QUASAR -- A quasar is a distant celestial object (at least four billion lightyears away) that provides a powerful source of radio energy. The Astronomical Journal (July 1995) re ported on a study of 90 quasars detected by a deep space survey. The survey enabled astronomers to measure several different quantitative characteristics of each quasar, including:

Y1 - Rest frame Equivalent Width

# Use R to perform a regression analysis on the QUASAR dataset

## **Evaluation of Explanatory Variables:**

#### X1 - Redshift

```
lm(formula = RFEWIDTH ~ REDSHIFT, data = QUASAR)
Residuals:
             1Q
                 Median
                              3Q
   Min
                 -8.504 24.590 166.590
-54.922 -36.077
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)
             112.115
                         70.151
                                   1.598
                                            0.124
REDSHIFT
              -7.013
                         20.477
                                  -0.342
                                            0.735
```

Residual standard error: 48.29 on 23 degrees of freedom

Multiple R-squared: 0.005073, Adjusted R-squared: -0.03818

F-statistic: 0.1173 on 1 and 23 DF, p-value: 0.7351

T-Test -> Negative value, so that means that the sample mean is less than the hypothesis

P value -> High P-Value (not less than .05)

R Squared -> .005073(.5%) of variability in Y1 (Rest Frame Equivalent width) is explained by our model. This is v ery low.

## X2 - Line Flux

```
lm(formula = RFEWIDTH ~ LINEFLUX, data = QUASAR)
Residuals:
             1Q
                 Median
                             3Q
   Min
                 -9.432 25.137 157.947
-59.053 -32.667
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                  1.181
(Intercept)
              665.77
                         563.70
                                            0.250
LINEFLUX
               41.83
                          40.83
                                  1.025
                                            0.316
Residual standard error: 47.35 on 23 degrees of freedom
Multiple R-squared: 0.04365,
                                   Adjusted R-squared: 0.002066
F-statistic: 1.05 on 1 and 23 DF, p-value: 0.3162
```

T-Test -> Fairly low value, tells us that the result have a low probability of being repeated

P value -> High P-Value (not less than .05)

R Squared -> .04 (4%) of variability in Y1 (Rest Frame Equivalent width) is explained by our model. This is low.

#### **X3** - Line Luminosity

```
call:
```

lm(formula = RFEWIDTH ~ LUMINOSITY, data = QUASAR)

Residuals:

Min 1Q Median 3Q Max -53.800 -30.427 -5.716 21.960 164.875

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) -1978.21 2226.43 -0.889 0.383
LUMINOSITY 45.78 49.32 0.928 0.363

Residual standard error: 47.53 on 23 degrees of freedom Multiple R-squared: 0.03611, Adjusted R-squared: -0.005803 F-statistic: 0.8615 on 1 and 23 DF, p-value: 0.3629

T-Test -> Fairly low value, tells us that the result have a low probability of being repeated

P value -> High P-Value (not less than .05)

R Squared -> .03611 (3.611%) of variability in Y1 (Rest Frame Equivalent width) is explained by our model. This is very low.

## X4 - AB1450 Magnitude

```
call:
```

lm(formula = RFEWIDTH ~ AB1450, data = QUASAR)

Residuals:

Min 1Q Median 3Q Max -50.630 -24.405 -3.409 7.946 144.479

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) -667.31 239.42 -2.787 0.0105 \*
AB1450 38.31 12.13 3.158 0.0044 \*\*

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

Residual standard error: 40.44 on 23 degrees of freedom Multiple R-squared: 0.3024, Adjusted R-squared: 0.2721 F-statistic: 9.972 on 1 and 23 DF, p-value: 0.004399

T-Test —> High T Value, tells us that the result have a good probability of being repeated

P value -> Low P-Value (under .05). This builds a strong case to reject the null hp

R Squared -> .3024 (30.24%) of variability in Y1 (Rest Frame Equivalent width) is explained by our model. This a relatively good for single variable, and should be looked at further.

# X5 - Absolute Magnitude

```
call:
lm(formula = RFEWIDTH \sim ABSMAG, data = QUASAR)
Residuals:
    Min
               10
                    Median
                                   3Q
-56.281 -22.287 -7.592
                             18.770 127.261
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                                         3.971 0.000605 ***
3.695 0.001197 **
(Intercept)
               1263.64
                              318.22
ABSMAG
                  44.63
                               12.08
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 38.36 on 23 degrees of freedom
Multiple R-squared: 0.3724, Adjusted R-squared F-statistic: 13.65 on 1 and 23 DF, p-value: 0.001197
                                          Adjusted R-squared:
T-Test -> High T Value, tells us that the result has a good probability
```

of being repeated

P value -> Low P-Value (under .05). This builds a strong case to reject the null hp

Adjusted R Squared -> .3724 (37.24%) of variability in Y1 (Rest Frame Equivalent width) is explained by our mod el. This a relatively good for single variable, and should be looked at further.

### Identifying the best model

The best model is for X5 – Absolute magnitude since it has the lowest p-value and the highest adjusted R squared. The low p value gives us the reasoning to reject the null hypothesis and accept the alternative. Since we are looking at one variable at a time we will be looking at R squared. The high R squared tells us that 37.24% of variability in Y is explained by the absolute magnitude.