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# Effect of Cognitive Behavioural and Family Treatment on Weight Changes in Anorexia nervosa



## Outline

- Introduction
- Data analysis by R
- Regression analysis by R
- Discussion and Conclusion



## What is anorexia nervosa?

- Serious psychiatric eating disorder
- Usually occurs in adolescent and adult female
- Clinical features
  - Severe weight loss
  - Malnutrition
  - Amenorrhea
  - Electrolyte and acid-base imbalance



# What is anorexia nervosa?

- Weight loss taken place by
  - diminished food intake caused by fear of overweight perception
- Treatment : psychiatrically pharmacological medication
- Poor prognosis
  - Patients usually deny eating
  - Causes severe malnutrition
  - End up with death from severe electrolyte and acid-base disturbance.





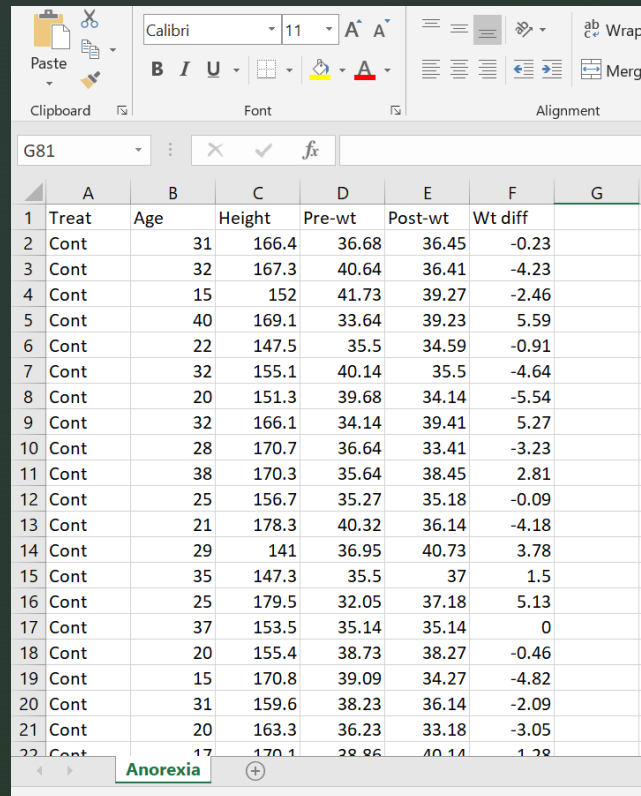
# Study Method

- Randomized control trial
- Separate patients into 3 groups
  - Control
  - Cognitive behavioural therapy (CBT)
  - Family treatment (FT)
- Sample size of each groups are 30 patients
- Patients are measure the weight before and after treatment
- Duration of treatment is 30 days



# Data Analysis

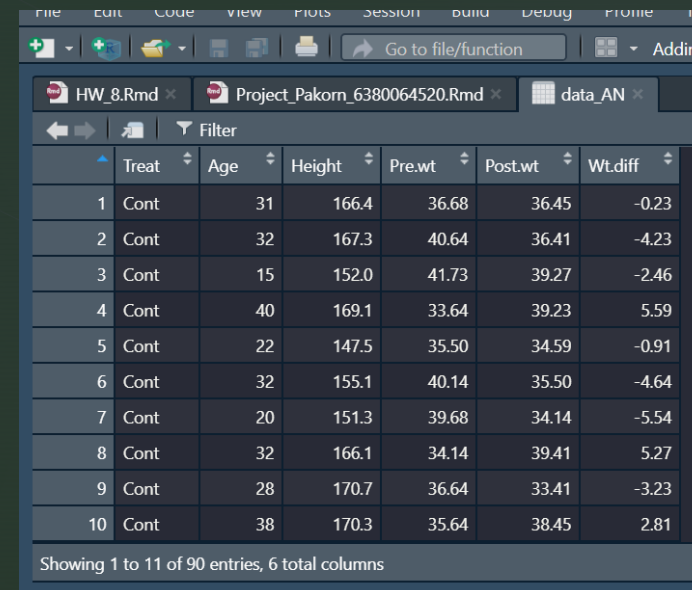
- Import data frame from pivot table in Excel



The Excel spreadsheet shows a pivot table with the following data:

	A	B	C	D	E	F	G
	Treat	Age	Height	Pre-wt	Post-wt	Wt diff	
1	Cont	31	166.4	36.68	36.45	-0.23	
2	Cont	32	167.3	40.64	36.41	-4.23	
3	Cont	15	152	41.73	39.27	-2.46	
4	Cont	40	169.1	33.64	39.23	5.59	
5	Cont	22	147.5	35.5	34.59	-0.91	
6	Cont	32	155.1	40.14	35.5	-4.64	
7	Cont	20	151.3	39.68	34.14	-5.54	
8	Cont	32	166.1	34.14	39.41	5.27	
9	Cont	28	170.7	36.64	33.41	-3.23	
10	Cont	38	170.3	35.64	38.45	2.81	
11	Cont	25	156.7	35.27	35.18	-0.09	
12	Cont	21	178.3	40.32	36.14	-4.18	
13	Cont	29	141	36.95	40.73	3.78	
14	Cont	35	147.3	35.5	37	1.5	
15	Cont	25	179.5	32.05	37.18	5.13	
16	Cont	37	153.5	35.14	35.14	0	
17	Cont	20	155.4	38.73	38.27	-0.46	
18	Cont	15	170.8	39.09	34.27	-4.82	
19	Cont	31	159.6	38.23	36.14	-2.09	
20	Cont	20	163.3	36.23	33.18	-3.05	
21	Cont	17	170.1	28.86	40.14	1.28	
22	Cont	17	170.1	28.86	40.14	1.28	

```
data_AN <- read.csv("D:\\Bioinfor MSc\\Stat for Bioinfo\\Project\\Pivot table
\\Anorexia.csv")
attach(data_AN)
head(data_AN)
```



The RStudio interface shows the data frame 'data\_AN' loaded from the Excel pivot table. The data is displayed in a table with the following columns: Treat, Age, Height, Pre.wt, Post.wt, and Wt.diff.

	Treat	Age	Height	Pre.wt	Post.wt	Wt.diff
1	Cont	31	166.4	36.68	36.45	-0.23
2	Cont	32	167.3	40.64	36.41	-4.23
3	Cont	15	152.0	41.73	39.27	-2.46
4	Cont	40	169.1	33.64	39.23	5.59
5	Cont	22	147.5	35.50	34.59	-0.91
6	Cont	32	155.1	40.14	35.50	-4.64
7	Cont	20	151.3	39.68	34.14	-5.54
8	Cont	32	166.1	34.14	39.41	5.27
9	Cont	28	170.7	36.64	33.41	-3.23
10	Cont	38	170.3	35.64	38.45	2.81

Showing 1 to 11 of 90 entries, 6 total columns



# Data Analysis

- General summary of dataframe

```
summary(data_AN)
```

```
##      Treat      Age      Height      Pre.wt
## Length:90      Min.   :15.00      Min.   :141.0      Min.   :31.82
## Class :character 1st Qu.:21.00      1st Qu.:156.0      1st Qu.:36.02
## Mode  :character Median :28.00      Median :165.5      Median :37.52
##                      Mean  :27.21      Mean  :163.0      Mean  :37.36
##                      3rd Qu.:32.75      3rd Qu.:170.4      3rd Qu.:39.03
##                      Max.   :40.00      Max.   :179.5      Max.   :43.14
##      Post.wt      Wt.diff
## Min.   :32.41      Min.   : -5.540
## 1st Qu.:36.21      1st Qu.: -0.595
## Median :38.62      Median :  1.150
## Mean   :39.03      Mean   :  1.668
## 3rd Qu.:41.63      3rd Qu.:  4.228
## Max.   :48.90      Max.   :14.880
```



# Data Analysis

- Test randomization
  - Age

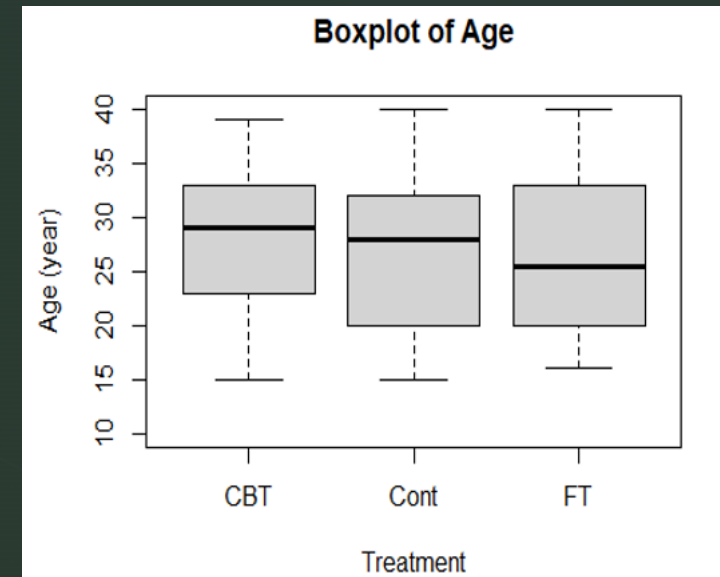
```
names(data_AN)
## [1] "Treat" "Age" "Height" "Pre.wt" "Post.wt" "Wt.diff"

plot(factor(Treat), Age, xlab="Treatment", ylab="Age (year)", main="Boxplot of Age", ylim=c(10, 40))
```

```
summary(aov(Age~Treat))

##           Df Sum Sq Mean Sq F value Pr(>F)
## Treat      2     38   19.03    0.367  0.694
## Residuals 87   4513   51.87
```

- Because p-value is more than 0.05
  - there are no differences in the mean of age in 3 groups
- Confirm randomization again by height approach.





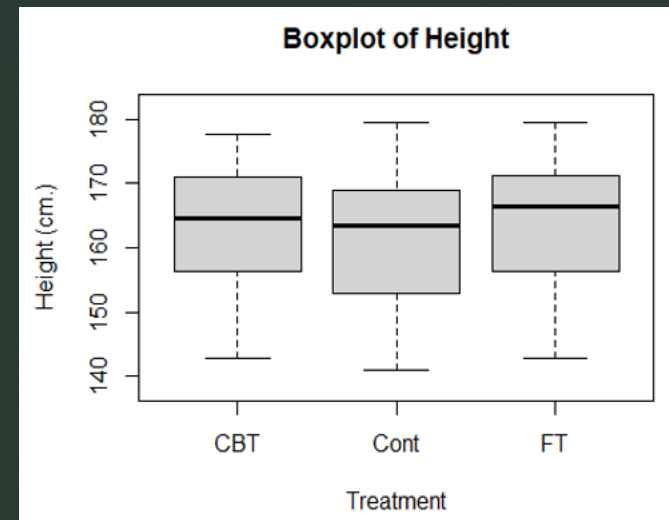
# Data Analysis

- Test randomization
  - Height

```
plot(factor(Treat),Height,xlab="Treatment",ylab="Height (cm.)", main="Boxplot  
of Height",ylim=c(138,182))
```

```
summary(aov(Height~Treat))
```

```
##           Df Sum Sq Mean Sq F value Pr(>F)  
## Treat      2    122    61.05    0.615  0.543  
## Residuals  87   8632    99.22
```



- p-value of ANOVA of height is also more than 0.05.
- It can be concluded that patients are randomly grouped



- Test normality
  - Control group

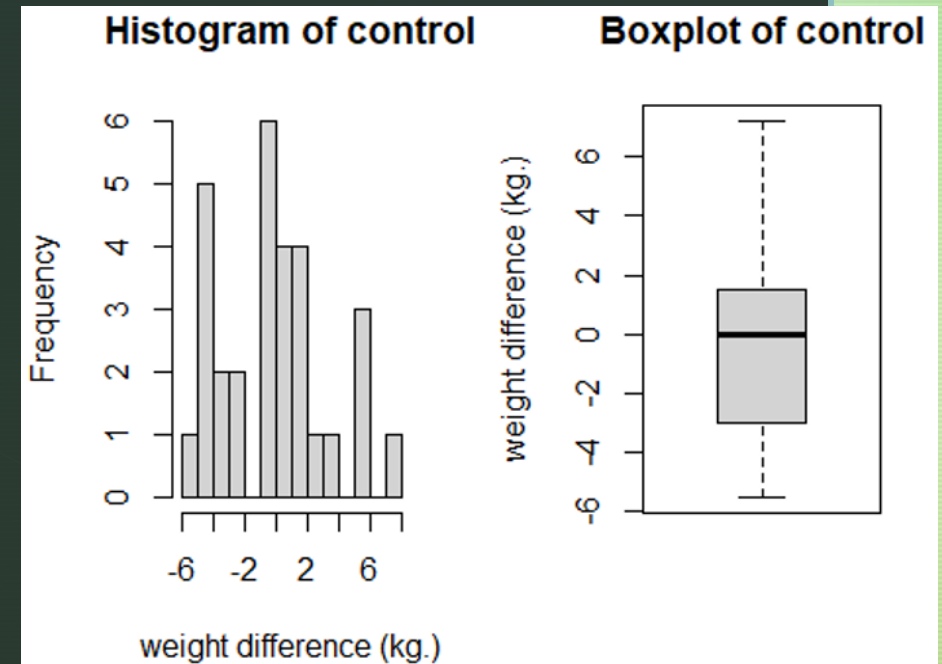
```
data_ctrl <- data_AN[1:30,6]
summary(data_ctrl)

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -5.540  -2.902  -0.025  -0.103   1.445   7.230

par(mfrow=c(1,2))
hist(data_ctrl,breaks=(-6:8),xlab="weight difference (kg.)",main="Histogram of control")
boxplot(data_ctrl,ylab="weight difference (kg.)",main="Boxplot of control")
```

- Histogram of control group looks like normal distribution
- As a result, it is tested by QQ plot.

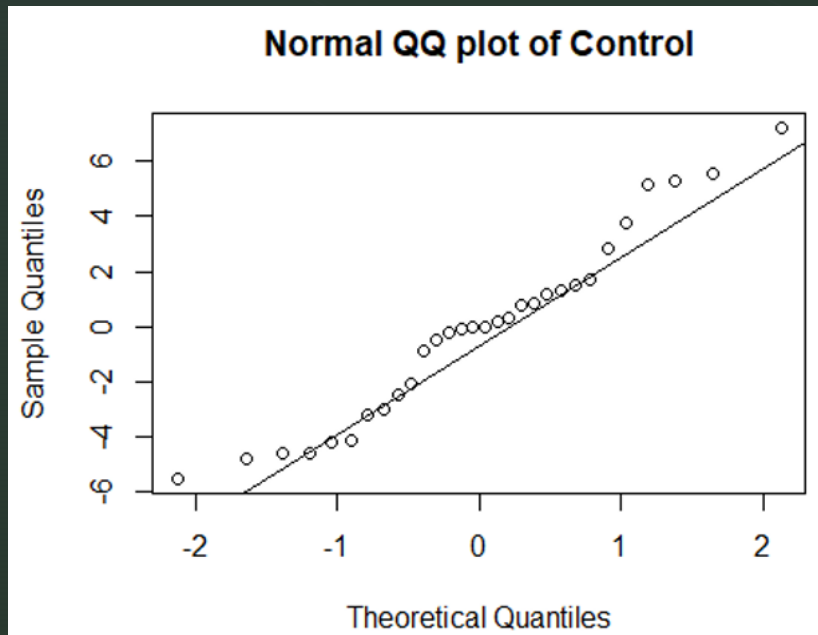
# Data Analysis



# Data Analysis

- Test normality
  - Control group

```
qqnorm(data_ctrl,main="Normal QQ plot of Control")  
qqline(data_ctrl)
```



```
shapiro.test(data_ctrl)  
  
##  
## Shapiro-Wilk normality test  
##  
## data:  data_ctrl  
## W = 0.95881, p-value = 0.2888  
  
library(nortest)  
ad.test(data_ctrl)  
  
##  
## Anderson-Darling normality test  
##  
## data:  data_ctrl  
## A = 0.41386, p-value = 0.3162
```

- QQ plot looks like normal distribution
  - Confirmed by two statistical tests (Shapiro-Wilk and Anderson test)
- No statistical significance in both normality tests.
  - Can conclude that data of weight change in control is normal distribution.



# Data Analysis

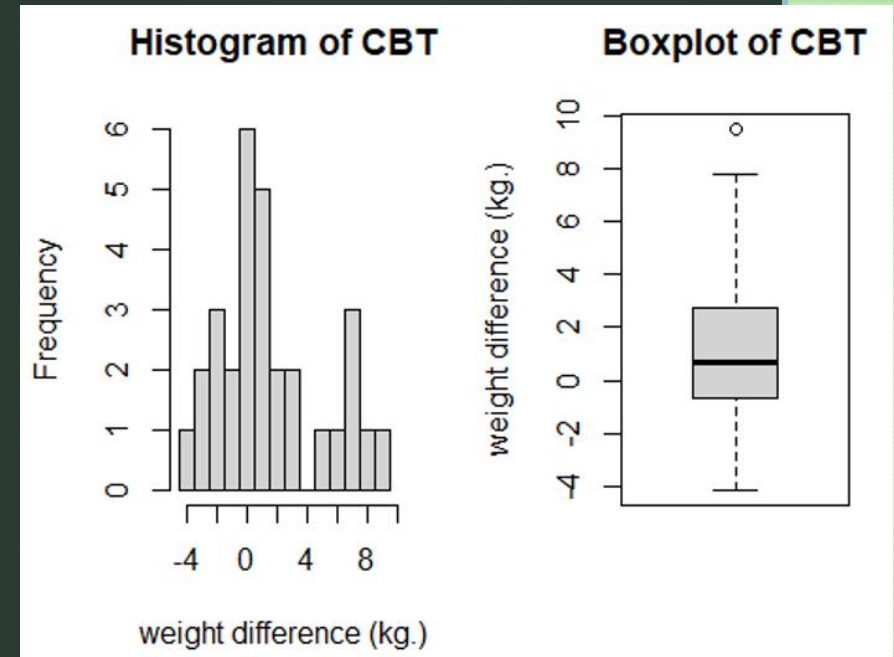
- Test normality
  - CBT group

```
data_cbt <- data_AN[31:60,6]
summary(data_cbt)

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -4.140  -0.570   0.685   1.499   2.708   9.500

par(mfrow=c(1,2))
hist(data_cbt,breaks=(-4.5:10),xlab="weight difference (kg.)",main="Histogram
of CBT")
boxplot(data_cbt,ylab="weight difference (kg.)",main="Boxplot of CBT")
```

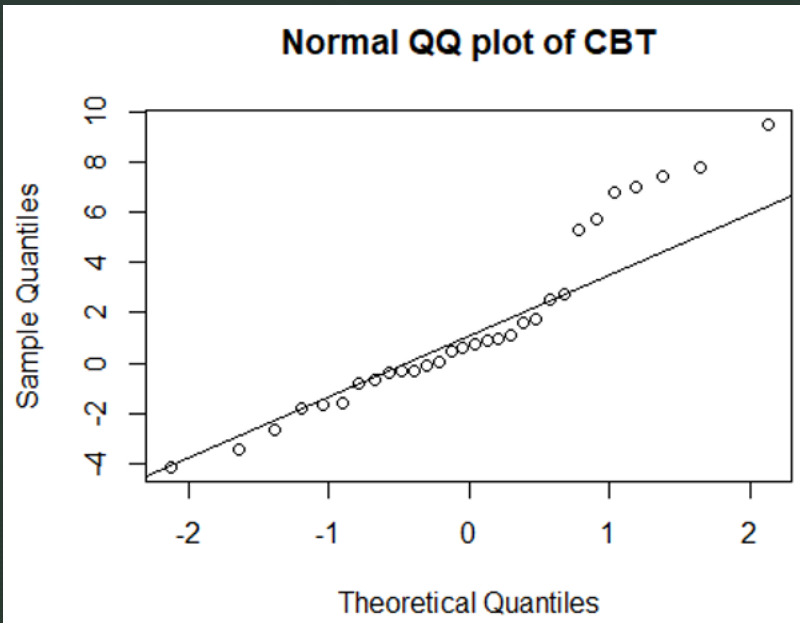
- Histogram of CBT is not normal distribution (right skewed pattern)
- Boxplot shows one upper outlier
  - Outlier cannot be deleted because weight before and after treatment are measured accurately.



# Data Analysis

- Test normality
  - CBT group

```
qqnorm(data_cbt, main="Normal QQ plot of CBT")  
qqline(data_cbt)
```



```
shapiro.test(data_cbt)  
  
##  
##  Shapiro-Wilk normality test  
##  
## data:  data_cbt  
## W = 0.92155, p-value = 0.02943  
  
ad.test(data_cbt)  
  
##  
##  Anderson-Darling normality test  
##  
## data:  data_cbt  
## A = 1.0383, p-value = 0.008431
```

- There is statistical significance in both normality test
- QQ plot and both normality tests tell weight change in CBT group is not normal distribution



# Data Analysis

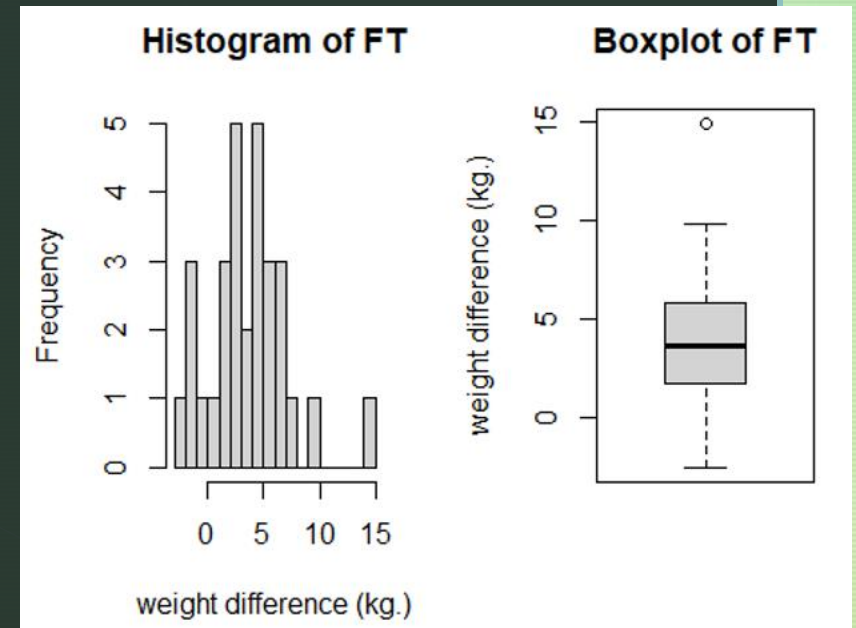
- Test normality
  - FT group

```
data_ft <- data_AN[61:90,6]
summary(data_ft)

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -2.480   1.790   3.675   3.608   5.690   14.880

par(mfrow=c(1,2))
hist(data_ft,breaks=(-3:15.5),xlab="weight difference (kg.)",main="Histogram
of FT")
boxplot(data_ft,ylab="weight difference (kg.)",main="Boxplot of FT")
```

- Histogram of FT is normal distribution when exclude outlier on right
- Boxplot shows one upper outlier
  - Nevertheless, outlier cannot be deleted because weight before and after treatment are measured accurately

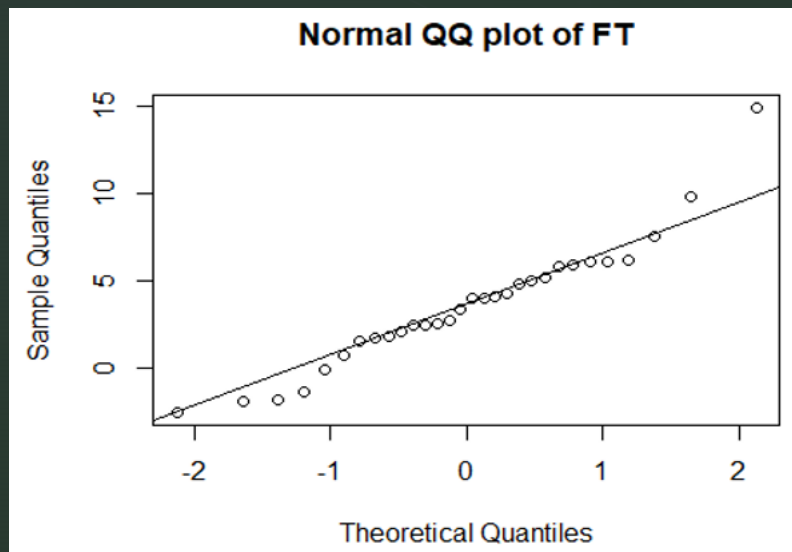




# Data Analysis

- Test normality
  - FT group

```
qqnorm(data_ft,main="Normal QQ plot of FT")  
qqline(data_ft)
```



```
shapiro.test(data_ft)  
  
##  
## Shapiro-Wilk normality test  
##  
## data: data_ft  
## W = 0.94048, p-value = 0.09373  
  
ad.test(data_ft)  
  
##  
## Anderson-Darling normality test  
##  
## data: data_ft  
## A = 0.44742, p-value = 0.2615
```

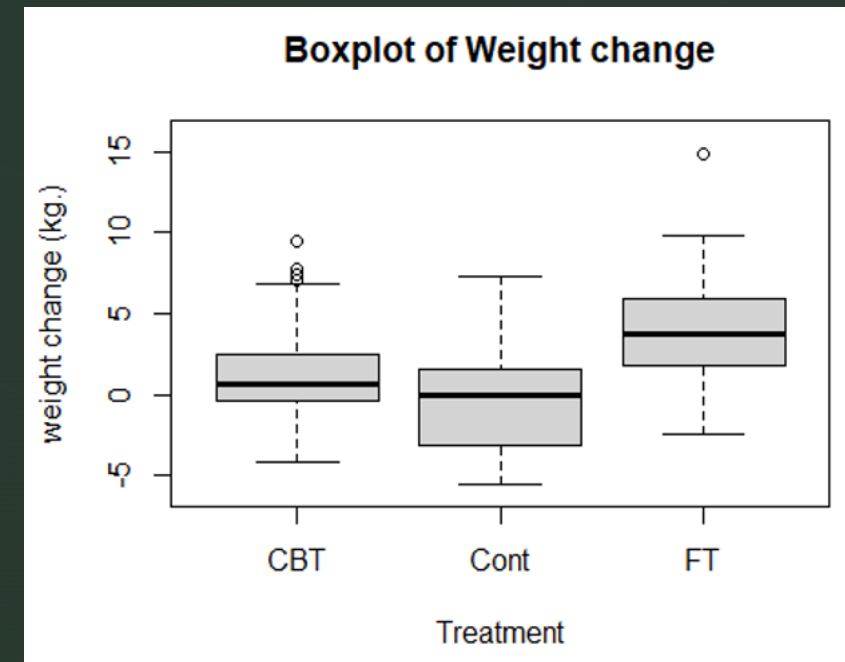
- QQ plot and both normality tests provide information that weight change in FT group is normal distribution



# Data Analysis

- Statistical analysis
  - Because data of weight difference of three groups are not normal distribution,
    - ANOVA cannot be used.
    - Kruskal-Willis test, non parametric test, should be used

```
plot(factor(Treat),Wt.diff,xlab="Treatment",ylab="weight change (kg.)", main="Boxplot of Weight change",ylim=c(-6,16))
```



# Data Analysis

- Statistical analysis

```
kruskal.test(Wt.diff ~ Treat, data = data_AN)

##
##  Kruskal-Wallis rank sum test
##
## data:  Wt.diff by Treat
## Kruskal-Wallis chi-squared = 14.645, df = 2, p-value = 0.0006605
```

- Test shows statistical significance in experiments
- Then, pairwise-comparison is performed



# Data Analysis

- Statistical analysis

```
pairwise.wilcox.test(Wt.diff,Treat,p.adj= "BH")

## Pairwise comparisons using Wilcoxon rank sum test with continuity correction
##
## data:  Wt.diff and Treat
##
##      CBT  Cont
## Cont 0.130 -
## FT   0.013 0.001
##
## P value adjustment method: BH
```

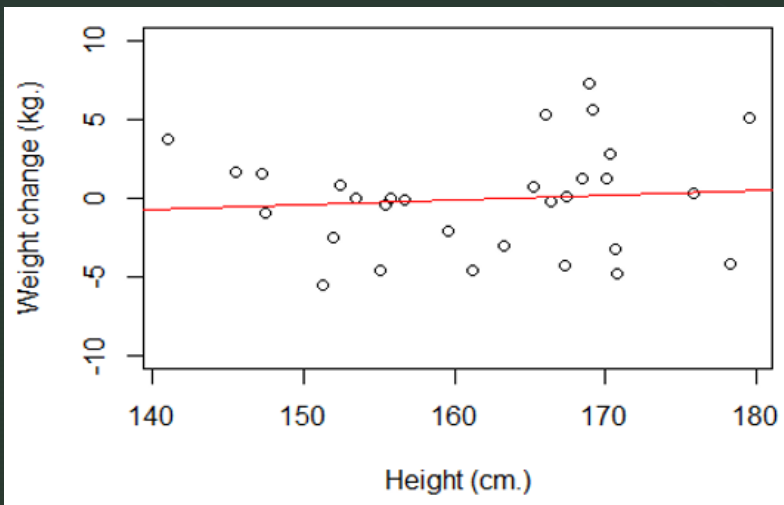
- Pairwise-comparison test gives information that Family treatment has impact on weight changes more than control group in AN patients in statistically significant level
- However, there are no statistically significant changes in weight between cognitive behavioural therapy and control group



# Regression Analysis

- Analyse relationship between weight changes, and age and height in each group of patients in experiment
- Control group

```
df_ctrl <- data_AN[1:30,]
plot(df_ctrl$Wt.diff ~ df_ctrl$Age,xlab="Age (year)",ylab="Weight change (kg.)",ylim=range(-10:10))
abline(lm(df_ctrl$Wt.diff ~ df_ctrl$Age),col="red")
```



```
summary(lm(df_ctrl$Wt.diff ~ df_ctrl$Height))

##
## Call:
## lm(formula = df_ctrl$Wt.diff ~ df_ctrl$Height)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.1291 -2.7624  0.1187  1.8213  7.1215
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -4.87573    10.19614  -0.478   0.636
## df_ctrl$Height  0.02951     0.06292   0.469   0.643
##
## Residual standard error: 3.433 on 28 degrees of freedom
## Multiple R-squared:  0.007794,    Adjusted R-squared:  -0.02764
## F-statistic: 0.2199 on 1 and 28 DF,  p-value: 0.6427
```

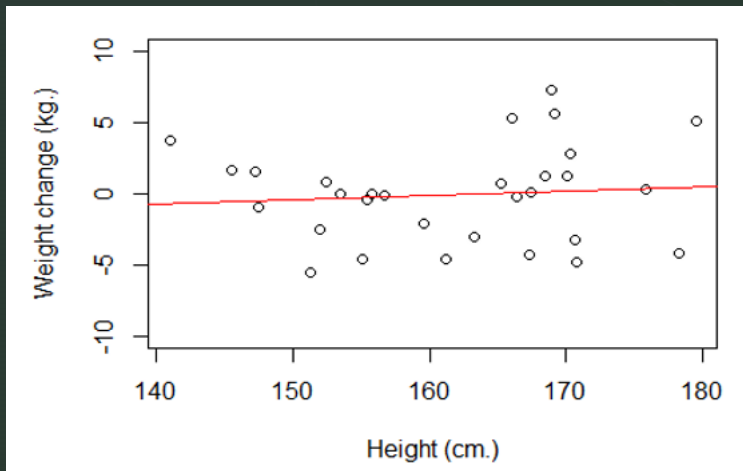
- Because R-squared is close to 0, linear model is not fit to data between weight change and height in control group.



# Regression Analysis

- Control group

```
plot(df_ctrl$Wt.diff ~ df_ctrl$Height,xlab="Height (cm.)",ylab="Weight change (kg.)",ylim=range(-10:10))
abline(lm(df_ctrl$Wt.diff ~ df_ctrl$Height),col="red")
```



- Because R-squared is close to 0
- Linear model is not fit to data between weight change and height in control group.

```
summary(lm(df_ctrl$Wt.diff ~ df_ctrl$Height))

##
## Call:
## lm(formula = df_ctrl$Wt.diff ~ df_ctrl$Height)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.1291 -2.7624  0.1187  1.8213  7.1215
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -4.87573    10.19614  -0.478   0.636
## df_ctrl$Height  0.02951     0.06292   0.469   0.643
##
## Residual standard error: 3.433 on 28 degrees of freedom
## Multiple R-squared:  0.007794,    Adjusted R-squared:  -0.02764
## F-statistic: 0.2199 on 1 and 28 DF,  p-value: 0.6427
```





# Regression Analysis

- Control group

```
summary(lm(df_ctrl$Wt.diff ~ df_ctrl$Age + df_ctrl$Height))

##
## Call:
## lm(formula = df_ctrl$Wt.diff ~ df_ctrl$Age + df_ctrl$Height)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.1834 -2.3607  0.1808  1.1927  8.0076
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -11.89570   10.27291  -1.158   0.2570
## df_ctrl$Age     0.16398    0.08084   2.028   0.0525 .
## df_ctrl$Height  0.04544    0.06021   0.755   0.4569
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.256 on 27 degrees of freedom
## Multiple R-squared:  0.139, Adjusted R-squared:  0.07522
## F-statistic: 2.179 on 2 and 27 DF, p-value: 0.1326
```

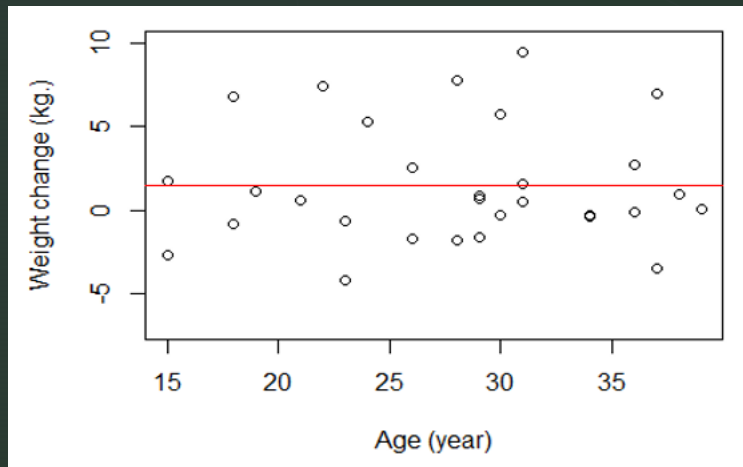
- In multiple linear regression (between weight change and age and height), there are weak correlations in model (r square is 0.14)



# Regression Analysis

- CBT group

```
df_cbt <- data_AN[31:60,]
plot(df_cbt$Wt.diff ~ df_cbt$Age,xlab="Age (year)",ylab="Weight change (kg.)"
,ylim=range(-7:10))
abline(lm(df_cbt$Wt.diff ~ df_cbt$Age),col="red")
```



```
summary(lm(df_cbt$Wt.diff ~ df_cbt$Age))

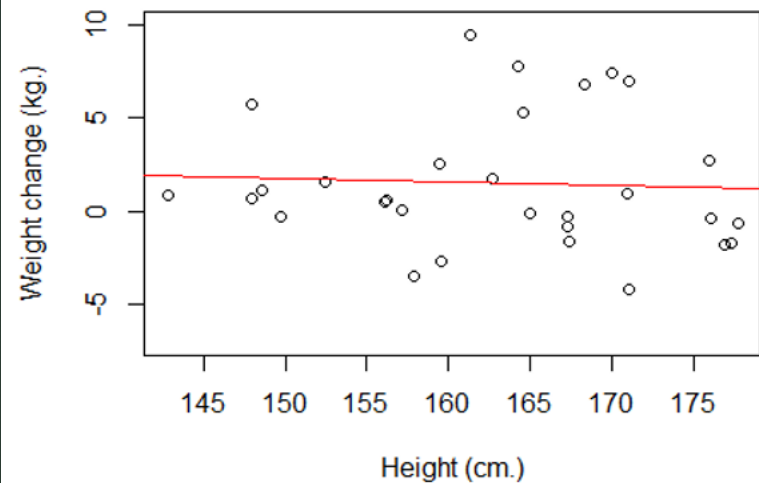
##
## Call:
## lm(formula = df_cbt$Wt.diff ~ df_cbt$Age)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.6257 -2.0633 -0.8063  1.1926  7.9920
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.421580    2.769284   0.513   0.612
## df_cbt$Age   0.002787    0.096412   0.029   0.977
##
## Residual standard error: 3.606 on 28 degrees of freedom
## Multiple R-squared:  2.984e-05, Adjusted R-squared:  -0.03568
## F-statistic: 0.0008355 on 1 and 28 DF,  p-value: 0.9771
```

- Because R-squared is close to 0
- Linear model is not fit to data between weight change and age in CBT group.



- CBT group

```
plot(df_cbt$Wt.diff ~ df_cbt$Height,xlab="Height (cm.)",ylab="Weight change (kg.)",ylim=range(-7:10))
abline(lm(df_cbt$Wt.diff ~ df_cbt$Height),col="red")
```



- Because R-squared is close to 0
- Linear model is not fit to data between weight change and height in CBT group.

# Regression Analysis

```
summary(lm(df_cbt$Wt.diff ~ df_cbt$Height))
```

```
##
## Call:
## lm(formula = df_cbt$Wt.diff ~ df_cbt$Height)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-5.485	-2.022	-1.040	1.377	7.967

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	4.65438	11.08223	0.420	0.678
df_cbt\$Height	-0.01935	0.06786	-0.285	0.778

```
##
## Residual standard error: 3.601 on 28 degrees of freedom
## Multiple R-squared: 0.002896, Adjusted R-squared: -0.03271
## F-statistic: 0.08134 on 1 and 28 DF, p-value: 0.7776
```



# Regression Analysis

- CBT group

```
summary(lm(df_cbt$Wt.diff ~ df_cbt$Age + df_cbt$Height))

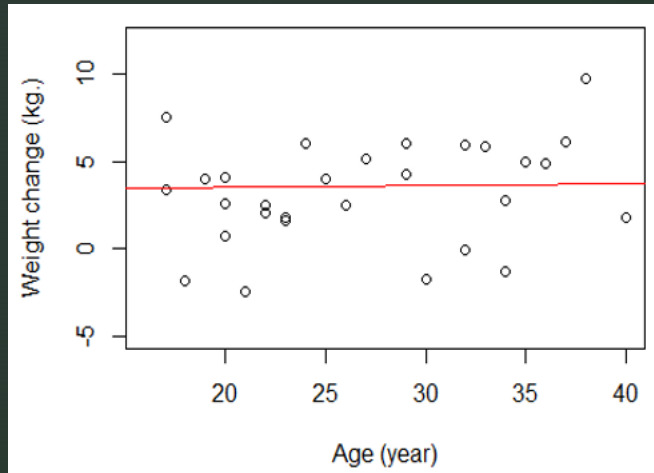
##
## Call:
## lm(formula = df_cbt$Wt.diff ~ df_cbt$Age + df_cbt$Height)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.466  -2.024  -1.046   1.357   7.956
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   4.565070   11.528676   0.396   0.695
## df_cbt$Age     0.003718    0.098094   0.038   0.970
## df_cbt$Height -0.019442    0.069142  -0.281   0.781
##
## Residual standard error: 3.667 on 27 degrees of freedom
## Multiple R-squared:  0.00295,    Adjusted R-squared:  -0.07091
## F-statistic: 0.03994 on 2 and 27 DF,  p-value: 0.9609
```

- In multiple linear regression (between weight change and age and height), there are weak correlations in model (r square is 0.003)



- FT group

```
df_ft <- data_AN[61:90,]
plot(df_ft$Wt.diff ~ df_ft$Age,xlab="Age (year)",ylab="Weight change (kg.)",y
lim=range(-5:12))
abline(lm(df_ft$Wt.diff ~ df_ft$Age),col="red")
```



# Regression Analysis

```
summary(lm(df_ft$Wt.diff ~ df_ft$Age))

##
## Call:
## lm(formula = df_ft$Wt.diff ~ df_ft$Age)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.0291 -1.9124  0.1259  2.0317 11.3829
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.33079    2.62274   1.270   0.215
## df_ft$Age     0.01040    0.09523   0.109   0.914
##
## Residual standard error: 3.66 on 28 degrees of freedom
## Multiple R-squared:  0.0004255, Adjusted R-squared:  -0.03527
## F-statistic: 0.01192 on 1 and 28 DF, p-value: 0.9138
```

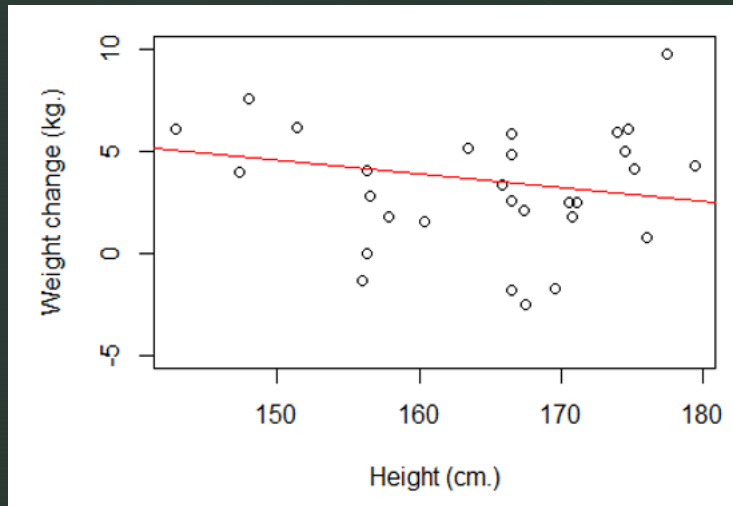
- Because R-squared is close to 0
- Linear model is not fit to data between weight change and age in FT group.



# Regression Analysis

- FT group

```
plot(df_ft$Wt.diff ~ df_ft$Height,xlab="Height (cm.)",ylab="Weight change (kg
.)",ylim=range(-5:10))
abline(lm(df_ft$Wt.diff ~ df_ft$Height),col="red")
```



```
summary(lm(df_ft$Wt.diff ~ df_ft$Height))

##
## Call:
## lm(formula = df_ft$Wt.diff ~ df_ft$Height)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.8651 -1.8713 -0.3912  1.7061 10.2856
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  14.63848   11.03692   1.326   0.195
## df_ft$Height -0.06718    0.06710  -1.001   0.325
##
## Residual standard error: 3.597 on 28 degrees of freedom
## Multiple R-squared:  0.03456,    Adjusted R-squared:  8.423e-05
## F-statistic: 1.002 on 1 and 28 DF,  p-value: 0.3253
```

- Because R-squared is close to 0
- Linear model is not fit to data between weight change and height in FT group.





# Regression Analysis

- FT group

```
summary(lm(df_ft$Wt.diff ~ df_ft$Age + df_ft$Height))

##
## Call:
## lm(formula = df_ft$Wt.diff ~ df_ft$Age + df_ft$Height)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.7305 -1.8126 -0.2373  1.6240 10.4964
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  14.37392   11.28357   1.274   0.214
## df_ft$Age     0.02267    0.09598   0.236   0.815
## df_ft$Height -0.06925    0.06882  -1.006   0.323
##
## Residual standard error: 3.659 on 27 degrees of freedom
## Multiple R-squared:  0.03656,    Adjusted R-squared:  -0.03481
## F-statistic: 0.5122 on 2 and 27 DF,  p-value: 0.6049
```

- In multiple linear regression (between weight change and age and height), there are weak correlations in model (r square is 0.037)



## Conclusion

- Patients are grouped randomly into 3 groups (control, CBT, and FT) because one-way ANOVA of age and height in groups is not statistically significant.
- Statistical analysis reveals that family treatment give result of weight change better than control but cognitive behavioural therapy does not give result better than control.
- In linear regression analysis, there are not any strong relationship between weight change, and age and height of patients in 3 groups.

