

homework4_3

syh

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```
# from HW4_2, we can get some features used to actually split a tree
# BsmtFinSF1 BsmtFinType1 BsmtFullBath BsmtUnfSF CentralAir Exterior1st
# Exterior2nd Fireplaces GarageArea GrLivArea KitchenQual LotArea
# Neighborhood OverallCond OverallQual TotalBsmtSF TotRmsAbvGrd WoodDeckSF
# X1stFlrSF YearBuilt YearRemodAdd
# Maybe we can use these features to build a linear regression model based on HW4

# read all data (train + prediction) without missing value
real_all_daat <- read.csv(file = "H:/kaggle/houseprice/data/real_all_data_hybrid.csv",
                          stringsAsFactors = FALSE)[-c(1,2)]

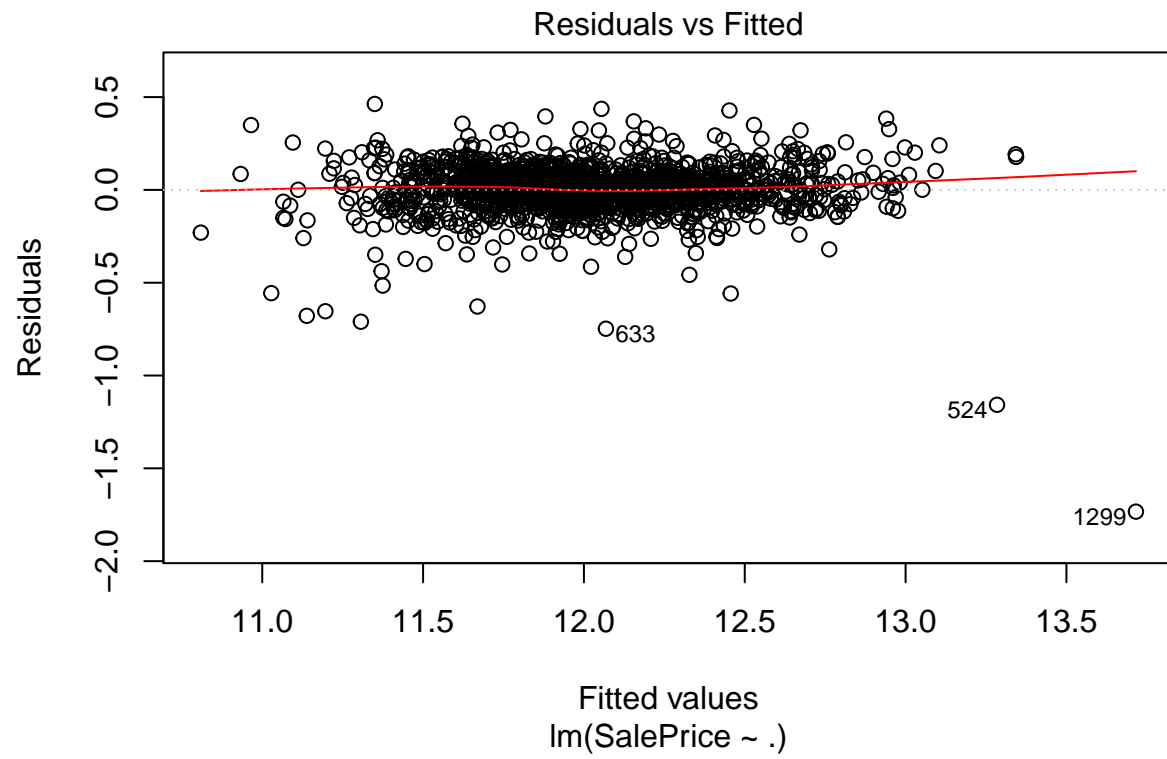
# transform sale price to log sale price
real_all_daat[, "SalePrice"] <- log(real_all_daat[, "SalePrice"])

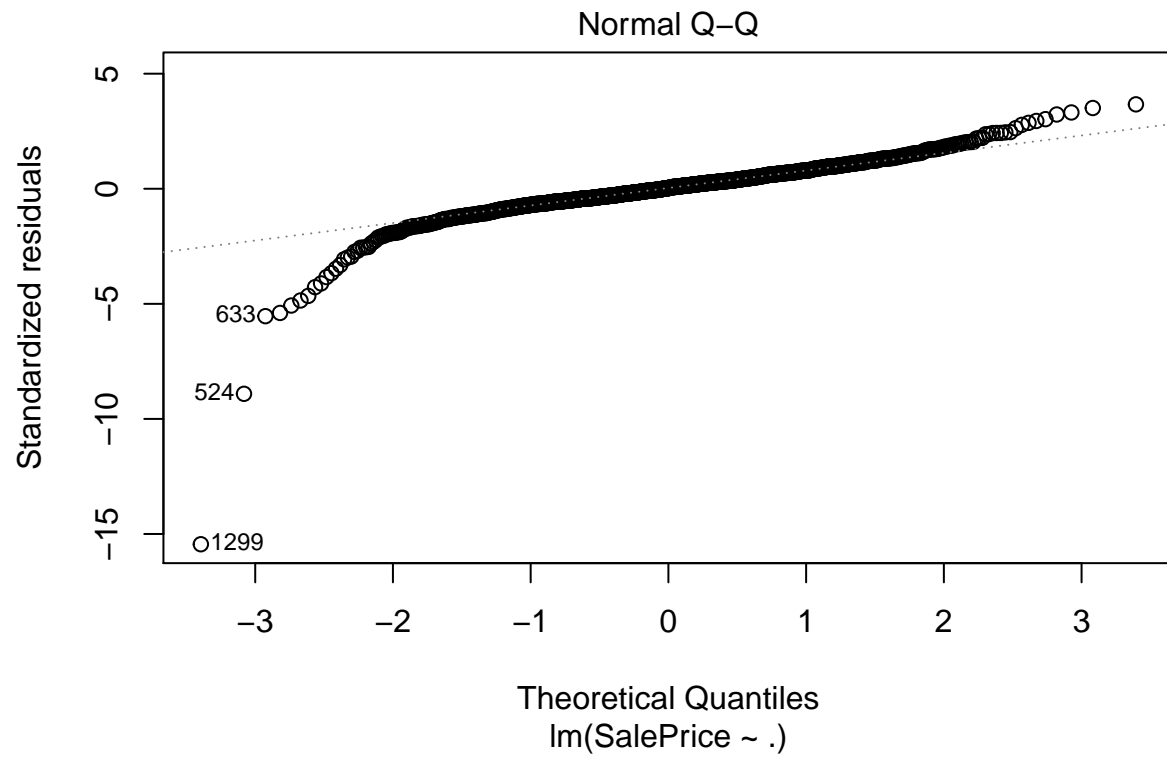
# just keep features above
keep_col <- c("BsmtFinSF1", "BsmtFinType1", "BsmtFullBath", "BsmtUnfSF", "CentralAir",
             "Exterior1st", "Exterior2nd", "Fireplaces", "GarageArea", "GrLivArea", "KitchenQual",
             "Neighborhood", "OverallCond", "OverallQual", "TotalBsmtSF", "TotRmsAbvGrd", "WoodDeckSF",
             "X1stFlrSF", "YearBuilt", "YearRemodAdd")

real_all_daat <- real_all_daat[, keep_col]

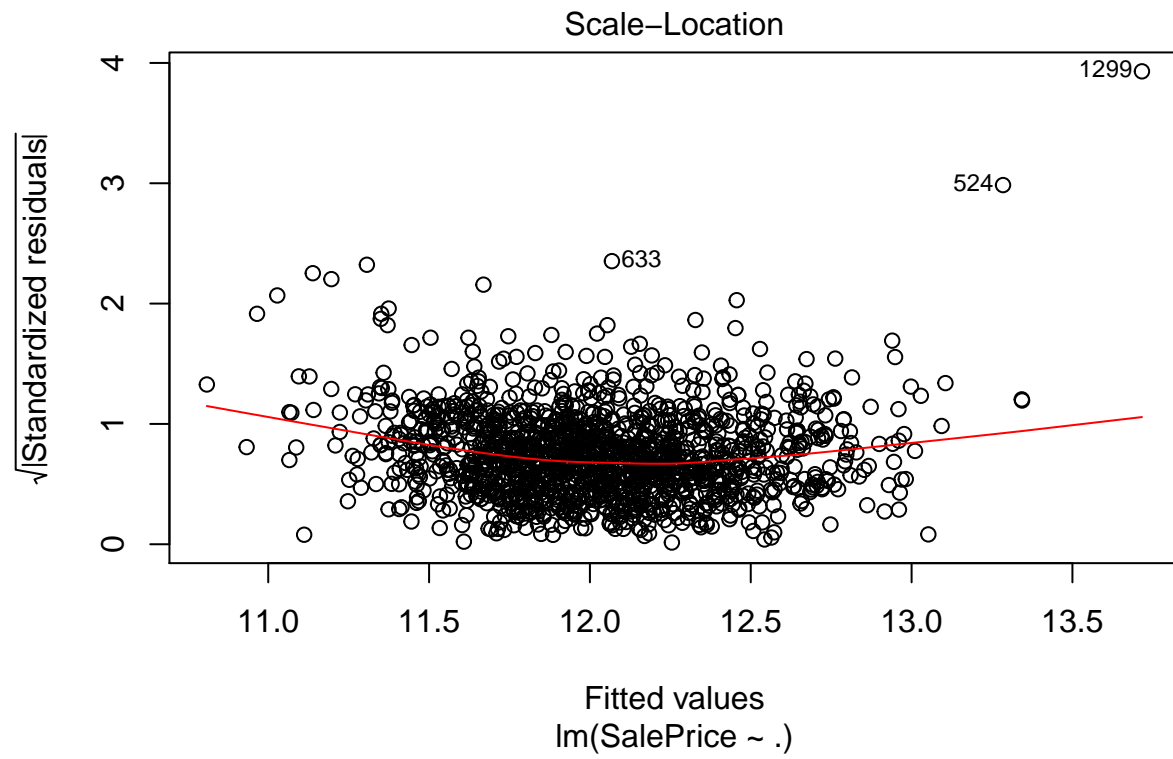
# train a simple linear model by lm()
# check if it's subject to linear regression
# find some outlier and delete them
simple_lm <- lm(formula = SalePrice ~., data = real_all_daat[1:1460,])
plot(simple_lm)

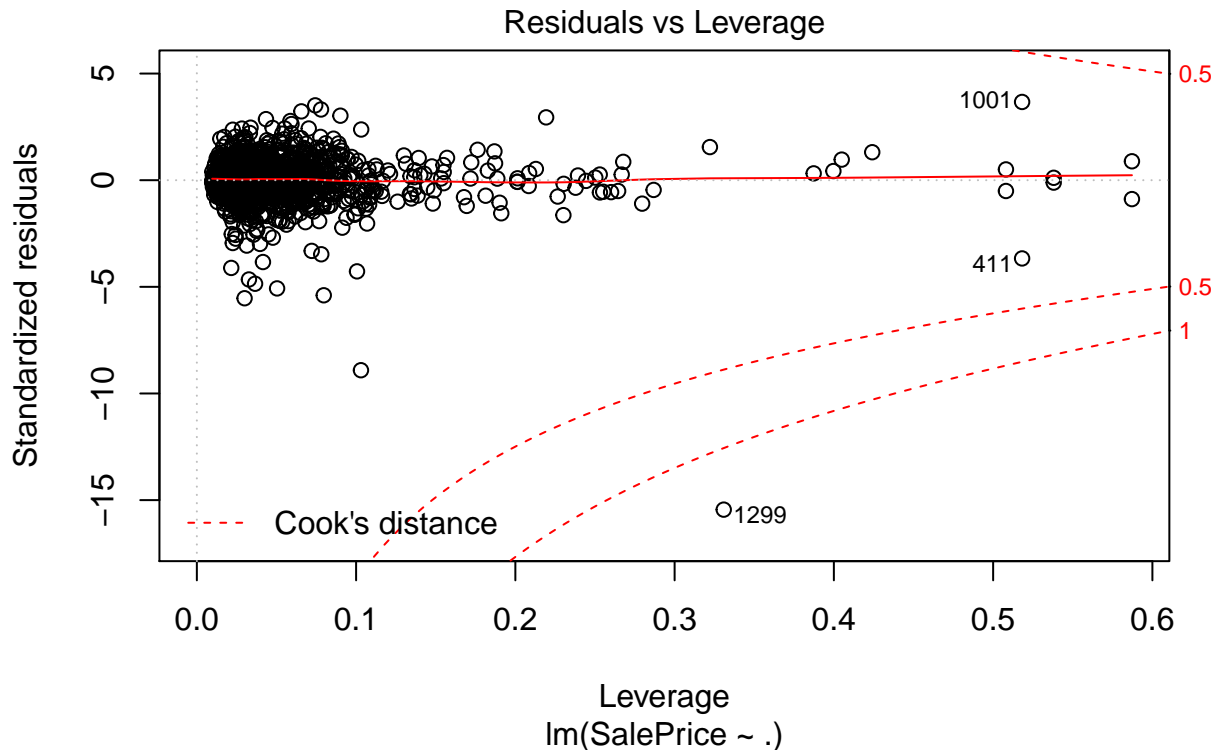
## Warning: not plotting observations with leverage one:
## 596, 1012, 1188, 1371
```





```
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## 596, 1012, 1188, 1371
```





```
# from above, we know basically it can subject to linear regression
# there are some outliers: 524, 1299, 633
# get rid of them now after converting data
```

```
# we would like to train a linear regression model with regulation
# 1. convert categorical ones to factors
for(i in 1:dim(real_all_daat)[2]){
  if(is.character(real_all_daat[,i])){
    real_all_daat[,i] <- as.factor(real_all_daat[,i])
  }
}
```

```
# 2. convert x to matrix
all_data_x_matrix <- model.matrix(~.-SalePrice, data = real_all_daat)
```

```
# 1. split all data into train and prediction
model_x <- all_data_x_matrix[1:1460,]
model_y <- real_all_daat[1:1460,"SalePrice"]
```

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pre_x <- all_data_x_matrix[-c(1:1460),]
```

```
# 2. get rid of outlier in train data
model_x <- model_x[-c(524, 1299, 633),]
model_y <- model_y[-c(524, 1299, 633)]
```

```
# 3. split model data into train and test
set.seed(1000)
```

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train_ind <- sample(1:dim(model_x)[1], size = dim(model_x)[1] * 0.7)
# train_ind <- read.csv(file = "H:/kaggle/houseprice/data/train_index.csv")
# train_ind <- train_ind[,1]

train_x <- model_x[train_ind,]
train_y <- model_y[train_ind]

test_x <- model_x[-train_ind,]
test_y <- model_y[-train_ind]

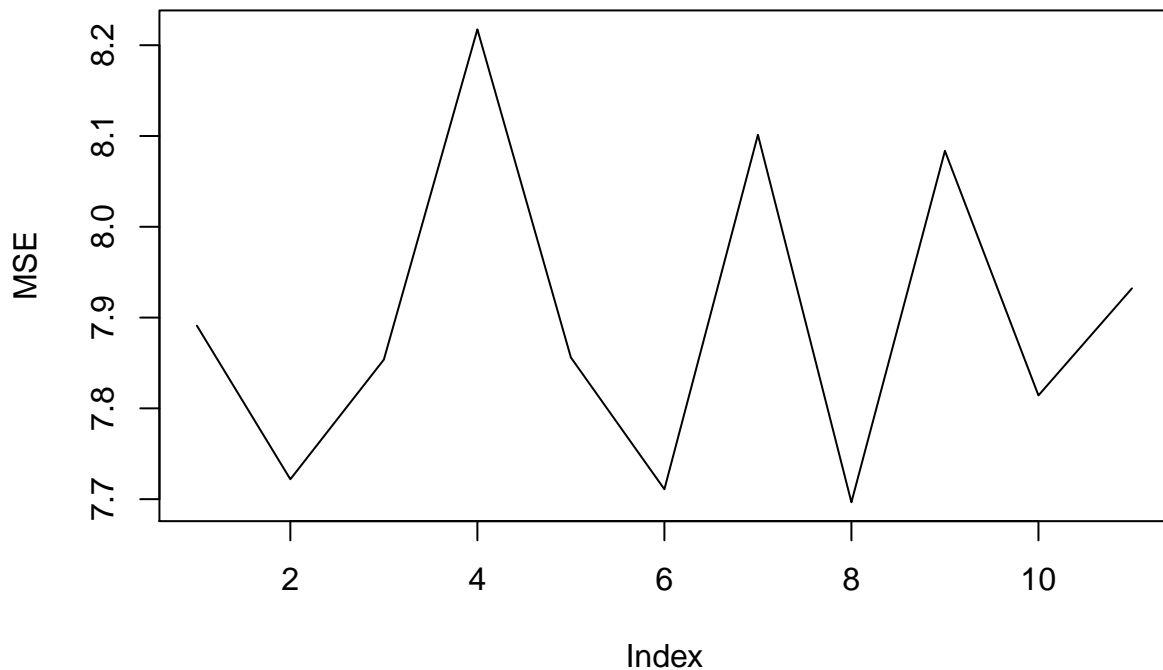
# use cross validation to find optimal alpha in penalty
library(glmnet)

## Loading required package: Matrix
## Loading required package: foreach
## Loaded glmnet 2.0-10

for(i in 0:10){
  assign(paste("fit", i, sep = ""),
        cv.glmnet(x = train_x, y = train_y, alpha = i/10,
                  type.measure = "mse", family = "gaussian"))
}

# find the optimal fit with least sse on test data
fit <- list(fit0, fit1, fit2, fit3, fit4, fit5, fit6, fit7, fit8, fit9, fit10)
mse <- NULL
for(i in 1:11){
  pre_y <- predict.cv.glmnet(object = fit[[i]], newx = test_x, s = fit[[i]]$lambda.1se)
  tmp_mse <- sum((pre_y - test_y) ^ 2)
  mse <- c(mse, tmp_mse)
}
plot(mse, type = 'l', ylab = "MSE")

```



```
# let's use this model to make prediction
opt_ind <- which.min(mse)
log_pre_y <- predict.cv.glmnet(object = fit[[opt_ind]],
                               newx = pre_x, s = fit[[opt_ind]]$lambda.1se)

# create submission
result <- data.frame(Id = c(1461:2919), SalePrice = exp(log_pre_y))
names(result) <- c("Id", "SalePrice")
write.csv(x = result, file = "H:/kaggle/houseprice/data/submission_5.csv",
          row.names = FALSE)

# Your submission scored 0.13634
# although score on kaggle doesn't improve, but from graph above,
# the range of MSE has decreased, I think this model is better.
```