

MSCI 446 - Assignment 2 - Question 3

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Include Packages

```
library('tidyverse')
library('caret')
library('gridExtra')
library('plotly')
library('ISLR')
library('AmesHousing')
library('leaps')

theme_set(theme_classic())
```

Import Dataset

```
ames <- AmesHousing::make_ames()
numericVars <- ames %>% summarize_all(is.numeric) %>% unlist()
ames <- ames[, numericVars]
head(ames)
```

```
## # A tibble: 6 x 35
##   Lot_Frontage Lot_Area Year_Built Year_Remod_Add Mas_Vnr_Area BsmtFin_SF_1
##         <dbl>   <int>    <int>         <int>         <dbl>       <dbl>
## 1          141   31770     1960         1960          112           2
## 2           80   11622     1961         1961           0           6
## 3           81   14267     1958         1958          108           1
## 4           93   11160     1968         1968           0           1
## 5           74   13830     1997         1998           0           3
## 6           78    9978     1998         1998           20           3
## # ... with 29 more variables: BsmtFin_SF_2 <dbl>, Bsmt_Unf_SF <dbl>,
## #   Total_Bsmt_SF <dbl>, First_Flr_SF <int>, Second_Flr_SF <int>,
## #   Low_Qual_Fin_SF <int>, Gr_Liv_Area <int>, Bsmt_Full_Bath <dbl>,
## #   Bsmt_Half_Bath <dbl>, Full_Bath <int>, Half_Bath <int>,
## #   Bedroom_AbvGr <int>, Kitchen_AbvGr <int>, TotRms_AbvGrd <int>,
## #   Fireplaces <int>, Garage_Cars <dbl>, Garage_Area <dbl>, Wood_Deck_SF <int>,
## #   Open_Porch_SF <int>, Enclosed_Porch <int>, Three_season_porch <int>,
## #   Screen_Porch <int>, Pool_Area <int>, Misc_Val <int>, Mo_Sold <int>,
## #   Year_Sold <int>, Sale_Price <int>, Longitude <dbl>, Latitude <dbl>
```

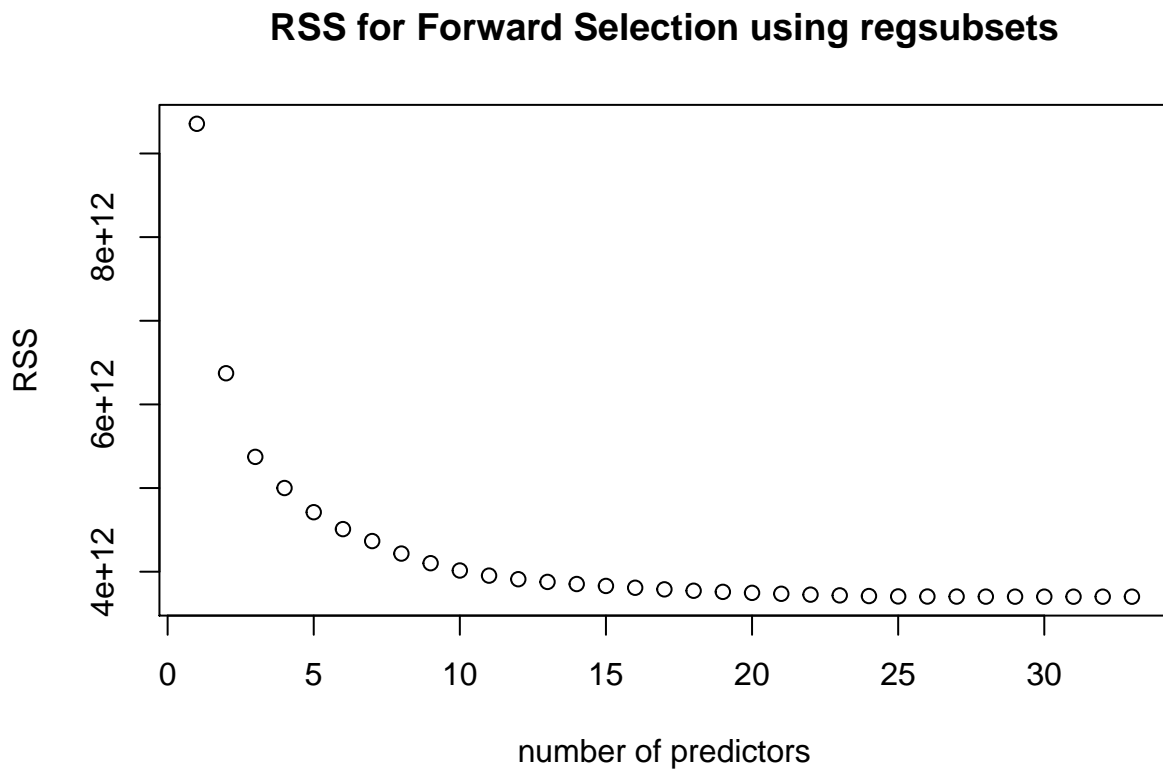
Forward Selection

```
NumCols <- ncol(ames)
res <- regsubsets(Sale_Price ~., data=ames, method='forward', nvmax=NumCols)
```

```
## Reordering variables and trying again:
```

```
smm <- summary(res)
```

```
plot(smm$rss, main="RSS for Forward Selection using regsubsets",
     xlab="number of predictors",ylab="RSS")
```



```
# Find number of predictors for smallest RSS value:
```

```
which.min(smm$rss)
```

```
## [1] 33
```

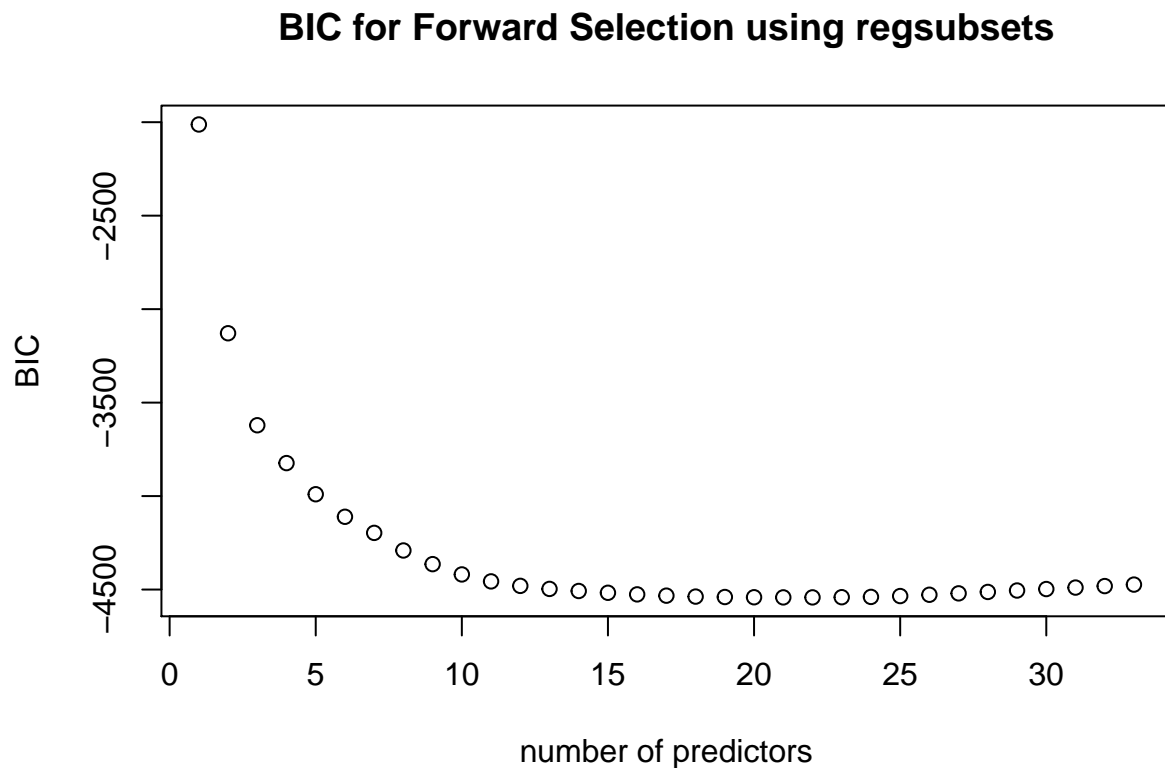
As seen above, using 33 predictors gives the smallest RSS value for Forward Selection. The values of these 33 parameters are shown below:

```
coef(res,33)
```

```
##      (Intercept)      Lot_Frontage      Lot_Area      Year_Built
##      -1.142977e+07      8.737532e+01      3.141331e-01      3.845931e+02
##      Year_Remod_Add      Mas_Vnr_Area      BsmtFin_SF_1      BsmtFin_SF_2
##      5.129858e+02      3.794721e+01      3.002994e+02      -1.338433e+01
##      Bsmt_Unf_SF      Total_Bsmt_SF      First_Flr_SF      Low_Qual_Fin_SF
##      -1.337146e+01      3.759189e+01      3.554565e-01      -4.417005e+01
##      Bsmt_Full_Bath      Bsmt_Half_Bath      Full_Bath      Half_Bath
##      6.504458e+03      -1.883312e+03      1.949198e+03      -3.471763e+03
##      Bedroom_AbvGr      Kitchen_AbvGr      TotRms_AbvGrd      Fireplaces
##      -1.034286e+04      -3.360632e+04      4.068734e+03      7.084818e+03
##      Garage_Cars      Garage_Area      Wood_Deck_SF      Open_Porch_SF
##      7.737977e+03      2.082670e+01      2.430170e+01      -4.100172e+00
##      Enclosed_Porch      Three_season_porch      Screen_Porch      Pool_Area
##      2.974408e+01      8.723251e+00      6.200042e+01      -6.447100e+01
##      Misc_Val      Mo_Sold      Year_Sold      Longitude
##      -9.497111e+00      2.762025e+01      -9.346976e+02      -1.299076e+04
##      Latitude      Gr_Liv_Area
##      2.464128e+05      6.324190e+01
```

Repeat Using BIC Metric

```
plot(smm$bic, main="BIC for Forward Selection using regsubsets",
     xlab="number of predictors", ylab="BIC")
```



```
which.min(smm$bic)
```

```
## [1] 21
```

21 predictors give the smallest BIC value when using Forward Selection. The reason fewer predictors optimize BIC as apposed to the 33 predictors needed to optimize RSS is due to the BIC calculation penalizing the number of predictors used in a model. The 21 parameter model is shown below:

```
coef(res, 21)
```

```
##      (Intercept)  Lot_Frontage      Lot_Area      Year_Built Year_Remod_Add
## -1.804094e+06   9.403297e+01   2.439368e-01   3.616190e+02   5.689112e+02
##   Mas_Vnr_Area  BsmtFin_SF_2    Bsmt_Unf_SF  Total_Bsmt_SF Bsmt_Full_Bath
##  4.363806e+01  -1.280552e+01  -1.309842e+01  4.126980e+01   6.192556e+03
## Bsmt_Half_Bath Kitchen_AbvGr TotRms_AbvGrd   Fireplaces   Garage_Cars
## -4.186852e+03  -3.385257e+04   5.606576e+02   9.867642e+03   1.004416e+04
##   Garage_Area  Wood_Deck_SF  Open_Porch_SF   Pool_Area   Misc_Val
##  2.165199e+01  1.963979e+01   1.895785e+00  -5.499532e+01  -9.029755e+00
##      Mo_Sold   Gr_Liv_Area
##  9.536313e+01  5.928065e+01
```

Backward Selection

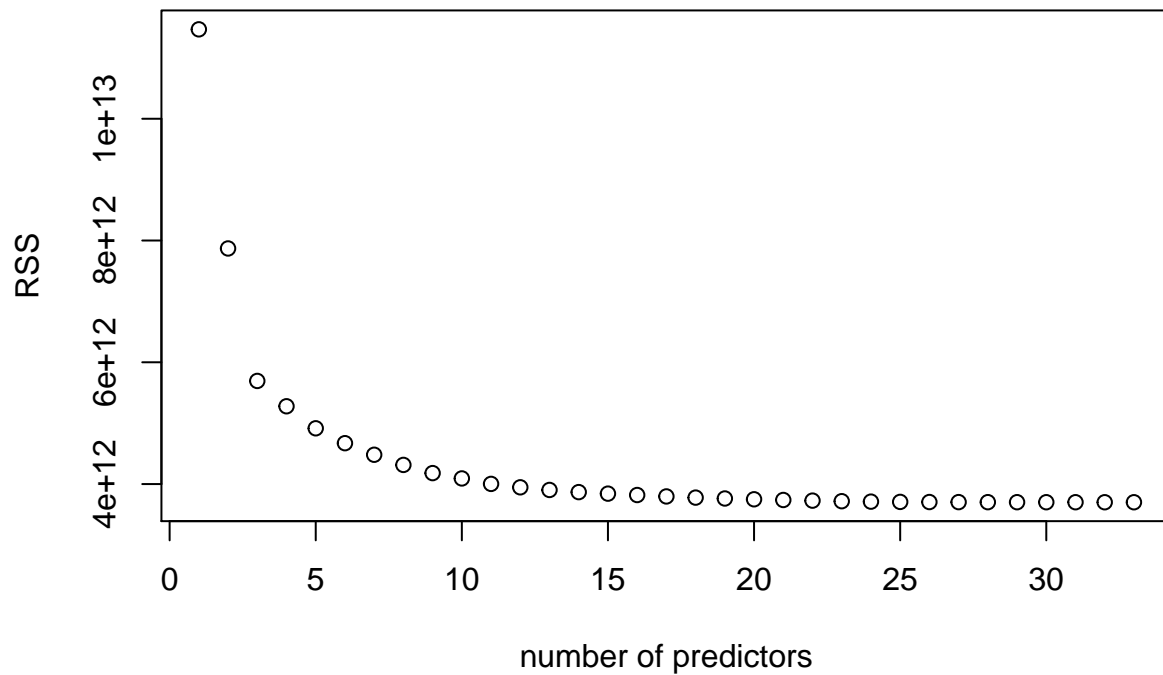
```
resbkw <- regsubsets(Sale_Price ~., data=ames, method='backward', nvmax=NumCols)
```

```
## Reordering variables and trying again:
```

```
smbbkw <- summary(resbkw)
```

```
plot(smbbkw$rss, main="RSS for Backward Selection using regsubsets",
     xlab="number of predictors", ylab="RSS")
```

RSS for Backward Selection using regsubsets



```
which.min(smbkw$rss)
```

```
## [1] 33
```

As seen in Forward Selection, the best RSS value for Backward Selection is at 33 parameters. This model is shown below:

```
coef(resbkw,33)
```

```
##      (Intercept)      Lot_Frontage      Lot_Area      Year_Built
##      -1.170805e+07      8.688692e+01      3.250816e-01      3.915167e+02
##      Year_Remod_Add      Mas_Vnr_Area      BsmtFin_SF_1      BsmtFin_SF_2
##      5.250215e+02      3.754647e+01      1.414811e+02      -1.391134e+01
##      Bsmt_Unf_SF      Total_Bsmt_SF      First_Flr_SF      Second_Flr_SF
##      -1.797736e+01      4.219896e+01      6.308277e+01      6.342274e+01
##      Low_Qual_Fin_SF      Bsmt_Half_Bath      Full_Bath      Half_Bath
##      1.994256e+01      -4.985513e+03      1.170822e+03      -3.889125e+03
##      Bedroom_AbvGr      Kitchen_AbvGr      TotRms_AbvGrd      Fireplaces
##      -1.045933e+04      -3.204082e+04      4.031002e+03      7.123055e+03
##      Garage_Cars      Garage_Area      Wood_Deck_SF      Open_Porch_SF
##      8.075298e+03      1.987748e+01      2.550571e+01      -2.347879e+00
##      Enclosed_Porch      Three_season_porch      Screen_Porch      Pool_Area
##      3.067302e+01      9.134332e+00      6.239160e+01      -6.435958e+01
##      Misc_Val      Mo_Sold      Year_Sold      Longitude
##      -9.835393e+00      4.225967e+01      -8.848423e+02      -1.570146e+04
```

```
##          Latitude          Gr_Liv_Area
##      2.437618e+05      0.000000e+00
```

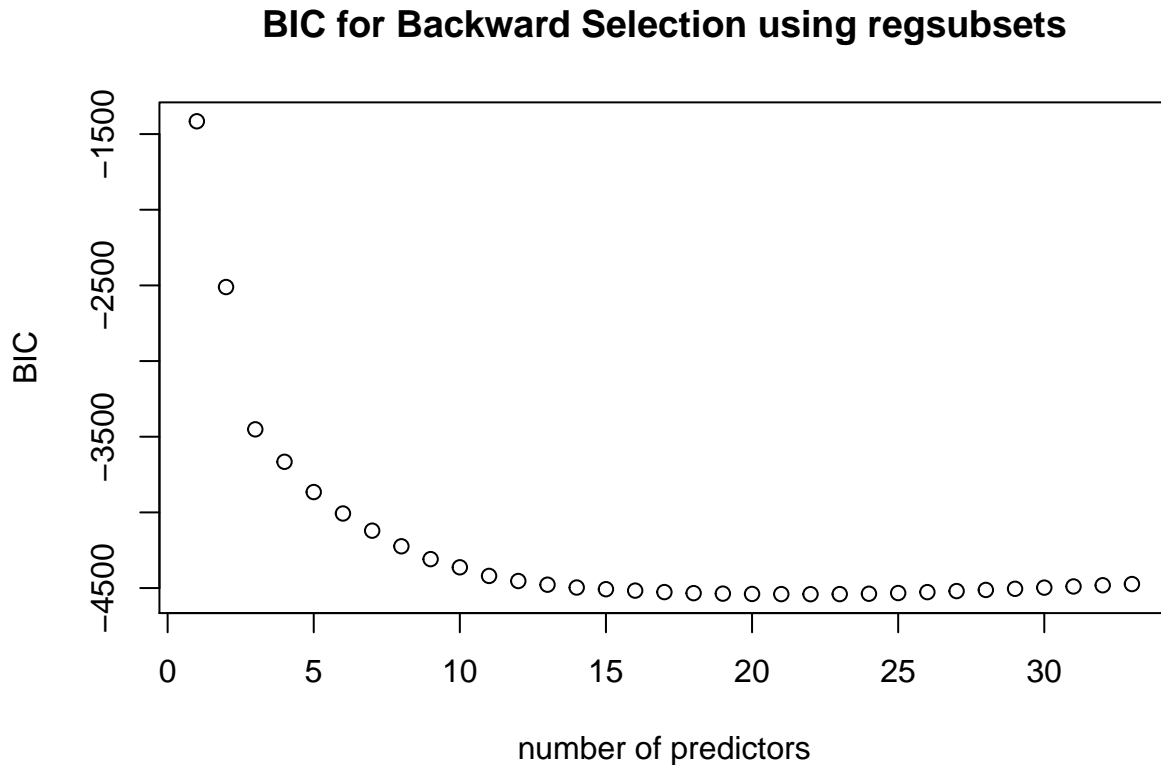
```
names(coef(resbkw,33)) == names(coef(res,33))
```

```
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE FALSE
## [13] FALSE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [25] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
```

It appears as though many of the same predictor variables are being used as well, with a few exceptions.

Repeat Using BIC Metric

```
plot(smbkw$bic, main="BIC for Backward Selection using regsubsets",
     xlab="number of predictors", ylab="BIC")
```



```
which.min(smbkw$bic)
```

```
## [1] 22
```

The model with minimum BIC value was found to have 22 parameters when using Backward Selection. In comparison, the model with minimum BIC value for Backward Selection had 22 parameters. The predictors and their respective coefficients of the 22 variable model is summarized below.

```
coef(resbkw, 22)
```

```
##      (Intercept)   Lot_Frontage      Lot_Area      Year_Built Year_Remod_Add
## -1.816554e+06    8.988071e+01    2.252629e-01    3.568215e+02    5.800933e+02
##   Mas_Vnr_Area   BsmtFin_SF_2    Bsmt_Unf_SF   Total_Bsmt_SF   First_Flr_SF
##  4.228330e+01   -1.357431e+01   -1.785513e+01   4.275903e+01   4.067518e+01
## Second_Flr_SF Bsmt_Half_Bath Kitchen_AbvGr TotRms_AbvGrd   Fireplaces
##  3.517944e+01   -7.244059e+03   -3.430550e+04   6.485669e+02   9.556181e+03
##   Garage_Cars   Garage_Area   Wood_Deck_SF   Open_Porch_SF   Pool_Area
##  1.024918e+04   2.015845e+01   2.092499e+01   3.072430e+00   -5.526520e+01
##      Misc_Val      Mo_Sold    Gr_Liv_Area
## -9.437835e+00   9.540143e+01   2.300514e+01
```