# Project 1

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ALY6000: Introduction to Analytics

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#### Introduction

This project acts as an introductory session to the R language and at the same time gives an idea as to how statistics can be seamlessly performed through the language. Vectors and Data Frames are the data types that have been stressed throughout the project.

Various handy, in-built functions such as "c", "seq"," rep", statistical functions like sum, mean, median, max, min are used. Accessing data from a csv file using "data\_csv()" function and then using functions like head(),name(), select(),arrange(),filter(),slice(),mutate() are implemented so as to have a clearer understanding of what the data frame is, which is almost always the very first step of EDA.

Function like the ggplot() is used to visualize a part of data so as to convey the story behind the data graphically.

### **Key Findings**

(1) Along with the use of sophisticated statistical functions, R can also be used to compute simple calculations as shown below.

```
R 4.2.2 D;/NEU/CPS-MPS Analytics/ALY6000/Assignment/Gokhale-Project1/ > 123*453
[1] 55719
> 5^2 * 40
[1] 1000
> TRUE & FALSE
[1] FALSE
> TRUE | FALSE
[1] TRUE
> 75 %% 10
[1] 5
> 75/10
[1] 7.5
```

(2) The code snippet below shows as to how the Broadcasting works in R.
The first line adds 20 to all the elements of the vector 'second\_vector' while the next line multiplies all the elements of the 'second\_vector' by 20.

The next two lines compare all the elements of the vector 'second\_vector' with 20 and accordingly return a Boolean vector with TRUE corresponding to 1 and FALSE corresponding to 0.

These lines in no way mutate or change the original vector 'second\_vector'.

```
Console Terminal ×
                Background Jobs
R 4.2.2 D:/NEU/CPS-MPS Analytics/ALY6000/Assignment/Gokhale-Project1/
> # 20 has been added to all the elements of the vector
> second vector+20
[1] 30 32 34 36 38 40 42 44 46 48 50
> # All the elements in the vector get multiplied by 20
> second vector*20
[1] 200 240 280 320 360 400 440 480 520 560 600
> # Returns the elements in the given vector that are greater than 20 as TRUE and remain
ing once as FALSE.
[1] FALSE FALSE FALSE FALSE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
> # Returns the elements in the given vector that are not equal to 20 as TRUE and the re
maining as FALSE.
> second_vector != 20
 [1] TRUE TRUE TRUE TRUE TRUE FALSE TRUE TRUE TRUE TRUE TRUE
```

(3) seq() function is used to generate a sequence of elements for a vector. The parameters like 'from' and 'to' act as starting and ending points for the sequence generation while 'by' can be used to provide an equal increment/decrement.

```
> reverse_numbers <- seq(from = 100, to = -100, by = -3)

> reverse_numbers

[1] 100 97 94 91 88 85 82 79 76 73 70 67 64 61 58 55 52 49 46 43

[21] 40 37 34 31 28 25 22 19 16 13 10 7 4 1 -2 -5 -8 -11 -14 -17

[41] -20 -23 -26 -29 -32 -35 -38 -41 -44 -47 -50 -53 -56 -59 -62 -65 -68 -71 -74 -77

[61] -80 -83 -86 -89 -92 -95 -98
```

(4) A Boolean vector is created using the combine function and 'first\_vector' as it's parameter.

The corresponding 'TRUE' values are returned.

```
> # The second and the Third element from the first_vector were returned.
> vector_from_boolean_brackets <- first_vector[c(FALSE,TRUE,FALSE, TRUE)]
> vector_from_boolean_brackets
[1] 12 5
```

(5) "second\_vector >= 20"returns a Boolean vector which states 'TRUE' for the values that are above 20 and 'FALSE' for the values less than 20.

```
> # Returns the elements from the second_vector which are above 20 as TRUE while others
as FALSE.
> second_vector >= 20
[1] FALSE FALSE FALSE FALSE TRUE TRUE TRUE TRUE TRUE TRUE
```

(6) 'ages\_vector' returns the values that are above 20 using Booleans as a mode of filtrations.

```
> # Returns the values which are greater then 20 from the 'ages_vector'
> ages_vector [ages_vector>=20]
[1] 20 22 24 26 28 30
```

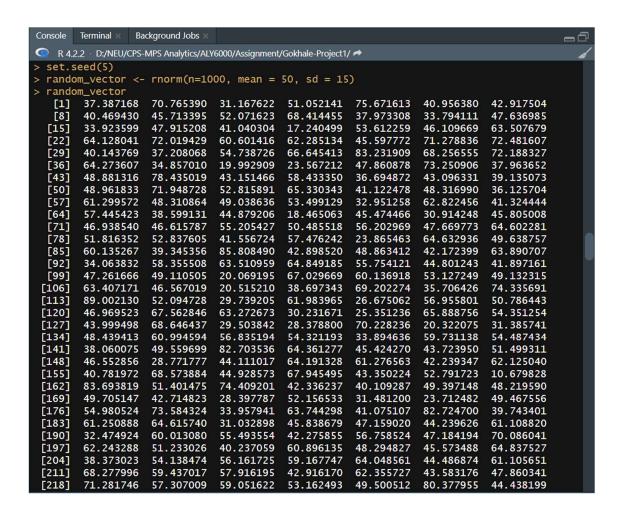
(7) By passing negative indexes to the combine function, we can drop the corresponding elements from the vector as displayed below.

```
R 4.2.2 · D:/NEU/CPS-MPS Analytics/ALY6000/Assignment/Gokhale-Project1/ > middle_grades_removed <- grades[c(-3,-4)] > middle_grades_removed [1] 96 100 81 72
```

(8) set.seed() is used to create the exact same random variables every time while the runif() is used to create a random variables with the arguments as minimum(starting point), maximum(end point) and the actual length of the vector.

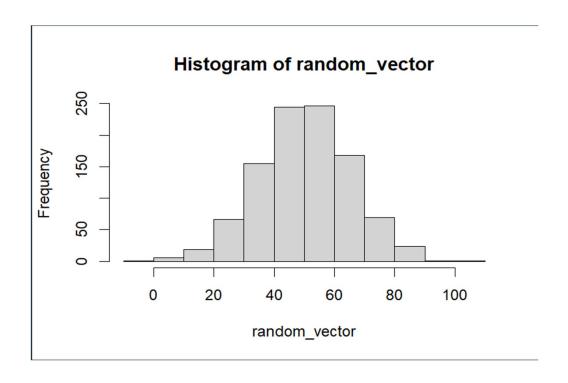
```
> # set.seed() is used to create the exact same Random Variables every time.
> # runif() is used to create the vector of given length, which each value being random.
> set.seed(5)
> random_vector <- runif(n=10, min = 0, max = 1000)
> random_vector
[1] 200.2145 685.2186 916.8758 284.3995 104.6501 701.0575 527.9600 807.9352 956.5001
[10] 110.4530
```

(9) The below code shows how to create random variables using normal distribution. Here a 1000 values with a mean value of 50 and the standard deviation of 15 are sampled out from a Gaussian Distribution using the "rnorm" function.



(10)R can be extensively used to create engaging visualizations, like histogram as shown below.

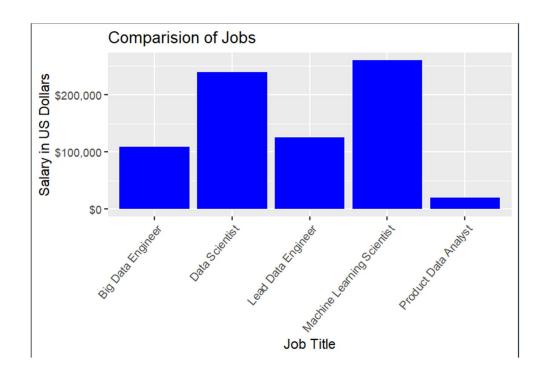
The hist function is passed with "random\_vector" to plot the histogram with the 1000 samples with the mean of 50 and SD of 15.



(11)The following lines of R code prove as to how flexible and versatile the R language is when it comes to analyzing the data frames. The readily available functions like head, select, arrange, filter, mutate and slice are extremely useful to get an overview as to how the data is structured and do the data conversions to gain insights out of it.

```
first_dataframe <- read_csv("ds_salaries.csv")</pre>
head(first_dataframe)
head(first_dataframe, n=7)
names(first_dataframe)
smaller_dataframe <- select(first_dataframe,job_title,salary_in_usd)</pre>
smaller_dataframe
better_smaller_dataframe <- arrange(smaller_dataframe,desc(salary_in_usd))</pre>
better_smaller_dataframe
better_smaller_dataframe <- filter(smaller_dataframe,salary_in_usd>80000)
better_smaller_dataframe
better_smaller_dataframe <- mutate(smaller_dataframe,salary_in_euros=salary_in_usd
better_smaller_dataframe
better_smaller_dataframe <- slice(smaller_dataframe,1,1,2,3,4,10,1)
better_smaller_dataframe
ggplot(better_smaller_dataframe)+
  geom_col(mapping = aes(x=job_title, y=salary_in_usd), fill="blue")+
  xlab("Job Title")+
  ylab("Salary in US Dollars")+
  labs(title = "Comparision of Jobs ")+
  scale_y_continuous(labels = scales::dollar)+
  theme(axis.text.x = element_text(angle = 50, hjust = 1))
```

(12) As mentioned previously R has extensive functions and ability to create visualization which truly convey insights in a compelling graphical manner. ggplot() from tidyverse package is extremely useful to create visualizations from the dataset or data frames. Various parameters like data-frame, labels, title, legend, color can be passed to create graphs. In the assignment a bar graph showing the "Comparison of the Job" is visualized with x-axis being "Job Title" while the y-axis being "Salary in USD"



## Conclusion

The assignment acted has a good exercise to gently introduce the R scripting language. It was quite good to learn and understand functions of R which have tremendous potential to not just to EDA but also carry out statistical analysis. R language also has a plethora of functions to visualize the data in a compelling manner. Overall, the assignment successfully helped me understand the basic functionality of R.

### References

https://www.rdocumentation.org/packages/pacman/versions/0.5.1

https://www.rdocumentation.org/packages/ggplot2/versions/0.9.0/topics/ggplot

https://www.rdocumentation.org/packages/tidyverse/versions/2.0.0