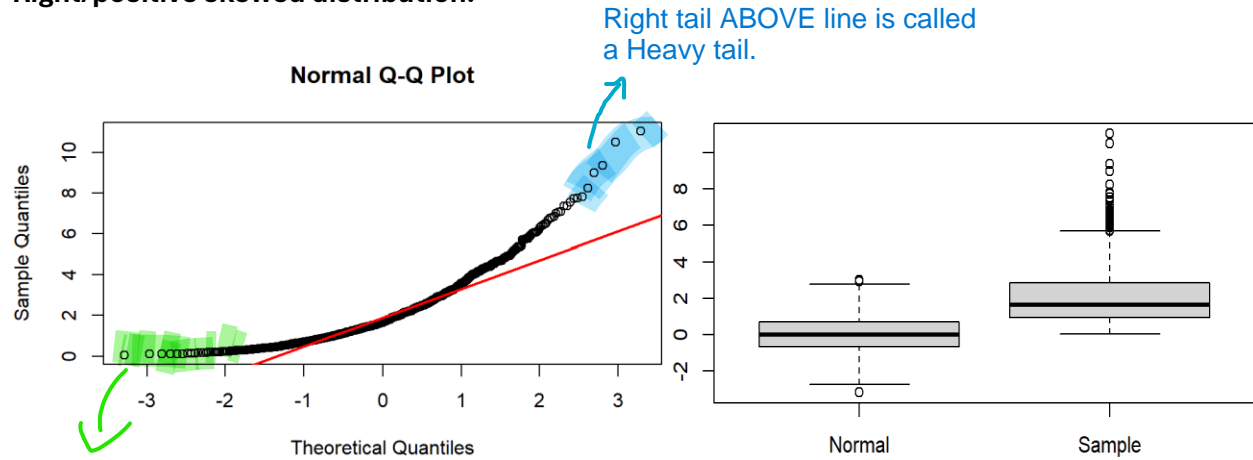


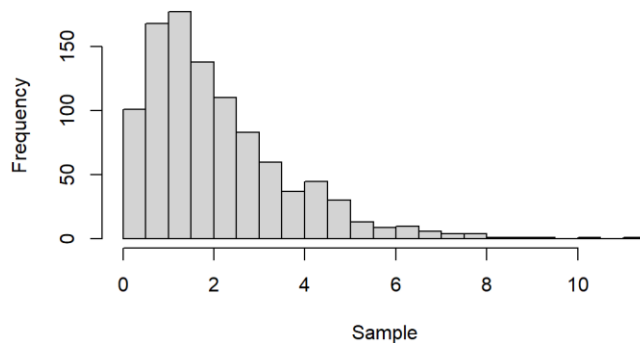
This file is made by me, to highlight different Q-Q plot shapes.

Part 1: Q-Q plots

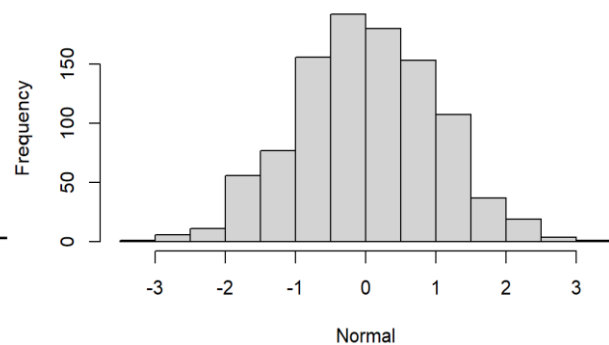
Right/positive skewed distribution:



Histogram of Sample



Histogram of Normal



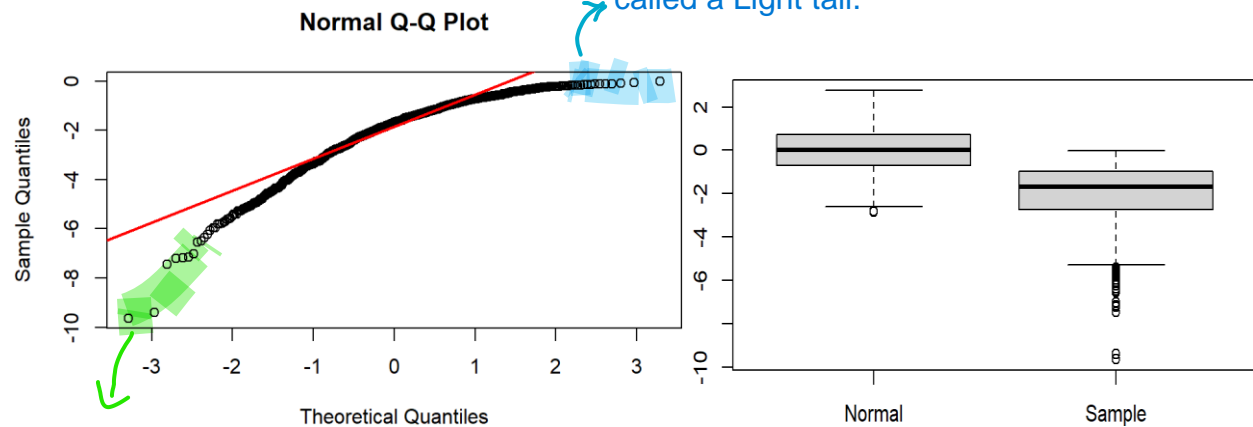
Normal	Sample
Min. : -2.85402	Min. : 0.02769
1st Qu.: -0.68906	1st Qu.: 0.92246
Median : 0.01147	Median : 1.65324
Mean : 0.01399	Mean : 1.93777
3rd Qu.: 0.73186	3rd Qu.: 2.58775
Max. : 2.75883	Max. : 9.76855

In this case, the Q-Q plot (backed up the histograms and boxplots) shows us that we observed more positive extreme values than would be expected under the Normal distribution.

Hence, distribution is right-skewed.

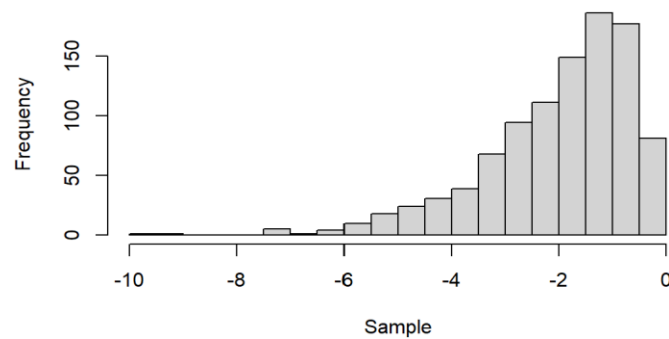
Left/Negative skewed distribution:

Right tail BELOW line is called a Light tail.

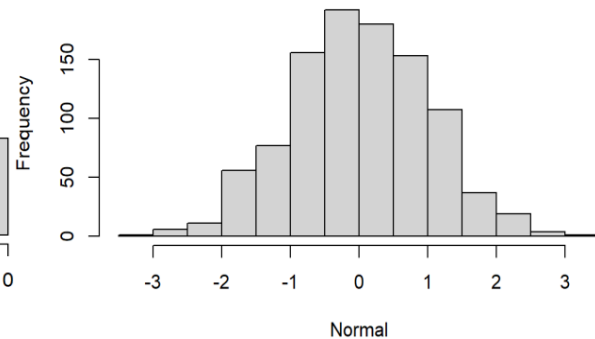


Left tail BELOW line is called a Heavy tail.

Histogram of Sample



Histogram of Normal

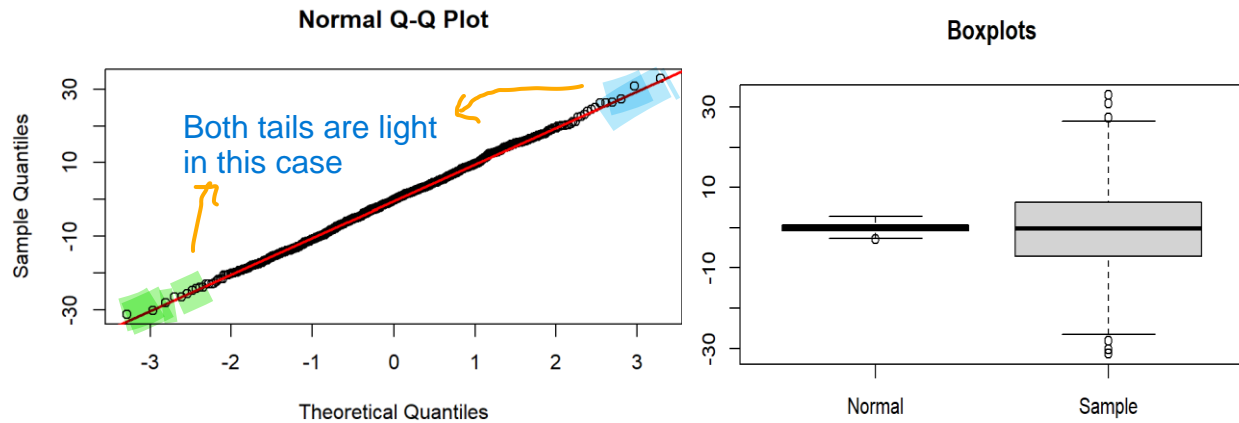


Normal	Sample
Min. : -2.85402	Min. : -9.64428
1st Qu.: -0.68906	1st Qu.: -2.73323
Median : 0.01147	Median : -1.68258
Mean : 0.01399	Mean : -2.01163
3rd Qu.: 0.73186	3rd Qu.: -0.97950
Max. : 2.75883	Max. : -0.02393

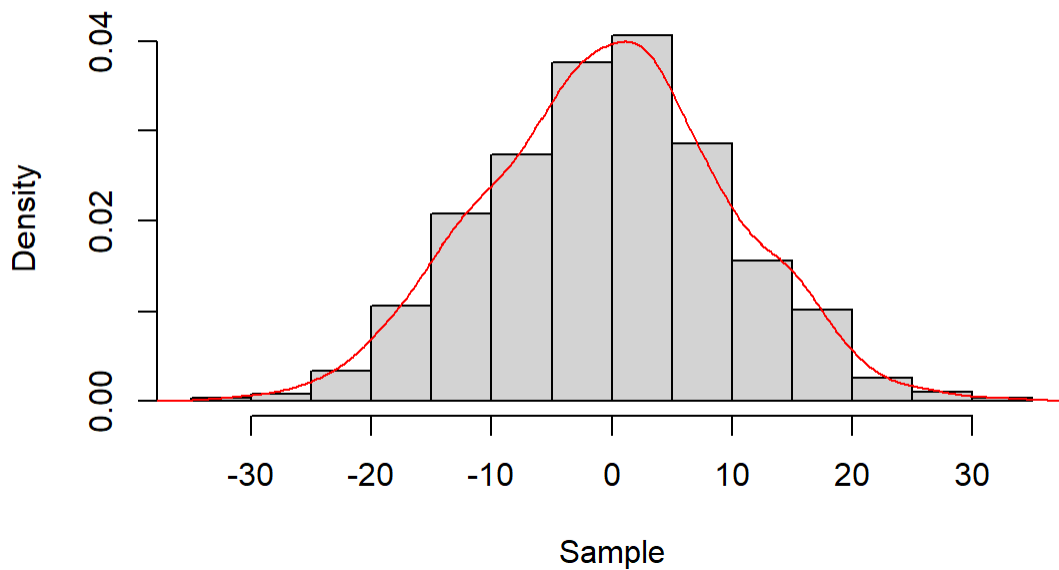
In this case, the Q-Q plot (backed up the histograms and boxplots) shows us that we observed more negative extreme values than would be expected under the Normal distribution.

Hence, distribution is left-skewed.

Light-tailed distribution:



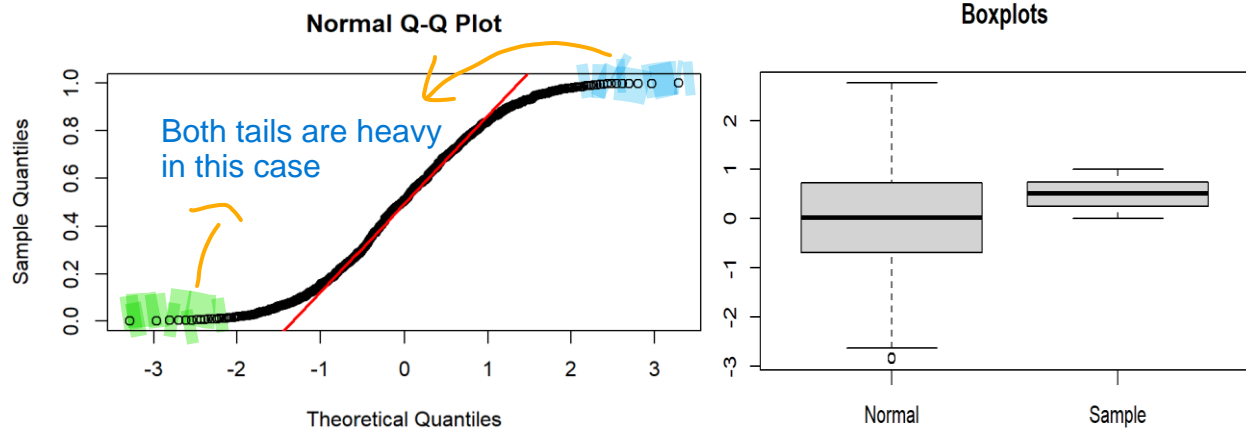
Histogram of Sample



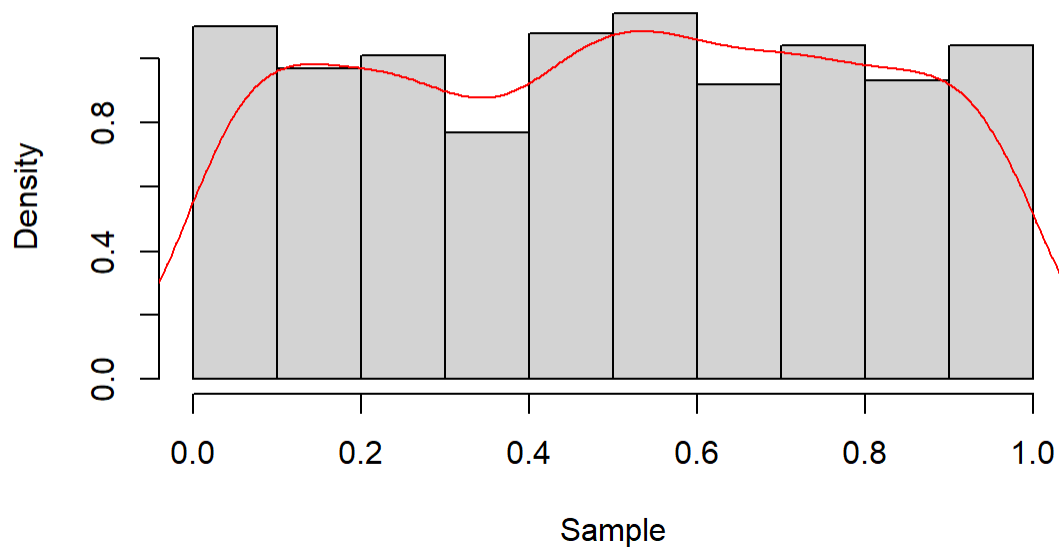
Normal	Sample
Min. : -2.85402	Min. : -31.2951
1st Qu.: -0.68906	1st Qu.: -7.1509
Median : 0.01147	Median : -0.1400
Mean : 0.01399	Mean : -0.3535
3rd Qu.: 0.73186	3rd Qu.: 6.2861
Max. : 2.75883	Max. : 32.8952

In this case, we don't observe as many extreme values as expected under its Normal distribution.

Heavy-tailed distribution:



Histogram of Sample

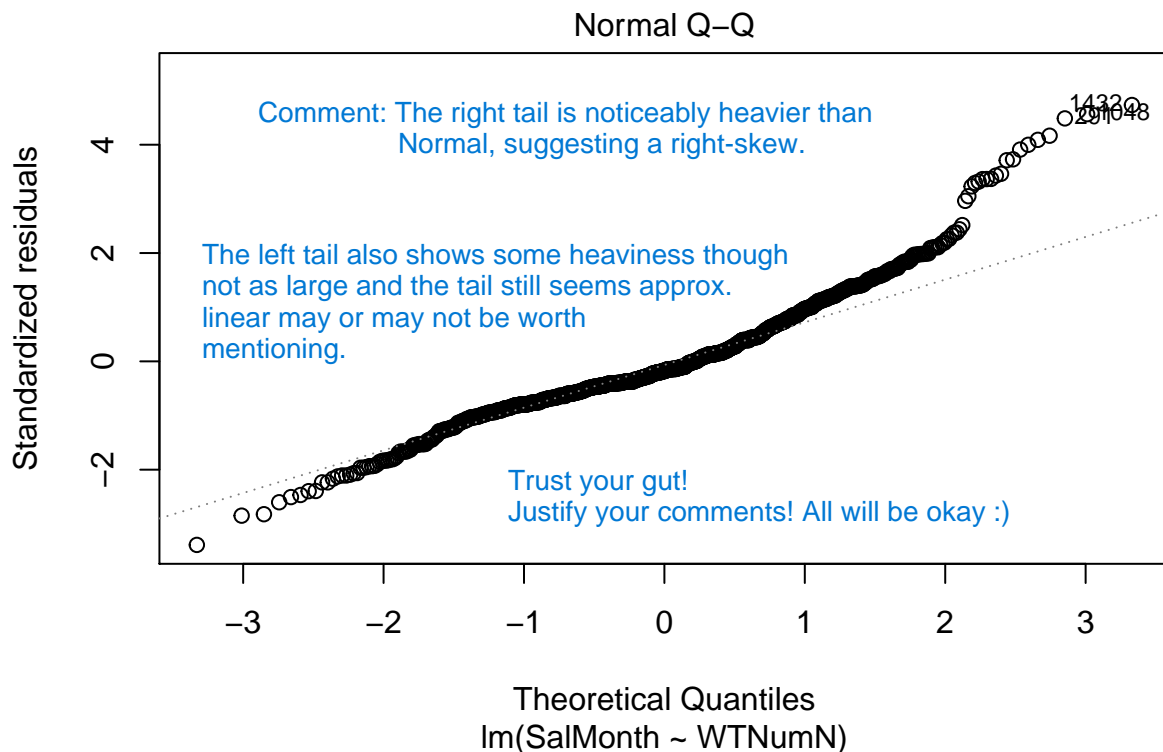
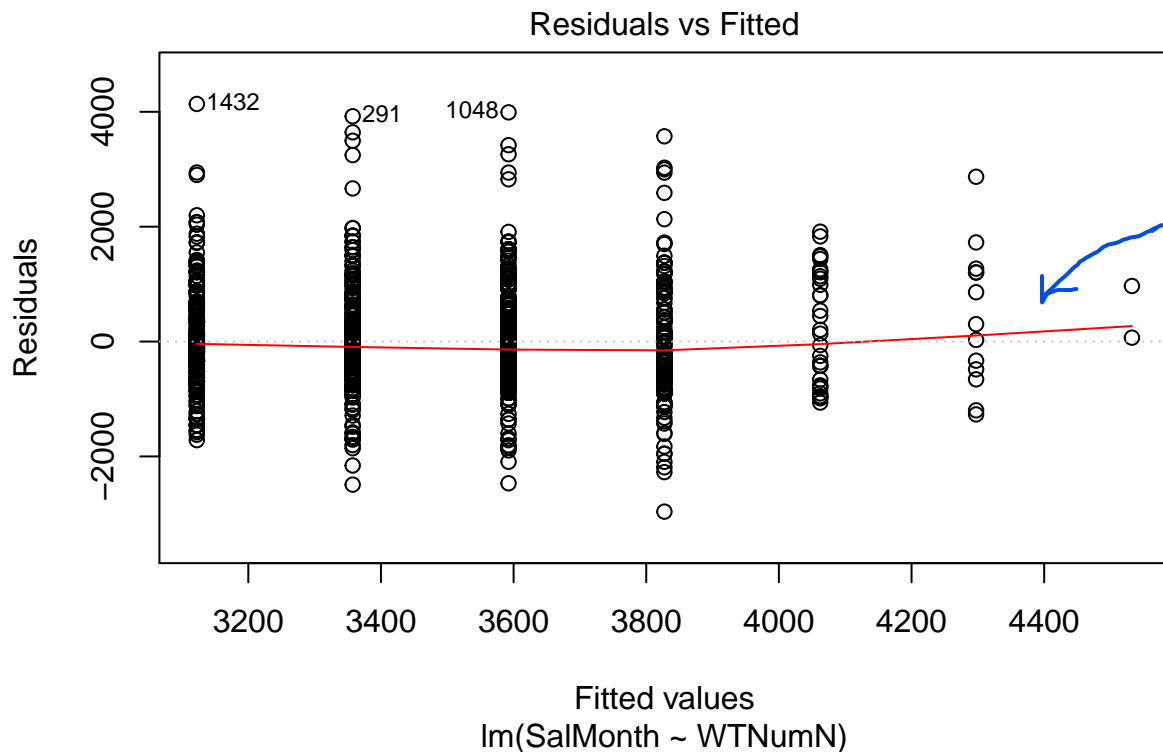


Normal	Sample
Min. : -2.85402	Min. : 0.0000412
1st Qu.: -0.68906	1st Qu.: 0.2425101
Median : 0.01147	Median : 0.4919487
Mean : 0.01399	Mean : 0.4965139
3rd Qu.: 0.73186	3rd Qu.: 0.7463338
Max. : 2.75883	Max. : 0.9999098

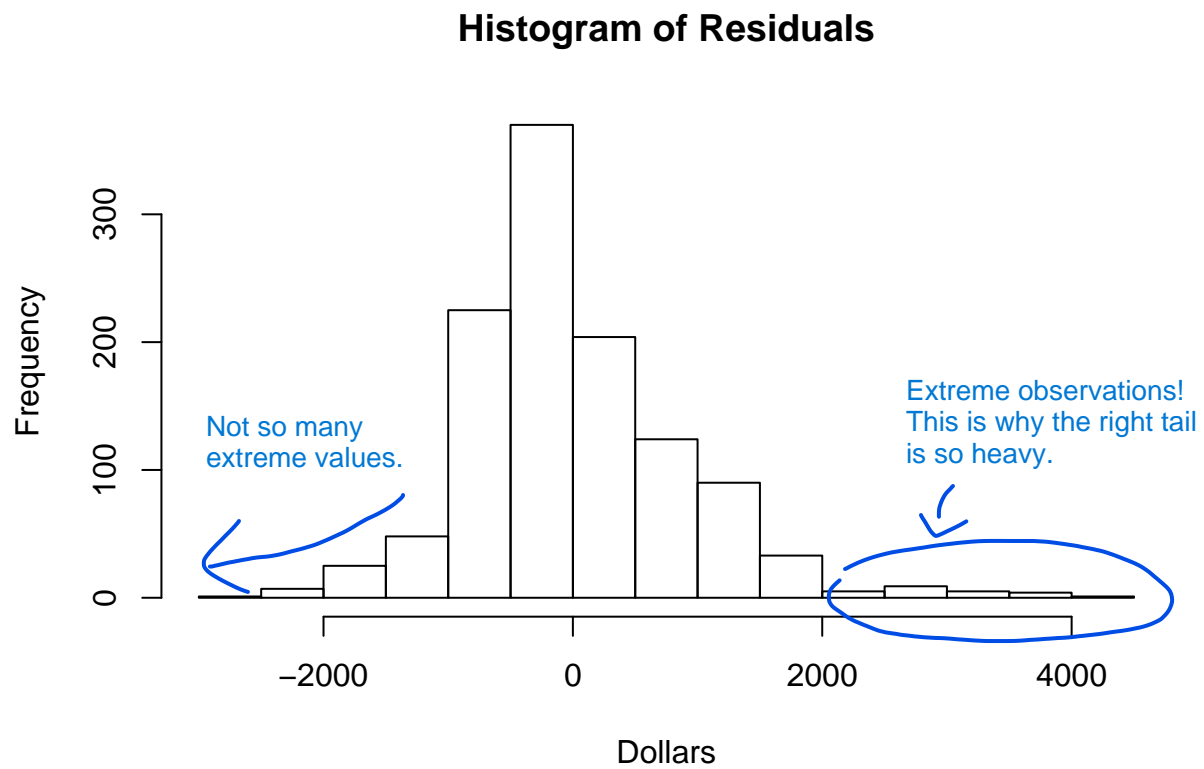
In this case, we observe too many extreme values than we'd expect under the Normal distribution.

Returning to the
Co-op Salary ~ Work term
model:

- No trends/patterns, so linearity assumptions seems valid
- Constant variance assumption looks good
- Residuals are scattered uniformly around 0, so normality seems ok.

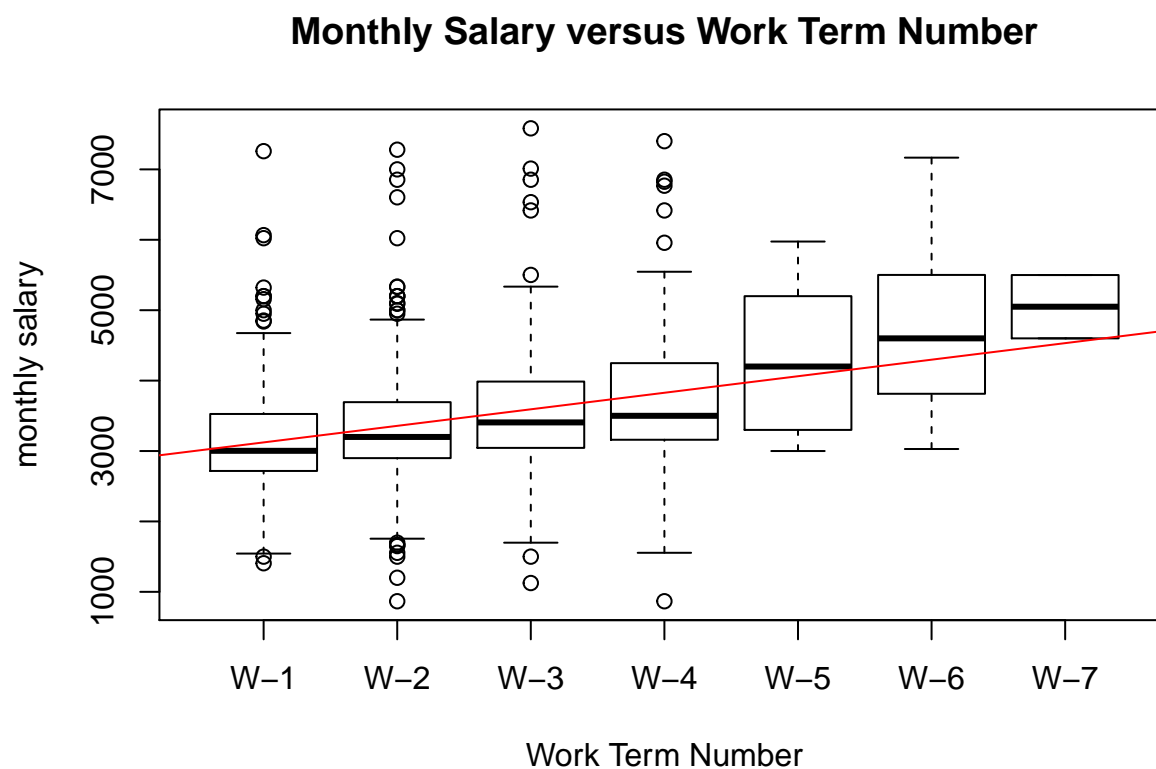


```
hist(resid(Sal.lm),main='Histogram of Residuals',xlab='Dollars')
```



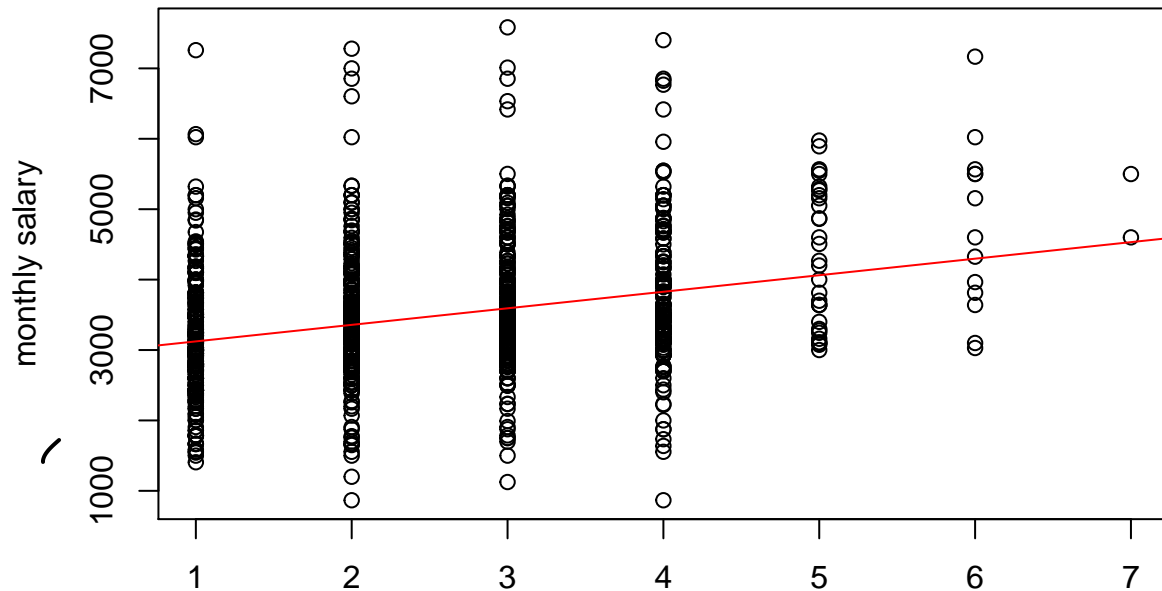
```
boxplot(salarynz$SalMonth~salarynz$WtNum, xlab='Work Term Number',ylab='monthly salary',  
        main='Monthly Salary versus Work Term Number')  
abline(Sal.lm,col=2)
```

The very first plot we looked at
in this section!



```
plot(salarynz$WtNumN, salarynz$SalMonth, xlab='Work Term Number', ylab='monthly salary',  
      main='Monthly Salary versus Work Term Number')  
abline(Sal.lm, col=2)
```

Monthly Salary versus Work Term Number



CI function in R

`confint(Sal.lm)`

this object is where we stored our `lm()` results.

```
##           2.5 %    97.5 %
## (Intercept) 2777.026 2997.7795
## WNumN       193.677  276.3077
```

α
13 } Present 95% CI for both parameters by default

`confint(Sal.lm, level=.99)`

```
##           0.5 %    99.5 %
## (Intercept) 2742.2548 3032.551
## WNumN       180.6617  289.323
```

If you want a different % CI, you can specify it.

`qt(.005,1149)` 99% critical value for t-distribution

```
## [1] -2.580115
```

`qnorm(.005)` 99% Critical value for $N(0,1)$

```
## [1] -2.575829
```

Given a critical value and `lm()` output, you should be able to manually compute CI'S for alpha and beta.

`anova(Sal.lm)`

```
## Analysis of Variance Table
```

```
##
```

```
## Response: SalMonth
```

```
##           Df      Sum Sq Mean Sq F value    Pr(>F)
## WNumN       1  95290565  95290565  124.54 < 2.2e-16 ***
## Residuals 1149  879171908   765163
```

```
##
```

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