

# LAB 2

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## Setup

Start by installing the necessary packages. In Lab 1, we installed “data.table”. Today we install two packages: **ggplot2** and **stargazer**. Once installed, you activate the packages using “the function”library”:

```
library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.1.3
library(stargazer)

##
## Please cite as:
## Hlavac, Marek (2022). stargazer: Well-Formatted Regression and Summary Statistics Tables.
## R package version 5.2.3. https://CRAN.R-project.org/package=stargazer
library(data.table)

## Warning: package 'data.table' was built under R version 4.1.3

Then set your working directory.
setwd("C:/TopicsDig/Labs")
# change the file's path to your own
```

## Data Analysis

We will use the same data set from Lab 1 to compute descriptive statistics. To start, load Lab 1’s data set and convert it to the data.table format:

```
load("C:/TopicsDig/Labs/datasets/ceosal2.RData")
dt.ceo.salaries <- data.table(data)
rm(data)
```

## Descriptive Statistics

### How many CEOS are in the sample?

As there are only CEOS in this data set, we can count the number of CEOS by simply counting the number of observations in the data set.

```
nrow(dt.ceo.salaries)
```

```
## [1] 177
```

### How many CEOS have a graduate degree?

Whenever a CEO has a graduate degree the variable “grad” takes the value 1 (0 otherwise). Thus, to count how many CEOs have a graduate degree we can simply compute the sum of the variable “grad”.

```
dt.ceo.salaries[, sum(grad)]
```

```
## [1] 94
```

Alternatively, we can count the number of rows in which the variable grad takes the value of 1.

```
nrow(dt.ceo.salaries[grad==1,])
```

```
## [1] 94
```

**What is the percentage of CEOs with graduate degrees?**

```
dt.ceo.salaries[, sum(grad)]/nrow(dt.ceo.salaries)
```

```
## [1] 0.5310734
```

Or we can simply calculate the mean of grad:

```
dt.ceo.salaries[, mean(grad)]
```

```
## [1] 0.5310734
```

**What is the average CEO salary?**

```
dt.ceo.salaries[, mean(salary)]
```

```
## [1] 865.8644
```

Alternatively:

```
mean(dt.ceo.salaries[, salary])
```

```
## [1] 865.8644
```

**What is the mean CEO salary for those with a graduate degree?**

```
dt.ceo.salaries[grad==1, mean(salary)]
```

```
## [1] 864.2128
```

**What is the mean CEO salary for those without a graduate degree?**

```
dt.ceo.salaries[grad==0, mean(salary)]
```

```
## [1] 867.7349
```

**How many CEOs are have/don't have a college degree?**

```
dt.ceo.salaries[, list(n_ceo=.N), by = college]
```

```
##      college n_ceo
## 1:         1    172
## 2:         0      5
```

**Can we say that the mean salary is statistically different from 800?**

```
t.test(dt.ceo.salaries[, salary], mu = 800)
```

```
##
## One Sample t-test
##
## data:  dt.ceo.salaries[, salary]
## t = 1.4913, df = 176, p-value = 0.1377
```

```
## alternative hypothesis: true mean is not equal to 800
## 95 percent confidence interval:
##  778.7015 953.0274
## sample estimates:
## mean of x
##  865.8644
```

The p-value is greater than 0.05 so we cannot reject the null that the population mean is 800. You can also see that 800 is contained in the confidence interval.

**Is the average salary different for CEOs with a graduate degree and those without?**

```
t.test(dt.ceo.salaries[, salary] ~ dt.ceo.salaries[, grad])
```

```
##
##  Welch Two Sample t-test
##
## data:  dt.ceo.salaries[, salary] by dt.ceo.salaries[, grad]
## t = 0.038973, df = 149.94, p-value = 0.969
## alternative hypothesis: true difference in means between group 0 and group 1 is not equal to 0
## 95 percent confidence interval:
##  -175.0489  182.0932
## sample estimates:
## mean in group 0 mean in group 1
##      867.7349      864.2128
```

Alternatively:

```
dt.ceo.salaries[, t.test (salary ~ grad)]
```

```
##
##  Welch Two Sample t-test
##
## data:  salary by grad
## t = 0.038973, df = 149.94, p-value = 0.969
## alternative hypothesis: true difference in means between group 0 and group 1 is not equal to 0
## 95 percent confidence interval:
##  -175.0489  182.0932
## sample estimates:
## mean in group 0 mean in group 1
##      867.7349      864.2128
```

Alternatively:

```
t.test(dt.ceo.salaries[grad==0, salary] , dt.ceo.salaries[grad==1, salary])
```

```
##
##  Welch Two Sample t-test
##
## data:  dt.ceo.salaries[grad == 0, salary] and dt.ceo.salaries[grad == 1, salary]
## t = 0.038973, df = 149.94, p-value = 0.969
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  -175.0489  182.0932
## sample estimates:
## mean of x mean of y
##  867.7349  864.2128
```

The p-value is greater than 0.05 so we cannot reject the null that the two population means are the same. You can also see that the value 0 (no difference between means) is contained in the confidence interval.

### Creating a table with descriptive statistics:

```
dt.ceo.salaries[, list( mean_salary = mean(salary)
                        , sd_salary = sd(salary)
                        , min_salary = min(salary)
                        , max_salary = max(salary)
                        , median_salary = median(salary))]
```

```
##      mean_salary sd_salary min_salary max_salary median_salary
## 1:      865.8644  587.5893         100        5299          707
```

You can also compute the summary statistics for different groups:

```
dt.ceo.salaries[, list( mean_salary = mean(salary)
                        , sd_salary = sd(salary)
                        , min_salary = min(salary)
                        , max_salary = max(salary)), by = list(grad, college)]
```

```
##      grad college mean_salary sd_salary min_salary max_salary
## 1:      1         1    864.2128  501.3924         100        2265
## 2:      0         1    853.0897  679.0268         174        5299
## 3:      0         0   1096.2000  633.4569         300        1738
```

Alternatively, you can use the stargazer function to get summary statistics for all variables in your data set:

```
stargazer(dt.ceo.salaries, type = "text")
```

```
##
## =====
## Statistic  N      Mean      St. Dev.    Min      Max
## -----
## salary    177  865.864    587.589    100    5,299
## age       177   56.429     8.422     33      86
## college   177    0.972     0.166      0       1
## grad      177    0.531     0.500      0       1
## comten    177   22.503    12.295      2       58
## ceoten    177    7.955     7.151      0       37
## sales     177  3,529.463  6,088.654    29   51,300
## profits   177   207.831   404.454   -463    2,700
## mktval    177  3,600.316  6,442.276   387   45,400
## lsalary   177    6.583     0.606    4.605    8.575
## lsales    177    7.231     1.432    3.367   10.845
## lmktval   177    7.399     1.133    5.958   10.723
## comtensq   177   656.684   577.123      4    3,364
## ceotensq   177   114.124   212.566      0    1,369
## profmarg   177    6.420    17.861  -203.077  47.458
## -----
```

You can also get summary statistics for a subset of your observations and for a specific list of variables.

```
stargazer(dt.ceo.salaries[grad==1, list(age, salary)], type = "text")
```

```
##
## =====
## Statistic N      Mean      St. Dev. Min  Max
## -----
```

```
## age      94 55.457  8.155  38  86
## salary   94 864.213 501.392 100 2,265
## -----
```

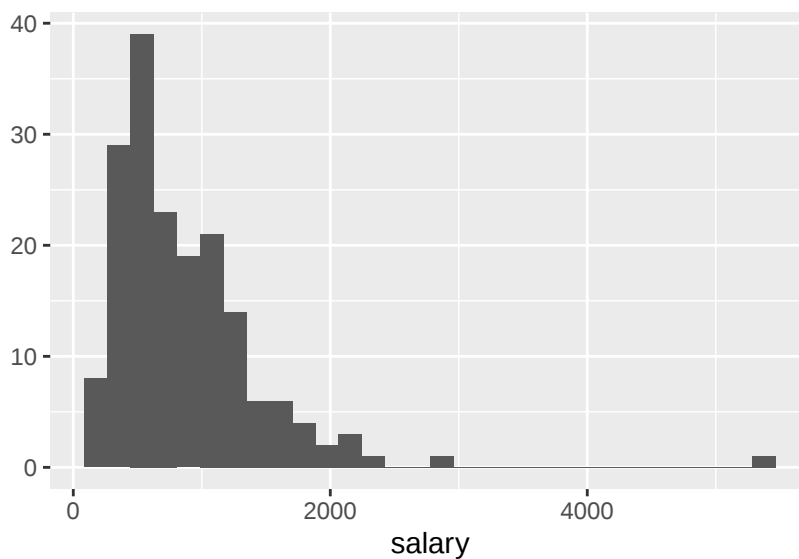
## Quick plots

### Histogram

#### Salary

```
qplot(  data = dt.ceo.salaries
, x = salary
, geom = "histogram")
```

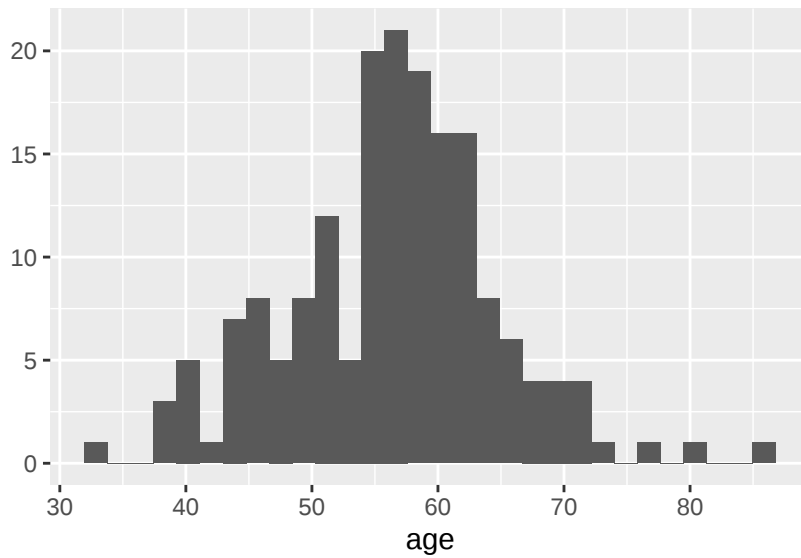
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



#### Age

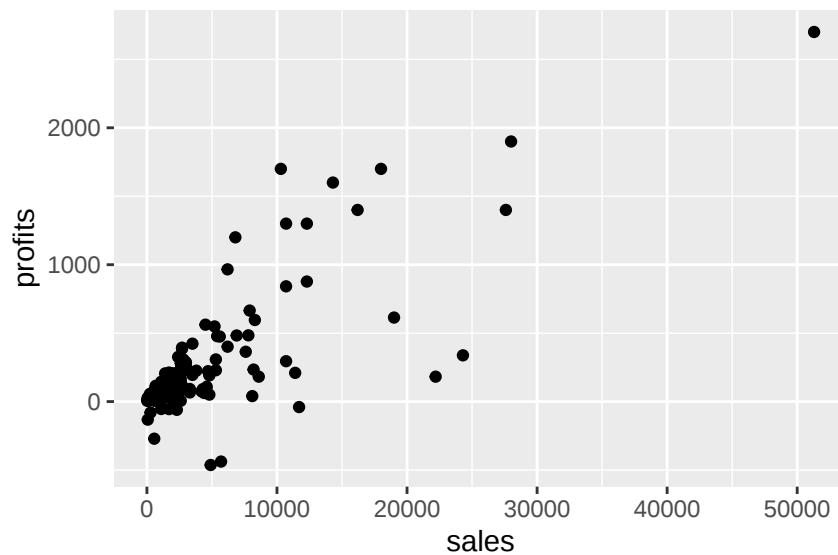
```
qplot(  data = dt.ceo.salaries
, x = age
, geom = "histogram")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



### Scatterplot

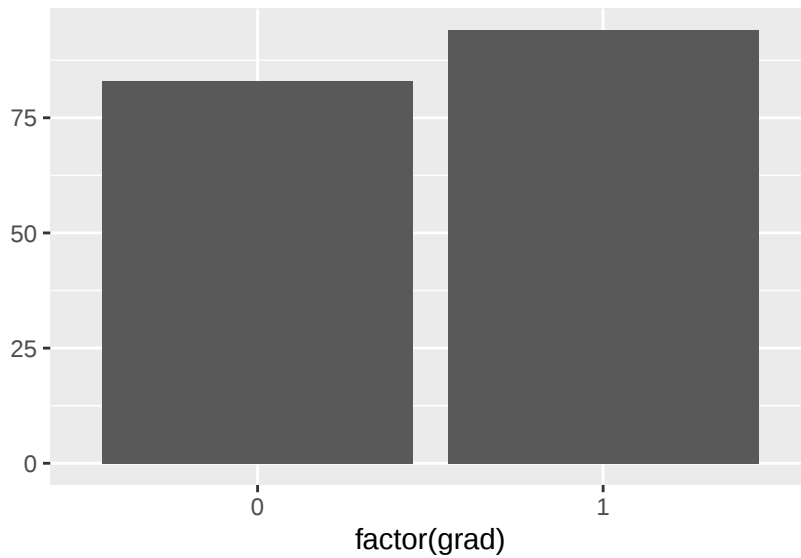
```
qplot(  data = dt.ceo.salaries
, x = sales
, y = profits
, geom = "point")
```



### Barplot

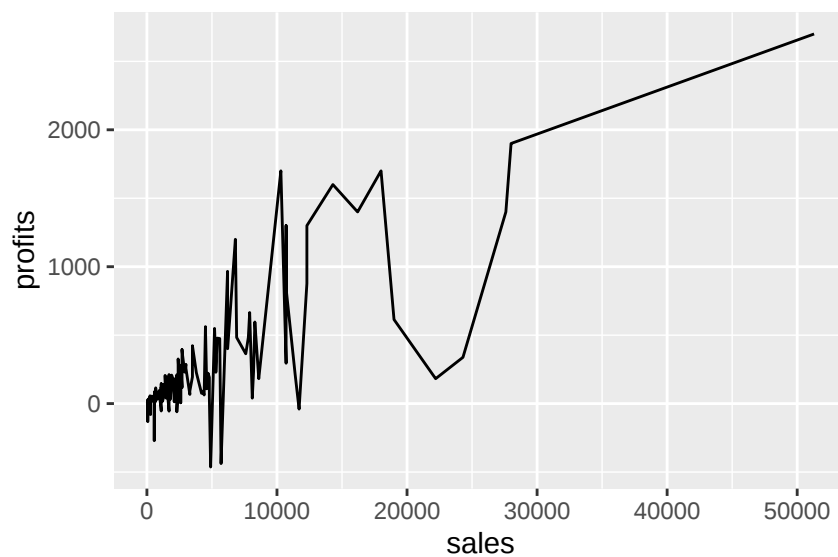
#### Graduate

```
qplot(  data = dt.ceo.salaries
, x = factor(grad)
, geom = "bar")
```



### Line

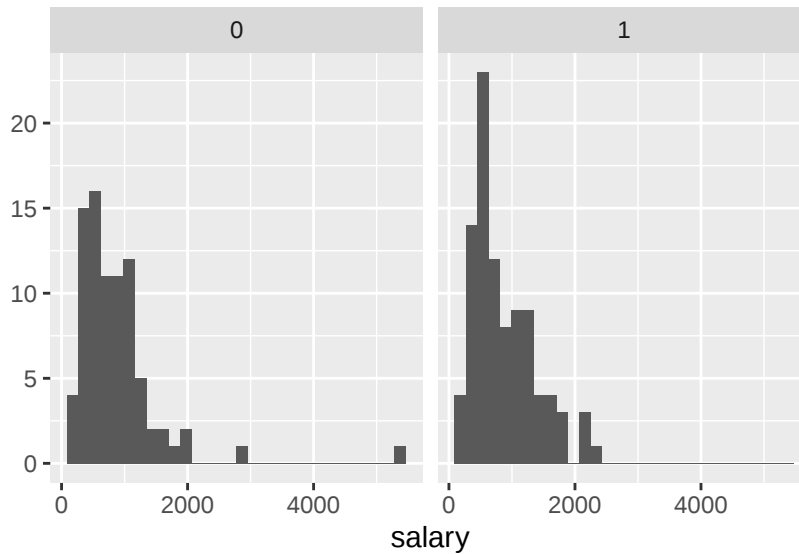
```
qplot(  data = dt.ceo.salaries
, x = sales
, y = profits
, geom = "line")
```



### Facet Wrap

```
qplot(  data = dt.ceo.salaries
, x = salary
, geom = "histogram") + facet_wrap(~ grad)
```

## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.



Customizing plots:

Salary

```
qplot(  data = dt.ceo.salaries
        , x = salary
        , geom = "histogram"
        , fill = factor(grad, levels = c(0,1), labels = c("Yes", "No"))) +
  theme_bw() +
  ylim(0,50) +
  xlim(0, 4000) +
  labs( title = "MY PLOT", x = "CEO Salary", y = "Number of CEOs", fill = "Grad. Degree")
```

## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.

## Warning: Removed 1 rows containing non-finite values (stat\_bin).

## Warning: Removed 4 rows containing missing values (geom\_bar).

MY PLOT



## **Acknowledgements and Thanks:**

This lab is based on material by M Godinho de Matos, R Belo and F Reis. Gratefully acknowledged!