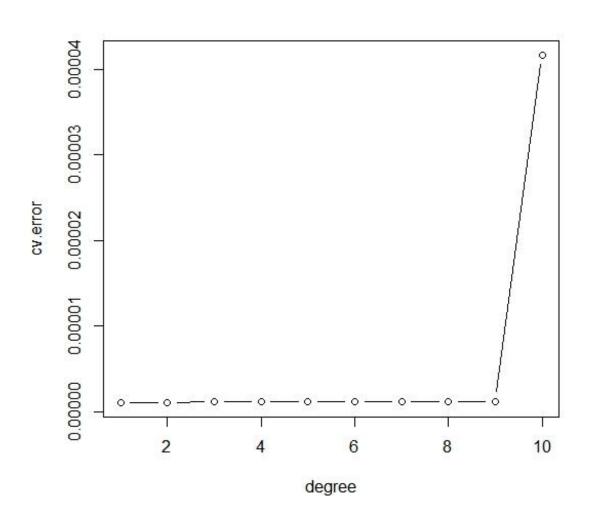
Following from our previous assignment we will now try to estimate the minimum MSE based on the degree of our linear model (For degree 2 to 10, with LOOCV method):

```
library(boot)
set.seed(1994)
cv.error=rep(0,10)
degree=1:10
for (d in degree) {
    glm.fit=glm(test$"100regist2013"~poly(test$"100labour2013",d), data=test)
    cv.error[d] = cv.glm(test, glm.fit)$delta[1]
}
cv.error [1]
0.000001039363
0.000001067604
0.000001128878
0.000001138033
0.000001138054
0.000001161906
0.000001175736
0.000001190911
0.000001195136
0.000041671210
plot(degree, cv.error, type="b")
```

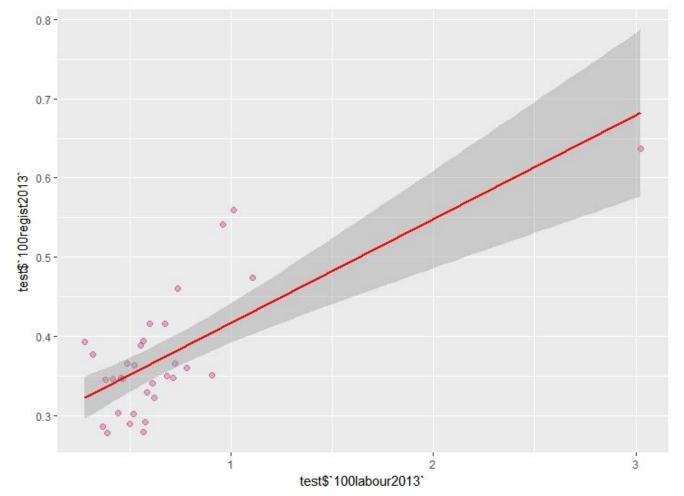


Also we will retrieve the summaries of Linear Models and as well as their plots ( $1^{st}$  to  $4^{th}$  degree):

# Linear (1<sup>st</sup> degree)

```
lm(formula = test$`100regist2013` ~ poly(test$`100labour2013`,
  1))
Residuals:
   Min
         1Q Median
                       3Q
-0.080301 -0.045076 -0.007448 0.035815 0.141547
Coefficients:
               Estimate Std. Error t value
(Intercept)
                 0.374262 0.009984 37.486
Pr(>|t|)
                 < 0.00000000000000002 ***
(Intercept)
Signif. codes: 0 \***' 0.001 \**' 0.01 \*' 0.05 \.' 0.1 \' 1
Residual standard error: 0.05648 on 30 degrees of freedom
```

Multiple R-squared: 0.5531, Adjusted R-squared: 0.5382 F-statistic: 37.13 on 1 and 30 DF, p-value: 0.000001073



# Quadratic (2<sup>nd</sup> degree)

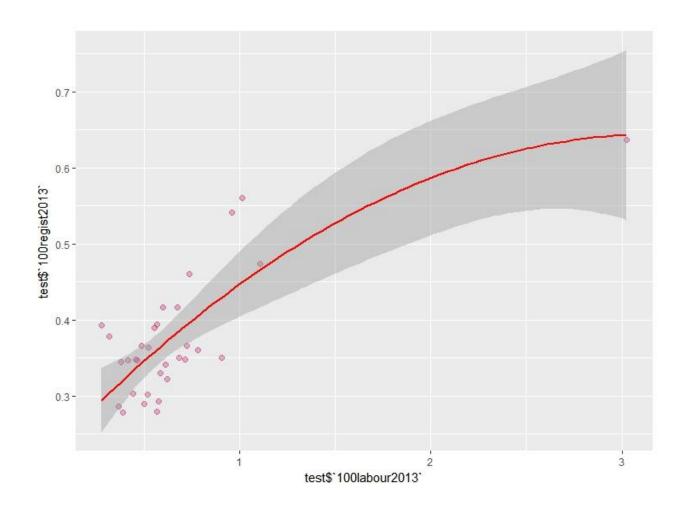
#### Residuals:

Min 1Q Median 3Q Max -0.08059 -0.04251 0.00083 0.03162 0.11089

#### Coefficients:

Signif. codes: 0 \\*\*\*' 0.001 \\*\*' 0.01 \\*' 0.05 \.' 0.1 \' 1

Residual standard error: 0.05466 on 29 degrees of freedom Multiple R-squared: 0.5953, Adjusted R-squared: 0.5674 F-statistic: 21.33 on 2 and 29 DF, p-value: 0.000002008



## Qubic (3<sup>rd</sup> degree)

```
Call:
```

lm(formula = test\$`100regist2013` ~ poly(test\$`100labour2013`,
 3))

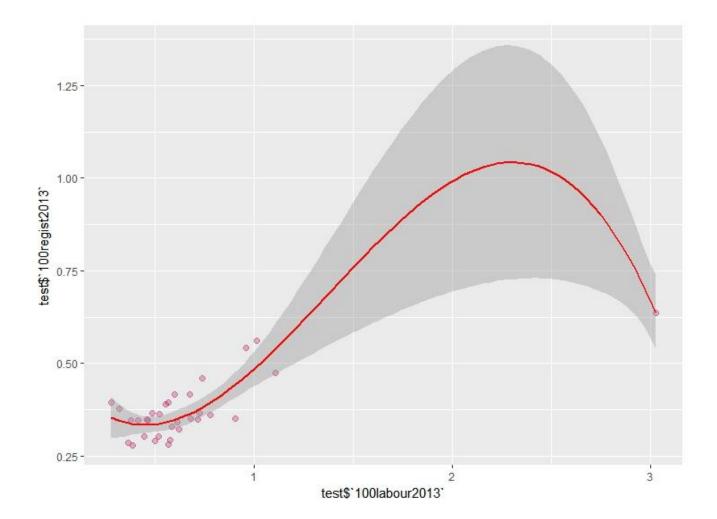
#### Residuals:

Min 1Q Median 3Q Max -0.090959 -0.033512 -0.004153 0.034715 0.079960

#### Coefficients:

Signif. codes: 0 \\*\*\*' 0.001 \\*\*' 0.01 \\*' 0.05 \.' 0.1 \' 1

Residual standard error: 0.04893 on 28 degrees of freedom Multiple R-squared: 0.6869, Adjusted R-squared: 0.6534 F-statistic: 20.48 on 3 and 28 DF, p-value: 0.0000003172



# Quartic (4<sup>th</sup> degree )

#### Call:

```
lm(formula = test$`100regist2013` ~ poly(test$`100labour2013`,
4))
```

### Residuals:

Min 1Q Median 3Q Max -0.105093 -0.035615 -0.004116 0.034964 0.071204

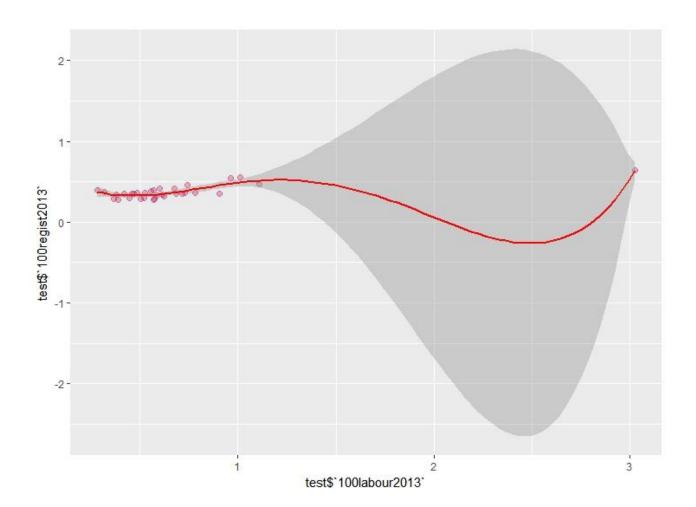
## Coefficients:

	Estimate Std. Error t value		Pr(> t )	
(Intercept)	0.374262 0.0	<mark>08615 43.445</mark>	5 < 0.00000	00000000002 ***
poly(test\$`100la	bour2013`, 4)1 0.34416	58 0.048732	7.063	0.000000136 ***
poly(test\$`100la	bour2013`, 4)2 -0.0950	63 0.048732	-1.951	0.06154.
poly(test\$`100la	bour2013`, 4)3 -0.1400	28 0.048732	-2.873	0.00782 **
poly(test\$`100la	bour2013`, 4)4 0.05412	25 0.048732	1.111	0.27650

---

Signif. codes: 0 `\*\*\*' 0.001 `\*\*' 0.05 `.' 0.1 ` ' 1

Residual standard error: 0.04873 on 27 degrees of freedom Multiple R-squared: 0.7006, Adjusted R-squared: 0.6562 F-statistic: 15.79 on 4 and 27 DF, p-value: 0.0000008896



## **Findings**

From the estimations of MSEs for different polynomial degrees, we can understand that our scenario can be best described with the use of a linear model ( $1^{st}$  degree). Because of insufficient number of observations, but also because of an extreme value (that one that has as labour ratio  $\sim 3.0$ ), any higher polynomial is not more accurate and seems that cannot offer any safer prediction.

It is recommended to have more observations in order to eliminate the problems that we have encountered because of the extreme value.

All in all, the MSE seems to be almost the same until the 9<sup>th</sup> polynomial degree.