

Final Project

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```
#import housing_data
housing_data = read.csv(file = '/Applications/MAT 342/housing_data.csv')

#load packages
pacman::p_load(tidyverse, mlr, mlr3, missForest, skimr, rpart, randomForest, data.table, dplyr, magrittr)

#picking certain data that does not apply in my opinion
housing_data_select = housing_data %>%
  select(approx_year_built, cats_allowed, coop_condo, dogs_allowed, dining_room_type, fuel_type,
         garage_exists, kitchen_type, maintenance_cost, num_bedrooms, num_floors_in_building,
         num_full_bathrooms, num_total_rooms, parking_charges, sale_price, sq_footage, total_taxes, walk_score)

setDT(housing_data_select)

dim(housing_data_select)

## [1] 2230  18

str(housing_data_select)

## Classes 'data.table' and 'data.frame':  2230 obs. of  18 variables:
## $ approx_year_built      : int  1955 1955 2004 2002 1949 1938 1950 1960 1960 2005 ...
## $ cats_allowed           : chr  "no" "no" "no" "no" ...
## $ coop_condo             : chr  "co-op" "co-op" "condo" "condo" ...
## $ dogs_allowed           : chr  "no" "no" "no" "no" ...
## $ dining_room_type       : chr  "combo" "formal" "combo" "combo" ...
## $ fuel_type              : chr  "gas" "oil" NA "gas" ...
## $ garage_exists          : chr  NA NA NA NA ...
## $ kitchen_type           : chr  "eat in" "eat in" "efficiency" "eat in" ...
## $ maintenance_cost       : chr  NA "$604 " NA NA ...
## $ num_bedrooms           : int  2 1 1 3 2 2 1 0 1 1 ...
## $ num_floors_in_building : int  6 7 1 NA 2 6 NA 2 NA 4 ...
## $ num_full_bathrooms     : int  1 1 1 2 1 1 1 1 1 1 ...
## $ num_total_rooms        : int  5 4 3 5 4 4 3 2 4 3 ...
## $ parking_charges        : chr  NA NA NA NA ...
## $ sale_price             : chr  "$228,000 " "$235,500 " "$137,550 " "$545,000 " ...
## $ sq_footage             : int  NA 890 550 NA 675 1000 NA 375 NA 681 ...
## $ total_taxes            : chr  NA NA "$5,500 " "$2,260 " ...
## $ walk_score             : int  82 89 90 94 71 90 72 93 70 98 ...
## - attr(*, ".internal.selfref")=<externalptr>

#removing any data with excessive NA
housing_data_drop = housing_data_select %>%
  select(-parking_charges, -sq_footage, -total_taxes, -num_floors_in_building)

#Adjusting the features so they can be used to run the algorithms
housing_data_new = housing_data_drop %>%
```

```

mutate(cats_allowed = ifelse(cats_allowed == "yes", 1, 0)) %>% #set data to binary
mutate(dogs_allowed = ifelse(dogs_allowed == "yes", 1, 0)) %>% #set data to binary
mutate(maintenance_cost = as.numeric(gsub('[$,]', '', housing_data_drop$maintenance_cost))) %>% #remove $ and , from maintenance_cost
mutate(sale_price = as.numeric(gsub('[$,]', '', housing_data_drop$sale_price))) %>% #remove $ and , from sale_price
mutate(coop_condo = factor(coop_condo, ordered = FALSE)) %>%
mutate(dining_room_type = factor(dining_room_type, ordered = FALSE)) %>%
mutate(fuel_type = factor(fuel_type, ordered = FALSE)) %>%
mutate(kitchen_type = factor(kitchen_type, ordered = FALSE)) %>%
mutate(garage_exists = ifelse(is.na(garage_exists), 0, 1)) #making sure NA is turned to 0

housing_data_new %>%
  filter(!is.na(sale_price))

```

```

##      approx_year_built cats_allowed coop_condo dogs_allowed dining_room_type
##  1:          1955          0      co-op          0          combo
##  2:          1955          0      co-op          0          formal
##  3:          2004          0      condo          0          combo
##  4:          2002          0      condo          0          combo
##  5:          1949          1      co-op          1          combo
## ---
## 524:          1950          0      co-op          0          <NA>
## 525:          1947          0      co-op          0          formal
## 526:          2010          0      condo          0          combo
## 527:          2006          0      condo          0          combo
## 528:          1958          0      co-op          0          other
##      fuel_type garage_exists kitchen_type maintenance_cost num_bedrooms
##  1:      gas          0      eat in          NA          2
##  2:      oil          0      eat in          604          1
##  3:    <NA>          0  efficiency          NA          1
##  4:      gas          0      eat in          NA          3
##  5:      gas          0      eat in          660          2
## ---
## 524:      gas          0      eat in          725          2
## 525:      gas          0      Combo          680          1
## 526:      gas          0      Eat In          NA          2
## 527: electric          0      Combo          NA          2
## 528:      other          0      eat in          659          2
##      num_full_bathrooms num_total_rooms sale_price walk_score
##  1:          1          5      228000          82
##  2:          1          4      235500          89
##  3:          1          3      137550          90
##  4:          2          5      545000          94
##  5:          1          4      241700          71
## ---
## 524:          1          4      216000          83
## 525:          1          5      232500          94
## 526:          2          5      428000          96
## 527:          2          4      635000          99
## 528:          1          4      310000          96

```

```
missing_data = tbl_df(apply(is.na(housing_data_new), 2, as.numeric))
```

```

## Warning: `tbl_df()` was deprecated in dplyr 1.0.0.
## Please use `tibble::as_tibble()` instead.

```

```

colnames(missing_data) = paste("missing_data_", colnames(housing_data_new), sep = "")
missing_data %<>%
  select_if(function(x){sum(x) > 0})
housing_imp = missForest(data.frame(housing_data_new))$ximp

## missForest iteration 1 in progress...done!
## missForest iteration 2 in progress...done!
## missForest iteration 3 in progress...done!
## missForest iteration 4 in progress...done!
## missForest iteration 5 in progress...done!
## missForest iteration 6 in progress...done!
## missForest iteration 7 in progress...done!

housing = cbind(housing_imp,missing_data)

#making train and test split
test_prop = 0.1

#test
test_indices = sample(1:nrow(housing), round((test_prop)*nrow(housing)))
housing_test = housing[test_indices,]
y_test = housing_test$sale_price
X_test = cbind(1, housing_test)
X_test$sale_price = NULL

#train
train_indices = setdiff(1:nrow(housing), test_indices)
housing_train = housing[train_indices,]
y_train = housing_train$sale_price
X_train = cbind(1, housing_train)
X_train$sale_price = NULL
n_train = nrow(X_train)

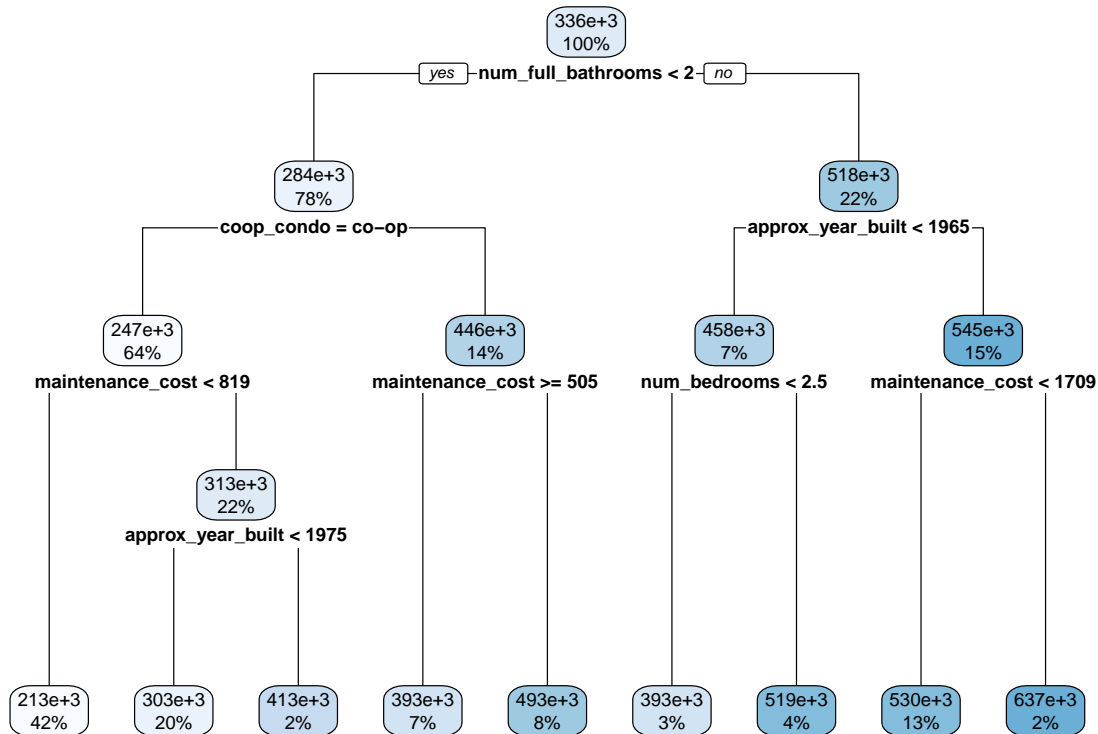
#Create Regression tree model

#use rpart, YARF not available for me
#In-sample Error
Reg_tree = rpart(y_train~., housing_train)
y_hat_train = predict(Reg_tree, housing_train)
e_in = y_train - y_hat_train
rsme_in = sd(e_in)
r_squared_in = (var(y_train)-var(e_in)) / var(y_train)

#OOSE
y_hat_test = predict(Reg_tree, housing_test)
e_oose = y_test - y_hat_test
rsme_oose = sd(e_oose)
rsquared_oose = (var(y_test) - var(e_oose)) / var(y_test)

fit_model = rpart(housing_train$sale_price~., data.frame(X_train), method="anova")
rpart.plot(fit_model)

```



```
fit_model
```

```
## n= 2007
##
## node), split, n, deviance, yval
##      * denotes terminal node
##
## 1) root 2007 4.228751e+13 335821.8
##    2) num_full_bathrooms< 1.5 1560 1.860696e+13 283659.1
##      4) coop_condo=co-op 1276 7.233930e+12 247489.9
##        8) maintenance_cost< 818.5 837 2.207091e+12 213224.0 *
##        9) maintenance_cost>=818.5 439 2.170322e+12 312821.6
##          18) approx_year_built< 1974.5 400 1.629533e+12 303042.3 *
##          19) approx_year_built>=1974.5 39 1.101891e+11 413121.9 *
##      5) coop_condo=condo 284 2.203753e+12 446165.9
##        10) maintenance_cost>=505.3131 133 8.476970e+11 393480.4 *
##        11) maintenance_cost< 505.3131 151 6.617111e+11 492571.0 *
##    3) num_full_bathrooms>=1.5 447 4.622290e+12 517865.8
##      6) approx_year_built< 1964.5 141 1.704860e+12 458049.5
##        12) num_bedrooms< 2.5 68 3.817784e+11 392813.1 *
##        13) num_bedrooms>=2.5 73 7.641167e+11 518817.6 *
##      7) approx_year_built>=1964.5 306 2.180469e+12 545428.2
##        14) maintenance_cost< 1709.249 261 1.279587e+12 529642.6 *
##        15) maintenance_cost>=1709.249 45 4.586285e+11 636984.6 *
```

```
#mlr attempt
```

```
mod_task = makeRegrTask(data = data.frame(X_train), target = 'housing_data$sale_price')
algor = makeLearner("regr.rpart")
valid = makeResampleDesc("CV", iteration = 5)
resample = resample(algor, mod_task, valid, measures = list(rmse))
resample
```

```

mean(resample$measures.test$rmse)

#Random Forest
random_forest = randomForest(sale_price~.,housing_imp)
random_forest

##
## Call:
##  randomForest(formula = sale_price ~ ., data = housing_imp)
##              Type of random forest: regression
##              Number of trees: 500
## No. of variables tried at each split: 4
##
##              Mean of squared residuals: 1531236666
##              % Var explained: 92.7

```