HW1 Question 1

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Question 1 - Box Office Ticket Sales

Introducing the Dataset

This section is dedicated to see and understand the data that was provided from the weekly reports about the box office ticket sales for plays in Broadway in New York. The data being observed is of the week of October 11- 17, 2004. The dataset contains data about the gross box office results for the current week October 11-17, 2004 and that of the previous week October 3-10, 2004.

The data table is as follows:

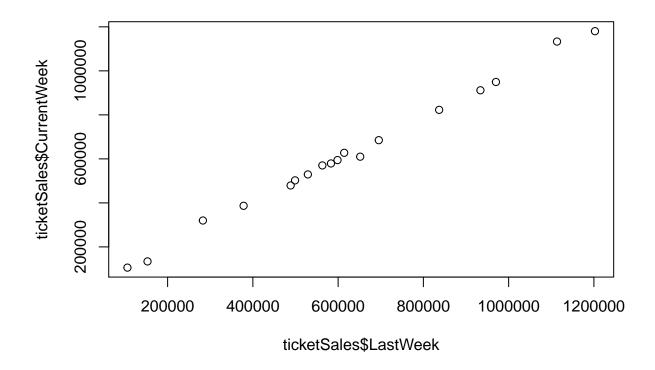
```
ticketSales = read.csv("playbill.csv", header=T)
ticketSales[,]
```

##			Production	${\tt CurrentWeek}$	LastWeek
##	1		42nd Street	684966	695437
##	2		Avenue Q	502367	498969
##	3		Beauty and Beast	594474	598576
##	4		Bombay Dreams	529298	528994
##	5		Chicago	570254	562964
##	6		Dracula	319959	282778
##	7		Fiddler on the Roof	579126	583177
##	8		Forever Tango	134042	152833
##	9		Golda's Balcony	105853	105698
##	10		Hairspray	822775	836959
##	11		Mamma Mia!	949462	970190
##	12		Movin' Out	610007	651808
##	13		Rent	386797	378238
##	14		The Lion King	1133034	1113510
##	15	The	Phantom of the Opera	627609	614246
##	16		The Producers	911727	933822
##	17		Wicked	1180266	1202536
##	18		Wonderful Town	479155	488624

Visualizing the Dataset

The dataset can be visalized through a scatterplot in the figure below:

```
plot(x = ticketSales$LastWeek, y = ticketSales$CurrentWeek)
```



Fitting a linear model

The linear trend in the scatterplot seems strong, which means that a linear regression model is appropriate. The linear model that we are trying to fit is of the form:

$$\hat{Y} = \hat{\beta_0} + \hat{\beta_1} X$$

here, X = Gross Box Office results for Previous Week Y = Gross Box Office results for Current Week.

Through this linear model we are trying to predict the value of actual gross box office results of current week (Y) using the gross box office results of the previous week (X). The following snippet of the R-Code fits a linear model on the data, prints out the values of the intercept and slope and also provides a summary of the fitted model.

```
mod = lm( ticketSales$CurrentWeek ~ ticketSales$LastWeek )
mod

##
## Call:
## lm(formula = ticketSales$CurrentWeek ~ ticketSales$LastWeek)
##
## Coefficients:
## (Intercept) ticketSales$LastWeek
## 6804.8860 0.9821
```

After fitting the model and observing the values of the parameters $\hat{\beta}_0$, $\hat{\beta}_1$ we find that:

$$\hat{\beta}_0 = 6804.8860$$

$$\hat{\beta}_1 = 0.9821$$

The summary of the model is as follows:

```
summary(mod)
```

```
##
## Call:
## lm(formula = ticketSales$CurrentWeek ~ ticketSales$LastWeek)
##
## Residuals:
##
     Min
             1Q Median
                           30
                                 Max
  -36926 -7525 -2581
                         7782
                               35443
## Coefficients:
##
                        Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                       6.805e+03 9.929e+03
                                              0.685
                                                       0.503
## ticketSales$LastWeek 9.821e-01 1.443e-02 68.071
                                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 18010 on 16 degrees of freedom
## Multiple R-squared: 0.9966, Adjusted R-squared: 0.9963
## F-statistic: 4634 on 1 and 16 DF, p-value: < 2.2e-16
```

The summary provides us with a lot of useful values that we will use in the upcoming sections to calculate the Confidence Intervals, Prediction Intervals and perform Hypothesis Testing. Some of the values are as follows:

Sum of Residuals Square (s) = 18010\ Degrees of freedom (n-2) = 16\

Combining all the information from the previous sections the final fitted model looks as follows:

$$Y = 6804.8860 + (0.9821 * X)$$

Finding a 95% Confidence Interval for β_1

A $100(1-\alpha)\%$ confidence interval for $\hat{\beta}_1$ is given by the following formula:

$$\hat{\beta_1} - t_{\frac{\alpha}{2},n-2} \tfrac{s}{\sqrt{S_{XX}}} \leq \beta_1 \leq \hat{\beta_1} + t_{\frac{\alpha}{2},n-2} \tfrac{s}{\sqrt{S_{XX}}}$$

The following snippet of code is used to find the 95% confidence interval for $\hat{\beta}_1$:

```
bet1_hat = 0.9821 #found using the linear model
x col = ticketSales$LastWeek
y_col = ticketSales$CurrentWeek
x_bar = mean(x_col)
\#x_bar = 622186.6
y_bar = mean(y_col)
#y_bar = 617842.8
sxx = sum((x_col-x_bar)^2)
\#sxx = 1.557916 * 10^12
sxy = sum((x_col-x_bar)*(y_col-y_bar))
\#sxy = 1.53 * 10^12
\#sxy/sxx = 0.9821
s = 18010 #found through the summary of the linear model
t_mult = 2.120 # found using the T Table, equal to t0.25 for 16 degrees of freedom
beta1CIlower = bet1_hat - (t_mult*(s/sqrt(sxx)))
beta1CIupper = bet1_hat + (t_mult*(s/sqrt(sxx)))
```

The values we found are as follows:

```
\bar{x} = 622186.6

\bar{y} = 617842.8

S_{XX} = 1.557916 * (10^{1}2)

S_{XY} = 1.53 * (10^{1}2)

\hat{\beta}_{1} = 0.9820815
```

The 95% confidence interval can thus be given as follows:

$$0.9515101 \le \beta_1 \le 1.01269$$

Yes we can say that 1 is a reasonable value for β_1 because 1 lies in the 95% Confidence Interval of β_1

Hypothesis Testing for β_0

We need to run a Hypothesis Test for β_0 given by:

$$Null(H_0)\beta_0 = 10000Alternate(H_1)\beta_0 \neq 10000$$

The hypothesis test for β_0 is given by the following formula:

$$\hat{T} = \frac{\hat{\beta_0} - \beta_0}{s\sqrt{\frac{1}{n} + \frac{\bar{x}^2}{S_{XX}}}}$$

To find the p-value we then use the t-statistic we got from the previous formula and find the probability as follows:

$$p - val = 2 \cdot P(t \ge |\hat{T}|)$$

The following snippet of code is used to perform the Hypothesis Test for β_0 :

```
bet0_hat = 6804.8860 #this value was found using the linear model through r in the previous section x\_col = ticketSales$LastWeek x\_bar = mean(x\_col) #x\_bar = 622186.6 sxx = sum((x_col-x_bar)^2) #sxx = 1.557916 * 10^12 s = 18010 #found through the summary of the linear model t_mult = 2.120 # found using the T Table, equal to t0.25 for 16 degrees of freedom n = length(x\_col) beta0test = ((bet0_hat - 10000)/(s * sqrt((1/n) + ((x_bar^2)/sxx)))) #beta0test = -0.3217422 p\_val = 2*pt(-abs(beta0test), df = n-2) #p\_val = 0.7518132
```

After performing the Hypothesis test for β_0 we find that the p-value=0.7518132 which is greater than 0.05, hence we cannot reject the null hypothesis.