IDS 702: Module 1.12

Bringing the MLR pieces together II (illustration)

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Back to the diamonds data

Let's try model selection for our diamonds example. We will do this on the log scale - recall our analysis in the previous module.

First, forward selection using AIC

```
diamonds <- read.csv("data/diamonds.csv", header= T,
                     colClasses = c("numeric","factor","factor","factor","numeric"))
diamonds$CaratsCent <- diamonds$Carats - mean(diamonds$Carats)
diamonds$CaratsCent2 <- diamonds$CaratsCent^2
NullModel <- lm(log(Price)~1,data=diamonds)
FullModel <- lm(log(Price)~CaratsCent+CaratsCent2+
                  Color*Clarity+Color*Certification+
                  Clarity*Certification,
                data=diamonds)
Model_forward <- step(NullModel, scope = formula(FullModel),direction="forward",trace=0)
# Remove the trace=0 option if you want to function to print the entire process
# Let's see the variables the model selected
Model_forward$call
```

```
## lm(formula = log(Price) ~ CaratsCent + CaratsCent2 + Color +
##     Clarity + Certification + Color:Clarity + Color:Certification,
##     data = diamonds)
```

```
#run summary(Model_forward) to see the results of the final model
```

Back to the diamonds data

Let's do the same using BIC

```
# use k = log(n) to use BIC instead.
n <- nrow(diamonds)
Model_forward <- step(NullModel, scope = formula(FullModel),direction="forward",trace=0,
                      k = log(n))
# Let's see the variables the model selected
Model_forward$call
```

```
## lm(formula = log(Price) ~ CaratsCent + CaratsCent2 + Color +
##     Clarity, data = diamonds)
```

```
#run summary(Model_forward) to see the results of the final model
```

Back to the diamonds data

Backward selection using AIC

```
Model_backward <- step(FullModel,direction="backward",trace=0)
# Let's see the variables the model selected
Model_backward$call
```

```
## lm(formula = log(Price) ~ CaratsCent + CaratsCent2 + Color +
##     Clarity + Certification + Color:Clarity + Color:Certification,
##     data = diamonds)
```

```
#run summary(Model_backward) to see the results of the final model
```

Same result as forward selection using AIC

Back to the diamonds data

Backward selection using BIC

```
Model_backward <- step(FullModel,direction="backward",trace=0,k = log(n))
# Let's see the variables the model selected
Model_backward$call
```

```
## lm(formula = log(Price) ~ CaratsCent + CaratsCent2 + Color +
##     Clarity, data = diamonds)
```

```
#run summary(Model_backward) to see the results of the final model
```

Same result as forward selection using BIC

Back to the diamonds data

Stepwise selection using AIC

```
Model_stepwise <- step(NullModel, scope = formula(FullModel),direction="both",trace=0)
# Let's see the variables the model selected
Model_stepwise$call
```

```
## lm(formula = log(Price) ~ CaratsCent + CaratsCent2 + Color +
##     Clarity + Certification + Color:Clarity + Color:Certification,
##     data = diamonds)
```

```
#run summary(Model_backward) to see the results of the final model
```

Same result as previous results using AIC

Back to the diamonds data

Stepwise selection using BIC

```
Model_stepwise <- step(NullModel, scope = formula(FullModel),direction="both",trace=0,
                       k = log(n))
# Let's see the variables the model selected
Model_stepwise$call
```

```
## lm(formula = log(Price) ~ CaratsCent + CaratsCent2 + Color +
##     Clarity, data = diamonds)
```

```
#run summary(Model_backward) to see the results of the final model
```

Same result as previous results using BIC

Back to the diamonds data

Let's use the regsubsets function.

```
library(leaps)
Model_forward <- regsubsets(log(Price)~CaratsCent+CaratsCent2+Color*Clarity+
                              Color*Certification+Clarity*Certification,data=diamonds,
                            method="forward")
Select_results <- summary(Model_forward)
coef(Model_forward, which.max(Select_results$adjr2)) # Adj R-sq
```

```
## (Intercept)  CaratsCent CaratsCent2      ColorG      ColorH      ColorI
##   8.6185951   3.0050895  -2.0109553  -0.1275071  -0.2147009  -0.3185926
##  ClarityVS1  ClarityVS2 ClarityVVS2
##  -0.1688242  -0.2525954  -0.1116575
```

```
coef(Model_forward, which.min(Select_results$bic)) #BIC
```

```
## (Intercept)  CaratsCent CaratsCent2      ColorG      ColorH      ColorI
##   8.6185951   3.0050895  -2.0109553  -0.1275071  -0.2147009  -0.3185926
##  ClarityVS1  ClarityVS2 ClarityVVS2
##  -0.1688242  -0.2525954  -0.1116575
```

Back to the diamonds data

```
Model_backward <- regsubsets(log(Price)~CaratsCent+CaratsCent2+Color*Clarity+
                               Color*Certification+Clarity*Certification,data=diamonds,
                             method="backward")
Select_results <- summary(Model_backward)
coef(Model_backward, which.max(Select_results$adjr2)) # Adj R-sq
```

```
## (Intercept)  CaratsCent CaratsCent2      ColorG      ColorH      ColorI
##   8.6185951   3.0050895  -2.0109553  -0.1275071  -0.2147009  -0.3185926
##  ClarityVS1  ClarityVS2 ClarityVVS2
##  -0.1688242  -0.2525954  -0.1116575
```

```
coef(Model_backward, which.min(Select_results$bic)) #BIC
```

```
## (Intercept)  CaratsCent CaratsCent2      ColorG      ColorH      ColorI
##   8.6185951   3.0050895  -2.0109553  -0.1275071  -0.2147009  -0.3185926
##  ClarityVS1  ClarityVS2 ClarityVVS2
##  -0.1688242  -0.2525954  -0.1116575
```

What's next?

Move on to the readings for the next module!