

# Data Science Salaries Project

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```
ds.salaries <- read.csv(file = 'ds_salaries.csv')
str(ds.salaries)

## 'data.frame':    607 obs. of  12 variables:
## $ X                : int  0 1 2 3 4 5 6 7 8 9 ...
## $ work_year        : int  2020 2020 2020 2020 2020 2020 2020 2020 2020 ...
## $ experience_level  : chr   "MI" "SE" "SE" "MI" ...
## $ employment_type   : chr   "FT" "FT" "FT" "FT" ...
## $ job_title         : chr   "Data Scientist" "Machine Learning Scientist" "Big Data Engineer" "Produ
## $ salary            : int  70000 260000 85000 20000 150000 72000 190000 11000000 135000 125000 ...
## $ salary_currency   : chr   "EUR" "USD" "GBP" "USD" ...
## $ salary_in_usd     : int  79833 260000 109024 20000 150000 72000 190000 35735 135000 125000 ...
## $ employee_residence: chr   "DE" "JP" "GB" "HN" ...
## $ remote_ratio      : int    0 0 50 0 50 100 100 50 100 50 ...
## $ company_location  : chr   "DE" "JP" "GB" "HN" ...
## $ company_size      : chr   "L" "S" "M" "S" ...
```

## Preprocessing and Exploratory Analysis

Since `salaries` is in different currencies we can drop it for in favor of `salary_in_usd`. Furthermore we can also drop variables like `X`, `company_location`, `employee_residence`, and `salary_currency`

```
cols = c("work_year", "experience_level", "employment_type", "job_title", "remote_ratio", "company_size", "salary_in_usd")
ds.salaries = subset(ds.salaries, select=cols)

# Treat the char variables as factors
ds.salaries$work_year = as.factor(ds.salaries$work_year)
ds.salaries$experience_level = as.factor(ds.salaries$experience_level)
ds.salaries$employment_type = as.factor(ds.salaries$employment_type)
ds.salaries$job_title = as.factor(ds.salaries$job_title)
ds.salaries$remote_ratio = as.factor(ds.salaries$remote_ratio)
ds.salaries$company_size = as.factor(ds.salaries$company_size)

# Treat employee_residence and company_location as continuous variables
ds.salaries$employee_residence = unclass(as.factor(ds.salaries$employee_residence))
ds.salaries$company_location = unclass(as.factor(ds.salaries$company_location))

# View the final data
str(ds.salaries)

## 'data.frame':    607 obs. of  9 variables:
## $ work_year        : Factor w/ 3 levels "2020","2021",...: 1 1 1 1 1 1 1 1 1 1 ...
```

```
## $ experience_level : Factor w/ 4 levels "EN","EX","MI",...: 3 4 4 3 4 1 4 3 3 4 ...
## $ employment_type : Factor w/ 4 levels "CT","FL","FT",...: 3 3 3 3 3 3 3 3 3 3 ...
## $ job_title       : Factor w/ 50 levels "3D Computer Vision Researcher",...: 23 41 8 48 38 13 35 2
## $ remote_ratio    : Factor w/ 3 levels "0","50","100": 1 1 2 1 2 3 3 2 3 2 ...
## $ company_size    : Factor w/ 3 levels "L","M","S": 1 3 2 3 1 1 3 1 1 3 ...
## $ employee_residence: int  15 33 21 24 56 56 56 26 56 42 ...
## .. attr(*, "levels")= chr [1:57] "AE" "AR" "AT" "AU" ...
## $ company_location : int  13 30 19 21 49 49 49 23 49 39 ...
## .. attr(*, "levels")= chr [1:50] "AE" "AS" "AT" "AU" ...
## $ salary_in_usd    : int  79833 260000 109024 20000 150000 72000 190000 35735 135000 125000 ...
```

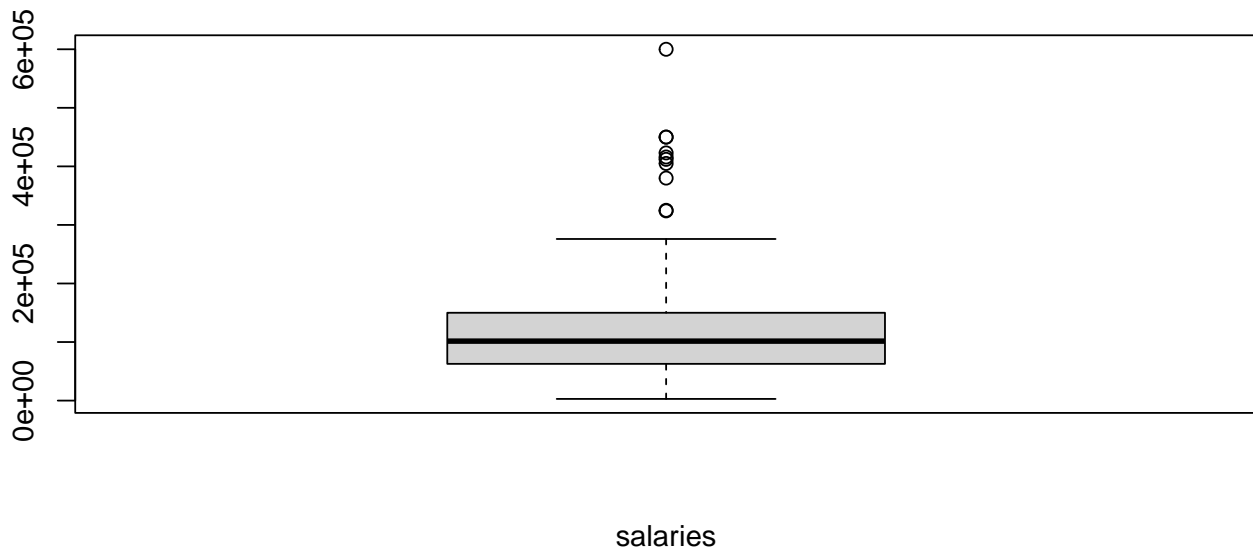
In this case we have factors `work_year`, `experience_level`, `employment_type`, `job_title`, `remote_ratio`, `work_year` and `company_size`.

`employee_residence` and `company_location` will be treated as continuous variables.

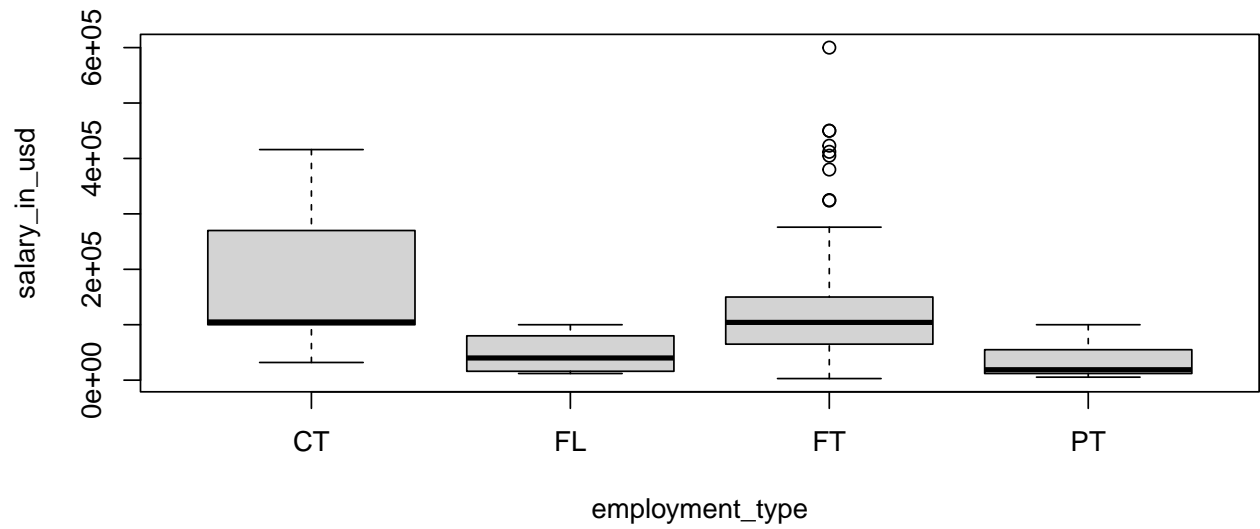
`job_title` has 50 levels

The response variable `salary_in_usd` is continuous.

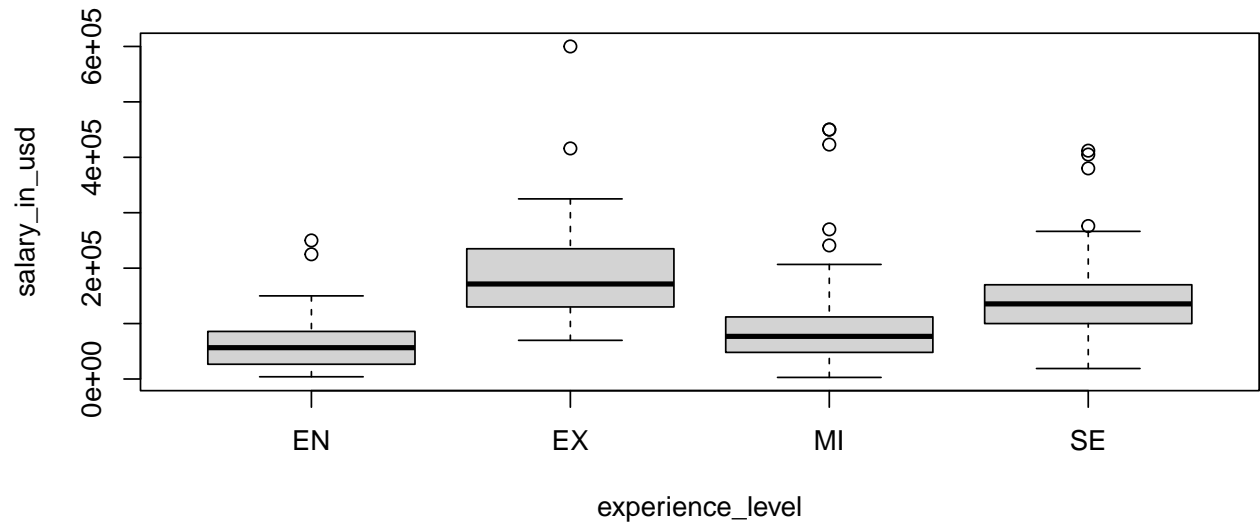
```
boxplot(ds.salaries$salary_in_usd, xlab="salaries")
```



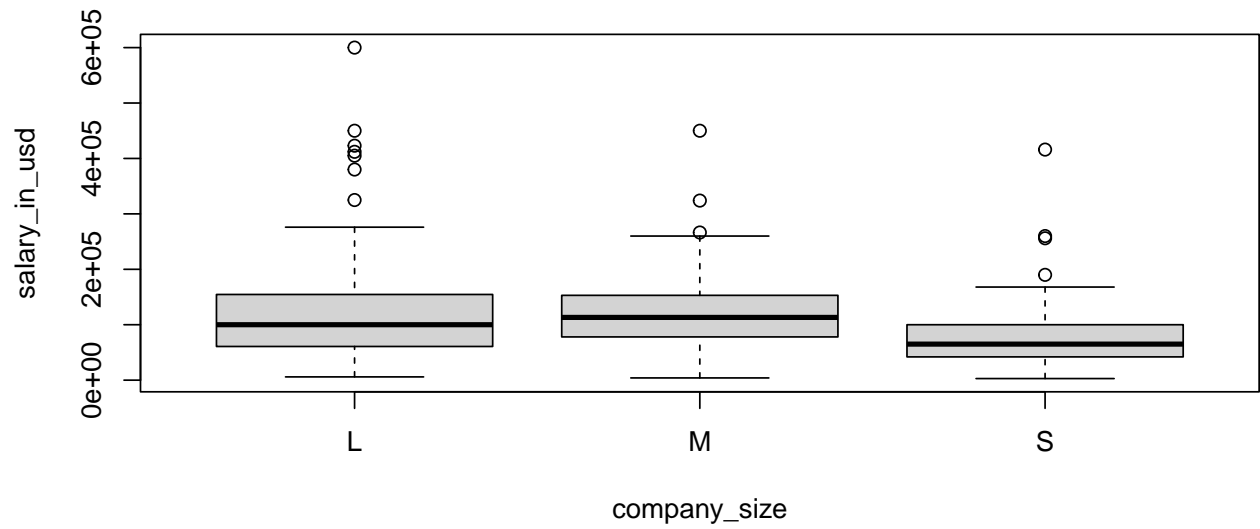
```
boxplot(salary_in_usd~employment_type,
        data=ds.salaries)
```



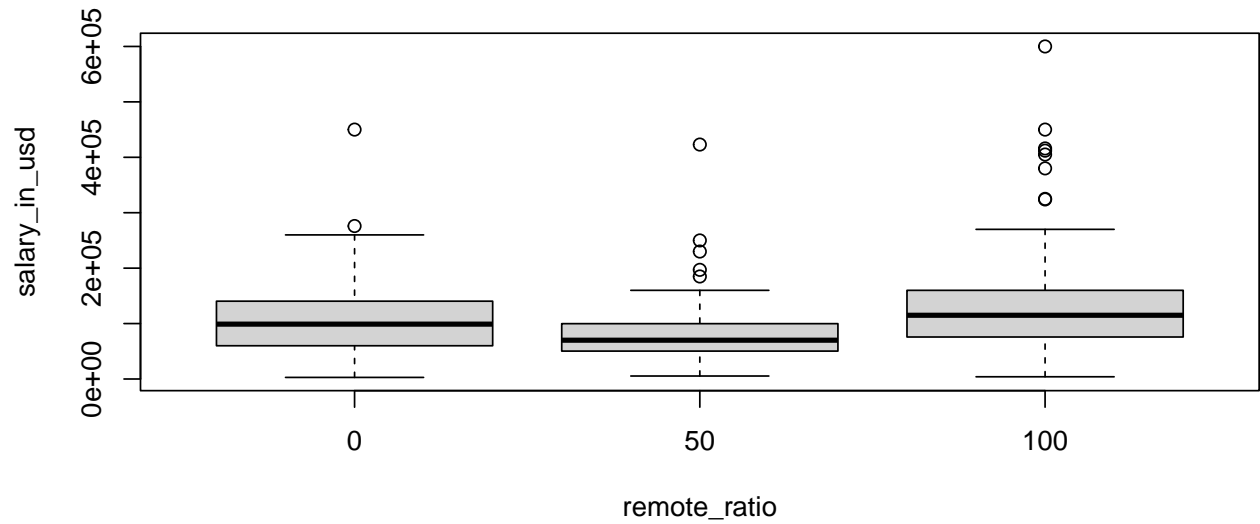
```
boxplot(salary_in_usd~experience_level,
        data=ds.salaries)
```



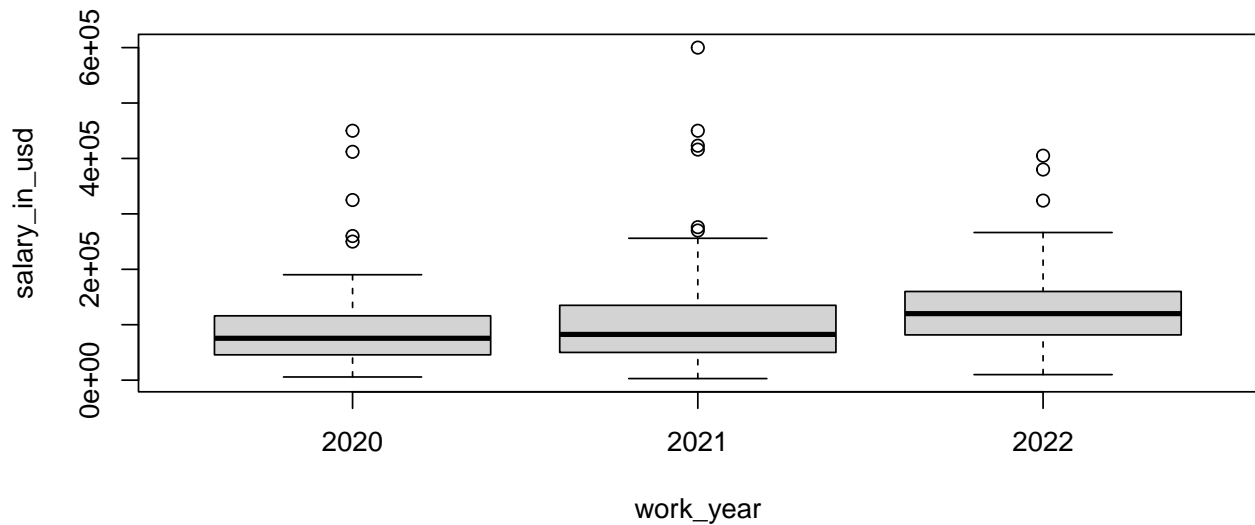
```
boxplot(salary_in_usd~company_size,
        data=ds.salaries)
```



```
boxplot(salary_in_usd~remote_ratio,  
        data=ds.salaries)
```



```
boxplot(salary_in_usd~work_year,  
        data=ds.salaries)
```



### Possible techniques

4-way anova? Tukey HSD to perform multiple comparison tests?

### Linear Regression for Model Selection

```
lm.salaries = lm(salary_in_usd~ work_year + experience_level + employment_type + job_title + employee_r
summary(lm.salaries)
```

```
##
## Call:
## lm(formula = salary_in_usd ~ work_year + experience_level + employment_type +
##     job_title + employee_residence + remote_ratio + company_location +
##     company_size, data = ds.salaries)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -163133  -26170   -3270    22055   289748
##
## Coefficients:
##
##              Estimate Std. Error t value
## (Intercept)      32151.9     62177.4    0.517
## work_year2021    -16632.1      7250.1   -2.294
## work_year2022     -4257.1      8074.4   -0.527
## experience_levelEX 116045.5    14781.2    7.851
## experience_levelMI  24148.2      7007.4    3.446
## experience_levelSE  53723.1      7345.0    7.314
## employment_typeFL -102687.0    39897.8   -2.574
## employment_typeFT -48827.3    28559.7   -1.710
## employment_typePT -53797.1    33626.8   -1.600
## job_titleAI Scientist    51734.3    56098.3    0.922
## job_titleAnalytics Engineer  42370.3    59833.2    0.708
## job_titleApplied Data Scientist 117663.8    58691.2    2.005
## job_titleApplied Machine Learning Scientist  86760.0    59787.4    1.451
## job_titleBI Data Analyst    26572.7    57814.4    0.460
## job_titleBig Data Architect   86030.2    73748.1    1.167
```

## job_titleBig Data Engineer	22786.6	56874.2	0.401
## job_titleBusiness Data Analyst	28778.8	58901.9	0.489
## job_titleCloud Data Engineer	88203.6	64985.2	1.357
## job_titleComputer Vision Engineer	25692.2	57136.8	0.450
## job_titleComputer Vision Software Engineer	104106.4	61913.2	1.681
## job_titleData Analyst	22742.1	54034.1	0.421
## job_titleData Analytics Engineer	10527.7	59507.8	0.177
## job_titleData Analytics Lead	297673.7	74177.4	4.013
## job_titleData Analytics Manager	29390.5	57112.6	0.515
## job_titleData Architect	97907.3	55940.1	1.750
## job_titleData Engineer	44313.3	53911.1	0.822
## job_titleData Engineering Manager	42154.5	58604.7	0.719
## job_titleData Science Consultant	51745.9	57609.3	0.898
## job_titleData Science Engineer	48934.6	61470.8	0.796
## job_titleData Science Manager	73156.2	55735.3	1.313
## job_titleData Scientist	52922.1	53933.7	0.981
## job_titleData Specialist	70048.7	73916.2	0.948
## job_titleDirector of Data Engineering	88145.9	64651.6	1.363
## job_titleDirector of Data Science	82599.7	58409.4	1.414
## job_titleETL Developer	28098.8	64660.1	0.435
## job_titleFinance Data Analyst	25545.9	73886.5	0.346
## job_titleFinancial Data Analyst	221491.0	64935.8	3.411
## job_titleHead of Data	65043.5	58678.0	1.108
## job_titleHead of Data Science	17441.2	60229.1	0.290
## job_titleHead of Machine Learning	-41690.9	75296.8	-0.554
## job_titleLead Data Analyst	23743.3	61367.5	0.387
## job_titleLead Data Engineer	74223.9	57863.8	1.283
## job_titleLead Data Scientist	66212.8	61455.0	1.077
## job_titleLead Machine Learning Engineer	41181.3	73999.7	0.557
## job_titleMachine Learning Developer	78995.3	61390.4	1.287
## job_titleMachine Learning Engineer	61926.8	54498.3	1.136
## job_titleMachine Learning Infrastructure Engineer	31960.2	61015.3	0.524
## job_titleMachine Learning Manager	80720.2	74353.7	1.086
## job_titleMachine Learning Scientist	103414.5	56785.8	1.821
## job_titleMarketing Data Analyst	41086.1	74117.6	0.554
## job_titleML Engineer	81362.8	57119.7	1.424
## job_titleNLP Engineer	-28553.8	74146.9	-0.385
## job_titlePrincipal Data Analyst	80577.3	64605.8	1.247
## job_titlePrincipal Data Engineer	214634.2	61432.8	3.494
## job_titlePrincipal Data Scientist	133372.5	57268.2	2.329
## job_titleProduct Data Analyst	-19522.6	65034.4	-0.300
## job_titleResearch Scientist	83879.9	55390.4	1.514
## job_titleStaff Data Scientist	-32699.8	78651.6	-0.416
## employee_residence	1028.7	285.1	3.608
## remote_ratio50	-12403.7	7722.7	-1.606
## remote_ratio100	-442.1	5444.4	-0.081
## company_location	354.5	319.8	1.109
## company_sizeM	-6078.9	5837.6	-1.041
## company_sizeS	-19540.3	7331.9	-2.665
##	Pr(> t )		
## (Intercept)	0.605297		
## work_year2021	0.022168 *		
## work_year2022	0.598246		
## experience_levelEX	2.22e-14 ***		

## experience_levelMI	0.000613	***
## experience_levelSE	9.34e-13	***
## employment_typeFL	0.010324	*
## employment_typeFT	0.087900	.
## employment_typePT	0.110219	
## job_titleAI Scientist	0.356830	
## job_titleAnalytics Engineer	0.479162	
## job_titleApplied Data Scientist	0.045481	*
## job_titleApplied Machine Learning Scientist	0.147318	
## job_titleBI Data Analyst	0.645973	
## job_titleBig Data Architect	0.243908	
## job_titleBig Data Engineer	0.688837	
## job_titleBusiness Data Analyst	0.625330	
## job_titleCloud Data Engineer	0.175254	
## job_titleComputer Vision Engineer	0.653134	
## job_titleComputer Vision Software Engineer	0.093243	.
## job_titleData Analyst	0.674007	
## job_titleData Analytics Engineer	0.859643	
## job_titleData Analytics Lead	6.84e-05	***
## job_titleData Analytics Manager	0.607038	
## job_titleData Architect	0.080645	.
## job_titleData Engineer	0.411456	
## job_titleData Engineering Manager	0.472264	
## job_titleData Science Consultant	0.369466	
## job_titleData Science Engineer	0.426344	
## job_titleData Science Manager	0.189884	
## job_titleData Scientist	0.326910	
## job_titleData Specialist	0.343716	
## job_titleDirector of Data Engineering	0.173322	
## job_titleDirector of Data Science	0.157891	
## job_titleETL Developer	0.664054	
## job_titleFinance Data Analyst	0.729668	
## job_titleFinancial Data Analyst	0.000696	***
## job_titleHead of Data	0.268145	
## job_titleHead of Data Science	0.772247	
## job_titleHead of Machine Learning	0.580020	
## job_titleLead Data Analyst	0.698979	
## job_titleLead Data Engineer	0.200132	
## job_titleLead Data Scientist	0.281772	
## job_titleLead Machine Learning Engineer	0.578094	
## job_titleMachine Learning Developer	0.198723	
## job_titleMachine Learning Engineer	0.256329	
## job_titleMachine Learning Infrastructure Engineer	0.600626	
## job_titleMachine Learning Manager	0.278127	
## job_titleMachine Learning Scientist	0.069137	.
## job_titleMarketing Data Analyst	0.579576	
## job_titleML Engineer	0.154898	
## job_titleNLP Engineer	0.700316	
## job_titlePrincipal Data Analyst	0.212857	
## job_titlePrincipal Data Engineer	0.000515	***
## job_titlePrincipal Data Scientist	0.020230	*
## job_titleProduct Data Analyst	0.764148	
## job_titleResearch Scientist	0.130521	
## job_titleStaff Data Scientist	0.677754	

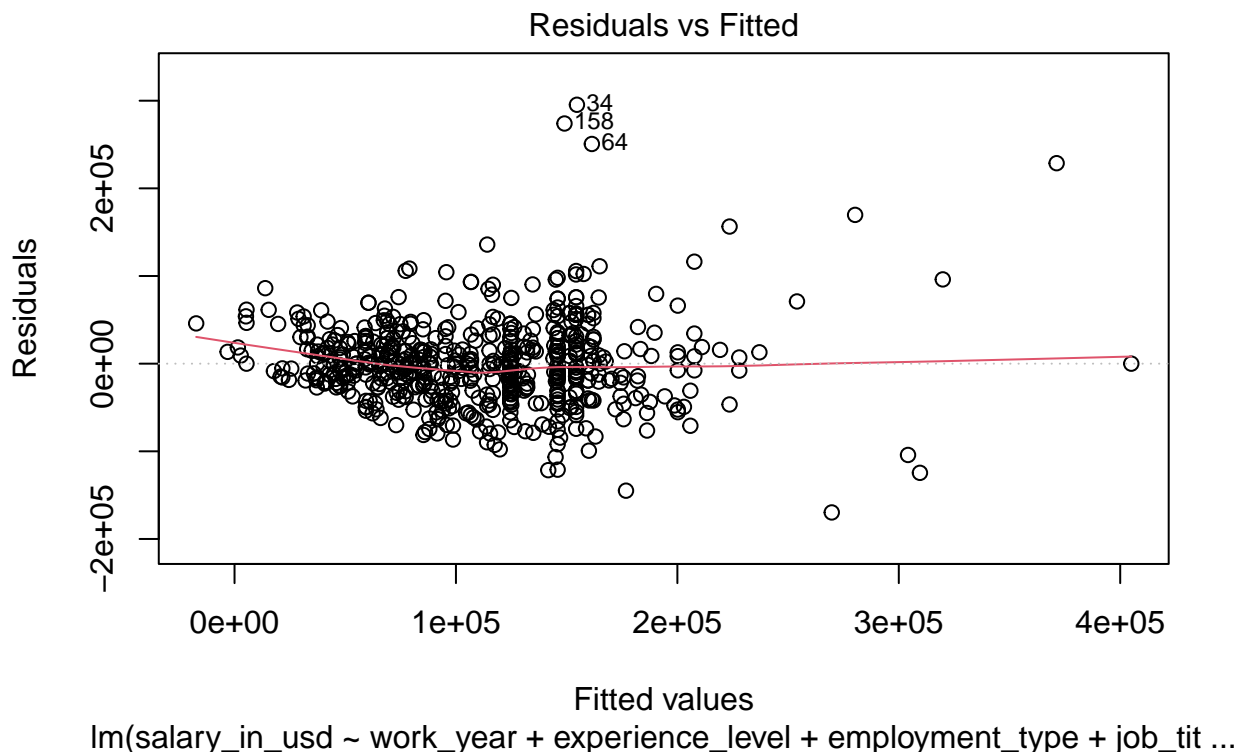
```
## employee_residence          0.000337 ***
## remote_ratio50              0.108825
## remote_ratio100            0.935306
## company_location            0.268062
## company_sizeM               0.298185
## company_sizeS               0.007926 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 50280 on 543 degrees of freedom
## Multiple R-squared:  0.5501, Adjusted R-squared:  0.4979
## F-statistic: 10.54 on 63 and 543 DF,  p-value: < 2.2e-16
```

We can see that the variables `work_year`, `experience_level`, `employment_type`, `job_title`, `company_size`, and `employee_residence` are significant. So we will keep them in the model

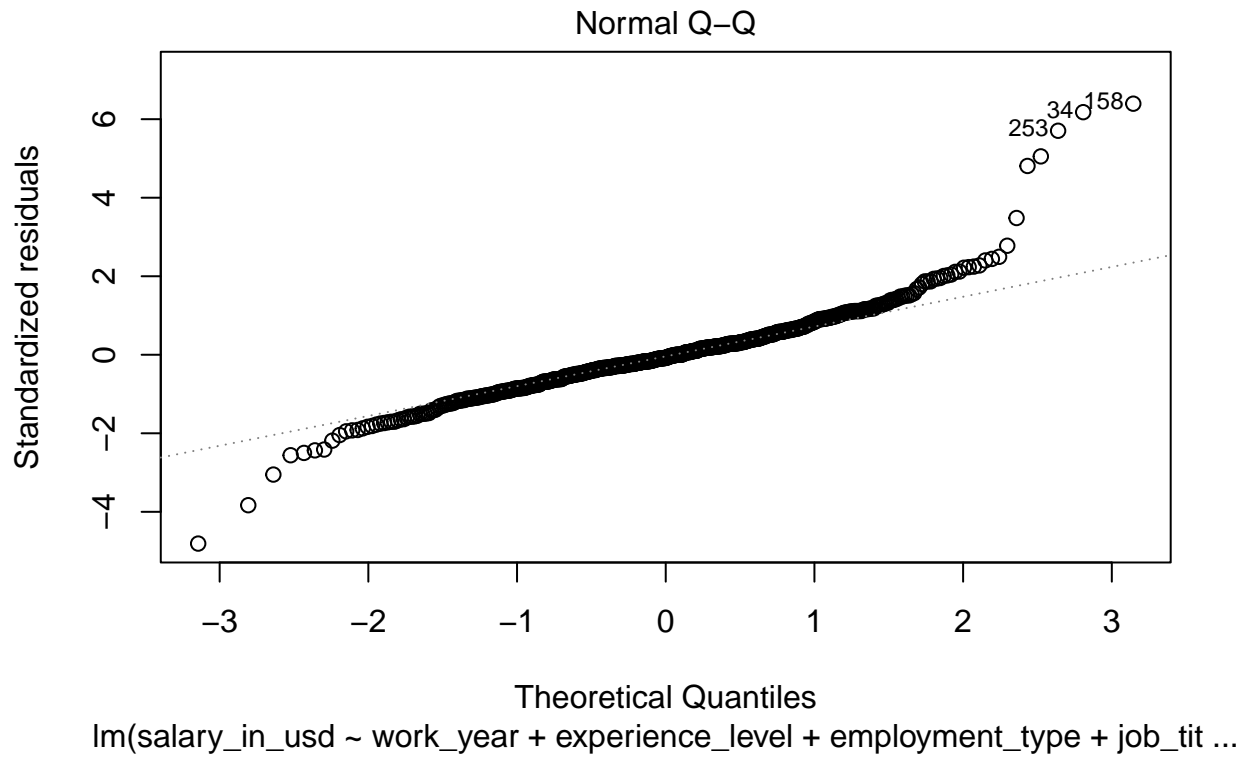
## ANOVA Assumptions

```
lm.salaries = lm(salary_in_usd~ work_year + experience_level + employment_type + job_title + employee_r
plot(lm.salaries,which=c(1,2))
```

```
## Warning: not plotting observations with leverage one:
## 91, 166, 256, 385, 456, 524
```







#### ANOVA 6-way

The following code takes a very long time to run. Is there a more efficient way to narrow down variables?

```
library(car)
```

```
## Loading required package: carData
```

```
# Anova(lm(salary_in_usd~work_year*experience_level*employment_type*job_title*employee_residence*compa
```