## Interaction Term Testing

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```
Loading the dataset
```

```
file_path = '/Users/Tarek/Documents/UCI_MDS_Coding/Stats210P/R_Statistical_Modeling/Datasets/MidwestSal
df = read.table(file_path, header=FALSE, sep="", dec=".")
```

Give labels to the columns of the dataset.

```
names(df)=c("id","price","sqft","bed","bath","ac","garage","pool","year","quality","style","lot","hwy")
```

Summary of data set

```
str(df)
```

```
## 'data.frame':
                 522 obs. of 13 variables:
## $ id : int 1 2 3 4 5 6 7 8 9 10 ...
## $ price : int 360000 340000 250000 205500 275500 248000 229900 150000 195000 160000 ...
## $ sqft : int 3032 2058 1780 1638 2196 1966 2216 1597 1622 1976 ...
## $ bed
          : int 4444443233...
## $ bath : int 4 2 3 2 3 3 2 1 2 3 ...
## $ ac
          : int 1 1 1 1 1 1 1 1 0 ...
   $ garage : int 2 2 2 2 2 5 2 1 2 1 ...
## $ pool
          : int 0000010000...
## $ year
          : int 1972 1976 1980 1963 1968 1972 1972 1955 1975 1918 ...
## $ quality: int 2 2 2 2 2 2 2 3 3 ...
## $ style : int 1 1 1 1 7 1 7 1 1 1 ...
## $ lot : int 22221 22912 21345 17342 21786 18902 18639 22112 14321 32358 ...
## $ hwy : int 0000000000...
```

Creating 2 different models with the same predictors except Model 2 will have an added interaction term.

$$\hat{Y} = Price$$
 
$$X_1 = sqft$$
 
$$X_2 = NumOfBedroomrs$$
 
$$X_1 * X_2 = sqft * NumOfBedroomrs$$

## Model 1

$$\hat{Y} = \hat{\beta_0} + \hat{\beta_1} X_1 + \hat{\beta_2} X_2$$

```
# Model 1
model_1 = lm(price ~ sqft + bed, data=df)
summary(model_1)
##
## Call:
## lm(formula = price ~ sqft + bed, data = df)
##
## Residuals:
##
       Min
                1Q Median
                                ЗQ
                                       Max
                     -8693
## -227961
           -38270
                             24670
                                    381949
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -66971.563 13404.170 -4.996
                                               8e-07 ***
## sqft
                  165.832
                               5.855 28.326
                                               <2e-16 ***
## bed
               -8647.511
                            4104.008 -2.107
                                              0.0356 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 78860 on 519 degrees of freedom
## Multiple R-squared: 0.6743, Adjusted R-squared: 0.6731
## F-statistic: 537.3 on 2 and 519 DF, p-value: < 2.2e-16
```

## Model 2

```
\hat{Y} = \hat{\beta}_0 + \hat{\beta}_1 X_1 + \hat{\beta}_2 X_2 + \hat{\beta}_3 (X_1 * X_2)
```

```
# Model 2
model_2 = lm(price ~ sqft + bed + sqft * bed, data=df)
summary(model_2)
##
## Call:
## lm(formula = price ~ sqft + bed + sqft * bed, data = df)
## Residuals:
##
      Min
               1Q Median
                               3Q
                                     Max
## -196725 -41800
                   -6185
                            27098 376624
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.033e+05 3.625e+04 -5.607 3.36e-08 ***
              2.281e+02 1.646e+01 13.857 < 2e-16 ***
## sqft
               2.754e+04 9.835e+03 2.800 0.00529 **
## bed
            -1.576e+01 3.904e+00 -4.037 6.23e-05 ***
## sqft:bed
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 77730 on 518 degrees of freedom
## Multiple R-squared: 0.6843, Adjusted R-squared: 0.6824
## F-statistic: 374.2 on 3 and 518 DF, p-value: < 2.2e-16
```