

# Linear Regression using R

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## Employee Performance Prediction

### Context

Given data about employee answers on different selection tools, predict their performance.

### Description of the features

The features and their meaning are provided below.

- ID: Unique identifier
- SJT: Situational judgement test (a type of selection tool that many organizations use)
- EmotionalIntelligence: inventory or test
- Proactivity: personality assessment test
- Performance: job evaluation ratings
- Turnover: whether an employee has left the company

Higher scores indicate high level and is assumed to have a positive impact.

### Prerequisites

The following libraries need to be installed in order to run the script. You can install them in case you don't have them.

```
install.packages('readr')  
install.packages('lessR')
```

When all is set up, include them in the project.

```
library(readr)  
library(lessR)
```

### Getting the data

Let's first create a variable that we'll treat as the path to our data.

```
DATA_PATH <- './SelectionExercise.csv'
```

Next, we load the data in a data frame by using the `read_csv()` function from the `readr` library.

```
df <- read_csv(DATA_PATH)
```

```
## Rows: 300 Columns: 6  
## -- Column specification -----  
## Delimiter: ","  
## db1 (6): ID, SJT, EmotionalIntelligence, Proactivity, Performance, Turnover  
##  
## i Use `spec()` to retrieve the full column specification for this data.
```

```
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

We have a total of 300 samples with no missing values. Here are the first 10 observations.

```
df
```

```
## # A tibble: 300 x 6
```

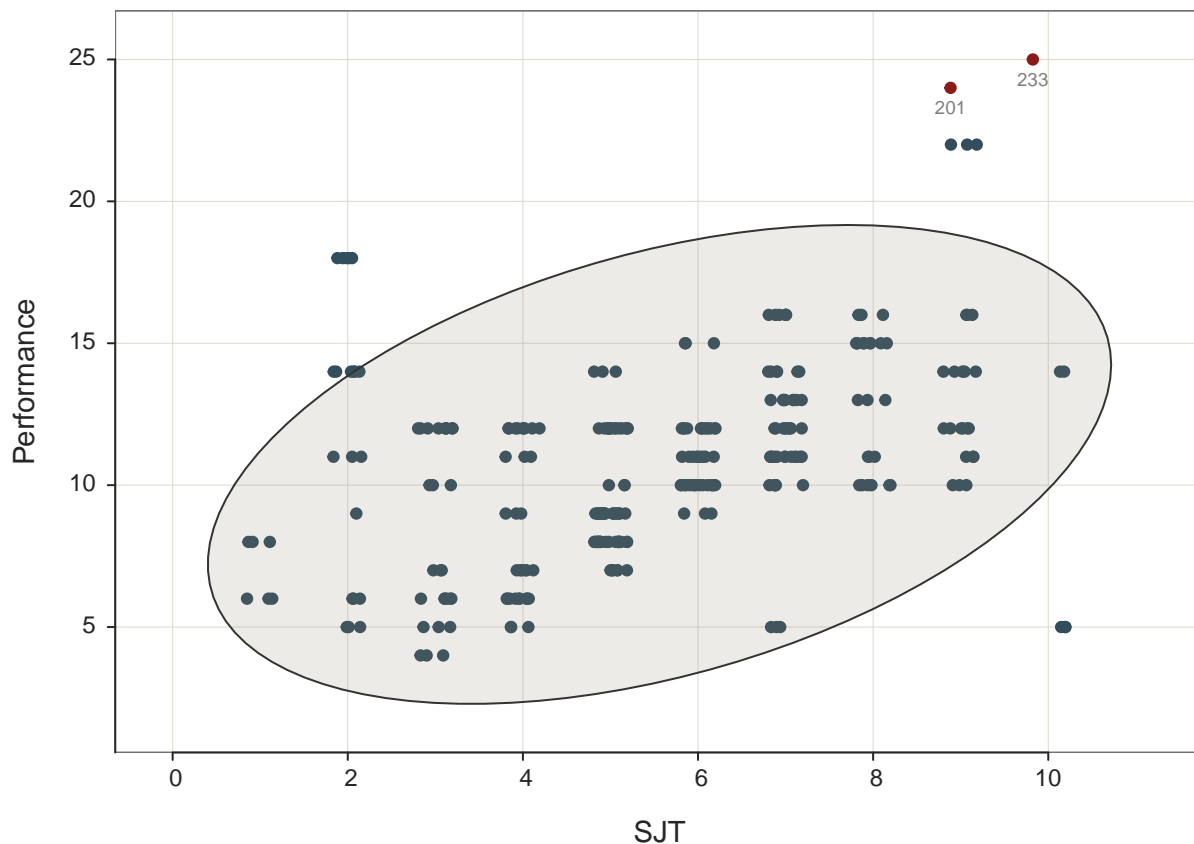
	ID	SJT	EmotionalIntelligence	Proactivity	Performance	Turnover
##	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
##	1	1	9	8	2	22
##	2	2	8	6	3	11
##	3	3	7	6	4	5
##	4	4	6	5	5	11
##	5	5	6	5	6	12
##	6	6	5	5	7	12
##	7	7	5	4	7	12
##	8	8	4	2	8	12
##	9	9	3	2	9	12
##	10	10	8	7	2	10

```
## # ... with 290 more rows
```

## Simple Linear Regression

Let's check whether we meet the assumptions of a linear regression model.

```
## [Ellipse with Murdoch and Chow's function ellipse from their ellipse package]
```



```
## >>> Suggestions
```

```
## Plot(SJT, Performance, enhance=TRUE) # many options
```

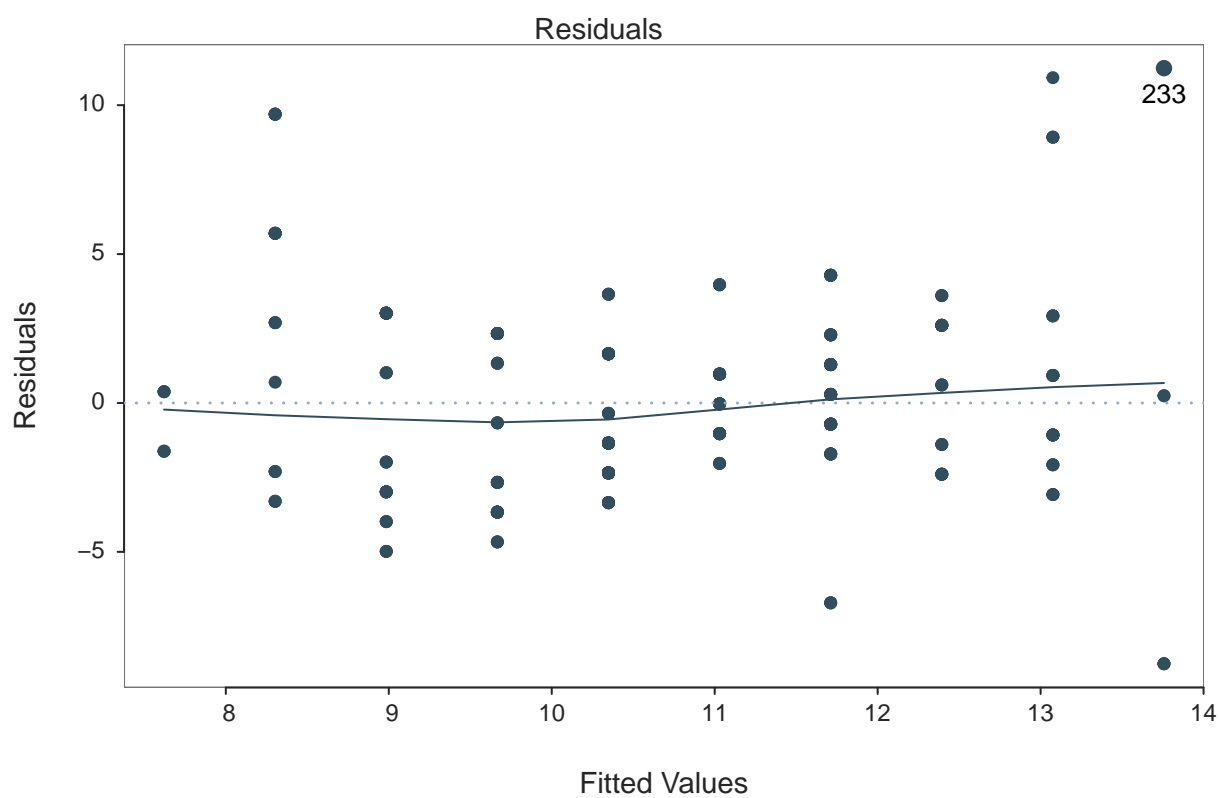
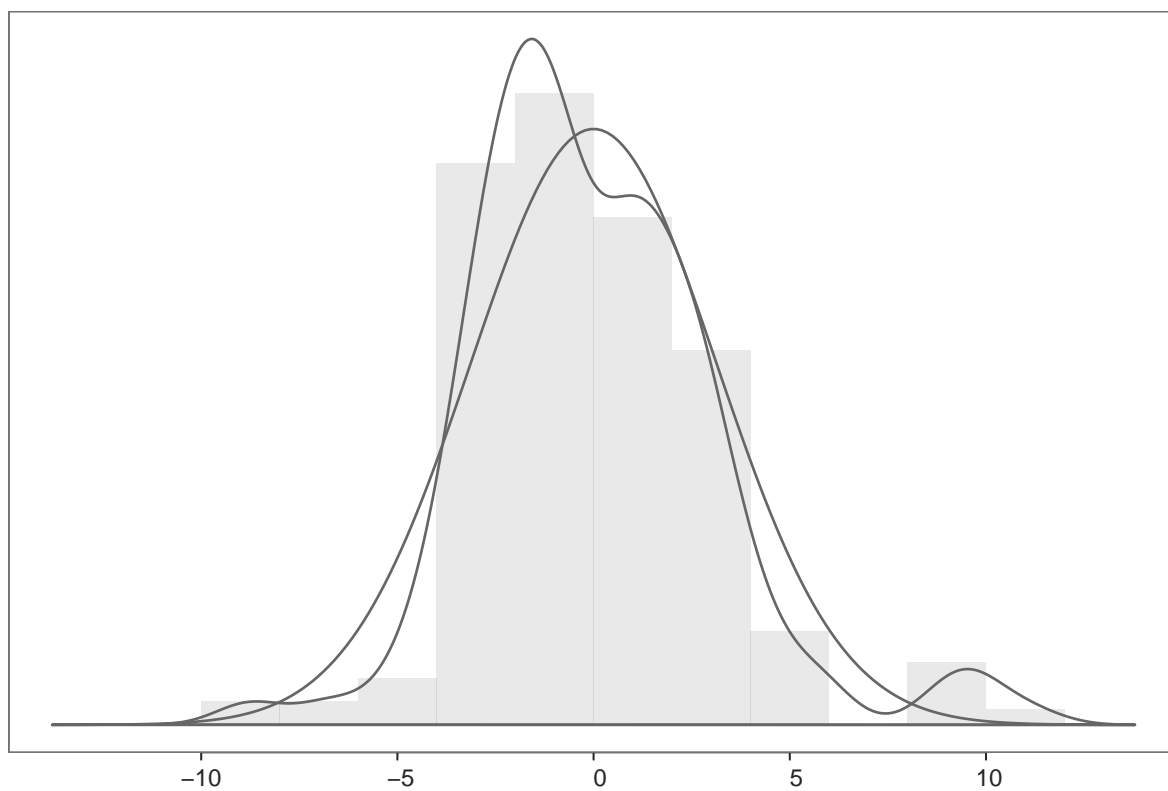
```
## Plot(SJT, Performance, color="red") # exterior edge color of points
```

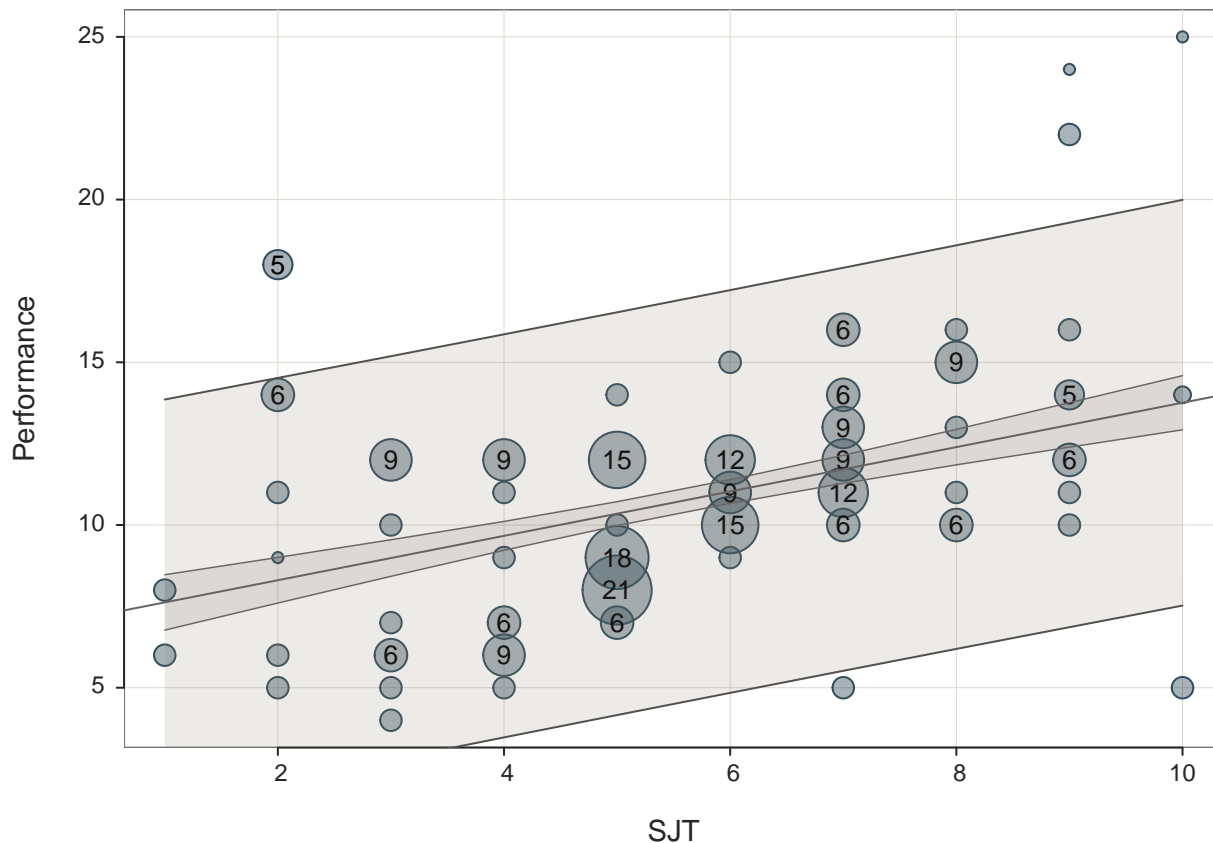
```
## Plot(SJT, Performance, fit="lm", fit_se=c(.90,.99)) # fit line, stnd errors
##
## >>> Pearson's product-moment correlation
##
## Number of paired values with neither missing, n = 300
## Sample Correlation of SJT and Performance: r = 0.417
##
## Hypothesis Test of 0 Correlation: t = 7.915, df = 298, p-value = 0.000
## 95% Confidence Interval for Correlation: 0.319 to 0.506
##
## >>> Outlier analysis with Mahalanobis Distance
##
##      MD  ID
## -----
## 17.31 233
## 14.82 201
##
## 12.43 20
## 12.43 69
## 12.43 120
## ... ...
##
## Some Parameter values (can be manually set)
## -----
## fill: #324E5C   filled color of the points
## color: #324E5C  edge color of the points
## size: 0.80     size of plotted points
## jitter_y: 0.00  random vertical movement of points
## jitter_x: 1.00  random horizontal movement of points
```

We meet the assumption of bivariate normal distribution. There is a linear relationship between SJT and Performance. The correlation coefficient is 0.417 which signals good relationship. The p-value is 0 which means that this is a statistically significant association. Nevertheless, there are some outliers (for example 201 and 233) and we will see how the model will perform if we remove them.

### Simple Regression from lessR

We perform linear regression by using the `Regression` function that comes with the `lessR` package. Conveniently, it also useful plots and analysis results.





```
## >>> Suggestion
## # Create an R markdown file for interpretative output with Rmd = "file_name"
## Regression(my_formula=Performance ~ SJT, data=df, Rmd="eg")
##
##
## BACKGROUND
##
## Data Frame: df
##
## Response Variable: Performance
## Predictor Variable: SJT
##
## Number of cases (rows) of data: 300
## Number of cases retained for analysis: 300
##
## BASIC ANALYSIS
##
##
```

	Estimate	Std Err	t-value	p-value	Lower 95%	Upper 95%
(Intercept)	6.939	0.512	13.552	0.000	5.932	7.947
SJT	0.682	0.086	7.915	0.000	0.512	0.851

```
##
## Standard deviation of Performance: 3.447
##
## Standard deviation of residuals: 3.139 for 298 degrees of freedom
## 95% range of residual variation: 12.354 = 2 * (1.968 * 3.139)
##
```

```

## R-squared:  0.174    Adjusted R-squared:  0.171    PRESS R-squared:  0.157
##
## Null hypothesis of all 0 population slope coefficients:
##   F-statistic: 62.654    df: 1 and 298    p-value:  0.000
##
## -- Analysis of Variance
##
##           df      Sum Sq   Mean Sq   F-value   p-value
## Model         1    617.264    617.264    62.654     0.000
## Residuals    298   2935.866     9.852
## Performance  299   3553.130    11.883
##
##
##   K-FOLD CROSS-VALIDATION
##
##
##   RELATIONS AMONG THE VARIABLES
##
##           Performance  SJT
## Performance         1.00 0.42
##           SJT         0.42 1.00
##
##
##   RESIDUALS AND INFLUENCE
##
## Data, Fitted, Residual, Studentized Residual, Dffits, Cook's Distance
##   [sorted by Cook's Distance]
##   [res_rows = 20, out of 300 rows of data, or do res_rows="all"]
## -----
##           SJT Performance fitted  resid rstdnt dffits cooks
## 233 10.000      25.000 13.757 11.243  3.691  0.502 0.121
## 201  9.000      24.000 13.075 10.925  3.570  0.398 0.076
##  70 10.000       5.000 13.757 -8.757 -2.849 -0.388 0.073
## 170 10.000       5.000 13.757 -8.757 -2.849 -0.388 0.073
## 270 10.000       5.000 13.757 -8.757 -2.849 -0.388 0.073
##  20  2.000      18.000  8.303  9.697  3.156  0.360 0.063
##  69  2.000      18.000  8.303  9.697  3.156  0.360 0.063
## 120  2.000      18.000  8.303  9.697  3.156  0.360 0.063
## 169  2.000      18.000  8.303  9.697  3.156  0.360 0.063
## 269  2.000      18.000  8.303  9.697  3.156  0.360 0.063
##   1  9.000      22.000 13.075  8.925  2.896  0.322 0.051
## 101  9.000      22.000 13.075  8.925  2.896  0.322 0.051
## 283  9.000      22.000 13.075  8.925  2.896  0.322 0.051
##  82  2.000      14.000  8.303  5.697  1.834  0.209 0.022
##  97  2.000      14.000  8.303  5.697  1.834  0.209 0.022
## 182  2.000      14.000  8.303  5.697  1.834  0.209 0.022
## 197  2.000      14.000  8.303  5.697  1.834  0.209 0.022
## 282  2.000      14.000  8.303  5.697  1.834  0.209 0.022
## 297  2.000      14.000  8.303  5.697  1.834  0.209 0.022
##   3  7.000       5.000 11.712 -6.712 -2.157 -0.151 0.011
##
##
##   PREDICTION ERROR
##

```

```
## Data, Predicted, Standard Error of Forecast,
## 95% Prediction Intervals
## [sorted by lower bound of prediction interval]
## [to see all intervals do pred_rows="all"]
## -----
##
##      SJT Performance    pred    sf pi.lwr pi.upr  width
## 31  1.000      6.000  7.621  3.168  1.386 13.856 12.471
## 41  1.000      8.000  7.621  3.168  1.386 13.856 12.471
## ...
## 299 5.000      8.000 10.348  3.144  4.160 16.536 12.376
## 4   6.000     11.000 11.030  3.144  4.842 17.218 12.375
## 5   6.000     12.000 11.030  3.144  4.842 17.218 12.375
## ...
## 170 10.000      5.000 13.757  3.167  7.524 19.990 12.466
## 233 10.000     25.000 13.757  3.167  7.524 19.990 12.466
## 270 10.000      5.000 13.757  3.167  7.524 19.990 12.466
##
## -----
## Plot 1: Distribution of Residuals
## Plot 2: Residuals vs Fitted Values
## Plot 3: Reg Line, Confidence & Prediction Intervals
## -----
```

The assumption of normally distributed residuals is met based on Plot 1: Distribution of Residuals. The assumptions of average residual error being (almost) 0 and homoscedasticity of variances are both met based on Plot 2: Residuals vs Fitted Values.

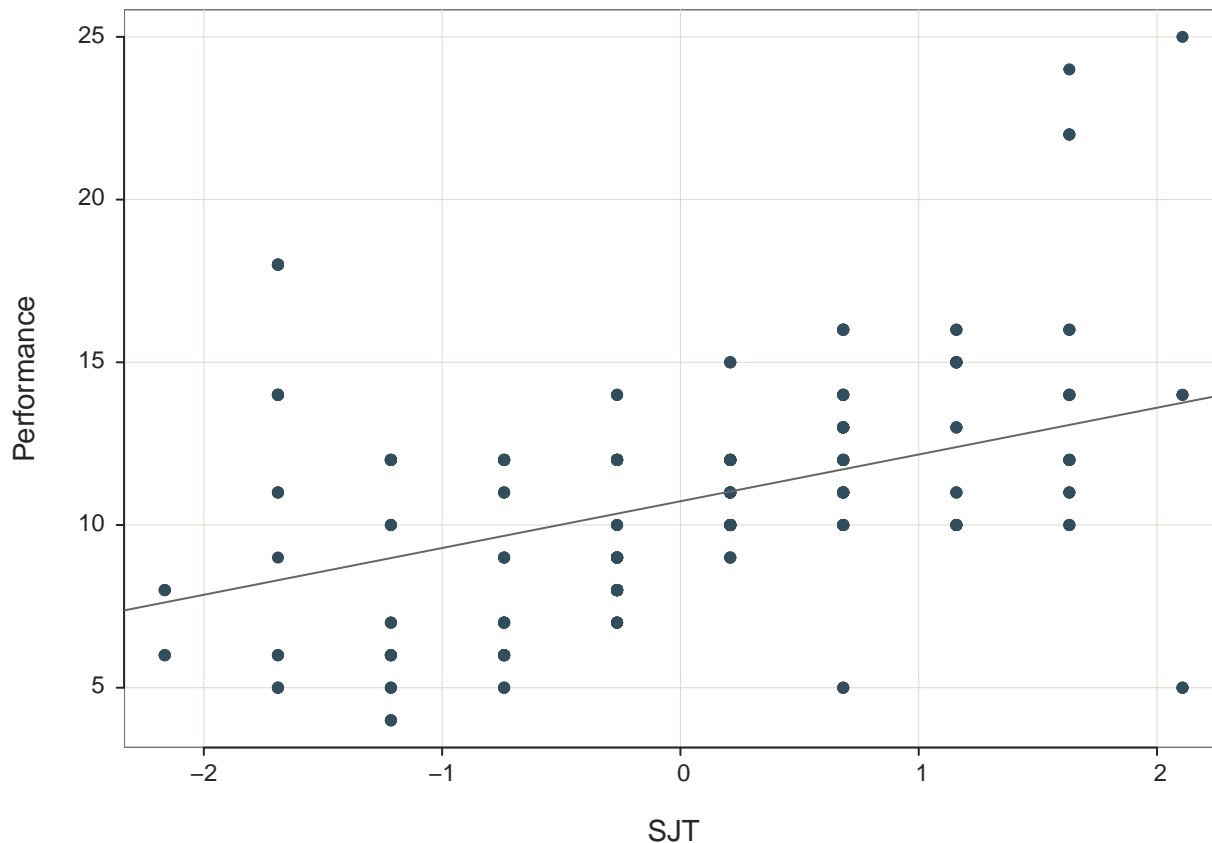
The BASIC ANALYSIS part of the output shows that for every 1 unit increase in SJT, Performance increases by 0.682 units. We can also construct the equation of the line:  $\text{Performance} = 6.939 + 0.682 * \text{SJT}$ .

The adjusted R-squared value is 0.171, i.e. SJT explains about 17% of the variability in Performance. The F-statistic shows that a model using SJT will outperform a null model, i.e. SJT is significant.

We can again see the potential outliers by looking at the part. The Cook's Distance of sample with id 233 is noticeably higher than the other samples: 0.121.

If we standardized our data, the results will not differ by much.

```
##
## Rescaled Data, First Six Rows
##      Performance    SJT
## 31      6 -2.164
## 41      8 -2.164
## 131     6 -2.164
## 141     8 -2.164
## 231     6 -2.164
## 241     8 -2.164
```



```
## >>> Suggestion
## # Create an R markdown file for interpretative output with Rmd = "file_name"
## reg(Performance ~ SJT, data=df, new_scale="z", Rmd="eg")
##
##
## BACKGROUND
##
## Data Frame: df
##
## Response Variable: Performance
## Predictor Variable: SJT
##
## Number of cases (rows) of data: 300
## Number of cases retained for analysis: 300
##
## Data are Standardized
##
##
## BASIC ANALYSIS
##
##
```

	Estimate	Std Err	t-value	p-value	Lower 95%	Upper 95%
(Intercept)	10.730	0.181	59.212	0.000	10.373	11.087
SJT	1.437	0.182	7.916	0.000	1.080	1.794

```
##
## Standard deviation of Performance: 3.447
##
## Standard deviation of residuals: 3.139 for 298 degrees of freedom
```



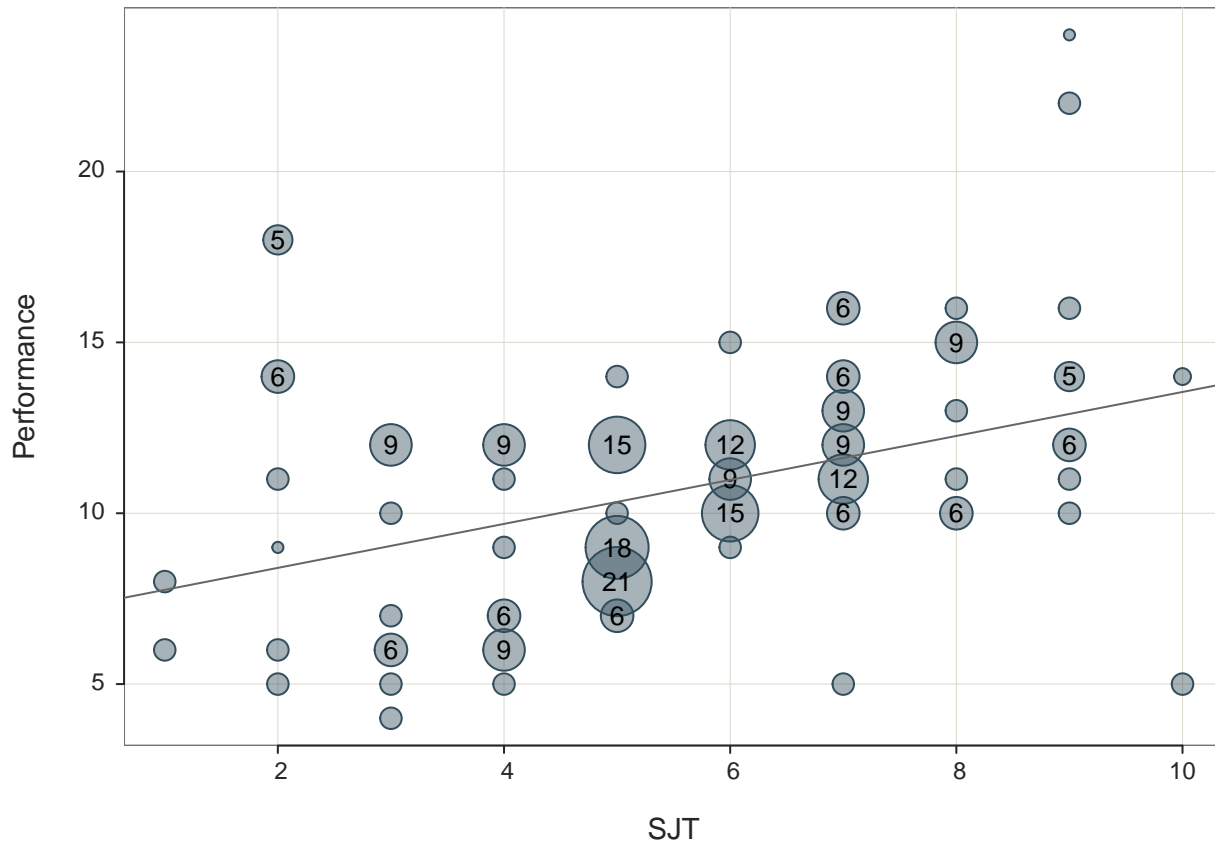
```

## 95% range of residual variation: 12.354 = 2 * (1.968 * 3.139)
##
## R-squared: 0.174    Adjusted R-squared: 0.171    PRESS R-squared: 0.157
##
## Null hypothesis of all 0 population slope coefficients:
##   F-statistic: 62.661    df: 1 and 298    p-value: 0.000
##
## -- Analysis of Variance
##
##           df    Sum Sq   Mean Sq   F-value   p-value
## Model         1    617.320    617.320    62.661    0.000
## Residuals    298   2935.810     9.852
## Performance  299   3553.130    11.883
##
##
##   K-FOLD CROSS-VALIDATION
##
##
##   RELATIONS AMONG THE VARIABLES
##
##
##   RESIDUALS AND INFLUENCE
##
##
##   PREDICTION ERROR

```

### Inspecting outliers

A model without the observation with id 233 does not perform that well. Here's what we would get.



```
## >>> Suggestion
## # Create an R markdown file for interpretative output with Rmd = "file_name"
## reg(Performance ~ SJT, data=df, rows=(ID != 233), Rmd="eg")
##
##
## BACKGROUND
##
## Data Frame: df
##
## Response Variable: Performance
## Predictor Variable: SJT
##
## Number of cases (rows) of data: 299
## Number of cases retained for analysis: 299
##
## BASIC ANALYSIS
##
##
```

	Estimate	Std Err	t-value	p-value	Lower 95%	Upper 95%
(Intercept)	7.114	0.504	14.122	0.000	6.123	8.105
SJT	0.644	0.085	7.570	0.000	0.476	0.811

```
##
## Standard deviation of Performance: 3.352
##
## Standard deviation of residuals: 3.074 for 297 degrees of freedom
## 95% range of residual variation: 12.101 = 2 * (1.968 * 3.074)
##
```

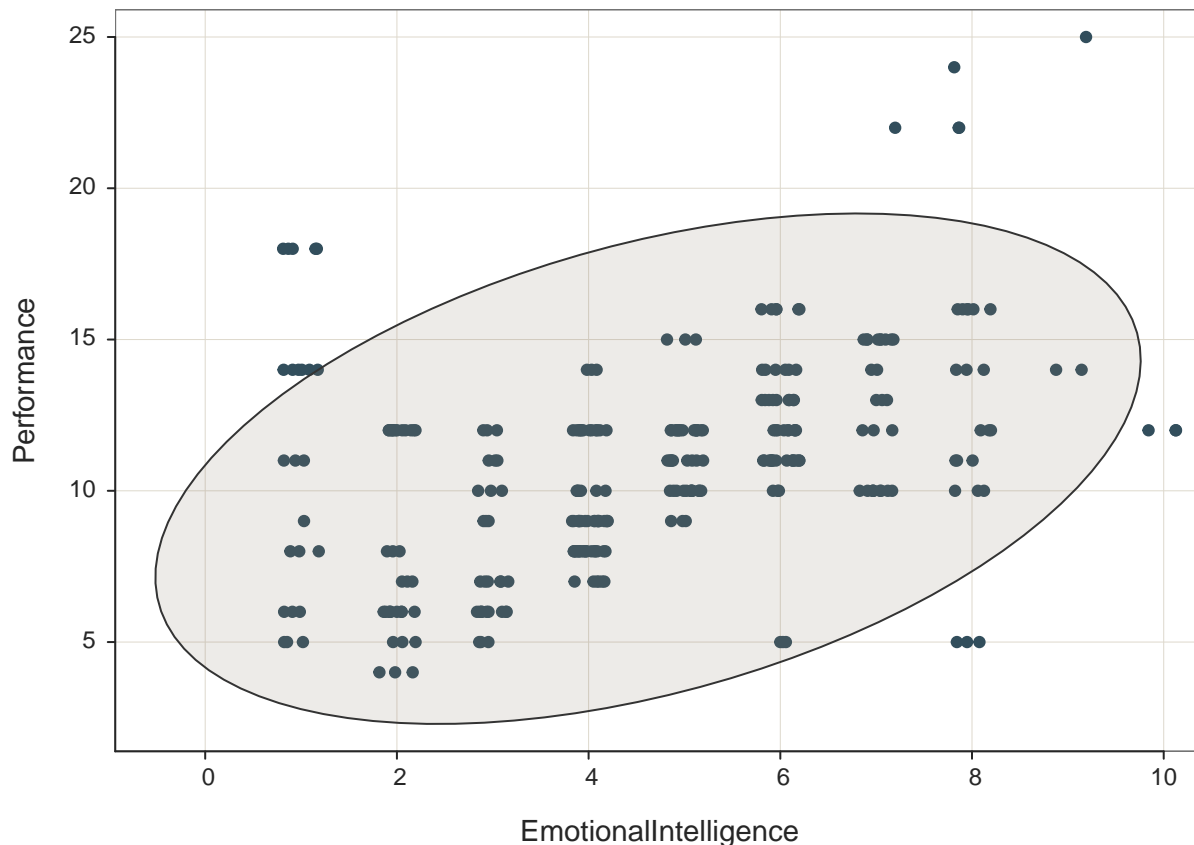
```
## R-squared: 0.162    Adjusted R-squared: 0.159    PRESS R-squared: 0.145
##
## Null hypothesis of all 0 population slope coefficients:
## F-statistic: 57.312    df: 1 and 297    p-value: 0.000
##
## -- Analysis of Variance
##
##          df    Sum Sq   Mean Sq   F-value   p-value
## Model         1   541.692   541.692    57.312    0.000
## Residuals    297  2807.124     9.452
## Performance  298  3348.816    11.238
##
##
## K-FOLD CROSS-VALIDATION
##
##
## RELATIONS AMONG THE VARIABLES
##
##
## RESIDUALS AND INFLUENCE
##
##
## PREDICTION ERROR
```

The new Adjusted R-squared value is 0.159 which is a slight decrease from 0.171 and therefore we have no reason to remove that observation.

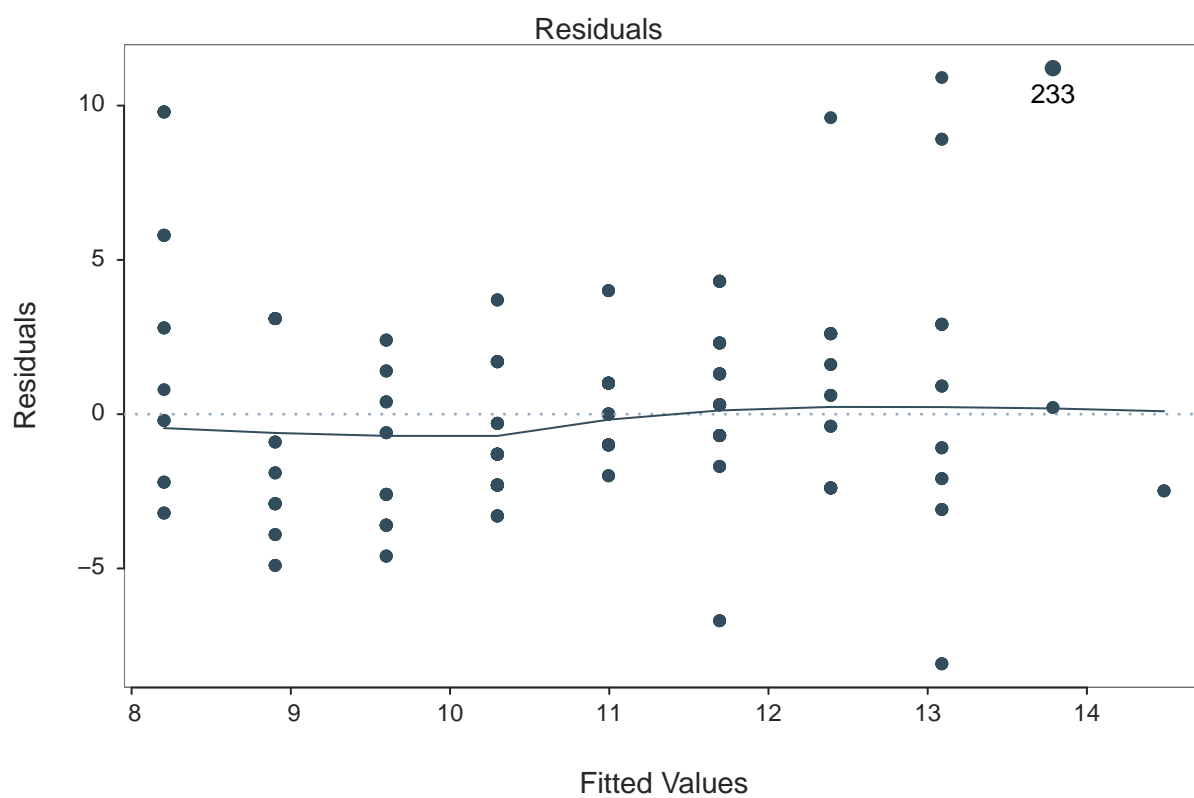
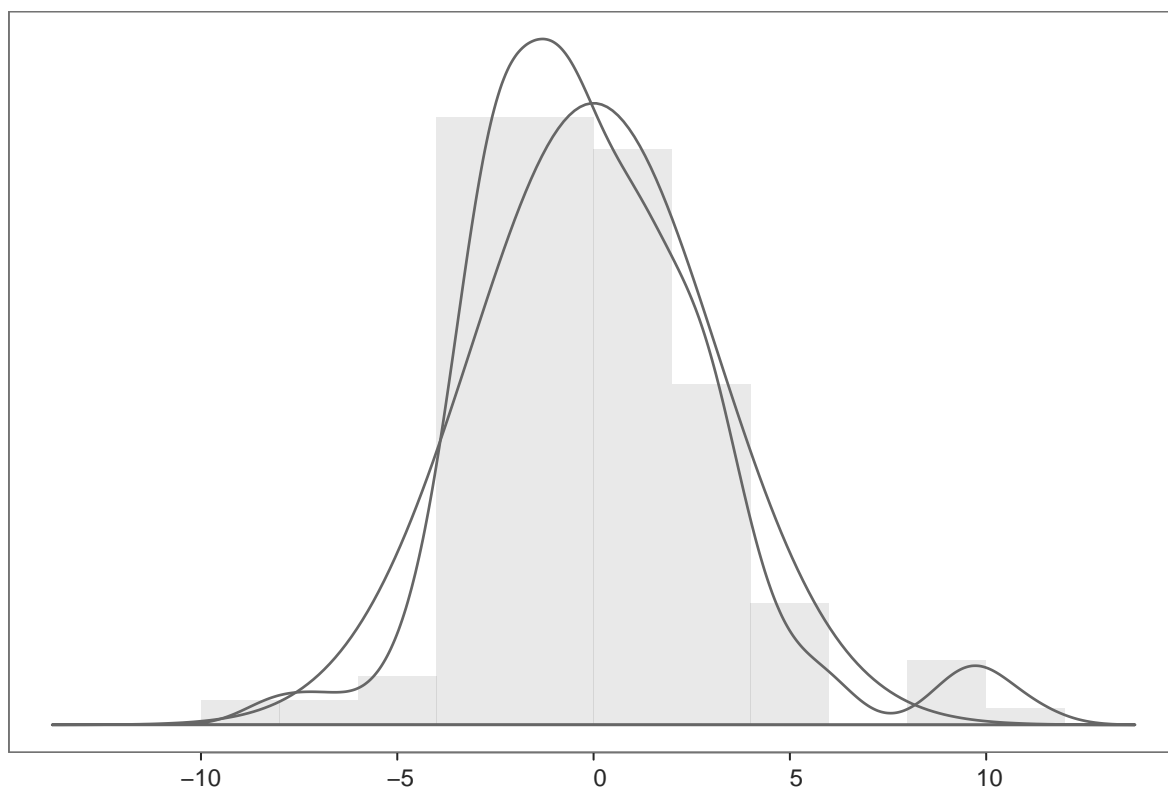
## Multiple Linear Regression

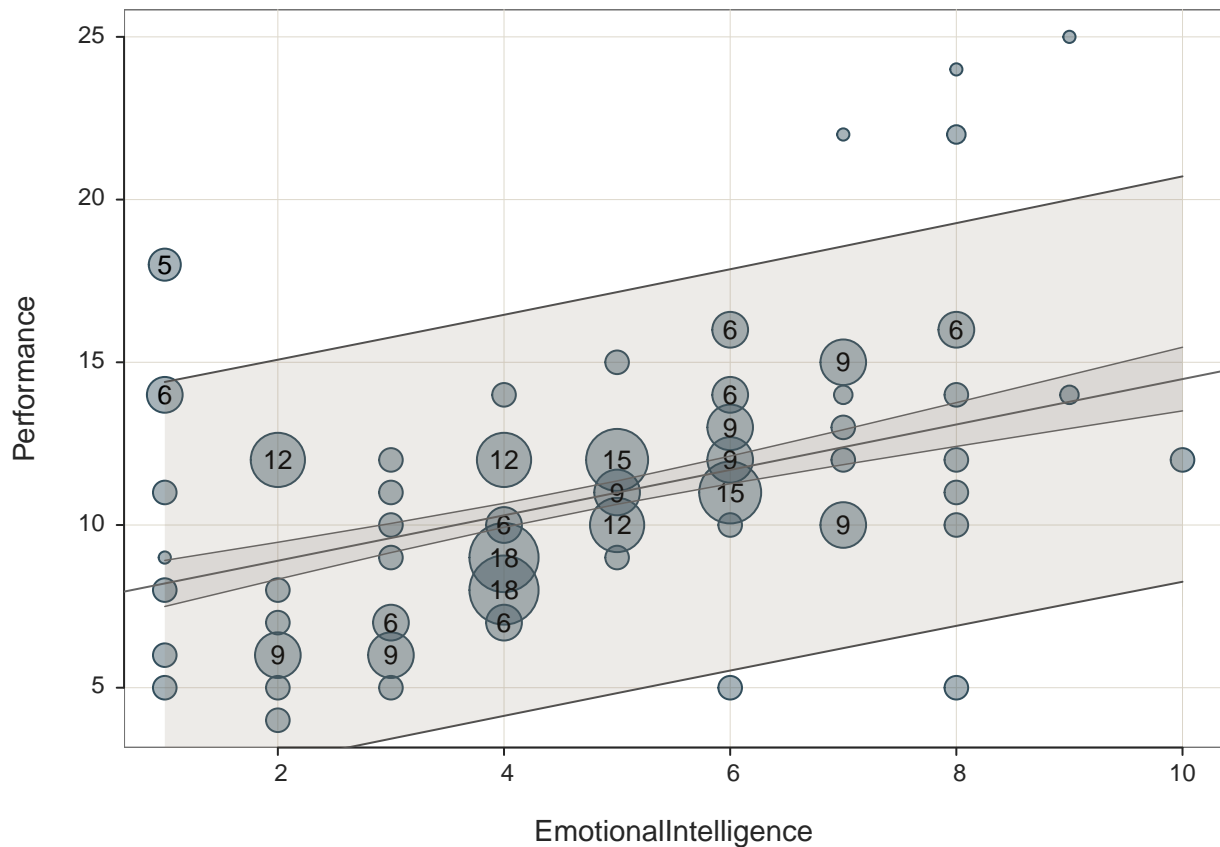
Firstly, we'll check how well the other features predict `Performance`. We start with `EmotionalIntelligence`.

```
## [Ellipse with Murdoch and Chow's function ellipse from their ellipse package]
```



```
## >>> Suggestions
## Plot(EmotionalIntelligence, Performance, enhance=TRUE) # many options
## Plot(EmotionalIntelligence, Performance, color="red") # exterior edge color of points
## Plot(EmotionalIntelligence, Performance, fit="lm", fit_se=c(.90,.99)) # fit line, std errors
## Plot(EmotionalIntelligence, Performance, out_cut=.10) # label top 10% from center as outliers
##
## >>> Pearson's product-moment correlation
##
## Number of paired values with neither missing, n = 300
## Sample Correlation of EmotionalIntelligence and Performance: r = 0.425
##
## Hypothesis Test of 0 Correlation: t = 8.108, df = 298, p-value = 0.000
## 95% Confidence Interval for Correlation: 0.328 to 0.514
##
## Some Parameter values (can be manually set)
## -----
## fill: #324E5C filled color of the points
## color: #324E5C edge color of the points
## size: 0.80 size of plotted points
## jitter_y: 0.00 random vertical movement of points
## jitter_x: 1.00 random horizontal movement of points
```





```
## >>> Suggestion
## # Create an R markdown file for interpretative output with Rmd = "file_name"
## Regression(my_formula=Performance ~ EmotionalIntelligence, data=df, Rmd="eg")
##
##
## BACKGROUND
##
## Data Frame: df
##
## Response Variable: Performance
## Predictor Variable: EmotionalIntelligence
##
## Number of cases (rows) of data: 300
## Number of cases retained for analysis: 300
##
## BASIC ANALYSIS
##
##
```

	Estimate	Std Err	t-value	p-value	Lower 95%	Upper 95%
(Intercept)	7.506	0.437	17.190	0.000	6.647	8.365
EmotionalIntelligence	0.698	0.086	8.108	0.000	0.528	0.867

```
##
## Standard deviation of Performance: 3.447
##
## Standard deviation of residuals: 3.125 for 298 degrees of freedom
## 95% range of residual variation: 12.301 = 2 * (1.968 * 3.125)
##
```

```

## R-squared:  0.181    Adjusted R-squared:  0.178    PRESS R-squared:  0.165
##
## Null hypothesis of all 0 population slope coefficients:
##   F-statistic: 65.739    df: 1 and 298    p-value:  0.000
##
## -- Analysis of Variance
##
##           df      Sum Sq   Mean Sq   F-value   p-value
## Model           1    642.161    642.161    65.739    0.000
## Residuals       298   2910.969     9.768
## Performance     299   3553.130    11.883
##
##
##   K-FOLD CROSS-VALIDATION
##
##
##   RELATIONS AMONG THE VARIABLES
##
##           Performance EmotionalIntelligence
##           Performance      1.00          0.43
##   EmotionalIntelligence      0.43          1.00
##
##
##   RESIDUALS AND INFLUENCE
##
## Data, Fitted, Residual, Studentized Residual, Dffits, Cook's Distance
##   [sorted by Cook's Distance]
##   [res_rows = 20, out of 300 rows of data, or do res_rows="all"]
## -----
##           EmotionalIntelligence Performance fitted   resid rstdnt dffits cooks
##   233              9.000      25.000 13.787 11.213  3.696  0.499 0.119
##   201              8.000      24.000 13.089 10.911  3.581  0.395 0.075
##    20              1.000      18.000  8.204  9.796  3.204  0.372 0.067
##    69              1.000      18.000  8.204  9.796  3.204  0.372 0.067
##   120              1.000      18.000  8.204  9.796  3.204  0.372 0.067
##   169              1.000      18.000  8.204  9.796  3.204  0.372 0.067
##   269              1.000      18.000  8.204  9.796  3.204  0.372 0.067
##     1              8.000      22.000 13.089  8.911  2.904  0.320 0.050
##   101              8.000      22.000 13.089  8.911  2.904  0.320 0.050
##    70              8.000       5.000 13.089 -8.089 -2.629 -0.290 0.041
##   170              8.000       5.000 13.089 -8.089 -2.629 -0.290 0.041
##   270              8.000       5.000 13.089 -8.089 -2.629 -0.290 0.041
##   283              7.000      22.000 12.391  9.609  3.132  0.275 0.037
##    82              1.000      14.000  8.204  5.796  1.875  0.217 0.023
##    97              1.000      14.000  8.204  5.796  1.875  0.217 0.023
##   182              1.000      14.000  8.204  5.796  1.875  0.217 0.023
##   197              1.000      14.000  8.204  5.796  1.875  0.217 0.023
##   282              1.000      14.000  8.204  5.796  1.875  0.217 0.023
##   297              1.000      14.000  8.204  5.796  1.875  0.217 0.023
##     3              6.000       5.000 11.693 -6.693 -2.160 -0.150 0.011
##
##
##   PREDICTION ERROR
##

```

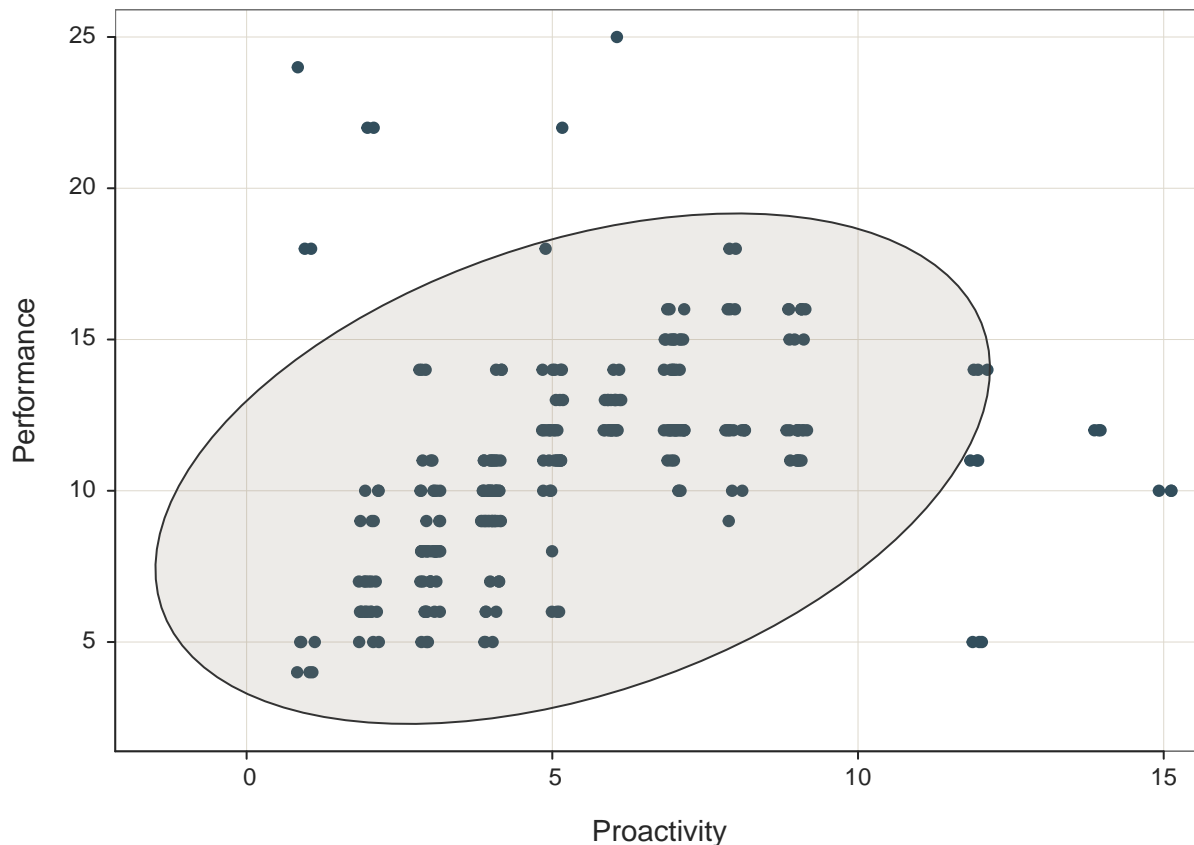
```
## Data, Predicted, Standard Error of Forecast,
## 95% Prediction Intervals
## [sorted by lower bound of prediction interval]
## [to see all intervals do pred_rows="all"]
## -----
##
##      EmotionalIntelligence Performance   pred   sf pi.lwr pi.upr width
##    20                1.000      18.000  8.204 3.146  2.012 14.395 12.383
##    40                1.000       5.000  8.204 3.146  2.012 14.395 12.383
##    55                1.000      11.000  8.204 3.146  2.012 14.395 12.383
##    ...
##   299                4.000       8.000 10.297 3.131  4.135 16.459 12.324
##     4                5.000      11.000 10.995 3.131  4.834 17.156 12.323
##     5                5.000      12.000 10.995 3.131  4.834 17.156 12.323
##    ...
##   233                9.000      25.000 13.787 3.153  7.581 19.992 12.411
##    84               10.000      12.000 14.484 3.165  8.256 20.712 12.456
##   184               10.000      12.000 14.484 3.165  8.256 20.712 12.456
##
## -----
## Plot 1: Distribution of Residuals
## Plot 2: Residuals vs Fitted Values
## Plot 3: Reg Line, Confidence & Prediction Intervals
## -----
```

The EmotionalIntelligence feature is statistically significantly associated in a positive direction with Performance.

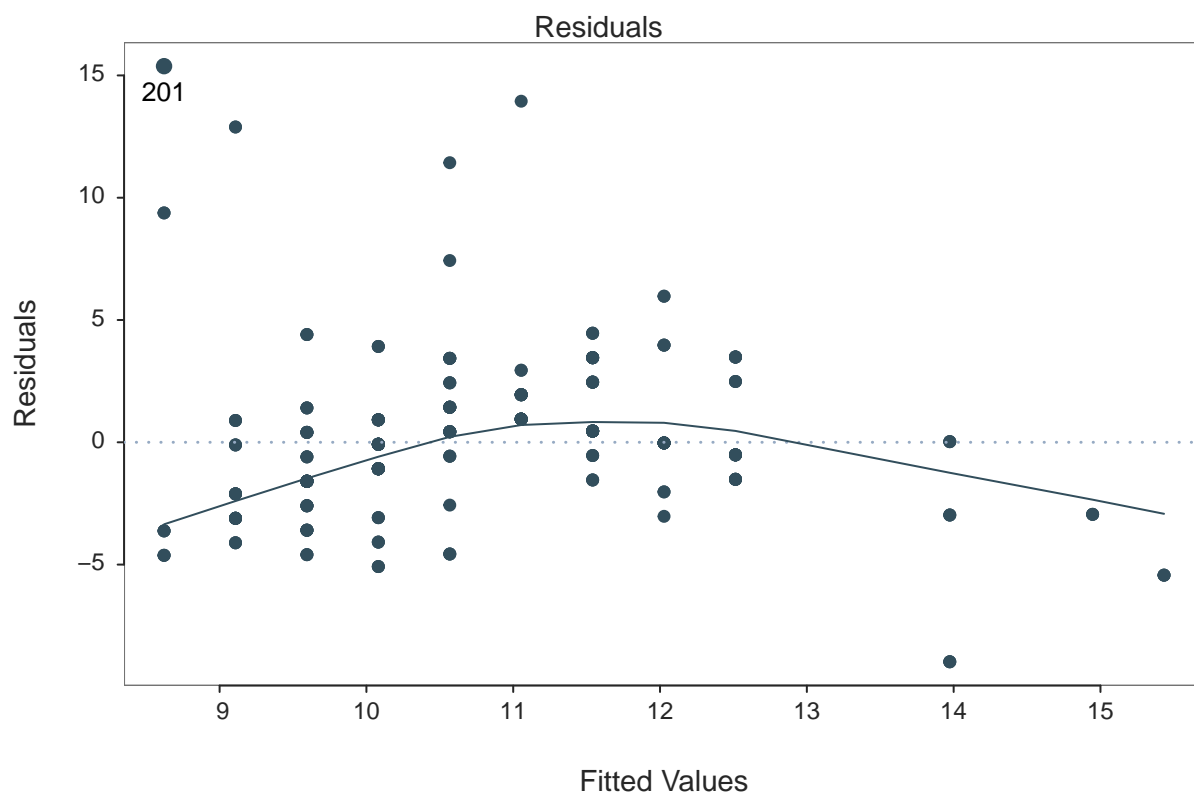
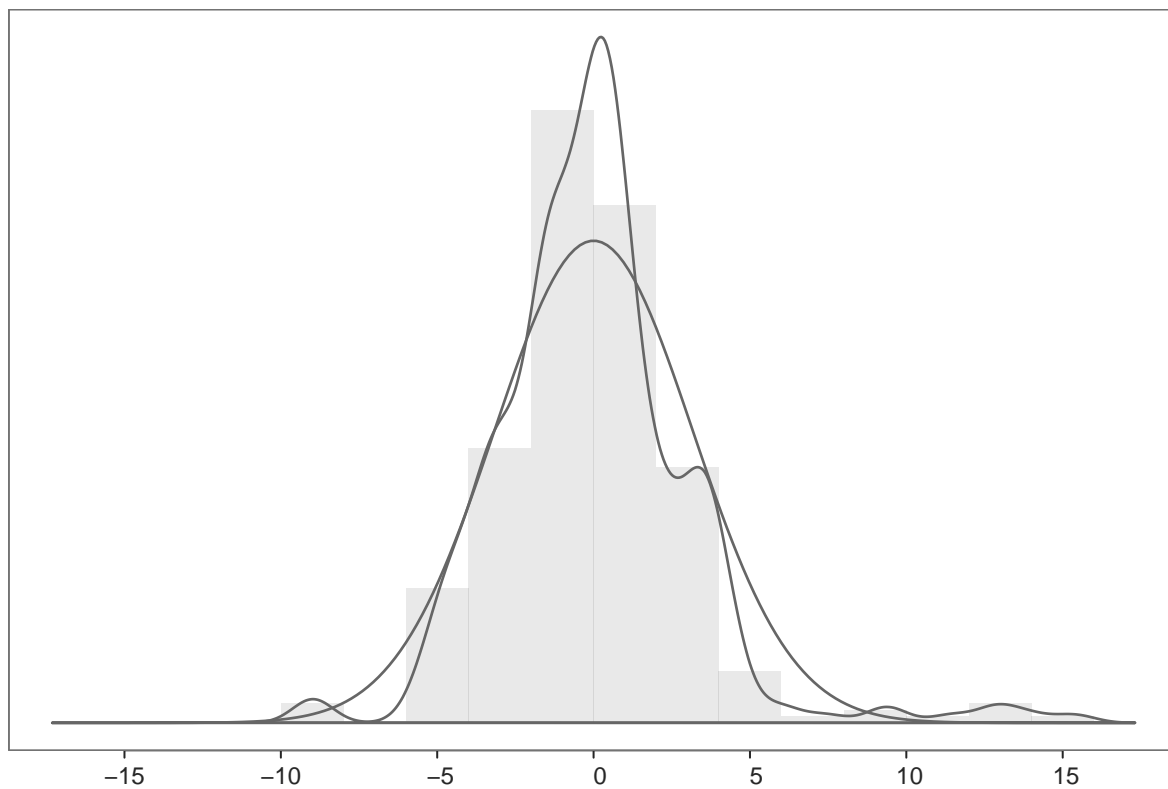
The same can be said for Proactivity.

```
## [Ellipse with Murdoch and Chow's function ellipse from their ellipse package]
```

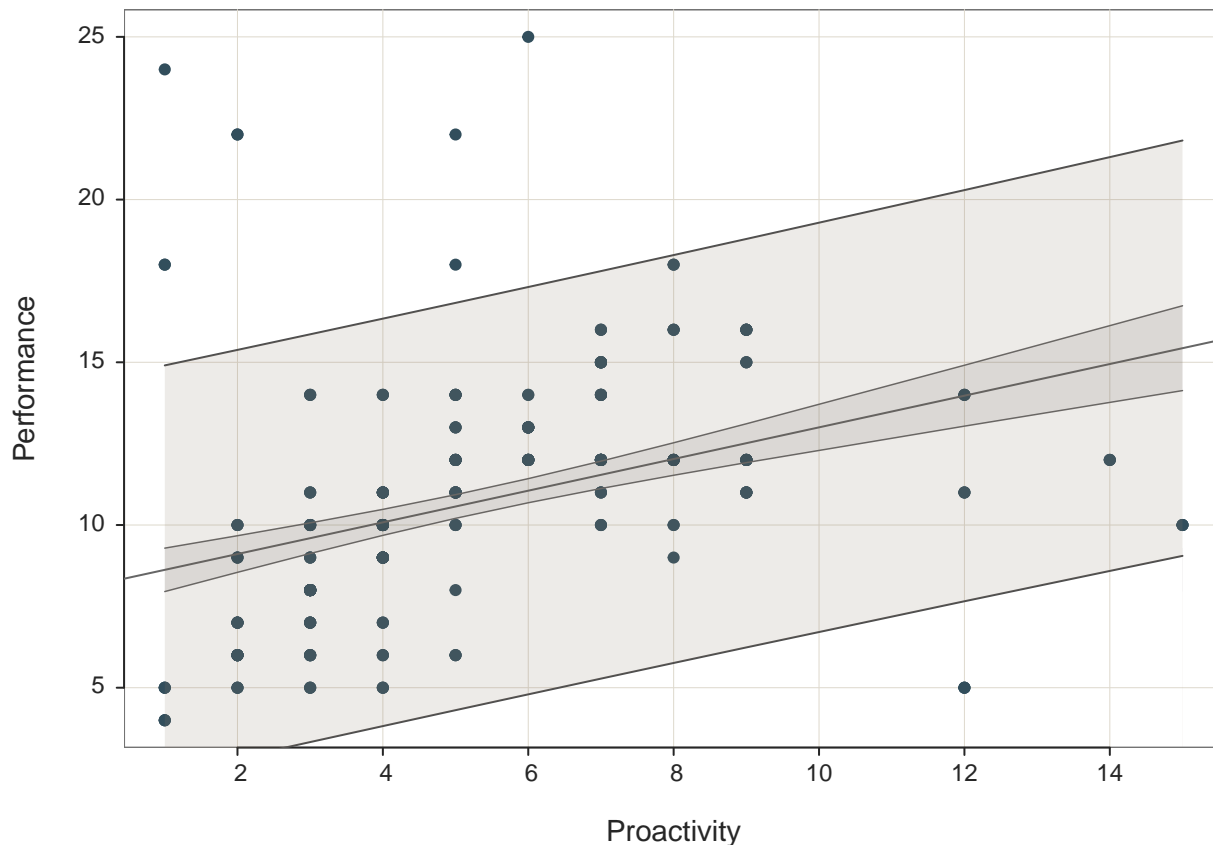




```
## >>> Suggestions
## Plot(Proactivity, Performance, enhance=TRUE) # many options
## Plot(Proactivity, Performance, color="red") # exterior edge color of points
## Plot(Proactivity, Performance, fit="lm", fit_se=c(.90,.99)) # fit line, std errors
## Plot(Proactivity, Performance, out_cut=.10) # label top 10% from center as outliers
##
## >>> Pearson's product-moment correlation
##
## Number of paired values with neither missing, n = 300
## Sample Correlation of Proactivity and Performance: r = 0.394
##
## Hypothesis Test of 0 Correlation: t = 7.390, df = 298, p-value = 0.000
## 95% Confidence Interval for Correlation: 0.293 to 0.485
##
## Some Parameter values (can be manually set)
## -----
## fill: #324E5C filled color of the points
## color: #324E5C edge color of the points
## size: 0.80 size of plotted points
## jitter_y: 0.00 random vertical movement of points
## jitter_x: 0.87 random horizontal movement of points
```



Point with largest Cook's Distance of 0.14 is labeled



```
## >>> Suggestion
## # Create an R markdown file for interpretative output with Rmd = "file_name"
## Regression(my_formula=Performance ~ Proactivity, data=df, Rmd="eg")
##
##
## BACKGROUND
##
## Data Frame: df
##
## Response Variable: Performance
## Predictor Variable: Proactivity
##
## Number of cases (rows) of data: 300
## Number of cases retained for analysis: 300
##
## BASIC ANALYSIS
##
##
```

	Estimate	Std Err	t-value	p-value	Lower 95%	Upper 95%
(Intercept)	8.135	0.396	20.539	0.000	7.356	8.915
Proactivity	0.487	0.066	7.390	0.000	0.357	0.616

```
##
## Standard deviation of Performance: 3.447
##
## Standard deviation of residuals: 3.174 for 298 degrees of freedom
## 95% range of residual variation: 12.494 = 2 * (1.968 * 3.174)
##
```

```

## R-squared:  0.155      Adjusted R-squared:  0.152      PRESS R-squared:  0.138
##
## Null hypothesis of all 0 population slope coefficients:
##   F-statistic: 54.606      df: 1 and 298      p-value:  0.000
##
## -- Analysis of Variance
##
##           df      Sum Sq   Mean Sq   F-value   p-value
## Model           1   550.256   550.256    54.606     0.000
## Residuals       298 3002.874    10.077
## Performance    299 3553.130    11.883
##
##
##   K-FOLD CROSS-VALIDATION
##
##
##   RELATIONS AMONG THE VARIABLES
##
##           Performance Proactivity
## Performance           1.00         0.39
## Proactivity           0.39         1.00
##
##
##   RESIDUALS AND INFLUENCE
##
## Data, Fitted, Residual, Studentized Residual, Dffits, Cook's Distance
##   [sorted by Cook's Distance]
##   [res_rows = 20, out of 300 rows of data, or do res_rows="all"]
## -----
##           Proactivity Performance fitted  resid rstdnt dffits cooks
## 201           1.000         24.000  8.622 15.378  5.070  0.545 0.137
## 70            12.000          5.000 13.973 -8.973 -2.894 -0.439 0.094
## 170           12.000          5.000 13.973 -8.973 -2.894 -0.439 0.094
## 270           12.000          5.000 13.973 -8.973 -2.894 -0.439 0.094
## 92            15.000         10.000 15.433 -5.433 -1.756 -0.375 0.070
## 192           15.000         10.000 15.433 -5.433 -1.756 -0.375 0.070
## 292           15.000         10.000 15.433 -5.433 -1.756 -0.375 0.070
## 1             2.000         22.000  9.108 12.892  4.189  0.379 0.068
## 101           2.000         22.000  9.108 12.892  4.189  0.379 0.068
## 69            1.000         18.000  8.622  9.378  3.011  0.324 0.051
## 169           1.000         18.000  8.622  9.378  3.011  0.324 0.051
## 233           6.000         25.000 11.054 13.946  4.544  0.270 0.034
## 283           5.000         22.000 10.568 11.432  3.683  0.215 0.022
## 28            14.000         12.000 14.947 -2.947 -0.945 -0.182 0.017
## 128           14.000         12.000 14.947 -2.947 -0.945 -0.182 0.017
## 228           14.000         12.000 14.947 -2.947 -0.945 -0.182 0.017
## 68            1.000          4.000  8.622 -4.622 -1.467 -0.158 0.012
## 168           1.000          4.000  8.622 -4.622 -1.467 -0.158 0.012
## 268           1.000          4.000  8.622 -4.622 -1.467 -0.158 0.012
## 20            8.000         18.000 12.027  5.973  1.896  0.152 0.011
##
##
##   PREDICTION ERROR
##

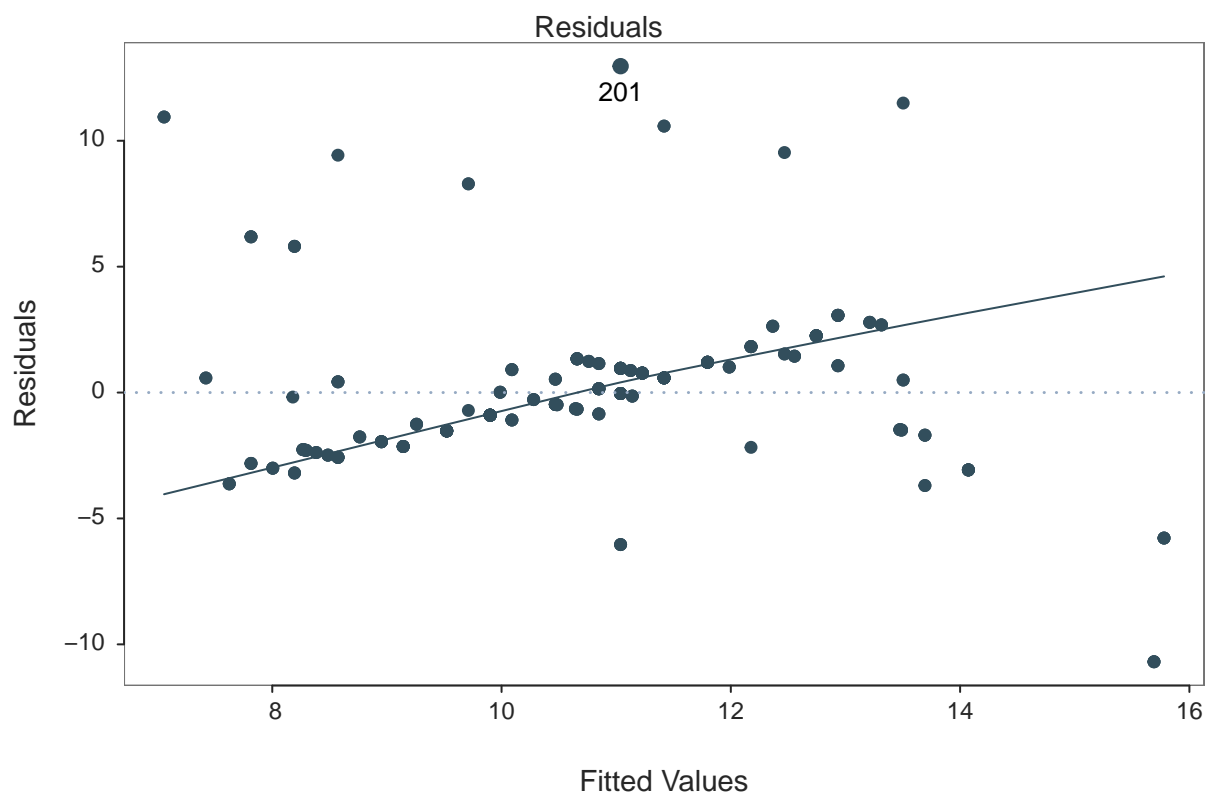
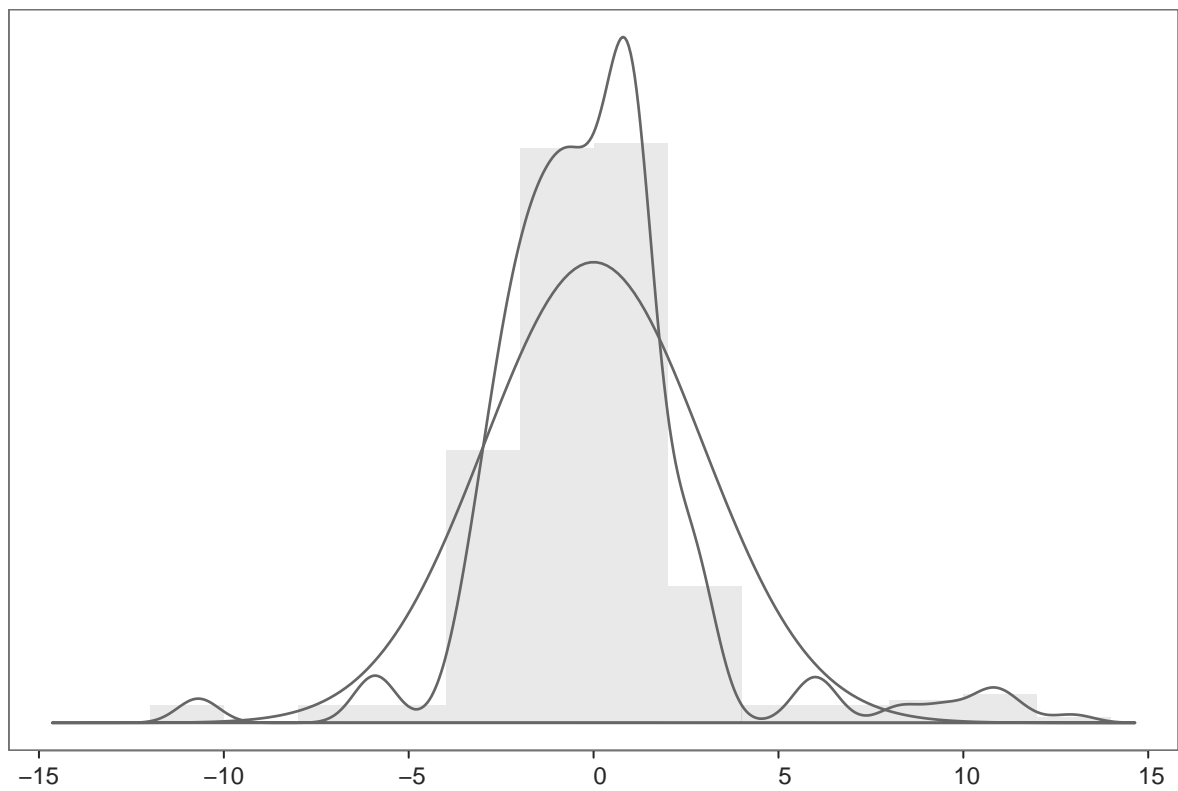
```

```
## Data, Predicted, Standard Error of Forecast,
## 95% Prediction Intervals
## [sorted by lower bound of prediction interval]
## [to see all intervals do pred_rows="all"]
## -----
##
##      Proactivity Performance   pred    sf pi.lwr pi.upr width
##    67          1.000         5.000 8.622 3.192  2.339 14.904 12.565
##    68          1.000         4.000 8.622 3.192  2.339 14.904 12.565
##    69          1.000        18.000 8.622 3.192  2.339 14.904 12.565
## ...
##   298          4.000         9.000 10.081 3.181  3.821 16.341 12.520
##     4          5.000        11.000 10.568 3.180  4.310 16.825 12.515
##    13          5.000        12.000 10.568 3.180  4.310 16.825 12.515
## ...
##   228         14.000        12.000 14.947 3.230  8.589 21.304 12.715
##    92         15.000        10.000 15.433 3.243  9.051 21.815 12.763
##   192         15.000        10.000 15.433 3.243  9.051 21.815 12.763
##
## -----
## Plot 1: Distribution of Residuals
## Plot 2: Residuals vs Fitted Values
## Plot 3: Reg Line, Confidence & Prediction Intervals
## -----
```

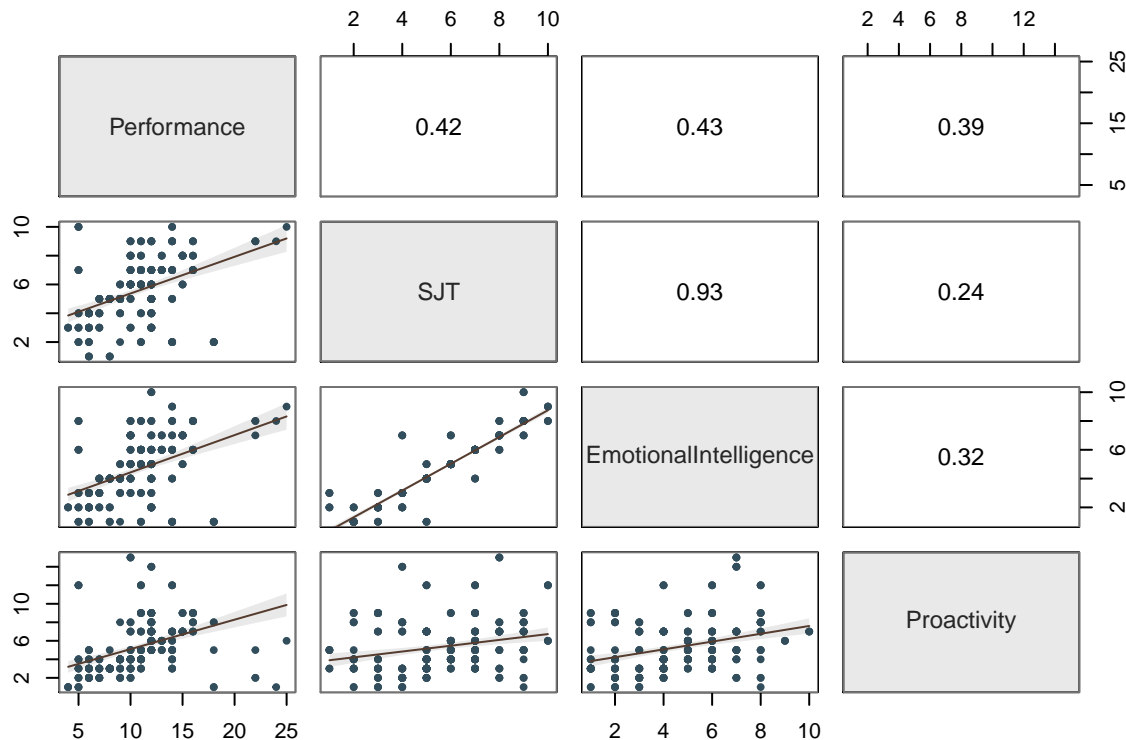
Proactivity is also statistically significantly associated in a positive direction with Performance. The residuals when using Proactivity don't have strong homoscedasticity though. Also, on average they are not 0.

Nevertheless, we do see a linear relationship between those features and Performance, so we'll try out a model by using them.

```
Regression(Performance ~ SJT + EmotionalIntelligence + Proactivity, data=df)
```



Point with largest Cook's Distance of 0.14 is labeled



```
## >>> Suggestion
## # Create an R markdown file for interpretative output with Rmd = "file_name"
## Regression(my_formula=Performance ~ SJT + EmotionalIntelligence + Proactivity, data=df, Rmd="eg")
##
##
## BACKGROUND
##
## Data Frame: df
##
## Response Variable: Performance
## Predictor Variable 1: SJT
## Predictor Variable 2: EmotionalIntelligence
## Predictor Variable 3: Proactivity
##
## Number of cases (rows) of data: 300
## Number of cases retained for analysis: 300
##
## BASIC ANALYSIS
##
##
```

	Estimate	Std Err	t-value	p-value	Lower 95%	Upper 95%
(Intercept)	5.627	0.569	9.893	0.000	4.507	6.746
SJT	0.481	0.229	2.104	0.036	0.031	0.932
EmotionalIntelligence	0.088	0.235	0.372	0.710	-0.375	0.550
Proactivity	0.379	0.066	5.755	0.000	0.250	0.509

```
##
## Standard deviation of Performance: 3.447
##
## Standard deviation of residuals: 2.968 for 296 degrees of freedom
## 95% range of residual variation: 11.683 = 2 * (1.968 * 2.968)
```

```
##
## R-squared: 0.266    Adjusted R-squared: 0.259    PRESS R-squared: 0.234
##
## Null hypothesis of all 0 population slope coefficients:
## F-statistic: 35.764    df: 3 and 296    p-value: 0.000
##
## -- Analysis of Variance
##
##              df      Sum Sq   Mean Sq   F-value   p-value
##          SJT      1    617.264    617.264    70.062    0.000
## EmotionalIntelligence  1     36.234     36.234     4.113    0.043
##          Proactivity  1    291.781    291.781    33.118    0.000
##
## Model              3    945.278    315.093    35.764    0.000
## Residuals          296   2607.852      8.810
## Performance        299   3553.130     11.883
##
##
## K-FOLD CROSS-VALIDATION
##
##
## RELATIONS AMONG THE VARIABLES
##
##              Performance  SJT EmotionalIntelligence Proactivity
##          Performance      1.00 0.42                  0.43      0.39
##              SJT          0.42 1.00                  0.93      0.24
## EmotionalIntelligence      0.43 0.93                  1.00      0.32
##          Proactivity      0.39 0.24                  0.32      1.00
##
##              Tolerance      VIF
##          SJT          0.127    7.887
## EmotionalIntelligence      0.121    8.278
##          Proactivity      0.873    1.146
##
## SJT EmotionalIntelligence Proactivity      adjr2      X's
## 1              0              1    0.261      2
## 1              1              1    0.259      3
## 0              1              1    0.250      2
## 1              1              0    0.178      2
## 0              1              0    0.178      1
## 1              0              0    0.171      1
## 0              0              1    0.152      1
##
## [based on Thomas Lumley's leaps function from the leaps package]
##
##
## RESIDUALS AND INFLUENCE
##
## Data, Fitted, Residual, Studentized Residual, Dffits, Cook's Distance
## [sorted by Cook's Distance]
## [res_rows = 20, out of 300 rows of data, or do res_rows="all"]
## -----
##          SJT EmotionalIntelligence Proactivity Performance fitted      resid rstdnt dffits cooks
## 201  9.000              8.000          1.000          24.000 11.038  12.962  4.576  0.775 0.141
```



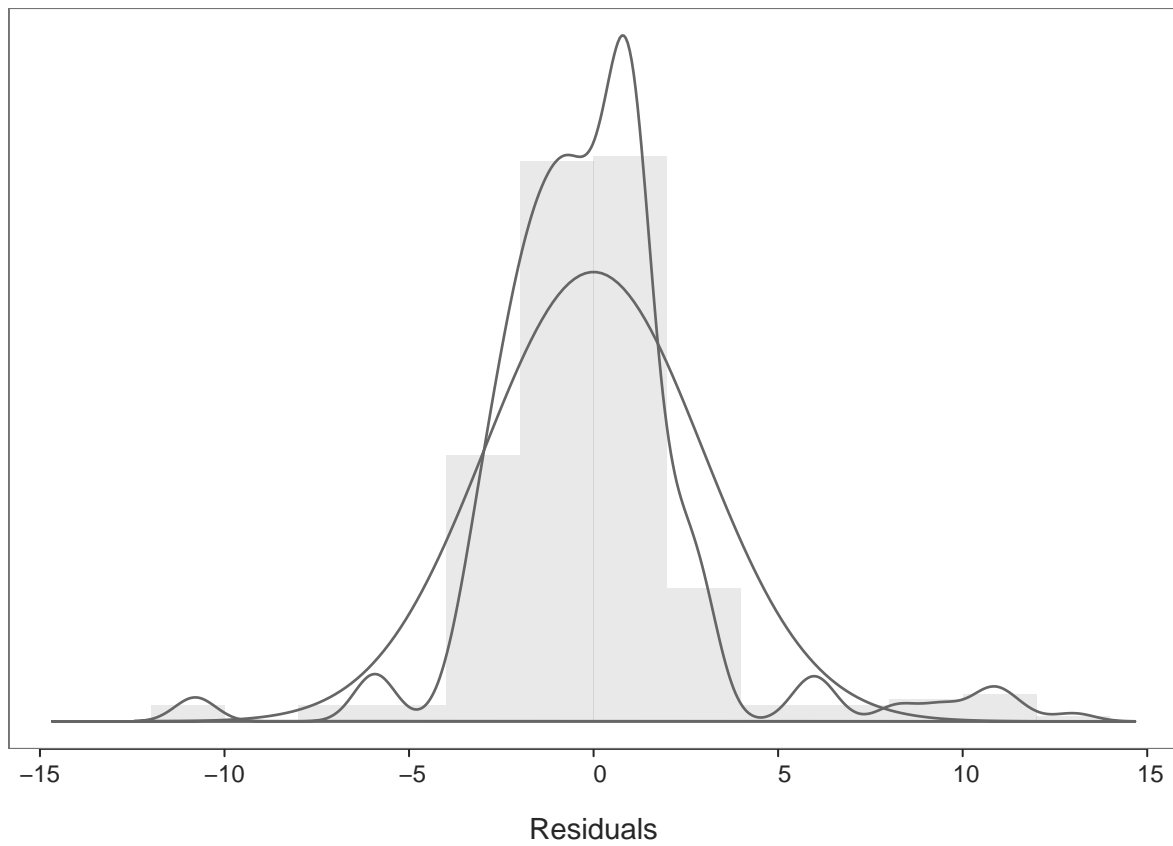
```
##      70 10.000      8.000      12.000      5.000 15.691 -10.691 -3.755 -0.758 0.138
##     170 10.000      8.000      12.000      5.000 15.691 -10.691 -3.755 -0.758 0.138
##     270 10.000      8.000      12.000      5.000 15.691 -10.691 -3.755 -0.758 0.138
##       1  9.000      8.000       2.000     22.000 11.418  10.582  3.682  0.563 0.076
##     101  9.000      8.000       2.000     22.000 11.418  10.582  3.682  0.563 0.076
##     233 10.000      9.000       6.000     25.000 13.503  11.497  4.009  0.558 0.074
##       69  2.000       1.000       1.000     18.000  7.056  10.944  3.804  0.509 0.062
##     169  2.000       1.000       1.000     18.000  7.056  10.944  3.804  0.509 0.062
##       92  8.000       7.000      15.000     10.000 15.779  -5.779  -2.003  -0.440 0.048
##     192  8.000       7.000      15.000     10.000 15.779  -5.779  -2.003  -0.440 0.048
##     292  8.000       7.000      15.000     10.000 15.779  -5.779  -2.003  -0.440 0.048
##       20  2.000       1.000       8.000     18.000  9.711   8.289  2.857  0.425 0.044
##     120  2.000       1.000       8.000     18.000  9.711   8.289  2.857  0.425 0.044
##     283  9.000       7.000       5.000     22.000 12.468   9.532  3.291  0.424 0.043
##     269  2.000       1.000       5.000     18.000  8.573   9.427  3.250  0.388 0.036
##       97  2.000       1.000       3.000     14.000  7.815   6.185  2.111  0.250 0.015
##     197  2.000       1.000       3.000     14.000  7.815   6.185  2.111  0.250 0.015
##     297  2.000       1.000       3.000     14.000  7.815   6.185  2.111  0.250 0.015
##       82  2.000       1.000       4.000     14.000  8.194   5.806  1.979  0.231 0.013
##
##
## PREDICTION ERROR
##
## Data, Predicted, Standard Error of Forecast,
## 95% Prediction Intervals
## [sorted by lower bound of prediction interval]
## [to see all intervals do pred_rows="all"]
## -----
##
##      SJT EmotionalIntelligence Proactivity Performance   pred    sf pi.lwr pi.upr  width
##     69 2.000      1.000      1.000     18.000  7.056 2.994  1.163 12.949 11.785
##    169 2.000      1.000      1.000     18.000  7.056 2.994  1.163 12.949 11.785
##     41 1.000      2.000      3.000      8.000  7.421 3.023  1.472 13.370 11.898
## ...
##    238 4.000      3.000      8.000     12.000 10.849 2.984  4.976 16.721 11.746
##      4 6.000      5.000      5.000     11.000 10.849 2.974  4.997 16.701 11.704
##     47 6.000      5.000      5.000     11.000 10.849 2.974  4.997 16.701 11.704
## ...
##    256 8.000      8.000      8.000     16.000 13.212 2.991  7.326 19.098 11.772
##     28 4.000      7.000     14.000     12.000 13.474 3.120  7.335 19.614 12.279
##    128 4.000      7.000     14.000     12.000 13.474 3.120  7.335 19.614 12.279
##
## -----
## Plot 1: Distribution of Residuals
## Plot 2: Residuals vs Fitted Values
## Plot 3: ScatterPlot Matrix
## -----
```

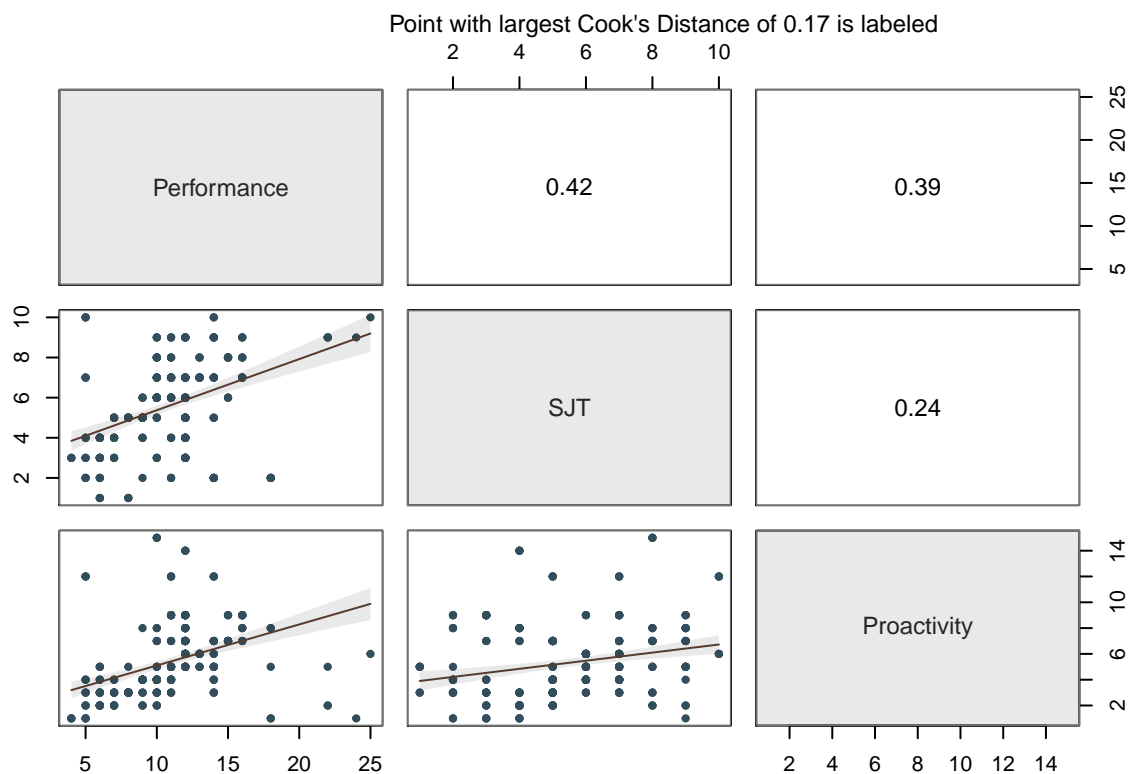
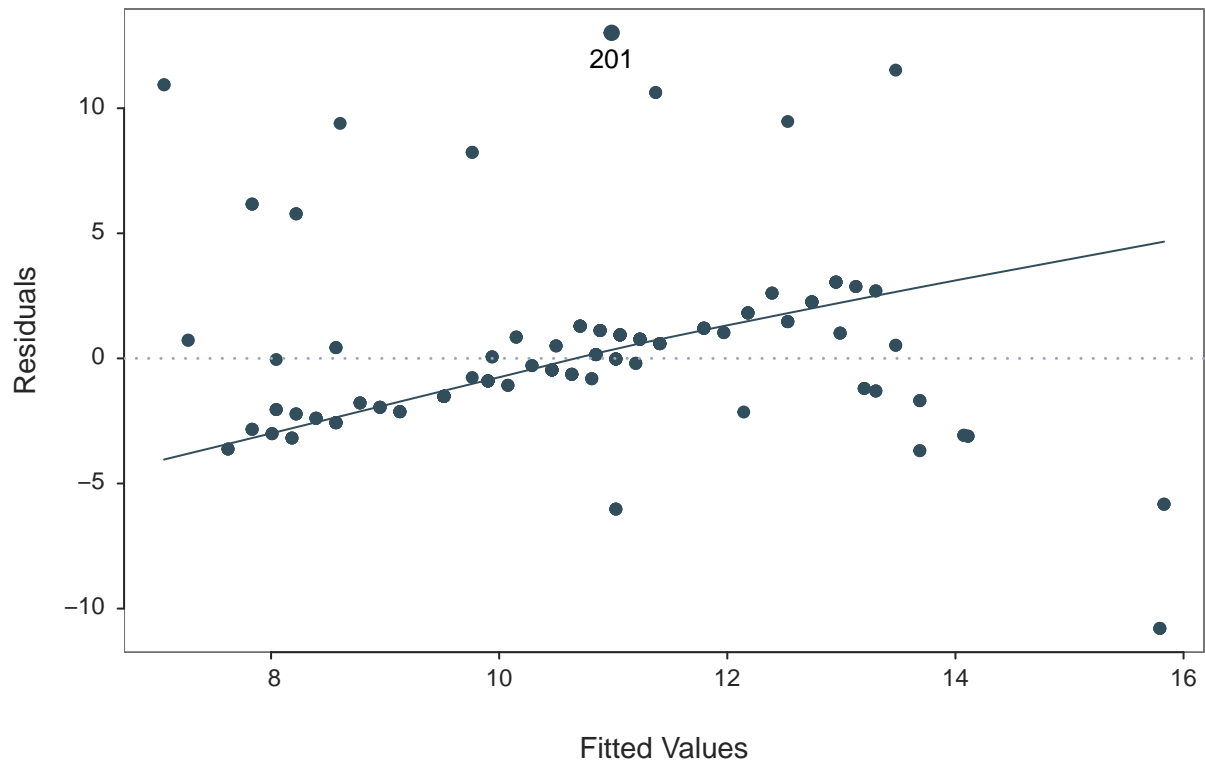
From the Collinearity part of the output we see that there is multicollinearity! SJT and EmotionalIntelligence are 0.93 correlated. Also the tolerance for them is below 0.2 (1.0 is optimal) and therefore we have to drop one of them in order to do proper regression. Because the p-value for EmotionalIntelligence is largest, we remove it.

We can also notice the higher Adjusted R-squared score of 0.259. We could say that this model performs better than the model with SJT only.

### Dropping the EmotionalIntelligence feature

```
Regression(Performance ~ SJT + Proactivity, data=df)
```





```
## >>> Suggestion
## # Create an R markdown file for interpretative output with Rmd = "file_name"
## Regression(my_formula=Performance ~ SJT + Proactivity, data=df, Rmd="eg")
##
##
```

```

## BACKGROUND
##
## Data Frame:  df
##
## Response Variable: Performance
## Predictor Variable 1: SJT
## Predictor Variable 2: Proactivity
##
## Number of cases (rows) of data:  300
## Number of cases retained for analysis:  300
##
## BASIC ANALYSIS
##
##           Estimate      Std Err  t-value  p-value   Lower 95%   Upper 95%
## (Intercept)    5.555      0.534   10.399   0.000     4.503     6.606
##           SJT      0.561      0.084    6.695   0.000     0.396     0.725
## Proactivity    0.386      0.063    6.099   0.000     0.261     0.510
##
## Standard deviation of Performance: 3.447
##
## Standard deviation of residuals:  2.964 for 297 degrees of freedom
## 95% range of residual variation:  11.666 = 2 * (1.968 * 2.964)
##
## R-squared:  0.266      Adjusted R-squared:  0.261      PRESS R-squared:  0.238
##
## Null hypothesis of all 0 population slope coefficients:
##   F-statistic: 53.733      df: 2 and 297      p-value:  0.000
##
## -- Analysis of Variance
##
##           df      Sum Sq   Mean Sq   F-value   p-value
##           SJT      1   617.264   617.264   70.265    0.000
## Proactivity      1   326.793   326.793   37.200    0.000
##
## Model           2   944.056   472.028   53.733    0.000
## Residuals      297  2609.074     8.785
## Performance    299  3553.130    11.883
##
##
## K-FOLD CROSS-VALIDATION
##
##
## RELATIONS AMONG THE VARIABLES
##
##           Performance  SJT Proactivity
## Performance      1.00  0.42      0.39
##           SJT      0.42  1.00      0.24
## Proactivity      0.39  0.24      1.00
##
##           Tolerance      VIF
##           SJT      0.944    1.060
## Proactivity      0.944    1.060
##

```

```

## SJT Proactivity   adjr2   X's
##      1           1   0.261   2
##      1           0   0.171   1
##      0           1   0.152   1
##
## [based on Thomas Lumley's leaps function from the leaps package]
##
##
## RESIDUALS AND INFLUENCE
##
## Data, Fitted, Residual, Studentized Residual, Dffits, Cook's Distance
## [sorted by Cook's Distance]
## [res_rows = 20, out of 300 rows of data, or do res_rows="all"]
## -----
##      SJT Proactivity Performance fitted   resid rstdnt dffits cooks
## 201  9.000           1.000      24.000 10.986  13.014  4.596  0.745 0.173
##  70 10.000           12.000       5.000 15.792 -10.792 -3.781 -0.675 0.145
## 170 10.000           12.000       5.000 15.792 -10.792 -3.781 -0.675 0.145
## 270 10.000           12.000       5.000 15.792 -10.792 -3.781 -0.675 0.145
## 233 10.000           6.000      25.000 13.476  11.524  4.023  0.551 0.096
##   1  9.000           2.000      22.000 11.372  10.628  3.701  0.544 0.094
## 101  9.000           2.000      22.000 11.372  10.628  3.701  0.544 0.094
##  69  2.000           1.000      18.000  7.062  10.938  3.807  0.509 0.083
## 169  2.000           1.000      18.000  7.062  10.938  3.807  0.509 0.083
##  92  8.000          15.000      10.000 15.829  -5.829 -2.022 -0.433 0.062
## 192  8.000          15.000      10.000 15.829  -5.829 -2.022 -0.433 0.062
## 292  8.000          15.000      10.000 15.829  -5.829 -2.022 -0.433 0.062
##  20  2.000           8.000      18.000  9.764   8.236  2.839  0.399 0.052
## 120  2.000           8.000      18.000  9.764   8.236  2.839  0.399 0.052
## 283  9.000           5.000      22.000 12.530   9.470  3.268  0.377 0.046
## 269  2.000           5.000      18.000  8.606   9.394  3.241  0.374 0.045
##  97  2.000           3.000      14.000  7.834   6.166  2.107  0.247 0.020
## 197  2.000           3.000      14.000  7.834   6.166  2.107  0.247 0.020
## 297  2.000           3.000      14.000  7.834   6.166  2.107  0.247 0.020
##  82  2.000           4.000      14.000  8.220   5.780  1.972  0.225 0.017
##
##
## PREDICTION ERROR
##
## Data, Predicted, Standard Error of Forecast,
## 95% Prediction Intervals
## [sorted by lower bound of prediction interval]
## [to see all intervals do pred_rows="all"]
## -----
##
##      SJT Proactivity Performance   pred    sf pi.lwr pi.upr  width
##  69  2.000           1.000      18.000 7.062 2.990   1.178 12.946 11.768
## 169  2.000           1.000      18.000 7.062 2.990   1.178 12.946 11.768
##  41  1.000           3.000       8.000 7.273 2.992   1.384 13.162 11.778
## ...
## 210  8.000           2.000      10.000 10.811 2.987   4.933 16.689 11.756
##   4  6.000           5.000      11.000 10.848 2.969   5.005 16.691 11.687
##  47  6.000           5.000      11.000 10.848 2.969   5.005 16.691 11.687
## ...

```

```
##      226 7.000      12.000      11.000 14.110 2.997  8.212 20.009 11.796
##      92 8.000      15.000      10.000 15.829 3.028  9.869 21.788 11.919
##     192 8.000      15.000      10.000 15.829 3.028  9.869 21.788 11.919
##
## -----
## Plot 1: Distribution of Residuals
## Plot 2: Residuals vs Fitted Values
## Plot 3: ScatterPlot Matrix
## -----
```

Now the non-collinearity assumption is met based on the **Tolerance** statistics. The residuals are normally distributed based on **Plot 1: Distribution of Residuals**. There is no evidence against heteroscedasticity in **Plot 2: Residuals vs Fitted Values**. Also, although there is a slight bend, the average error is not that far from 0.

The **Adjusted R-squared** value is 0.261, i.e. we keep improving the model and explaining more and more variance.

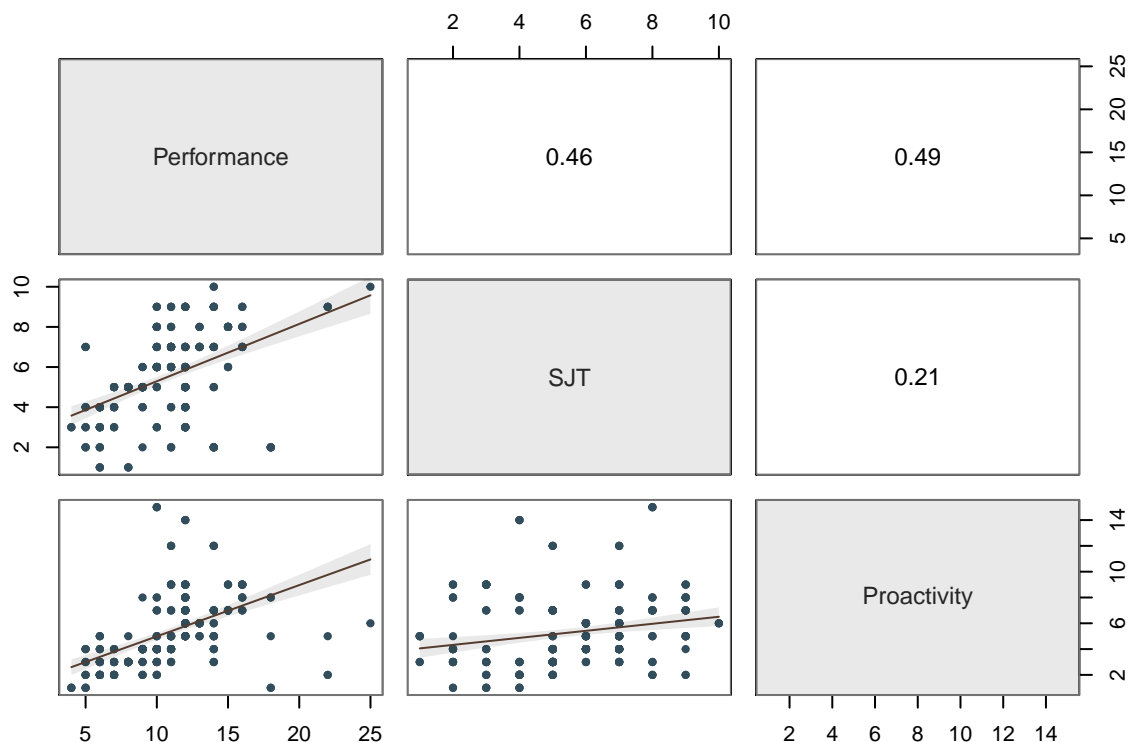
Looking at the **BASIC ANALYSIS** part of the output, we see that when **Proactivity** is fixed, for every 1 unit increase in **SJT**, there is a 0.561 increase in **Performance**. When **SJT** is fixed, for every 1 unit increase in **Proactivity**, there is a 0.386 increase in **Performance**. This goes to say that an increase in **Proactivity** is better (more important) than an increase in **SJT**.

This is the line we would get:  $\text{Performance} = 5.555 + 0.561 \cdot \text{SJT} + 0.386 \cdot \text{Proactivity}$ .

### Removing potential outliers

From the **RESIDUALS AND INFLUENCE** part of the output, we see that observations 201, 70, 170, and 270 have a large Cook's Distance, i.e. our model performed poorly on them. We can test to see if removing them yields a better result.

```
reg_brief(Performance ~ SJT + Proactivity, data=df, rows = (!ID %in% c(201, 70, 170, 270)))
```



```

## >>> Suggestion
## # Create an R markdown file for interpretative output with Rmd = "file_name"
## reg(Performance ~ SJT + Proactivity, data=df, rows=(!ID %in% c(201, 70, 170, 270)), Rmd="eg")
##
##
## BACKGROUND
##
## Data Frame: df
##
## Response Variable: Performance
## Predictor Variable 1: SJT
## Predictor Variable 2: Proactivity
##
## Number of cases (rows) of data: 296
## Number of cases retained for analysis: 296
##
##
## BASIC ANALYSIS
##
##           Estimate      Std Err  t-value  p-value  Lower 95%  Upper 95%
## (Intercept)    4.734      0.491    9.649   0.000    3.769    5.700
## SJT            0.608      0.076    7.966   0.000    0.458    0.758
## Proactivity    0.504      0.058    8.688   0.000    0.390    0.619
##
## Standard deviation of Performance: 3.334
##
## Standard deviation of residuals: 2.646 for 293 degrees of freedom
## 95% range of residual variation: 10.417 = 2 * (1.968 * 2.646)
##
## R-squared: 0.374    Adjusted R-squared: 0.370    PRESS R-squared: 0.351
##
## Null hypothesis of all 0 population slope coefficients:
## F-statistic: 87.570    df: 2 and 293    p-value: 0.000
##
## -- Analysis of Variance
##
##           df      Sum Sq  Mean Sq  F-value  p-value
## SJT        1    697.913   697.913   99.656   0.000
## Proactivity 1    528.628   528.628   75.484   0.000
##
## Model      2   1226.542   613.271   87.570   0.000
## Residuals 293  2051.945    7.003
## Performance 295  3278.486   11.114
##
##
## K-FOLD CROSS-VALIDATION
##
##
## RELATIONS AMONG THE VARIABLES
##
##
## RESIDUALS AND INFLUENCE
##
##

```

## ## PREDICTION ERROR

The new Adjusted R-squared is 0.370 which means that with this data it is better to remove those observations but we should be careful before drawing conclusions. Those samples may not be true outliers, but rather the result of poor sampling.