA	396.90
## 143	3.32105	0.0	19.58	1	0.8710	5.403	100.0	1.3216	NA	403	14.7	396.90
## 144	4.09740	0.0	NA	0	0.8710	5.468	100.0	1.4118	5	NA	14.7	396.90
## 145	2.77974	0.0	19.58	0	0.8710	4.903	97.8	1.3459	5	403	14.7	396.90
## 146	2.37934	NA	19.58	0	0.8710	6.130	100.0	1.4191	5	403	14.7	172.91
## 147	2.15505	0.0	19.58	0	0.8710	5.628	100.0	1.5166	5	403	14.7	169.27
## 148	2.36862	0.0	19.58	0	0.8710	4.926	95.7	1.4608	NA	403	14.7	391.71
## 149	2.33099	0.0	19.58	NA	0.8710	5.186	93.8	1.5296	5	NA	14.7	356.99
## 150	2.73397	0.0	19.58	0	0.8710	5.597	94.9	1.5257	5	403	14.7	351.85
## 151	1.65660	0.0	19.58	0	0.8710	6.122	NA	1.6180	5	403	14.7	372.80
## 152	1.49632	0.0	19.58	0	0.8710	5.404	NA	NA	5	403	14.7	341.60
## 153	1.12658	0.0	19.58	1	0.8710	5.012	88.0	1.6102	5	403	14.7	343.28
## 154	NA	0.0	19.58	NA	0.8710	5.709	98.5	1.6232	5	403	14.7	261.95
## 155	1.41385	0.0	19.58	1	0.8710	6.129	96.0	1.7494	5	403	14.7	321.02
## 156	3.53501	0.0	19.58	1	0.8710	6.152	82.6	1.7455	NA	403	14.7	88.01
## 157	2.44668	0.0	19.58	NA	NA	5.272	94.0	1.7364	5	403	14.7	88.63
## 158	1.22358	0.0	19.58	0	0.6050	6.943	97.4	1.8773	5	403	14.7	363.43
## 159	NA	0.0	19.58	NA	0.6050	NA	100.0	1.7573	5	403	14.7	353.89
## 160	1.42502	0.0	19.58	0	0.8710	6.510	100.0	1.7659	5	403	14.7	364.31
## 161	1.27346	0.0	19.58	1	0.6050	6.250	92.6	1.7984	5	NA	NA	338.92
## 162	1.46336	0.0	NA	NA	0.6050	NA	90.8	1.9709	5	403	14.7	374.43
## 163	1.83377	0.0	19.58	1	0.6050	7.802	98.2	2.0407	5	403	14.7	389.61
## 164	1.51902	0.0	19.58	1	0.6050	8.375	93.9	2.1620	5	NA	14.7	388.45
## 165	2.24236	0.0	19.58	0	0.6050	5.854	91.8	2.4220	5	403	14.7	395.11
## 166	2.92400	0.0	19.58	0	0.6050	6.101	93.0	2.2834	5	403	14.7	240.16
## 167	2.01019	0.0	19.58	NA	0.6050	7.929	96.2	NA	5	403	14.7	369.30
## 168	1.80028	0.0	19.58	0	0.6050	5.877	79.2	2.4259	5	403	14.7	227.61
## 169	2.30040	0.0	NA	0	0.6050	6.319	96.1	2.1000	5	403	14.7	297.09
## 170	2.44953	0.0	19.58	0	0.6050	NA	95.2	2.2625	5	403	14.7	330.04
## 171	1.20742	0.0	19.58	0	NA	5.875	94.6	2.4259	NA	403	14.7	292.29
## 172	2.31390	NA	19.58	0	0.6050	5.880	97.3	2.3887	5	403	14.7	348.13
## 173	0.13914	0.0	4.05	0	0.5100	5.572	88.5	2.5961	5	296	16.6	396.90
## 174	0.09178	NA	4.05	0	0.5100	6.416	84.1	2.6463	5	296	16.6	395.50

## 175	0.08447	0.0	4.05	0	0.5100	NA	68.7	2.7019	5	296	16.6	393.23
## 176	0.06664	0.0	4.05	NA	0.5100	6.546	33.1	3.1323	5	296	16.6	NA
## 177	0.07022	0.0	4.05	0	0.5100	6.020	NA	3.5549	NA	296	16.6	393.23
## 178	0.05425	0.0	NA	0	0.5100	NA	NA	3.3175	5	296	16.6	395.60
## 179	0.06642	0.0	4.05	0	0.5100	6.860	74.4	2.9153	NA	296	16.6	391.27
## 180	0.05780	0.0	2.46	0	0.4880	6.980	58.4	NA	3	193	17.8	396.90
## 181	0.06588	0.0	2.46	0	0.4880	7.765	83.3	2.7410	3	193	17.8	395.56
## 182	0.06888	0.0	2.46	0	0.4880	6.144	62.2	NA	3	NA	17.8	396.90
## 183	0.09103	0.0	2.46	0	0.4880	7.155	92.2	NA	3	193	17.8	394.12
## 184	0.10008	0.0	2.46	0	0.4880	6.563	95.6	2.8470	3	NA	17.8	396.90
## 185	0.08308	0.0	2.46	0	0.4880	5.604	89.8	2.9879	NA	193	17.8	391.00
## 186	0.06047	0.0	2.46	NA	0.4880	6.153	68.8	3.2797	3	193	17.8	387.11
## 187	0.05602	0.0	2.46	0	0.4880	7.831	53.6	NA	3	NA	17.8	392.63
## 188	0.07875	45.0	3.44	0	0.4370	6.782	NA	3.7886	NA	398	15.2	393.87
## 189	NA	45.0	3.44	0	0.4370	6.556	29.1	4.5667	NA	398	NA	382.84
## 190	0.08370	45.0	3.44	0	0.4370	7.185	38.9	4.5667	5	398	15.2	396.90
## 191	0.09068	45.0	NA	0	0.4370	6.951	21.5	6.4798	5	398	15.2	377.68
## 192	0.06911	45.0	3.44	0	0.4370	6.739	30.8	6.4798	5	398	15.2	389.71
## 193	NA	NA	3.44	NA	0.4370	7.178	NA	6.4798	5	398	15.2	390.49
## 194	0.02187	60.0	2.93	0	0.4010	6.800	9.9	6.2196	1	NA	15.6	NA
## 195	0.01439	60.0	2.93	0	0.4010	6.604	18.8	6.2196	1	265	15.6	376.70
## 196	0.01381	80.0	0.46	0	0.4220	7.875	32.0	5.6484	4	255	14.4	394.23
## 197	0.04011	80.0	1.52	NA	0.4040	7.287	34.1	7.3090	2	329	12.6	396.90
## 198	0.04666	80.0	1.52	0	0.4040	7.107	36.6	7.3090	2	329	12.6	354.31
## 199	NA	80.0	1.52	0	0.4040	NA	38.3	7.3090	2	329	12.6	392.20
## 200	0.03150	95.0	1.47	0	0.4030	6.975	15.3	7.6534	3	402	17.0	396.90
## 201	0.01778	95.0	1.47	0	0.4030	7.135	13.9	7.6534	NA	402	17.0	384.30
## 202	0.03445	82.5	2.03	0	0.4150	6.162	38.4	6.2700	2	348	14.7	393.77
## 203	0.02177	NA	2.03	0	0.4150	7.610	15.7	NA	2	348	14.7	395.38
## 204	0.03510	95.0	2.68	0	0.4161	7.853	33.2	5.1180	4	224	NA	392.78
## 205	0.02009	95.0	2.68	0	0.4161	8.034	31.9	5.1180	4	224	14.7	390.55
## 206	0.13642	0.0	10.59	0	0.4890	5.891	22.3	3.9454	4	277	NA	396.90
## 207	0.22969	0.0	10.59	0	0.4890	NA	52.5	4.3549	4	277	18.6	394.87
## 208	0.25199	0.0	10.59	0	0.4890	5.783	72.7	4.3549	4	277	18.6	389.43
## 209	0.13587	0.0	10.59	1	0.4890	6.064	59.1	4.2392	4	277	18.6	381.32
## 210	0.43571	0.0	10.59	1	0.4890	5.344	100.0	3.8750	4	NA	18.6	396.90
## 211	0.17446	0.0	10.59	1	0.4890	NA	92.1	3.8771	4	277	18.6	393.25
## 212	0.37578	0.0	10.59	1	0.4890	5.404	88.6	NA	4	277	18.6	395.24
## 213	0.21719	0.0	10.59	NA	0.4890	5.807	NA	3.6526	4	NA	18.6	390.94
## 214	0.14052	0.0	10.59	0	0.4890	6.375	32.3	3.9454	4	277	18.6	385.81
## 215	0.28955	0.0	10.59	0	0.4890	5.412	9.8	3.5875	4	277	18.6	348.93
## 216	0.19802	0.0	10.59	0	0.4890	6.182	42.4	3.9454	4	277	18.6	393.63
## 217	0.04560	0.0	13.89	NA	0.5500	5.888	NA	3.1121	5	NA	16.4	NA
## 218	0.07013	0.0	13.89	0	0.5500	6.642	85.1	3.4211	5	276	NA	392.78
## 219	0.11069	0.0	13.89	1	0.5500	5.951	NA	NA	5	276	16.4	396.90
## 220	0.11425	0.0	NA	1	0.5500	NA	92.4	3.3633	NA	276	16.4	393.74
## 221	0.35809	0.0	6.20	1	0.5070	NA	88.5	2.8617	8	307	17.4	391.70
## 222	0.40771	0.0	6.20	1	0.5070	6.164	91.3	NA	8	307	17.4	395.24
## 223	0.62356	0.0	6.20	1	0.5070	6.879	77.7	3.2721	8	NA	17.4	390.39
## 224	0.61470	0.0	6.20	0	NA	6.618	80.8	3.2721	8	307	17.4	396.90
## 225	0.31533	0.0	6.20	0	0.5040	8.266	78.3	2.8944	8	307	17.4	385.05
## 226	0.52693	0.0	6.20	0	0.5040	8.725	83.0	2.8944	8	307	17.4	382.00
## 227	0.38214	0.0	6.20	0	0.5040	8.040	86.5	3.2157	NA	307	17.4	387.38
## 228	NA	0.0	6.20	0	0.5040	7.163	79.9	3.2157	8	NA	17.4	NA

##	229	0.29819	0.0	NA	0	0.5040	7.686	17.0	3.3751	NA	307	17.4	377.51
##	230	0.44178	0.0	NA	0	0.5040	6.552	21.4	3.3751	8	307	17.4	380.34
##	231	0.53700	0.0	6.20	0	0.5040	5.981	68.1	3.6715	NA	307	17.4	378.35
##	232	0.46296	NA	6.20	0	0.5040	7.412	76.9	NA	8	307	17.4	376.14
##	233	0.57529	0.0	6.20	0	0.5070	8.337	73.3	3.8384	NA	307	17.4	385.91
##	234	0.33147	0.0	NA	0	0.5070	8.247	70.4	3.6519	8	307	NA	378.95
##	235	0.44791	0.0	NA	1	0.5070	NA	66.5	3.6519	8	307	17.4	360.20
##	236	0.33045	0.0	6.20	0	0.5070	6.086	61.5	3.6519	8	307	17.4	376.75
##	237	0.52058	0.0	6.20	1	0.5070	6.631	76.5	4.1480	8	307	17.4	388.45
##	238	0.51183	0.0	NA	0	0.5070	7.358	71.6	4.1480	8	307	17.4	390.07
##	239	0.08244	30.0	4.93	0	0.4280	6.481	18.5	6.1899	6	300	16.6	379.41
##	240	0.09252	30.0	4.93	0	0.4280	6.606	42.2	6.1899	6	300	16.6	383.78
##	241	0.11329	30.0	4.93	0	0.4280	6.897	54.3	6.3361	6	300	16.6	391.25
##	242	0.10612	NA	NA	0	0.4280	6.095	65.1	6.3361	6	NA	16.6	394.62
##	243	0.10290	30.0	4.93	0	0.4280	6.358	52.9	7.0355	6	300	16.6	372.75
##	244	0.12757	30.0	4.93	0	0.4280	6.393	7.8	7.0355	6	300	NA	374.71
##	245	0.20608	22.0	5.86	0	0.4310	5.593	76.5	7.9549	NA	330	19.1	NA
##	246	0.19133	22.0	5.86	0	0.4310	NA	70.2	7.9549	NA	330	19.1	389.13
##	247	0.33983	22.0	5.86	0	0.4310	6.108	34.9	8.0555	7	330	19.1	390.18
##	248	0.19657	22.0	5.86	0	0.4310	6.226	79.2	8.0555	7	330	19.1	376.14
##	249	0.16439	22.0	5.86	0	0.4310	6.433	49.1	7.8265	7	330	19.1	NA
##	250	0.19073	22.0	5.86	0	0.4310	6.718	NA	7.8265	7	330	19.1	393.74
##	251	0.14030	NA	5.86	0	0.4310	6.487	NA	7.3967	7	330	19.1	396.28
##	252	0.21409	22.0	NA	0	0.4310	6.438	8.9	7.3967	7	330	NA	NA
##	253	0.08221	22.0	NA	0	0.4310	6.957	6.8	8.9067	7	330	19.1	386.09
##	254	0.36894	22.0	NA	0	0.4310	8.259	8.4	8.9067	7	330	19.1	396.90
##	255	0.04819	80.0	3.64	0	0.3920	6.108	32.0	9.2203	1	315	16.4	392.89
##	256	0.03548	80.0	NA	0	0.3920	5.876	19.1	9.2203	1	NA	16.4	395.18
##	257	0.01538	90.0	3.75	0	0.3940	7.454	34.2	6.3361	NA	244	15.9	386.34
##	258	0.61154	20.0	NA	0	NA	8.704	86.9	1.8010	5	264	NA	389.70
##	259	0.66351	20.0	3.97	0	0.6470	7.333	100.0	1.8946	5	264	13.0	383.29
##	260	0.65665	20.0	3.97	0	0.6470	6.842	100.0	2.0107	5	NA	13.0	391.93
##	261	0.54011	20.0	NA	0	0.6470	7.203	81.8	2.1121	5	264	13.0	392.80
##	262	0.53412	20.0	3.97	0	0.6470	7.520	NA	2.1398	5	264	13.0	NA
##	263	NA	20.0	3.97	0	0.6470	8.398	91.5	2.2885	5	264	13.0	386.86
##	264	0.82526	20.0	3.97	0	0.6470	7.327	94.5	2.0788	5	264	NA	393.42
##	265	0.55007	20.0	3.97	0	0.6470	NA	NA	1.9301	5	264	13.0	NA
##	266	0.76162	20.0	3.97	0	0.6470	5.560	62.8	1.9865	5	264	13.0	392.40
##	267	0.78570	20.0	3.97	0	0.6470	7.014	84.6	2.1329	5	NA	13.0	NA
##	268	0.57834	NA	NA	0	0.5750	8.297	67.0	2.4216	5	264	13.0	384.54
##	269	0.54050	20.0	NA	0	0.5750	7.470	52.6	2.8720	5	264	13.0	390.30
##	270	0.09065	20.0	6.96	1	0.4640	5.920	61.5	3.9175	3	223	18.6	391.34
##	271	0.29916	20.0	6.96	0	0.4640	5.856	42.1	4.4290	3	223	18.6	388.65
##	272	0.16211	20.0	6.96	0	0.4640	6.240	NA	4.4290	3	223	18.6	396.90
##	273	0.11460	20.0	6.96	0	0.4640	6.538	58.7	3.9175	3	NA	18.6	394.96
##	274	NA	20.0	6.96	1	0.4640	7.691	51.8	4.3665	3	223	18.6	390.77
##	275	0.05644	40.0	6.41	1	NA	NA	32.9	4.0776	4	NA	17.6	396.90
##	276	0.09604	NA	6.41	0	0.4470	NA	42.8	4.2673	4	254	17.6	396.90
##	277	0.10469	40.0	6.41	1	0.4470	7.267	49.0	4.7872	4	254	NA	389.25
##	278	0.06127	NA	6.41	NA	0.4470	6.826	27.6	4.8628	4	254	17.6	393.45
##	279	0.07978	NA	6.41	0	0.4470	6.482	NA	4.1403	4	254	17.6	396.90
##	280	0.21038	20.0	3.33	0	NA	6.812	32.2	4.1007	5	216	14.9	396.90
##	281	0.03578	20.0	3.33	0	0.4429	7.820	64.5	NA	5	NA	14.9	387.31
##	282	0.03705	NA	3.33	0	0.4429	NA	37.2	NA	5	216	14.9	392.23

## 283	0.06129	20.0	3.33	1	0.4429	7.645	49.7	5.2119	5	NA	14.9	377.07
## 284	0.01501	90.0	NA	1	0.4010	7.923	24.8	5.8850	1	198	13.6	395.52
## 285	0.00906	90.0	2.97	0	0.4000	7.088	20.8	NA	1	285	15.3	394.72
## 286	0.01096	55.0	NA	0	0.3890	6.453	31.9	7.3073	1	300	NA	394.72
## 287	0.01965	80.0	1.76	0	0.3850	NA	31.5	9.0892	1	241	18.2	341.60
## 288	NA	NA	NA	0	0.4050	6.209	31.3	7.3172	NA	293	16.6	396.90
## 289	0.04590	52.5	5.32	0	0.4050	6.315	45.6	7.3172	6	293	16.6	396.90
## 290	0.04297	52.5	5.32	0	0.4050	6.565	22.9	7.3172	6	293	16.6	NA
## 291	NA	80.0	4.95	NA	0.4110	6.861	NA	NA	4	245	19.2	396.90
## 292	0.07886	80.0	4.95	NA	0.4110	7.148	27.7	5.1167	4	245	NA	396.90
## 293	0.03615	80.0	4.95	0	0.4110	6.630	23.4	5.1167	4	245	19.2	396.90
## 294	0.08265	0.0	13.92	0	0.4370	6.127	18.4	5.5027	4	289	16.0	396.90
## 295	0.08199	0.0	13.92	0	NA	6.009	42.3	5.5027	4	289	16.0	396.90
## 296	0.12932	0.0	13.92	0	NA	6.678	31.1	5.9604	4	289	16.0	396.90
## 297	0.05372	0.0	13.92	0	0.4370	6.549	51.0	NA	4	289	16.0	392.85
## 298	0.14103	0.0	13.92	NA	0.4370	5.790	58.0	6.3200	4	289	16.0	396.90
## 299	0.06466	70.0	2.24	0	0.4000	6.345	NA	7.8278	5	358	14.8	NA
## 300	0.05561	70.0	2.24	0	NA	7.041	10.0	7.8278	5	358	14.8	371.58
## 301	0.04417	70.0	2.24	NA	0.4000	6.871	47.4	7.8278	5	358	14.8	390.86
## 302	0.03537	34.0	6.09	0	0.4330	6.590	40.4	NA	7	329	NA	395.75
## 303	NA	34.0	6.09	0	0.4330	6.495	18.4	5.4917	7	329	16.1	383.61
## 304	0.10000	34.0	6.09	NA	0.4330	6.982	17.7	5.4917	NA	329	NA	390.43
## 305	NA	33.0	NA	0	0.4720	7.236	41.1	4.0220	NA	222	18.4	393.68
## 306	NA	33.0	2.18	NA	0.4720	6.616	58.1	3.3700	NA	222	NA	393.36
## 307	0.07503	33.0	2.18	0	NA	7.420	71.9	3.0992	7	222	18.4	396.90
## 308	0.04932	33.0	2.18	0	0.4720	6.849	70.3	3.1827	7	222	18.4	396.90
## 309	0.49298	0.0	9.90	0	NA	6.635	82.5	3.3175	4	304	NA	396.90
## 310	0.34940	0.0	9.90	0	0.5440	5.972	76.7	3.1025	4	304	18.4	396.24
## 311	2.63548	0.0	9.90	0	0.5440	4.973	NA	2.5194	4	304	18.4	350.45
## 312	0.79041	0.0	9.90	0	0.5440	6.122	52.8	2.6403	4	304	18.4	396.90
## 313	NA	0.0	9.90	NA	0.5440	6.023	90.4	2.8340	4	304	18.4	396.30
## 314	0.26938	0.0	9.90	0	NA	6.266	82.8	3.2628	4	304	18.4	393.39
## 315	0.36920	0.0	9.90	NA	0.5440	6.567	87.3	3.6023	4	304	NA	395.69
## 316	0.25356	0.0	9.90	NA	NA	5.705	77.7	3.9450	4	304	18.4	396.42
## 317	0.31827	0.0	9.90	0	0.5440	5.914	83.2	3.9986	4	304	18.4	390.70
## 318	0.24522	0.0	9.90	0	0.5440	5.782	71.7	4.0317	NA	304	18.4	396.90
## 319	0.40202	0.0	9.90	0	0.5440	6.382	67.2	3.5325	4	304	18.4	395.21
## 320	0.47547	0.0	9.90	0	0.5440	6.113	NA	4.0019	4	304	18.4	396.23
## 321	0.16760	NA	7.38	0	NA	6.426	NA	4.5404	5	287	NA	396.90
## 322	0.18159	0.0	7.38	0	0.4930	6.376	54.3	4.5404	5	287	19.6	396.90
## 323	0.35114	0.0	NA	0	0.4930	6.041	NA	4.7211	5	287	19.6	396.90
## 324	0.28392	0.0	7.38	0	0.4930	5.708	NA	4.7211	5	287	19.6	391.13
## 325	0.34109	0.0	7.38	NA	0.4930	6.415	40.1	4.7211	5	287	19.6	NA
## 326	0.19186	0.0	7.38	0	0.4930	6.431	14.7	5.4159	5	287	19.6	393.68
## 327	0.30347	0.0	7.38	0	0.4930	6.312	28.9	5.4159	5	NA	19.6	396.90
## 328	0.24103	NA	7.38	0	NA	6.083	43.7	5.4159	5	287	19.6	396.90
## 329	0.06617	NA	NA	0	0.4600	5.868	25.8	5.2146	4	430	16.9	382.44
## 330	0.06724	0.0	3.24	0	0.4600	6.333	17.2	5.2146	4	NA	16.9	NA
## 331	0.04544	0.0	NA	0	0.4600	6.144	32.2	5.8736	4	430	16.9	NA
## 332	0.05023	35.0	6.06	0	0.4379	5.706	28.4	6.6407	1	304	16.9	394.02
## 333	0.03466	35.0	6.06	NA	0.4379	6.031	23.3	NA	1	304	16.9	362.25
## 334	0.05083	0.0	5.19	0	NA	6.316	38.1	6.4584	5	224	20.2	389.71
## 335	0.03738	0.0	5.19	0	0.5150	6.310	38.5	6.4584	5	224	20.2	389.40
## 336	NA	0.0	5.19	0	0.5150	6.037	34.5	5.9853	5	224	20.2	396.90

## 337	0.03427	0.0	5.19	0	0.5150	5.869	46.3	5.2311	5	224	20.2	396.90
## 338	0.03041	0.0	5.19	0	0.5150	5.895	59.6	5.6150	5	224	NA	NA
## 339	0.03306	0.0	5.19	0	0.5150	6.059	37.3	4.8122	5	224	20.2	396.14
## 340	NA	0.0	5.19	0	0.5150	5.985	45.4	4.8122	5	224	20.2	396.90
## 341	NA	NA	NA	0	0.5150	5.968	58.5	NA	5	224	20.2	396.90
## 342	0.01301	NA	1.52	NA	0.4420	7.241	49.3	7.0379	1	284	15.5	394.74
## 343	NA	NA	1.89	NA	0.5180	NA	59.7	6.2669	1	422	15.9	389.96
## 344	0.02543	55.0	3.78	0	0.4840	6.696	56.4	5.7321	5	370	17.6	396.90
## 345	0.03049	55.0	3.78	0	0.4840	6.874	28.1	6.4654	5	370	17.6	387.97
## 346	0.03113	0.0	4.39	0	0.4420	6.014	48.5	8.0136	NA	352	18.8	385.64
## 347	0.06162	0.0	4.39	0	NA	5.898	52.3	8.0136	NA	352	18.8	364.61
## 348	0.01870	85.0	4.15	0	0.4290	6.516	27.7	8.5353	4	351	17.9	392.43
## 349	0.01501	80.0	2.01	NA	NA	6.635	29.7	8.3440	4	280	17.0	NA
## 350	0.02899	40.0	1.25	0	0.4290	6.939	NA	8.7921	1	NA	19.7	389.85
## 351	0.06211	40.0	1.25	0	NA	6.490	44.4	8.7921	1	335	19.7	396.90
## 352	0.07950	60.0	NA	0	0.4110	6.579	35.9	NA	4	411	18.3	370.78
## 353	0.07244	60.0	1.69	0	NA	5.884	18.5	10.7103	4	411	18.3	NA
## 354	0.01709	90.0	2.02	0	0.4100	6.728	NA	12.1265	5	187	NA	384.46
## 355	0.04301	80.0	1.91	0	0.4130	5.663	21.9	10.5857	4	334	22.0	382.80
## 356	0.10659	80.0	1.91	NA	0.4130	5.936	19.5	10.5857	4	NA	22.0	376.04
## 357	8.98296	NA	18.10	1	0.7700	6.212	97.4	2.1222	24	666	20.2	377.73
## 358	3.84970	0.0	18.10	1	0.7700	NA	NA	2.5052	24	666	20.2	391.34
## 359	5.20177	NA	18.10	1	0.7700	6.127	83.4	2.7227	24	666	20.2	395.43
## 360	4.26131	0.0	18.10	0	0.7700	6.112	81.3	2.5091	24	666	20.2	390.74
## 361	4.54192	0.0	18.10	0	0.7700	6.398	88.0	2.5182	24	666	20.2	374.56
## 362	3.83684	0.0	NA	NA	0.7700	NA	91.1	2.2955	24	666	20.2	350.65
## 363	3.67822	0.0	18.10	0	0.7700	5.362	96.2	2.1036	24	666	20.2	380.79
## 364	4.22239	NA	18.10	1	0.7700	5.803	89.0	1.9047	24	666	20.2	353.04
## 365	NA	0.0	NA	1	0.7180	NA	82.9	1.9047	24	666	20.2	354.55
## 366	NA	NA	18.10	0	0.7180	3.561	NA	1.6132	NA	666	NA	354.70
## 367	3.69695	0.0	18.10	0	0.7180	4.963	91.4	1.7523	24	666	20.2	316.03
## 368	13.52220	0.0	18.10	0	0.6310	NA	100.0	1.5106	24	666	20.2	131.42
## 369	4.89822	0.0	18.10	0	0.6310	4.970	100.0	1.3325	NA	666	20.2	375.52
## 370	5.66998	0.0	18.10	1	0.6310	6.683	96.8	1.3567	24	666	20.2	375.33
## 371	6.53876	0.0	18.10	1	0.6310	7.016	97.5	1.2024	24	666	20.2	392.05
## 372	9.23230	0.0	18.10	NA	0.6310	6.216	100.0	1.1691	24	666	20.2	366.15
## 373	8.26725	0.0	18.10	1	0.6680	5.875	89.6	1.1296	NA	666	NA	347.88
## 374	11.10810	0.0	NA	0	0.6680	4.906	100.0	1.1742	24	666	20.2	396.90
## 375	18.49820	0.0	18.10	0	0.6680	4.138	100.0	1.1370	24	666	20.2	396.90
## 376	19.60910	0.0	18.10	0	0.6710	7.313	97.9	1.3163	24	NA	20.2	396.90
## 377	15.28800	NA	18.10	0	0.6710	6.649	93.3	1.3449	24	666	20.2	363.02
## 378	9.82349	0.0	18.10	0	0.6710	6.794	98.8	NA	24	666	20.2	396.90
## 379	23.64820	0.0	18.10	0	0.6710	NA	96.2	1.3861	24	666	20.2	396.90
## 380	NA	0.0	18.10	0	0.6710	6.223	100.0	1.3861	NA	666	20.2	393.74
## 381	88.97620	NA	18.10	0	0.6710	6.968	91.9	1.4165	24	666	20.2	396.90
## 382	15.87440	0.0	NA	0	0.6710	NA	99.1	1.5192	24	666	NA	396.90
## 383	9.18702	0.0	NA	0	0.7000	5.536	100.0	1.5804	24	666	20.2	396.90
## 384	7.99248	0.0	18.10	0	0.7000	NA	100.0	1.5331	24	666	20.2	396.90
## 385	20.08490	0.0	18.10	0	0.7000	4.368	91.2	1.4395	24	666	20.2	285.83
## 386	16.81180	0.0	18.10	0	0.7000	5.277	98.1	1.4261	24	666	20.2	396.90
## 387	24.39380	0.0	18.10	0	0.7000	4.652	100.0	1.4672	24	666	20.2	396.90
## 388	22.59710	0.0	NA	0	NA	5.000	89.5	1.5184	24	666	20.2	396.90
## 389	14.33370	0.0	18.10	0	0.7000	4.880	100.0	1.5895	24	666	20.2	372.92
## 390	8.15174	NA	18.10	0	0.7000	5.390	98.9	1.7281	24	666	20.2	396.90

## 391	6.96215	0.0	18.10	NA	0.7000	NA	97.0	1.9265	24	666	20.2	394.43
## 392	5.29305	0.0	NA	0	NA	6.051	82.5	2.1678	24	666	20.2	378.38
## 393	11.57790	0.0	18.10	0	0.7000	5.036	97.0	1.7700	24	666	NA	396.90
## 394	8.64476	NA	18.10	0	0.6930	6.193	92.6	1.7912	24	666	20.2	396.90
## 395	13.35980	0.0	18.10	0	0.6930	5.887	NA	1.7821	24	666	20.2	396.90
## 396	8.71675	NA	18.10	0	NA	NA	98.8	1.7257	24	666	20.2	391.98
## 397	5.87205	0.0	18.10	0	0.6930	6.405	96.0	1.6768	24	666	20.2	396.90
## 398	7.67202	0.0	18.10	NA	0.6930	5.747	98.9	1.6334	24	666	NA	393.10
## 399	38.35180	0.0	18.10	0	0.6930	5.453	100.0	NA	24	666	20.2	396.90
## 400	9.91655	0.0	18.10	0	0.6930	5.852	77.8	1.5004	24	666	20.2	338.16
## 401	25.04610	0.0	18.10	0	0.6930	5.987	100.0	1.5888	24	NA	20.2	396.90
## 402	14.23620	0.0	18.10	0	0.6930	6.343	100.0	1.5741	24	666	NA	NA
## 403	9.59571	0.0	18.10	0	0.6930	NA	100.0	1.6390	24	666	20.2	376.11
## 404	24.80170	NA	18.10	0	0.6930	5.349	96.0	NA	24	666	20.2	NA
## 405	41.52920	0.0	18.10	0	0.6930	NA	85.4	1.6074	24	666	20.2	329.46
## 406	67.92080	0.0	18.10	0	0.6930	NA	NA	1.4254	24	666	NA	NA
## 407	20.71620	0.0	18.10	0	0.6590	4.138	100.0	1.1781	24	666	NA	370.22
## 408	11.95110	0.0	18.10	0	0.6590	5.608	100.0	1.2852	24	666	20.2	332.09
## 409	7.40389	0.0	18.10	0	0.5970	5.617	97.9	1.4547	24	666	20.2	314.64
## 410	14.43830	0.0	18.10	0	NA	6.852	NA	1.4655	24	666	20.2	179.36
## 411	51.13580	0.0	18.10	0	0.5970	5.757	100.0	1.4130	24	666	20.2	2.60
## 412	NA	0.0	NA	0	NA	6.657	100.0	1.5275	24	666	20.2	35.05
## 413	18.81100	0.0	18.10	0	NA	4.628	NA	1.5539	24	666	20.2	28.79
## 414	28.65580	0.0	18.10	0	0.5970	5.155	100.0	1.5894	NA	666	NA	210.97
## 415	45.74610	0.0	18.10	0	0.6930	NA	100.0	1.6582	24	666	20.2	88.27
## 416	18.08460	0.0	18.10	0	0.6790	6.434	100.0	1.8347	24	666	NA	27.25
## 417	10.83420	0.0	18.10	0	0.6790	6.782	90.8	NA	24	666	20.2	21.57
## 418	25.94060	NA	18.10	NA	0.6790	NA	89.1	NA	24	666	20.2	NA
## 419	73.53410	0.0	NA	0	0.6790	5.957	100.0	1.8026	24	666	20.2	16.45
## 420	11.81230	0.0	18.10	0	0.7180	6.824	76.5	NA	24	666	20.2	48.45
## 421	11.08740	NA	18.10	0	0.7180	6.411	100.0	1.8589	NA	666	20.2	318.75
## 422	NA	0.0	18.10	0	0.7180	6.006	95.3	1.8746	24	666	20.2	319.98
## 423	12.04820	0.0	18.10	0	NA	5.648	87.6	1.9512	24	666	20.2	291.55
## 424	7.05042	0.0	18.10	0	0.6140	6.103	85.1	2.0218	24	666	20.2	NA
## 425	NA	0.0	18.10	0	NA	5.565	70.6	NA	24	666	20.2	3.65
## 426	15.86030	0.0	NA	0	0.6790	5.896	95.4	1.9096	24	666	20.2	7.68
## 427	NA	NA	18.10	0	0.5840	5.837	NA	1.9976	NA	666	20.2	24.65
## 428	37.66190	0.0	18.10	0	0.6790	6.202	NA	1.8629	24	666	20.2	18.82
## 429	7.36711	0.0	18.10	0	0.6790	6.193	78.1	1.9356	24	666	20.2	96.73
## 430	9.33889	0.0	18.10	0	0.6790	6.380	95.6	1.9682	24	666	20.2	60.72
## 431	8.49213	0.0	18.10	0	0.5840	6.348	86.1	2.0527	24	666	20.2	83.45
## 432	10.06230	0.0	18.10	0	0.5840	6.833	94.3	2.0882	24	666	20.2	NA
## 433	NA	0.0	18.10	0	0.5840	6.425	74.8	2.2004	24	666	20.2	NA
## 434	5.58107	NA	18.10	0	0.7130	6.436	87.9	2.3158	24	666	20.2	100.19
## 435	13.91340	0.0	18.10	0	0.7130	6.208	95.0	2.2222	24	666	20.2	100.63
## 436	11.16040	0.0	18.10	0	NA	6.629	94.6	2.1247	24	666	20.2	109.85
## 437	14.42080	0.0	18.10	0	0.7400	6.461	93.3	2.0026	24	666	20.2	NA
## 438	15.17720	0.0	18.10	0	0.7400	6.152	100.0	1.9142	24	666	20.2	9.32
## 439	13.67810	0.0	18.10	0	0.7400	5.935	87.9	1.8206	NA	666	20.2	68.95
## 440	NA	0.0	18.10	0	0.7400	5.627	93.9	1.8172	NA	666	20.2	396.90
## 441	22.05110	0.0	18.10	0	0.7400	5.818	92.4	1.8662	NA	666	20.2	391.45
## 442	9.72418	0.0	18.10	0	0.7400	6.406	97.2	2.0651	24	666	20.2	NA
## 443	5.66637	0.0	18.10	0	0.7400	NA	100.0	2.0048	24	666	20.2	395.69
## 444	9.96654	0.0	18.10	0	0.7400	6.485	100.0	1.9784	24	666	20.2	NA

## 445	12.80230	0.0	18.10	0	0.7400	5.854	96.6	1.8956	24	666	20.2	240.52
## 446	10.67180	0.0	NA	0	0.7400	NA	94.8	1.9879	24	666	20.2	43.06
## 447	6.28807	0.0	18.10	0	0.7400	6.341	96.4	2.0720	24	666	NA	318.01
## 448	9.92485	0.0	NA	0	0.7400	6.251	96.6	2.1980	24	666	20.2	NA
## 449	9.32909	0.0	18.10	NA	NA	6.185	98.7	2.2616	24	666	20.2	396.90
## 450	NA	0.0	18.10	0	NA	6.417	NA	2.1850	24	NA	20.2	304.21
## 451	NA	NA	18.10	0	0.7130	6.749	92.6	2.3236	24	666	20.2	0.32
## 452	5.44114	0.0	18.10	0	0.7130	6.655	98.2	2.3552	24	666	20.2	355.29
## 453	NA	0.0	18.10	0	0.7130	6.297	91.8	2.3682	24	666	20.2	NA
## 454	8.24809	0.0	NA	0	0.7130	7.393	99.3	2.4527	24	666	20.2	NA
## 455	9.51363	0.0	NA	0	0.7130	NA	94.1	2.4961	24	666	20.2	6.68
## 456	4.75237	0.0	18.10	0	0.7130	6.525	86.5	2.4358	24	666	NA	NA
## 457	4.66883	NA	18.10	0	0.7130	5.976	87.9	2.5806	24	666	20.2	10.48
## 458	8.20058	0.0	18.10	0	0.7130	5.936	NA	2.7792	24	666	20.2	3.50
## 459	7.75223	0.0	18.10	0	0.7130	6.301	83.7	2.7831	NA	666	20.2	272.21
## 460	6.80117	0.0	18.10	0	0.7130	6.081	84.4	2.7175	24	666	20.2	396.90
## 461	4.81213	0.0	18.10	0	0.7130	6.701	90.0	2.5975	24	666	20.2	255.23
## 462	NA	0.0	18.10	0	0.7130	NA	NA	2.5671	24	666	20.2	391.43
## 463	6.65492	NA	18.10	NA	0.7130	6.317	83.0	2.7344	24	666	20.2	NA
## 464	5.82115	0.0	18.10	0	0.7130	6.513	89.9	2.8016	24	666	20.2	393.82
## 465	7.83932	0.0	NA	0	0.6550	6.209	65.4	2.9634	24	666	20.2	396.90
## 466	3.16360	0.0	18.10	0	0.6550	5.759	48.2	3.0665	24	666	20.2	334.40
## 467	NA	0.0	18.10	0	0.6550	NA	84.7	2.8715	24	666	20.2	22.01
## 468	4.42228	0.0	18.10	0	0.5840	6.003	94.5	NA	24	666	20.2	331.29
## 469	15.57570	0.0	18.10	0	0.5800	5.926	71.0	2.9084	24	666	20.2	368.74
## 470	13.07510	0.0	NA	NA	0.5800	5.713	56.7	2.8237	24	666	20.2	396.90
## 471	4.34879	0.0	18.10	0	NA	6.167	84.0	3.0334	24	666	20.2	396.90
## 472	4.03841	0.0	18.10	0	NA	6.229	90.7	3.0993	24	666	20.2	395.33
## 473	3.56868	0.0	18.10	0	0.5800	6.437	75.0	2.8965	24	666	20.2	NA
## 474	4.64689	0.0	18.10	0	0.6140	NA	67.6	2.5329	24	666	20.2	374.68
## 475	8.05579	0.0	18.10	0	0.5840	5.427	95.4	2.4298	24	666	20.2	352.58
## 476	NA	0.0	18.10	0	0.5840	6.162	97.4	2.2060	24	666	20.2	302.76
## 477	NA	0.0	18.10	0	0.6140	NA	93.6	2.3053	NA	666	20.2	396.21
## 478	15.02340	0.0	NA	0	NA	5.304	97.3	2.1007	24	666	20.2	349.48
## 479	NA	0.0	18.10	0	0.6140	NA	NA	2.1705	NA	666	20.2	379.70
## 480	14.33370	0.0	NA	0	0.6140	6.229	88.0	1.9512	24	666	20.2	383.32
## 481	5.82401	0.0	18.10	0	0.5320	6.242	64.7	3.4242	24	666	20.2	396.90
## 482	5.70818	NA	18.10	0	0.5320	6.750	74.9	3.3317	24	666	20.2	NA
## 483	5.73116	0.0	18.10	0	0.5320	7.061	77.0	NA	NA	666	20.2	395.28
## 484	2.81838	0.0	18.10	0	0.5320	5.762	40.3	4.0983	24	666	20.2	392.92
## 485	2.37857	0.0	18.10	0	0.5830	5.871	41.9	3.7240	24	666	20.2	370.73
## 486	3.67367	0.0	18.10	0	0.5830	6.312	51.9	3.9917	24	666	20.2	388.62
## 487	5.69175	0.0	18.10	0	0.5830	6.114	79.8	NA	NA	666	20.2	392.68
## 488	4.83567	0.0	18.10	0	0.5830	5.905	53.2	3.1523	24	666	20.2	388.22
## 489	0.15086	0.0	27.74	0	NA	5.454	92.7	1.8209	4	NA	20.1	NA
## 490	0.18337	0.0	27.74	0	0.6090	5.414	98.3	1.7554	4	711	20.1	344.05
## 491	0.20746	0.0	27.74	0	0.6090	5.093	98.0	1.8226	NA	711	20.1	318.43
## 492	0.10574	0.0	27.74	0	0.6090	5.983	98.8	1.8681	4	711	20.1	390.11
## 493	0.11132	0.0	27.74	0	0.6090	5.983	83.5	2.1099	4	711	20.1	NA
## 494	0.17331	0.0	9.69	0	0.5850	5.707	54.0	2.3817	6	391	19.2	396.90
## 495	0.27957	0.0	9.69	0	0.5850	5.926	42.6	2.3817	6	391	19.2	396.90
## 496	0.17899	0.0	NA	0	0.5850	5.670	NA	2.7986	6	391	NA	393.29
## 497	0.28960	0.0	9.69	0	0.5850	5.390	72.9	2.7986	6	391	19.2	396.90
## 498	0.26838	0.0	9.69	0	0.5850	5.794	70.6	2.8927	6	391	19.2	396.90

##	499	0.23912	0.0	9.69	0	0.5850	6.019	65.3	2.4091	6	391	19.2	396.90
##	500	0.17783	0.0	9.69	0	0.5850	5.569	73.5	2.3999	6	391	19.2	395.77
##	501	0.22438	0.0	NA	0	0.5850	6.027	79.7	NA	6	391	19.2	396.90
##	502	0.06263	0.0	11.93	0	0.5730	6.593	69.1	2.4786	NA	273	21.0	391.99
##	503	0.04527	0.0	11.93	0	0.5730	6.120	76.7	2.2875	1	273	21.0	396.90
##	504	0.06076	0.0	11.93	0	0.5730	6.976	91.0	2.1675	1	273	NA	396.90
##	505	0.10959	0.0	11.93	NA	NA	6.794	89.3	2.3889	1	NA	21.0	393.45
##	506	0.04741	NA	11.93	NA	0.5730	6.030	80.8	2.5050	1	273	21.0	396.90
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##	20		11.28										
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##	90	5.70
##	91	8.81
##	92	8.20
##	93	NA
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##	96	6.65
##	97	11.34
##	98	4.21
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## 153 12.12
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239 6.36
240 7.37
241 11.38
242 12.40
243 11.22
244 5.19
245 12.50
246 18.46
247 9.16
248 10.15
249 9.52
250 6.56
251 5.90
252 3.59
253 3.53
254 3.54
255 6.57
256 9.25
257 3.11
258 5.12
259 7.79
260 6.90
261 9.59

262 7.26
263 5.91
264 11.25
265 8.10
266 10.45
267 14.79
268 7.44
269 3.16
270 13.65
271 13.00
272 6.59
273 7.73
274 6.58
275 3.53
276 2.98
277 6.05
278 4.16
279 7.19
280 4.85
281 3.76
282 4.59
283 3.01
284 3.16
285 7.85
286 8.23
287 NA
288 7.14
289 7.60
290 9.51
291 3.33
292 3.56
293 4.70
294 8.58
295 10.40
296 6.27
297 NA
298 15.84
299 4.97
300 4.74
301 6.07
302 9.50
303 NA
304 4.86
305 6.93
306 8.93
307 6.47
308 7.53
309 4.54
310 NA
311 12.64
312 5.98
313 11.72
314 7.90
315 9.28

316 11.50
317 18.33
318 NA
319 10.36
320 12.73
321 7.20
322 6.87
323 7.70
324 11.74
325 6.12
326 5.08
327 6.15
328 12.79
329 9.97
330 7.34
331 9.09
332 12.43
333 7.83
334 NA
335 6.75
336 8.01
337 9.80
338 NA
339 8.51
340 9.74
341 NA
342 5.49
343 8.65
344 7.18
345 4.61
346 10.53
347 12.67
348 6.36
349 5.99
350 5.89
351 5.98
352 NA
353 7.79
354 4.50
355 8.05
356 5.57
357 17.60
358 13.27
359 11.48
360 12.67
361 7.79
362 14.19
363 10.19
364 14.64
365 5.29
366 7.12
367 14.00
368 13.33
369 3.26

370 3.73
371 2.96
372 9.53
373 8.88
374 34.77
375 37.97
376 13.44
377 23.24
378 21.24
379 23.69
380 21.78
381 17.21
382 21.08
383 23.60
384 24.56
385 30.63
386 30.81
387 28.28
388 31.99
389 30.62
390 20.85
391 17.11
392 18.76
393 25.68
394 15.17
395 16.35
396 17.12
397 19.37
398 19.92
399 NA
400 29.97
401 26.77
402 20.32
403 20.31
404 19.77
405 27.38
406 22.98
407 23.34
408 12.13
409 26.40
410 19.78
411 10.11
412 21.22
413 NA
414 20.08
415 36.98
416 29.05
417 25.79
418 26.64
419 20.62
420 NA
421 15.02
422 15.70
423 14.10

424 23.29
425 17.16
426 24.39
427 NA
428 14.52
429 21.52
430 24.08
431 17.64
432 19.69
433 12.03
434 NA
435 15.17
436 NA
437 18.05
438 26.45
439 34.02
440 22.88
441 22.11
442 19.52
443 16.59
444 18.85
445 23.79
446 23.98
447 17.79
448 16.44
449 18.13
450 19.31
451 17.44
452 17.73
453 17.27
454 16.74
455 18.71
456 18.13
457 19.01
458 16.94
459 NA
460 14.70
461 16.42
462 14.65
463 13.99
464 10.29
465 13.22
466 14.13
467 17.15
468 21.32
469 18.13
470 14.76
471 16.29
472 12.87
473 14.36
474 11.66
475 18.14
476 NA
477 18.68

```
## 478    NA
## 479    NA
## 480 13.11
## 481    NA
## 482  7.74
## 483  7.01
## 484 10.42
## 485 13.34
## 486 10.58
## 487 14.98
## 488 11.45
## 489 18.06
## 490 23.97
## 491 29.68
## 492 18.07
## 493 13.35
## 494 12.01
## 495 13.59
## 496 17.60
## 497 21.14
## 498 14.10
## 499 12.92
## 500 15.10
## 501 14.33
## 502  9.67
## 503  9.08
## 504  5.64
## 505  6.48
## 506  7.88
```

Use a random forest modeling procedure to iteratively fill in the NA's by predicting each feature of X using every other feature of X. You need to start by filling in the holes to use RF. So fill them in with the average of the feature.

```
#Getting error while knitting

#pacman::p_load(randomForest)

n = nrow(X_fill)
p = ncol(X_fill)

for(i in 1:n){
  for(j in 1:p){
    if(is.na(X_fill[i,j])){
      X_fill_2 = X_fill %>%
        replace_na(as.list(colMeans(X_fill, na.rm = TRUE)))
      rf_mod = randomForest(X_fill_2[,j] ~., data = X_fill_2, ntree = 100)
      X_fill[i,j] =predict(rf_mod, X_fill_2[i,])
    }
  }
}
```

#X_fill