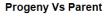
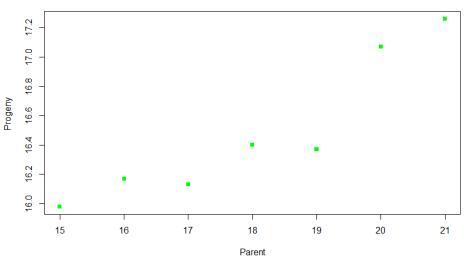
# **STAT 8030 HW 4**

## 1.1 (7.7.1):

Scatter Plot is shown below:





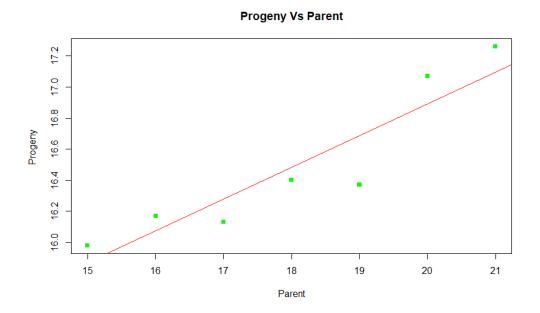
As we see above, there are very few data points. Hence, we cannot have any assumption and conclusion for a pattern.

### 1.2 (7.7.2):

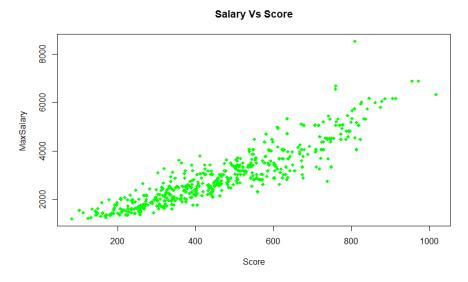
We can see below the summary for computing weighted regression:

#### > summary(ElmWeighted) call: lm(formula = Progeny ~ Parent, data = galtonpeas, weights = 1/SD^2) Weighted Residuals: 3 2 Coefficients: Estimate Std. Error t value Pr(>|t|) 0.68112 18.787 7.87e-06 \*\*\* (Intercept) 12.79642 5.368 0.00302 \*\* 0.03815 Parent 0.20480 Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.05 '.' 0.1 ' '1 Residual standard error: 0.11 on 5 degrees of freedom Multiple R-squared: 0.8521, Adjusted R-squared: F-statistic: 28.81 on 1 and 5 DF, p-value: 0.003021

And below is the scatter plot for fitting regression:



**2.1**Scatter Plot for Maxsalary vs Score is shown below:



As we see above, the pattern does not look to be linear since the datapoints seems to be making a curve, though we can see an uphill pattern in the datapoints.

Summary and scatter plot for Linear Model is shown below:

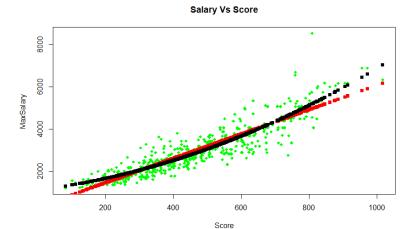
## 

Salary Vs Score

Summary and scatter plot (in black) for Model with degree 2 is shown below:

Score

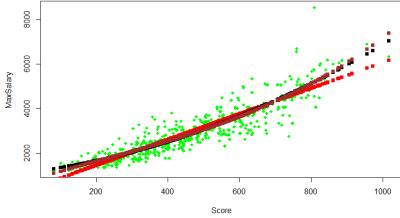
```
> summary(1m_2)
call:
lm(formula = MaxSalary ~ Score + I(Score^2), data = salarygov)
Residuals:
              1Q Median
    Min
                                 3Q
                                          Max
-1877.0 -251.8
                    -67.2
                              251.2 3344.7
Coefficients:
               Estimate Std. Error t value Pr(>|t|)
(Intercept) 1.102e+03 1.320e+02 8.345 7.23e-16 ***
Score 2.007e+00 5.613e-01 3.575 0.000384 ***
I(Score^2) 3.750e-03 5.484e-04
                                       6.838 2.39e-11 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 485.1 on 492 degrees of freedom
Multiple R-squared: 0.8325, Adjusted R-squared: 0.8318
F-statistic: 1222 on 2 and 492 DF, p-value: < 2.2e-16
```



## Summary and scatter plot (in brown) for Model with degree 3 is shown below:

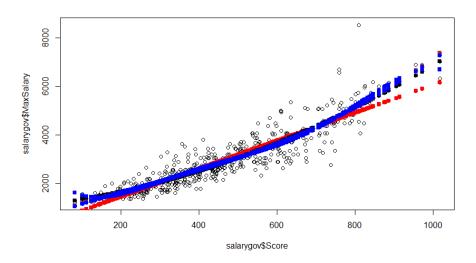
```
lm(formula = MaxSalary ~ Score + I(Score^2) + I(Score^3), data = salarygov)
Residuals:
Min 1Q Median
-1842.8 -257.8 -45.7
                            3Q Max
245.0 3343.5
Coefficients:
               Estimate Std. Error t value Pr(>|t|)
(Intercept)
              7.077e+02
                          2.560e+02
                                        2.764
                                                0.00593 **
                                                0.00449 **
score
              4.956e+00
                          1.736e+00
                                        2.855
I(Score^2)
                          3.602e-03
             -2.640e-03
                                       -0.733
                                                0.46398
I(Score^3)
             4.139e-06 2.306e-06
                                       1.795
                                               0.07334 .
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 484 on 491 degrees of freedom
Multiple R-squared: 0.8336, Adjusted R-squared: 0.8325
F-statistic: 819.7 on 3 and 491 DF, p-value: < 2.2e-16
```





Summary and scatter plot (in blue) for Model with degree 5 is shown below:

```
> summary(1m_5)
call:
Im(formula = MaxSalary ~ Score + I(Score^2) + I(Score^3) + I(Score^4) +
I(Score^5), data = salarygov)
Residuals:
    Min
               1Q Median
                                  3Q
                                          мах
-1831.3 -274.9
                     -56.1
                              236.6
Coefficients:
                Estimate Std. Error t value Pr(>|t|)
(Intercept) 2.896e+03 8.195e+02
                                         3.534 0.000449 ***
score
              -2.562e+01
                            1.040e+01
                                         -2.464 0.014080 *
I(Score^2)
                                          3.056 0.002366 **
               1.470e-01
                            4.812e-02
                                         -3.174 0.001597 **
I(Score^3)
              -3.245e-04
                           1.022e-04
I(Score^4) 3.300e-07 1.008e-07 3.273 0.001141 **
I(Score^5) -1.231e-10 3.732e-11 -3.297 0.001047 **
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 479.7 on 489 degrees of freedom
Multiple R-squared: 0.8372, Adjusted R-squared: 0.8
F-statistic: 502.9 on 5 and 489 DF, p-value: < 2.2e-16
                                     Adjusted R-squared: 0.8355
```



#### 2.3

## F-Tests:

For Fitted model with d=1 and d=2

Null Hypothesis => Ho:  $\beta$ 2 = 0 Alternate Hypothesis => Ha:  $\beta$ 2!=0

We can see F Statistic: F-Value = 46.754 and P-Value = 2.393e-11

Now, considering alpha = 0.05, we see that p-value is less than alpha. Hence, we reject the null hypothesis. Thus, adding the quadratic term in model can improves its accuracy.

For Fitted model with d=2 and d=3

Null Hypothesis => Ho:  $\beta$ 3 = 0 Alternate Hypothesis => Ha:  $\beta$ 3!=0

We can see F Statistic: F-Value = 3.2205 and P-Value = 0.07334

Now, considering alpha = 0.05, we see that p-value is more than alpha. Hence, we fail to reject the null hypothesis. Thus, adding the quadratic term in model will not improve its accuracy.

For Fitted model with d=1 and d=5

Null Hypothesis => Ho:  $\beta$ 5 = 0 Alternate Hypothesis => Ha:  $\beta$ 5!=0

We can see F Statistic: F-Value = 4.7229 and P-Value = 0.002931

Now, considering alpha = 0.05, we see that p-value is less than alpha. Hence, we reject the null hypothesis. Thus, adding the quadratic term in model can improves its accuracy.

## R-squared values for the models:

```
> summary(N_lm)$r.squared
[1] 0.8165477
> summary(lm_2)$r.squared
[1] 0.8324681
> summary(lm_3)$r.squared
[1] 0.8335598
> summary(lm_5)$r.squared
[1] 0.8371856
> |
```

As we see above, R-squared value for model d=5 is changed the max. Thus, confirming that the model with d=5 is the most accurate model and the coefficients for the same (d=5) is shown below:

Scatter Plot with fitted mean function is shown below:



#### 2.4

We can use the NE column to find the weights for the total number of employees currently employed in specific job class.

```
> AW = salarygov$NE
```

#### 2.5

Coefficients for weighted model is shown below:

```
2 - -
> #2.5
> MW <- lm(MaxSalary~Score, data = salarygov, weights = AW)
> summary(MW)
lm(formula = MaxSalary ~ Score, data = salarygov, weights = AW)
Weighted Residuals:
  Min 1Q Median
                        3Q
                              Max
-6549.1 -558.7 -83.7 497.8 10445.6
Coefficients:
         Estimate Std. Error t value Pr(>|t|)
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 1282 on 493 degrees of freedom
Multiple R-squared: 0.8337, Adjusted R-squared: 0.8334
F-statistic: 2472 on 1 and 493 DF, p-value: < 2.2e-16
```

Coefficients for unweighted model is shown below:

Seeing above, we can say that the increase in maximum salary is 5.5961\$ for weighted model and is 5.76\$ for the unweighted model.