

# Ch 6 Example - Dwaine Studio Sales

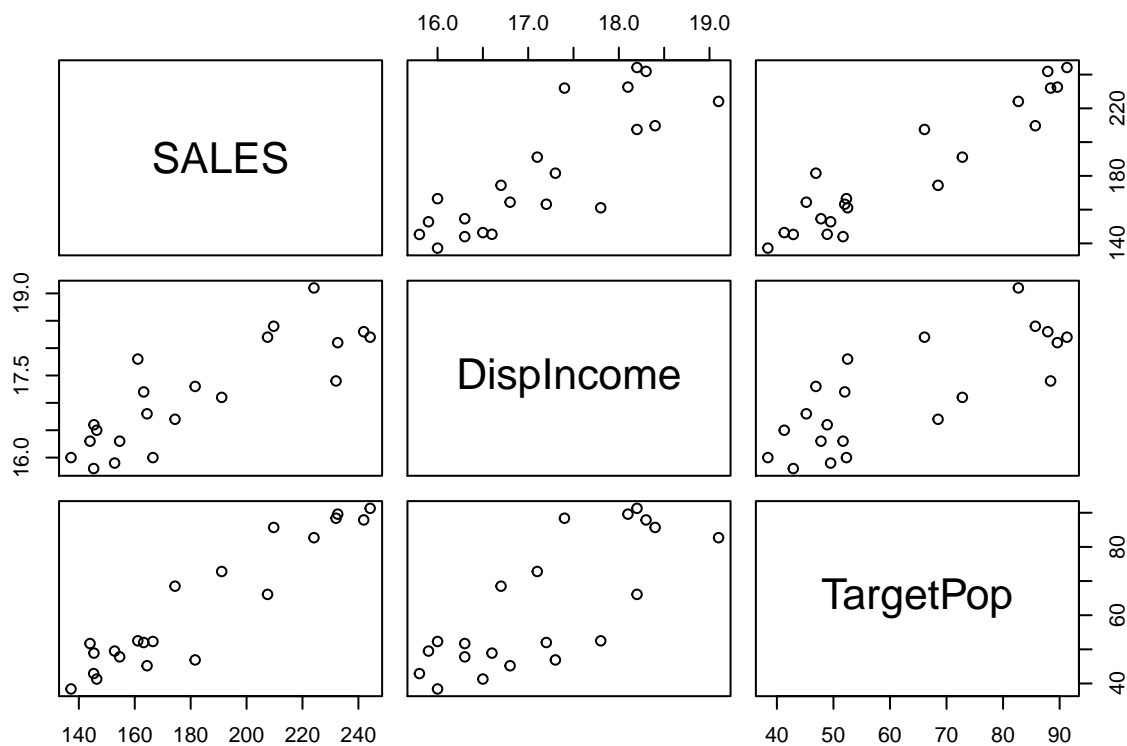
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*Sunday, April 03, 2016*

```
library("ggplot2")
setwd("C:\\Users\\AdamMcQuistan\\Documents\\ISQA 8340\\ch 6")
df <- read.csv("data/ch 6 sales.csv")
df$id = NULL
df <- with(df, data.frame(SALES, DispIncome, TargetPop))
```

## View General Linearity Plots Among Predictors and Outcome Vars

```
pairs(df, pch=21)
```



Sales appears to have a linear relationship with the target population (individuals less than or equal to 16 yrs old) and the disposable income of the cities.

## Building the Model

```

result <- lm(SALES ~ DispIncome + TargetPop, data=df)
result_smry <- summary(result)
F_stat <- round(as.numeric(result_smry$fstatistic["value"]),1)
F_crit <- round(qf(0.95, df1=2, df2=result_smry$df[2]),1)
result_smry

##
## Call:
## lm(formula = SALES ~ DispIncome + TargetPop, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -18.4239  -6.2161   0.7449   9.4356  20.2151
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -68.8571    60.0170  -1.147   0.2663
## DispIncome   9.3655     4.0640   2.305   0.0333 *
## TargetPop    1.4546     0.2118   6.868   2e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.01 on 18 degrees of freedom
## Multiple R-squared:  0.9167, Adjusted R-squared:  0.9075
## F-statistic: 99.1 on 2 and 18 DF,  p-value: 1.921e-10

```

Analyze and test for the relatedness of sales to the target population and per capita disposable income. F critical is 3.6.

$H_0 : \beta_1 = 0 \text{ and } \beta_2 = 0 \text{ for } F^* \leq F \text{ critical}$

$H_a : \text{not both } \beta_1 \text{ and } \beta_2 \text{ equal zero for } F^* > F \text{ critical}$

Since  $F^* (99.1) > F \text{ critical } (3.6)$  we conclude that sales are related to target population and per capital income but, the usefulness of the model for making predictions have yet to be determined.

The P-value is  $< 0.05$  further indicating the rejection of  $H_0$ .

## Part C. Calculation coefficient of multiple determination.

The Multiple  $R^2$  value is 0.9167 which translates into the model explaining about 92% of the variation in sales. The Adjusted  $R^2$  value of 0.9075 shows the correction for more than on predictor variable overfitting which is a very small change from the standard multiple r-squared value indicating little effects of overfitting.

To analyze and test the individual parameters of the regression model we use the students t-test such that:

$H_0 : \beta_p = 0$  indicating no relationship

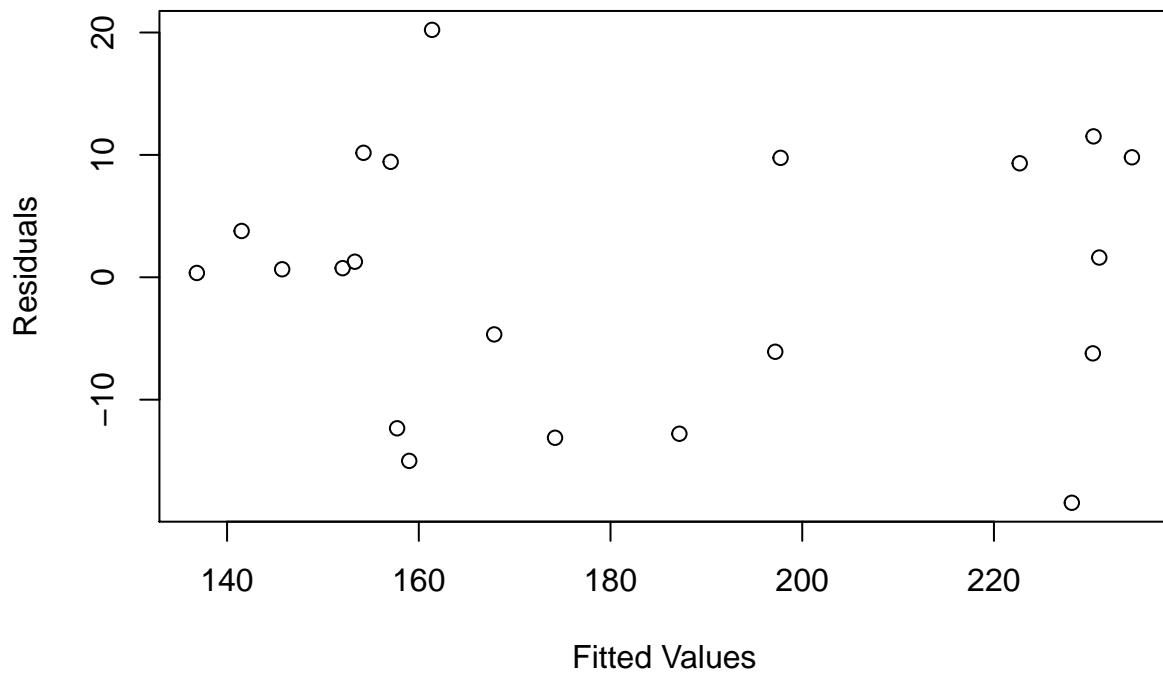
$H_a : \beta_p \neq 0$  indicating there is a relationship

Since the p-values for both parameters (Disposable income and Target population) are less than 0.05 we conclude each individual parameter is significant and related to sales.

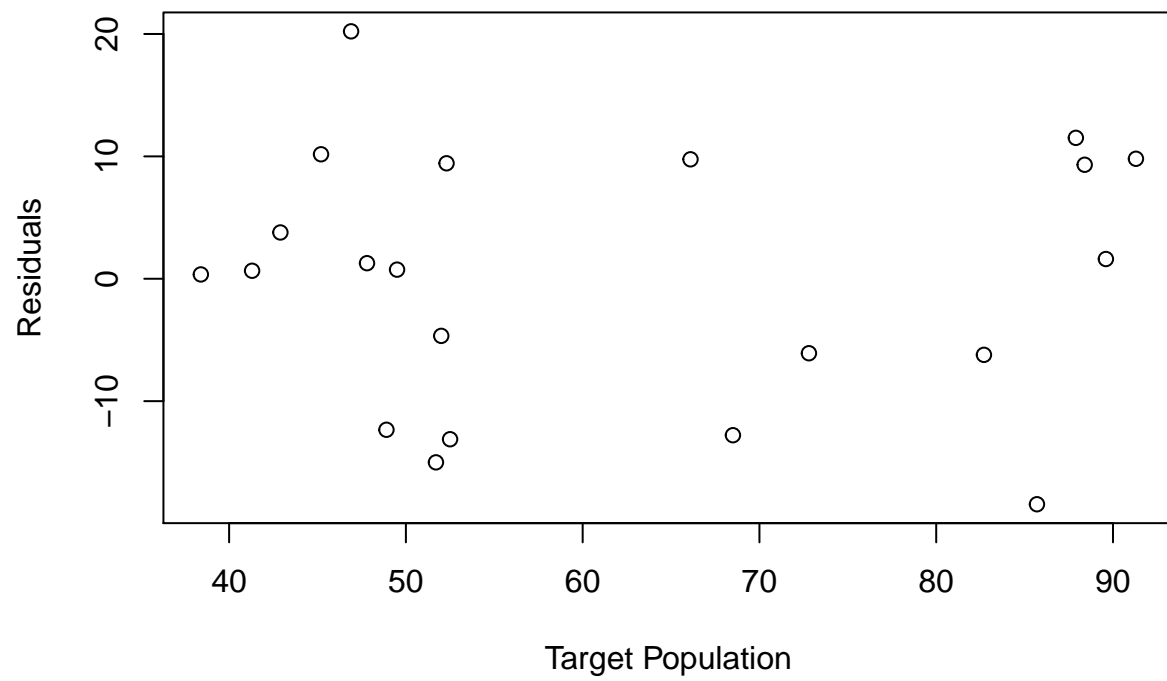
## Graphical Assessment of the Model (Residual Analysis)

```
df_model <- result$model[,1:3]
df_model$FittedVals <- result$fitted.values
df_model$Residuals <- result$residuals

plot(x=df_model$FittedVals, y=df_model$Residuals, xlab="Fitted Values", ylab="Residuals")
```



```
plot(x=df_model$TargetPop, y=df_model$Residuals, xlab="Target Population", ylab="Residuals")
```



```
plot(x=df_model$DispIncome, y=df_model$Residuals, xlab="Disposable Income", ylab="Residuals")
```

