

# Assignment 1

## Assignment 1

My first assignment has three parts:

**(a) a brief summary about “Veri Bilimi ve Endüstri Mühendisliği Üzerine Sohbetler - Baykal Hafizoğlu & Erdi Daşdemir”:**

Baykal Hafizoğlu is the guest of our lecturer Erdi Daşdemir’s talk about data science and industrial engineering. In the beginning of the talk Hafizoğlu defines himself as an OR Scientist and Optimization Scientist. He indicates that his job is to write and solve OR models (mathematical models).

Hafizoğlu starts with OR and indicates that the classic definition of OR has changed a lot in the past few years. It was about optimization and statistics. However, it’s more about artificial intelligence and machine learning nowadays. That’s why Hafizoğlu prefers to call OR “analytics”.

It is possible to divide analytics into four:

- 1-Descriptive analytics: Deals with defining the problem.
- 2-Diagnostic analytics: Deals with what the problem is and what the reasons are.
- 3-Predictive analytics: Deals with the future.
- 4-Prescriptive analytics: Deals with suggesting a solution.

He highlights that what students learn in undergrad is very precious and they should be aware of it. In the end of undergrad, students gain a good knowledge about analytics.

Later in his speech, he talks about the lessons he learned from his own experiences. Firstly, he says that “All projects start with a clear problem definition.”. It means that even about the simplest thing he needs to deal with, the problem definition must be clear and concise. A good, compatible KPI (Key Performance Indicator) needs to be chosen in order to present the problem, interpret and explain it to the user. After that, a success criteria needs to be defined in order to understand whether we have achieved success or not.

Another lesson that he learned from his own experiences is the importance of deployment. He thinks delivering the model to the user is always better than keeping it in your computer. In this case, “How to deploy” and “where to deploy” are important questions.

Hafizoğlu highlights the importance of the user interface and early prototype. User satisfaction is the most important thing that’s why the user needs to be understood well. Without keeping the user waiting, early prototype must be delivered for an early feedback. The user should not struggle to understand the user interface. If the user has to put effort, it shows that the user interface is not good enough.

It's very crucial for users to understand and own the model they will use. Because they wouldn't want to use models they don't understand. Therefore, we should be able to explain the analytical model we have established.

In the end, we need to show the solution's effect mathematically and refer to the problem and KPIs. Conflicting KPIs need to be considered because there can be a trade-off between KPIs. When one thing gets better, another thing can get worse.

## **(b) exploring statistical summaries with custom functions and loops:**

```
data(mtcars)#mtcars dataset

custom_summary<-function(cars){
  mean_cars<-mean(cars)
  median_cars<-median(cars)
  standard_deviation_cars<-sd(cars)
  minimum_cars<-min(cars)
  maximum_cars<-max(cars)

  result<-c("mean"=mean_cars,
            "median"=median_cars,
            "standard deviation"=standard_deviation_cars,
            "minimum"=minimum_cars,
            "maximum"=maximum_cars)

  return(result)
}

cars<-c(1, 2, 3)#numeric vector

#writing a custom summary function
print("writing a custom summary function")
```

```
[1] "writing a custom summary function"
```

```
print("mpg")
```

```
[1] "mpg"
```

```
custom_summary(mtcars$mpg[cars])
```

mean	median	standard deviation	minimum
21.60000	21.00000	1.03923	21.00000
maximum			
22.80000			

```
print("cyl")
```

```
[1] "cyl"
```

```
custom_summary(mtcars$cyl[cars])
```

	mean	median	standard deviation	minimum
	5.333333	6.000000	1.154701	4.000000
maximum	6.000000			

```
print("disp")
```

```
[1] "disp"
```

```
custom_summary(mtcars$disp[cars])
```

	mean	median	standard deviation	minimum
	142.66667	160.00000	30.02221	108.00000
maximum	160.00000			

```
print("hp")
```

```
[1] "hp"
```

```
custom_summary(mtcars$hp[cars])
```

	mean	median	standard deviation	minimum
	104.333333	110.000000	9.814955	93.000000
maximum	110.000000			

```
print("drat")
```

```
[1] "drat"
```

```
custom_summary(mtcars$drat[cars])
```

	mean	median	standard deviation	minimum
	3.88333333	3.90000000	0.02886751	3.85000000
maximum	3.90000000			

```
print("wt")
```

```
[1] "wt"
```

```
custom_summary(mtcars$wt[cars])
```

	mean	median	standard deviation	minimum
	2.6050000	2.6200000	0.2778039	2.3200000
maximum	2.8750000			

```
print("qsec")
```

```
[1] "qsec"
```

```
custom_summary(mtcars$qsec[cars])
```

	mean	median	standard deviation	minimum
	17.363333	17.020000	1.115362	16.460000
maximum	18.610000			

```
print("vs")
```

```
[1] "vs"
```

```
custom_summary(mtcars$vs[cars])
```

	mean	median	standard deviation	minimum
	0.3333333	0.0000000	0.5773503	0.0000000
maximum	1.0000000			

```
print("am")
```

```
[1] "am"
```

```
custom_summary(mtcars$am[cars])
```

	mean	median	standard deviation	minimum
	1	1	0	1
maximum	1			

```
print("gear")
```

```
[1] "gear"
```

```
custom_summary(mtcars$gear[cars])
```

	mean	median	standard deviation	minimum
	4	4	0	4
maximum	4			

```
print("carb")
```

```
[1] "carb"
```

```
custom_summary(mtcars$carb[cars])
```

	mean	median	standard deviation	minimum
	3.000000	4.000000	1.732051	1.000000
maximum	4.000000			

```
#applying the function using a loop
print("applying the function using a loop")
```

```
[1] "applying the function using a loop"
```

```
for(column_name in colnames(mtcars)){
  column_data<-mtcars[[column_name]][cars]
  print(column_name)
  print(custom_summary(column_data))
}
```

```
[1] "mpg"
```

	mean	median	standard deviation	minimum
	21.60000	21.00000	1.03923	21.00000
maximum	22.80000			

```
[1] "cyl"
```

	mean	median	standard deviation	minimum
	5.333333	6.000000	1.154701	4.000000
maximum	6.000000			

```
[1] "disp"
```

	mean	median	standard deviation	minimum
	142.66667	160.00000	30.02221	108.00000
maximum	160.00000			

```
[1] "hp"
```

	mean	median	standard deviation	minimum
	104.333333	110.000000	9.814955	93.000000

```

        maximum
110.000000
[1] "drat"
        mean          median standard deviation          minimum
3.883333333          3.900000000          0.02886751          3.850000000
        maximum
3.900000000
[1] "wt"
        mean          median standard deviation          minimum
2.6050000          2.6200000          0.2778039          2.3200000
        maximum
2.8750000
[1] "qsec"
        mean          median standard deviation          minimum
17.363333          17.020000          1.115362          16.460000
        maximum
18.610000
[1] "vs"
        mean          median standard deviation          minimum
0.3333333          0.0000000          0.5773503          0.0000000
        maximum
1.0000000
[1] "am"
        mean          median standard deviation          minimum
1              1              0              1
        maximum
1
[1] "gear"
        mean          median standard deviation          minimum
4              4              0              4
        maximum
4
[1] "carb"
        mean          median standard deviation          minimum
3.000000          4.000000          1.732051          1.000000
        maximum
4.000000

```

```

#an alternative approach with apply
print("an alternative approach with apply")

```

```

[1] "an alternative approach with apply"

```

```

apply(mtcars[cars, ], 2, custom_summary)

```

	mpg	cyl	disp	hp	drat	wt
mean	21.60000	5.333333	142.66667	104.333333	3.88333333	2.6050000
median	21.00000	6.000000	160.00000	110.000000	3.90000000	2.6200000
standard deviation	1.03923	1.154701	30.02221	9.814955	0.02886751	0.2778039
minimum	21.00000	4.000000	108.00000	93.000000	3.85000000	2.3200000
maximum	22.80000	6.000000	160.00000	110.000000	3.90000000	2.8750000

	qsec	vs	am	gear	carb
mean	17.363333	0.3333333	1	4	3.000000
median	17.020000	0.0000000	1	4	4.000000
standard deviation	1.115362	0.5773503	0	0	1.732051
minimum	16.460000	0.0000000	1	4	1.000000
maximum	18.610000	1.0000000	1	4	4.000000

### (c) counting NA values and substituting with the number 660:

```
#install.packages("dslabs")
library(dslabs)
data(na_example)
```

```
#total count of NA values
sum(is.na(na_example))
```

[1] 145

```
#substituting the NA values with the number 660 and saving it as a new dataframe
no_nas<-ifelse(is.na(na_example), 660, na_example)
```

```
#total count of NA values in the new dataframe
sum(is.na(no_nas))
```

[1] 0

```
count<-0
#total count of the number 660 in the new dataframe
for(i in 1:1000)
  if(no_nas[i]==660)
    count<-count+1
count
```

[1] 145

Dataset with NA values

Dataset with the number 660

[1]	2	1	3	2	1	3	1	4	3	2	2	NA	2	2	1	4	NA	1	1	2	1	2	2	1	2	5	NA	2	2	3	1	
[32]	2	4	1	1	1	4	5	2	3	4	1	2	4	1	1	2	1	5	NA	NA	NA	1	1	5	1	3	1	NA	4	4	7	
[63]	3	2	NA	NA	1	NA	4	1	2	2	3	2	1	2	2	4	3	4	2	3	1	3	2	1	1	1	3	1	NA	3	1	
[94]	2	2	1	2	2	1	1	4	1	1	2	3	3	2	2	3	3	3	4	1	1	1	2	NA	4	3	4	3	1	2	1	
[125]	NA	NA	NA	NA	1	5	1	2	1	3	5	3	2	2	NA	NA	NA	NA	3	5	3	1	1	4	2	4	3	3	NA	2	3	
[156]	2	6	NA	1	1	2	2	1	3	1	1	5	NA	NA	2	4	NA	2	5	1	4	3	3	NA	4	3	1	4	1	1	3	
[187]	1	1	NA	NA	3	5	2	2	2	3	1	2	2	3	2	1	NA	2	NA	1	NA	NA	2	1	1	NA	3	NA	1	2	2	
[218]	1	3	2	2	1	1	2	3	1	1	1	4	3	4	2	2	1	4	1	NA	5	1	4	NA	3	NA	NA	1	1	5	2	
[249]	3	3	2	4	NA	3	2	5	NA	2	3	4	6	2	2	2	NA	2	NA	2	NA	3	3	2	2	4	3	1	4	2	NA	
[280]	2	4	NA	6	2	3	1	NA	2	2	NA	1	1	3	2	3	3	1	NA	1	4	2	1	1	3	2	1	2	3	1	NA	
[311]	2	3	3	2	1	2	3	5	5	1	2	3	3	1	NA	NA	1	2	4	NA	2	1	1	1	3	2	1	1	3	4	NA	
[342]	1	2	1	1	3	3	NA	1	1	3	5	3	2	3	4	1	4	3	1	NA	2	1	2	2	1	2	2	6	1	2	4	
[373]	5	NA	3	4	2	1	1	4	2	1	1	1	1	2	1	4	4	1	3	NA	3	3	NA	2	NA	1	2	1	1	4	2	
[404]	1	4	4	NA	1	2	NA	3	2	2	2	1	4	3	6	1	2	3	1	3	2	2	2	1	1	3	2	1	1	1	3	
[435]	2	2	NA	4	4	4	1	1	NA	4	3	NA	1	3	1	3	2	4	2	2	2	3	2	1	4	3	NA	1	4	3	1	
[466]	3	2	NA	3	NA	1	3	1	4	1	1	1	2	4	3	1	2	2	2	3	2	3	1	1	NA	3	2	1	1	2	NA	
[497]	2	2	2	3	3	1	1	2	NA	1	2	1	1	3	3	1	3	1	1	1	1	1	1	2	5	1	1	2	2	1	1	NA
[528]	1	4	1	2	4	1	3	2	NA	1	1	NA	2	1	1	4	2	3	3	1	5	3	1	1	2	NA	1	1	3	1	3	
[559]	2	4	NA	2	3	2	1	2	1	1	1	2	2	3	1	5	2	NA	2	NA	3	2	2	2	1	5	3	2	3	1	NA	
[590]	3	1	2	2	2	1	2	2	4	NA	6	1	2	NA	1	1	2	2	3	NA	3	2	3	3	4	2	NA	2	NA	4	NA	
[621]	1	1	2	2	3	1	1	1	3	NA	2	5	NA	7	1	NA	4	3	3	1	NA	1	1	1	1	3	2	4	2	2	3	
[652]	NA	NA	1	4	3	2	2	2	3	2	4	2	2	4	NA	NA	NA	6	3	3	1	4	4	2	1	NA	1	6	NA	3	3	
[683]	2	1	1	6	NA	1	5	1	NA	2	6	2	NA	4	1	3	1	2	NA	1	1	3	1	2	4	2	1	3	2	4	3	
[714]	2	2	1	1	5	6	4	2	2	2	2	4	NA	1	2	2	2	2	4	5	NA	NA	NA	4	3	3	3	2	4	2	4	
[745]	NA	NA	NA	NA	2	1	NA	2	4	3	2	NA	2	3	1	3	4	NA	1	2	1	2	NA	3	1	2	1	2	1	2	1	
[776]	2	2	2	2	1	1	3	3	1	3	4	3	NA	NA	4	2	3	2	1	3	2	4	2	2	3	1	2	4	3	3	4	
[807]	NA	1	4	2	1	1	1	3	1	5	2	2	4	2	NA	1	3	1	2	NA	1	2	1	2	1	NA	1	3	2	3	2	
[838]	NA	2	1	4	2	NA	NA	NA	2	4	2	NA	NA	3	1	NA	5	5	2	2	2	NA	2	1	3	1	3	2	4	2	4	
[869]	NA	4	1	2	3	2	3	3	2	3	2	2	2	1	3	2	4	2	NA	3	3	2	2	NA	NA	3	2	1	2	4	1	
[900]	1	1	1	4	3	2	3	3	2	NA	1	NA	3	2	1	1	1	2	NA	2	2	3	3	2	NA	NA	4	5	2	2	2	
[931]	1	2	3	1	3	3	4	3	NA	1	1	1	NA	4	3	5	1	1	2	NA	2	2	2	2	5	2	2	3	1	2	3	
[962]	NA	1	2	NA	NA	2	NA	3	1	1	2	5	3	5	1	1	4	NA	2	1	3	1	1	2	4	3	3	3	NA	1	1	
[993]	2	2	1	1	2	2	NA	2																								

[1]	2	1	3	2	1	3	1	4	3	2	2	660	2	2	1	4	660	1	1	2	1	2	2	1								
[25]	2	5	660	2	2	3	1	2	4	1	1	1	4	5	2	3	4	1	2	4	1	1	2	1								
[49]	5	660	660	660	1	1	5	1	3	1	660	4	4	7	3	2	660	660	1	660	4	1	2	2								
[73]	3	2	1	2	2	4	3	4	2	3	1	3	2	1	1	1	3	1	660	3	1	2	2	1								
[97]	2	2	1	1	4	1	1	2	3	3	2	2	3	3	4	1	1	1	2	660	4	3	4									
[121]	3	1	2	1	660	660	660	660	1	5	1	2	1	3	5	3	2	2	660	660	660	660	3	5								
[145]	3	1	1	4	2	4	3	3	660	2	3	2	6	660	1	1	2	2	1	3	1	1	5	660								
[169]	660	2	4	660	2	5	1	4	3	3	660	4	3	1	4	1	1	3	1	1	660	660	3	5								
[193]	2	2	2	3	1	2	2	3	2	1	660	2	660	2	660	1	660	660	2	1	1	660	3	660	1	2						
[217]	2	1	3	2	2	1	1	2	3	1	1	1	4	3	4	2	2	1	4	1	660	5	1	4								
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[289]	2	660	1	1	3	2	3	3	1	660	1	4	2	1	1	3	2	1	2	3	1	660	2	3								
[313]	3	2	1	2	3	5	5	1	2	3	3	1	660	660	1	2	4	660	2	1	1	1	3	2								
[337]	1	1	3	4	660	1	2	1	1	3	3	660	1	1	3	5	3	2	3	4	1	4	3	1								
[361]	660	2	1	2	2	1	2	2	6	1	2	4	5	660	3	4	2	1	1	4	2	1	1	1								
[385]	1	2	1	4	4	1	3	660	3	3	660	2	660	1	2	1	1	4	2	1	4	4	660	1								
[409]	2	660	3	2	2	2	1	4	3	6	1	2	3	1	3	2	2	2	1	1	3	2	1	1								
[433]	1	3	2	2	660	4	4	4	1	1	660	4	3	660	1	3	1	3	2	4	2	2	2	3								
[457]	2	1	4	3	660	1	4	3	1	3	2	660	3	660	1	3	1	4	1	1	1	2	4	3								
[481]	1	2	2	2	3	2	3	1	1	660	3	2	1	1	2	660	2	2	2	3	3	1	1	2								
[505]	660	1	2	1	1	3	3	1	3	1	1	1	1	1	2	5	1	1	2	2	1	1	660	1								
[529]	4	1	2	4	1	3	2	660	1	1	660	2	1	1	4	2	3	3	1	5	3	1	1	2								
[553]	660	1	1	3	1	3	2	4	660	2	3	2	1	2	1	1	1	2	2	3	1	5	2	660								
[577]	2	660	3	2	2	2	1	5	3	2	3	1	660	3	1	2	2	2	1	2	2	4	660	6								
[601]	1	2	660	1	1	2	2	3	660	3	2	3	3	4	2	660	2	660	4	660	1	1	2	2								
[625]	3	1	1	1	3	660	2	5	660	7	1	660	4	3	3	1	660	1	1	1	1	3	2	4								
[649]	2	2	3	660	660	1	4	3	2	2	2	3	2	4	2	2	4	660	660	660	6	3	3	1								
[673]	4	4	2	1	660	1	6	660	3	3	2	1	1	6	660	1	5	1	660	2	6	2	660	4								
[697]	1	3	1	2	660	1	1	3	1	2	4	2	1	3	2	4	3	2														



[793]	2	1	3	2	4	2	2	3	1	2	4	3	3	4	660	1	4	2	1	1	1	3	1	5
[817]	2	2	4	2	660	1	3	1	2	660	1	2	1	2	1	660	1	3	2	3	2	660	2	1
[841]	4	2	660	660	660	2	4	2	660	660	3	1	660	5	5	2	2	2	660	2	1	3	1	3
[865]	2	4	2	4	660	4	1	2	3	2	3	3	2	3	2	2	2	1	3	2	4	2	660	3
[889]	3	2	2	660	660	3	2	1	2	4	1	1	1	1	4	3	2	660	3	2	660	1	660	3
[913]	2	1	1	1	2	660	2	2	3	3	2	660	660	4	5	2	2	2	1	2	3	1	3	3
[937]	4	3	660	1	1	1	660	4	3	5	1	1	2	660	2	2	2	2	5	2	2	3	1	2
[961]	3	660	1	2	660	660	2	660	3	1	1	2	5	3	5	1	1	4	660	2	1	3	1	1
[985]	2	4	3	3	3	660	1	1	2	2	1	1	2	2	660	2								