Assignment 2 - Data Analysis using R Programming

Group 5

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Data Loading

```
# Load the dataset df <-
read.csv("employee_salary_dataset.csv")
```

Structure and Overview

1. Print the structure of your dataset

```
str(df)
```

```
## 'data.frame':    50 obs. of 9 variables:
## $ EmployeeID      : int 1 2 3 4 5 6 7 8 9 10 ...
## $ Name            : chr "Employee_1" "Employee_2" "Employee_3" "Employee_4" ...
## $ Department      : chr "Marketing" "Operations" "IT" "Operations" ...
## $ Experience_Years: int 15 7 12 8 15 3 14 17 4 18 ...
## $ Education_Level : chr "Master" "Bachelor" "High School" "PhD" ...
## $ Age             : int 53 25 51 44 36 50 57 34 53 28 ...
## $ Gender          : chr "Female" "Female" "Female" "Male" ...
## $ City            : chr "Delhi" "Bangalore" "Hyderabad" "Delhi" ...
## $ Monthly_Salary : int 111416 95271 69064 95091 132450 65818 70525 44830 42429 31893 ...
```

2. List the variables in your dataset

```
names(df)
```

```
## [1] "EmployeeID"      "Name"            "Department"      "Experience_Years"
## [5] "Education_Level" "Age"             "Gender"          "City"
## [9] "Monthly_Salary"
```

3. Print the top 15 rows of your dataset

```
head(df, 15)
```

```
##    EmployeeID        Name Department Experience_Years Education_Level Age
## 1           1  Employee_1  Marketing               15          Master  53
## 2           2  Employee_2 Operations                7        Bachelor  25
## 3           3  Employee_3         IT               12     High School  51
```

## 4	4	Employee_4	Operations		8	PhD 44
## 5	5	Employee_5	Operations		15	Master 36
## 6	6	Employee_6	Finance		3	High School 50
## 7	7	Employee_7	IT		14	PhD 57
## 8	8	Employee_8	IT		17	PhD 34
## 9	9	Employee_9	IT		4	Bachelor 53
## 10	10	Employee_10	Operations		18	High School 28
## 11	11	Employee_11	Marketing		8	PhD 43
## 12	12	Employee_12	IT		4	Master 49
## 13	13	Employee_13	Operations		2	Master 23
## 14	14	Employee_14	Finance		6	Bachelor 27
## 15	15	Employee_15	Marketing		10	PhD 49

	Gender	City	Monthly_Salary
## 1	Female	Delhi	111416
## 2	Female	Bangalore	95271
## 3	Female	Hyderabad	69064
## 4	Male	Delhi	95091
## 5	Female	Delhi	132450
## 6	Male	Mumbai	65818
## 7	Male	Mumbai	70525
## 8	Female	Bangalore	44830
## 9	Male	Hyderabad	42429
## 10	Male	Mumbai	31893
## 11	Male	Delhi	141381
## 12	Female	Hyderabad	104909
## 13	Female	Hyderabad	72333
## 14	Male	Delhi	28436
## 15	Female	Mumbai	99290

User Defined Function

4. Write a user defined function using any of the variables from the data set.

```r
# Function to categorize experience level
categorize_experience <- function(years) {
  if (years < 5) {
    return("Junior")
  } else if (years >= 5 & years <= 10) { return("Mid-Level")
  } else {
    return("Senior")
  }
}

# Apply the function to the first few rows to demonstrate sapply(head(df$Experience_Years), categorize_experience)
```

```
## [1] "Senior"    "Mid-Level" "Senior"    "Mid-Level" "Senior"    "Junior"
```

Data Manipulation and Filtering

5. Use data manipulation techniques and filter rows based on any logical criteria

```r
# Filter employees with more than 10 years of experience and are from IT department
filtered_df <- df %>% filter(Experience_Years > 10 & Department
== "IT") head(filtered_df)
```

```
##   EmployeeID       Name Department Experience_Years Education_Level Age Gender
## 1          3 Employee_3         IT               12     High School  51 Female
## 2          7 Employee_7         IT               14             PhD  57   Male
## 3          8 Employee_8         IT               17             PhD  34 Female
## 4         26 Employee_26         IT               14          Master  24   Male
## 5         31 Employee_31         IT               15        Bachelor  54 Female
## 6         35 Employee_35         IT               13          Master  53   Male
##       City Monthly_Salary
## 1 Hyderabad         69064
## 2   Mumbai         70525
## 3 Bangalore         44830
## 4 Hyderabad         30600
## 5 Hyderabad         70714
## 6 Bangalore        130983
```

Reshaping and Joining

6. Identify the dependent & independent variables and use reshaping techniquesand create a new data frame by joining those variables from your dataset.

```r
# Dependent variable: Monthly_Salary
# Independent variables: Experience_Years, Age
```

```
# Create two separate dataframes to demonstrate joining
df_salary <- df %>% select(EmployeeID, Monthly_Salary) df_details <- df %>% select(EmployeeID, Experience_Years,
Age) # Join them back together joined_df <- left_join(df_details, df_salary, by = "EmployeeID") head(joined_df)
```

```
##   EmployeeID Experience_Years Age Monthly_Salary
## 1          1              15 53          111416
## 2          2               7 25           95271
## 3          3              12 51           69064
## 4          4               8 44           95091
## 5          5              15 36          132450
## 6          6               3 50           65818
```

Data Cleaning

7. Remove missing values in your dataset.

```
# Check for missing values sum(is.na(df))
```

```
## [1] 0
```

```
# Remove missing values (if any) df_clean <-
na.omit(df)
```

8. Identify and remove duplicated data in your dataset

```
# Check for duplicates sum(duplicated(df_clean))
```

```
## [1] 0
```

```
# Remove duplicates df_clean <- df_clean
%>% distinct()
```

Reordering and Renaming

9. Reorder multiple rows in descending order

```
# Reorder by Monthly_Salary in descending order
df_sorted <- df_clean %>% arrange(desc(Monthly_Salary)) head(df_sorted)
```

```
##   EmployeeID          Name Department Experience_Years Education_Level Age Gender
## 1         38 Employee_38 Operations                 9         Master 23   Male
## 2         11 Employee_11 Marketing                  8            PhD 43   Male
## 3         34 Employee_34         HR                15       Bachelor 53 Female
```

```
## 4          25 Employee_25          HR          8    High School 34 Female
## 5           5 Employee_5 Operations          15          Master 36 Female
## 6          35 Employee_35          IT          13          Master 53   Male
##          City Monthly_Salary

## 1    Mumbai          149123

## 2     Delhi          141381
## 3     Delhi          134616
## 4 Bangalore          132455
## 5     Delhi          132450
## 6 Bangalore          130983
```

10. Rename some of the column names in your dataset

```
# Rename 'Monthly_Salary' to 'Salary' and 'Experience_Years' to 'Experience'
df_renamed <- df_sorted %>% rename(Salary =
  Monthly_Salary,
          Experience = Experience_Years)

names(df_renamed)
```

```
## [1] "EmployeeID"      "Name"            "Department"      "Experience"
## [5] "Education_Level" "Age"             "Gender"          "City"
## [9] "Salary"
```

New Variables

11. Add new variables in your data frame by using a mathematical function

```
# Add a new variable 'Annual_Salary' (Monthly_Salary * 12)
df_final <- df_renamed %>% mutate(Annual_Salary =
Salary * 12) head(df_final)
```

```
##   EmployeeID          Name Department Experience Education_Level Age Gender
## 1         38 Employee_38 Operations          9          Master 23   Male
## 2         11 Employee_11 Marketing          8           PhD 43   Male
## 3         34 Employee_34         HR         15          Bachelor 53 Female
## 4         25 Employee_25         HR          8    High School 34 Female
## 5          5 Employee_5 Operations          15          Master 36 Female
## 6         35 Employee_35         IT          13          Master 53   Male
##          City Salary Annual_Salary

## 1    Mumbai 149123       1789476

## 2     Delhi 141381       1696572

## 3     Delhi 134616       1615392
## 4 Bangalore 132455       1589460
## 5     Delhi 132450       1589400
## 6 Bangalore 130983       1571796
```

Training Set

12. Create a training set using random number generator engine.

```
set.seed(123) # Set seed for reproducibility
sample_index <- sample(1:nrow(df_final), 0.7 * nrow(df_final)) training_set <-
df_final[sample_index, ] testing_set <- df_final[-sample_index, ] dim(training_set)
```

```
## [1] 35 10
```

Summary Statistics

13. Print the summary statistics of your dataset

```
summary(df_final)
```

```
##       EmployeeID         Name            Department          Experience
## Min.        : 1.00   Length:50          Length:50          Min.       : 1.00
## 1st Qu.:13.25          Class :character   Class :character   1st Qu.: 5.25
## Median :25.50          Mode :character    Mode :character    Median :10.00
## Mean        :25.50                                           Mean       : 9.90
## 3rd Qu.:37.75                                                3rd Qu.:14.75
## Max.        :50.00                                           Max.       :19.00
## Education_Level          Age            Gender             City
## Length:50          Min.        :22.00   Length:50          Length:50
## Class :character     1st Qu.:28.25   Class :character   Class :character
## Mode :character      Median :43.50   Mode :character    Mode :character
##                      Mean        :39.76
##                        3rd Qu.:49.00
##                      Max.        :57.00
##          Salary          Annual_Salary
## Min.        : 28420   Min.       : 341040
## 1st Qu.: 59424        1st Qu.: 713088
## Median : 73890        Median : 886686
## Mean       : 82289    Mean      : 987466
## 3rd Qu.:107219        3rd Qu.:1286628
## Max.        :149123   Max.      :1789476
```

14. Use any of the numerical variables from the dataset and perform the following statistical functions

```
# Using 'Salary' variable
salary_mean <- mean(df_final$Salary) salary_median <-
median(df_final$Salary) salary_range <- range(df_final$Salary)

# Calculate Mode
get_mode <- function(v) { uniqv <-
    unique(v)

    uniqv[which.max(tabulate(match(v, uniqv)))]
} salary_mode <- get_mode(df_final$Salary)
cat("Mean Salary:", salary_mean, "\n")
```

Mean Salary: 82288.8

```
cat("Median Salary:", salary_median, "\n")
```

Median Salary: 73890.5

```
cat("Mode Salary:", salary_mode, "\n")
```

Mode Salary: 149123

```
cat("Range Salary:", salary_range, "\n")
```

Range Salary: 28420 149123

Visualization

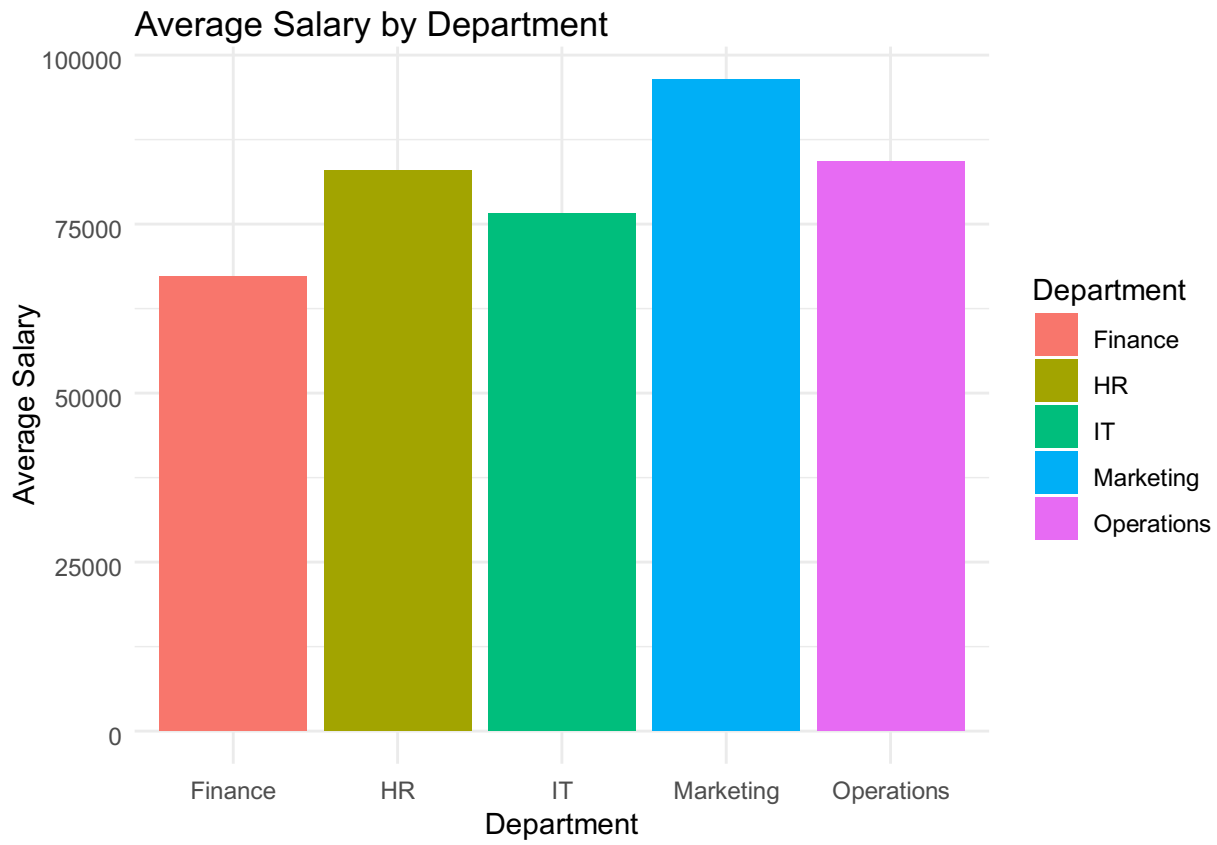
15. Plot a scatter plot for any 2 variables in your dataset

```
ggplot(df_final, aes(x = Experience, y = Salary)) + geom_point(color =
  "blue") + labs(title = "Scatter Plot of Salary vs Experience", x =
  "Experience (Years)", y = "Monthly Salary") +
  theme_minimal()
```

16. Plot a bar plot for any 2 variables in your dataset

```
# Average salary by Department
avg_salary_dept <- df_final %>% group_by(Department)
  %>% summarise(Avg_Salary = mean(Salary))

ggplot(avg_salary_dept, aes(x = Department, y = Avg_Salary, fill = Department)) + geom_bar(stat = "identity") +
  labs(title = "Average Salary by Department", x =
         "Department", y = "Average Salary") +
  theme_minimal()
```



Correlation

17. Find the correlation between any 2 variables by applying Pearson correlation

```
correlation <- cor(df_final$Experience, df_final$Salary, method = "pearson") cat("Pearson correlation between Experience and Salary:", correlation, "\n")
```

Pearson correlation between Experience and Salary: 0.07422086