

Course: Data Exploration and Preparation

Course Code: CAP482

CA₂

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Submitted By

Name: Aditya Kumar Yadav Name: Deepak Kuamr Rajbhar

Roll No: 42 **Roll No:** 5 **Reg:** 12307896 **Reg:** 12307910

Section: De225, Group: 2 Section: De225, Group: 1

<u>Submitted to</u> Ms. Ranjit Kaur Walia

UID: 28632 Assistant Professor SCA, LPU

Lovely Faculty of Technology & Sciences
School of Computer Applications
Lovely Professional University
Punjab

Title:- Analyzing Student Depression Using R

Project Overview

This project studies depression in students using a dataset. Many students feel mental pressure because of studies, personal life, and social expectations. The aim is to find the main reasons behind student depression, see common patterns, and understand how depression levels change among different students. This can help in knowing the real problems students face and suggest ways to improve their mental health.

Dataset Used

The dataset has student details like age, gender, academic performance, habits, sleep, and stress levels. It was downloaded from Kaggle: <u>Student Depression Dataset</u>. The columns "Work Pressure" and "Job Satisfaction" have been removed since these students do not have jobs.

Objectives

- 1. Understand the dataset (columns, missing values, data types).
- 2. Do basic analysis using filtering, grouping, and summarizing.
- 3. Find useful patterns in student depression and stress factors.
- 4. Gain insights to help in mental health support for students.

Levels of Data Analysis

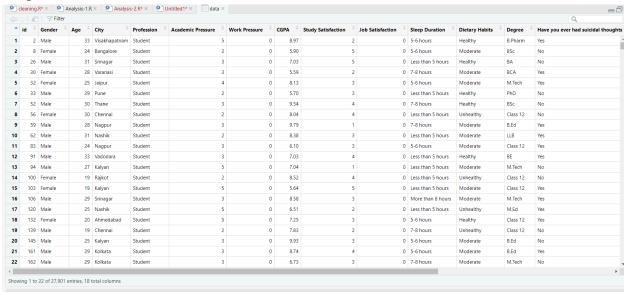
Level 1: Basic Exploration

Key Insights:

- 1. The dataset contained several missing values, which could affect the accuracy of the analysis. Removing them helped improve data reliability.
- 2. Basic statistical exploration provided an overview of the dataset, helping us understand which variables might be significant.
- 3. Calculating the percentage of students with depression gives us a strong metric that will be useful for deeper analysis and predictions.

Load required libraries library(readr) library(dplyr) library(tidyr)

#0: Load data set data <- read_csv("C:/Users/Aditya Yadav/Downloads/student depression.csv") View(data)



1. What are the column names and data types?

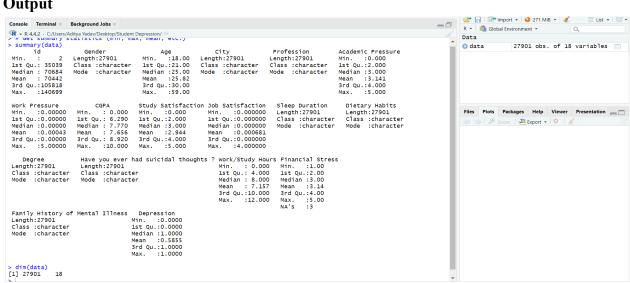
#1: Data understanding

Check structure of dataset (data types of each column) str(data)

Get summary statistics (min, max, mean, etc.) summary(data)

Get data set dimensions (total rows and columns) dim(data)

Output



Interpretation of Data Understanding:

1. **Dataset Size:** The dataset contains 27,901 rows and 18 columns, meaning we have data from 27,901 students with 18 features related to their academic and mental health conditions.

2. Column Types:

- Numerical Data: Includes Age, Academic Pressure, CGPA, Study Satisfaction, Work/Study Hours, Financial Stress, and Depression (where 1 = Depressed, 0 = Not Depressed).
- o Categorical Data: Includes Gender, City, Profession, Sleep Duration, Dietary Habits, Degree, and Family History of Mental Illness.
- 3. Columns to Remove: Work Pressure and Job Satisfaction: These columns are irrelevant since students generally don't have jobs.

2. Are there any missing values?

#2: Missing Values

Count total missing values sum(is.na(data))

Column-wise count of missing values colSums(is.na(data))

Output

3. Are there any unnecessary columns?

#3: Clean Data

Remove unnecessary columns data <- data %>% select(-`Work Pressure`, -`Job Satisfaction`) #Remove rows with NA data<-na.omit(data)

Output

```
> data <- data %>% select(-`Work Pressure`, -`Job Satisfaction`)
> #Remove rows with NA
> data<-na.omit(data)
> #2: Missing Values
> # Count total missing values
> sum(is.na(data))
[1] 0
```

Interpretation for Missing Values and Data Cleaning:

• Only 3 missing values found in the Financial Stress column.

Data Cleaning Steps Taken:

- Removed "Work Pressure" & "Job Satisfaction" as they are irrelevant for student depression analysis.
- Deleted rows with missing values (NA) to ensure clean data.

Final Dataset:

- No missing values.
- Refined for further analysis.

4. What is the percentage of students with depression?

#4: Calculate the percentage of students with depression

```
percentage_depressed <- mean(data$Depression) * 100
print(paste("Percentage of students with depression:", round(percentage_depressed, 2), "%"))
```

Output

```
> percentage_depressed <- mean(data$Depression) * 100
> print(paste("Percentage of students with depression:", round(percentage_depressed, 2), "%"))
[1] "Percentage of students with depression: 58.55 %"
```

Interpretation of Depression Percentage

1. High Depression Rate:

o 58.55% of students in the dataset show signs of depression, indicating a major concern.

2. Significance:

• More than half of the students are struggling with depression, which could be linked to factors like academic pressure, financial stress, or personal issues.

3. Next Steps:

- Further analysis is needed to identify key factors contributing to this high depression rate.
- Possible interventions can be explored for mental health support.

Level 2: Identifying Patterns

Key Insights from Level 2:

- 1. Academic Pressure vs. Financial Stress:
- One of these factors plays a bigger role in student depression.
- Solutions can be provided based on which factor is dominant.

2. Depressed Students List:

- Helps analyze common traits in depressed students.
- Can be used for targeted support programs.

3. Low CGPA & Depression Connection:

• If many students with low CGPA are depressed, universities should address academic-related stress and provide additional support.

5. Find the most common stress factors among students.

```
# Mean calculation with NA handling
financial_stress_mean <- mean(data$`Financial Stress`, na.rm = TRUE)
academic_pressure_mean <- mean(data$`Academic Pressure`, na.rm = TRUE)

# Comparison using if-else
if (academic_pressure_mean > financial_stress_mean) {
    print("Academic pressure is higher than financial stress for depression")
} else if (financial_stress_mean > academic_pressure_mean) {
    print("Financial stress is higher than academic pressure for depression")
} else {
    print("Both academic pressure and financial stress are equal for depression")
}
```

Output

Interpretation:

- Academic pressure is the primary stressor among students, slightly exceeding financial stress in its contribution to depression.
- The small difference (3.141 vs. 3.139) suggests that both factors are almost equally significant, indicating a need for balanced mental health support addressing both academic and financial concerns.
- Universities and institutions should focus on reducing academic pressure while also offering financial aid and counseling services to help students manage stress effectively.

6. List students who have depression (Depression = 1).

```
students_with_depression <- data %>% filter(Depression == 1) head(students with depression)
```

Output

```
Console Terminal × Background Jobs ×
R 4.4.2 · C:/Users/Aditya Yadav/Desktop/Student Depression/
> students_with_depression <- data %>% filter(Depression == 1)
> head(students_with_depression)
# A tibble: 6 x 16
      id Gender Age City
                                           Profession `Academic Pressure` CGPA `Study Satisfaction` `Sleep Duration`
   <db1> <chr> <db1> <chr>
       2 Male
                     33 Visakhapatnam Student
                                                                               5 8.97
                                                                                                                2 5-6 hours
                                                                                                                2 7-8 hours
1 7-8 hours
      30 Female
                      28 Varanasi Student
                                                                               3 5.59
                  28 Nagpur
31 Nashik
                                                                                  9.79
      59 Male
                                           Student
      62 Male
                                           Student
                                                                                  8.38
                                                                                                                 3 Less than 5 hours
                    24 Nagpur
27 Kalyan
      83 Male
                                           Student
                                                                                  6.1
                                                                                                                 3 5-6 hours
                                                                                                                 1 Less than 5 hours
                                           Student
# i 7 more variables: Dietary Habits' <chr>, Degree <chr>, 'Have you ever had suicidal thoughts?' <chr>, 'Barily History of Mental Illness' <chr>, Degreesion <dbl>, 'Family History of Mental Illness' <chr>, Degreesion <dbl>
```

Interpretation

- The dataset contains students diagnosed with depression, showing varied academic pressure, CGPA, and study satisfaction levels.
- **Sleep duration is inconsistent**, with some students getting less than 5 hours of sleep, which may contribute to mental health issues.

7. Identify students with low academic performance and depression.

```
# Total depressed students with CGPA below average
low_performance_depressed <- data %>%
filter(CGPA < mean(CGPA) & Depression == 1)%>%
nrow()

# Total students with CGPA below average
low_cgpa_students <- data %>%
filter(CGPA < mean(CGPA)) %>%
nrow()

# Calculate percentage
percentage_low_cgpa_depressed <- (low_performance_depressed / low_cgpa_students) * 100

# Print result
print(paste("Percentage of students with low CGPA who are depressed:",
round(percentage_low_cgpa_depressed, 2), "%"))
```

Output

```
Console Terminal × Background Jobs ×

R +R 4.4.2 · C:/Users/Aditya Yadaw/Desktop/Student Depression/ 
> low_performance_depressed

[1] 7622
> low_cgpa_students

[1] 13410
> percentage_low_cgpa_depressed

[1] 56.83818
> # Print result
> print(paste("Percentage of students with low CGPA who are depressed:", round(percentage_low_cgpa_depressed, 2), "%"))

[1] "Percentage of students with low CGPA who are depressed: "56.84 %"
```

Interpretation

- 56.84% of students with below-average CGPA are depressed, indicating a strong link between academic performance and mental health.
- Students struggling with grades may experience higher academic pressure, stress, or self-doubt, leading to depression.
- This insight suggests the need for mental health support, counseling, or academic assistance for students with lower CGPAs.

Level 3: Grouping & Summarization

Key Insights from Level 3:

1. **Depression by Age Group:**

Older students may experience more depression due to increasing pressure.

2. CGPA & Depression relation:

Students with low CGPA (0-4) are most affected by depression.

Academic stress plays a big role in students' mental health.

8. Find the percentage of students with depression by age group.

```
# 8: Group data by Age and calculate percentage of depressed students in each group
age_group_depression <- data %>%
group_by(Age) %>%
summarise(
   total_students = n(),
   depressed_students = sum(Depression),
   percentage_depressed = (depressed_students / total_students) * 100
)
# Print result
print(age_group_depression)
```

```
Console Terminal × Background Jobs ×
R 4.4.2 · C:/Users/Aditya Yadav/Desktop/Student Depression/
> # 8: Group data by Age and calculate percentage of depressed students in each group
> age_group_depression <- data %>%
     group_by(Age) %>%
  summarise(
      total_students = n(),
depressed_students = sum(Depression),
percentage_depressed = (depressed_students / total_students) * 100
  print(age_group_depression)
      Age total_students depressed_students percentage_depressed
                       1587
                                              1216
                       1560
                                              1100
                       <u>2</u>236
                       1726
                                              <u>1</u>169
                                               701
                       1160
                                                                        60.4
                       <u>2</u>258
                                              1509
                       1784
                                              1082
                                                                        60.7
                       <u>1</u>155
                      <u>1</u>462
# i 24 more rows
# i Use `print(n = ...) ` to see more rows
```

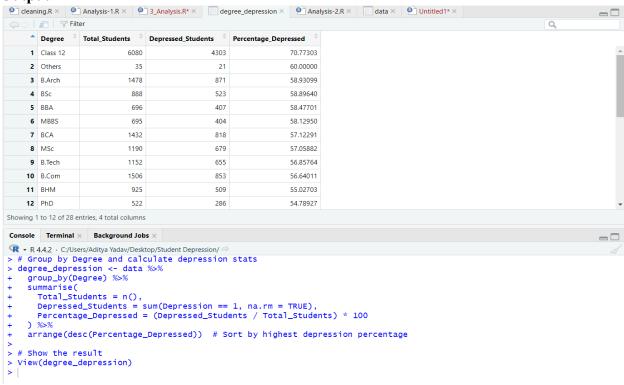
- Teenagers (18-20 years) have the highest depression rates (above 70%), likely due to academic pressure, career uncertainty, and social stress.
- Depression decreases slightly with age, with students aged 26+ having lower depression rates (~57-60%), possibly due to better coping mechanisms or life stability.
- These insights highlight the need for early mental health support, stress management programs, and counseling for younger students.
- 9. Analyze the relationship between CGPA and depression by grouping students into CGPA categories (0-4, 5-7, 8-10).

- Low CGPA (0–4): 44.4% of students are depressed relatively high, but this group has very few students (only 9), so the result might not be fully reliable.
- Average CGPA (5–7): 57.4% of students are depressed shows that a majority of average performers are experiencing depression.
- **High CGPA (8–10)**: 59.2% of students are depressed surprisingly, even top performers have high depression levels, possibly due to pressure to maintain performance.

10. Which degree program has the highest number and percentage of depressed students?

```
# Group by Degree and calculate depression stats
degree_depression <- data %>%
group_by(Degree) %>%
summarise(
    Total_Students = n(),
    Depressed_Students = sum(Depression == 1, na.rm = TRUE),
    Percentage_Depressed = (Depressed_Students / Total_Students) * 100
) %>%
arrange(desc(Percentage_Depressed)) # Sort by highest depression percentage
# Show the result
```

View(degree depression)



- Class 12 students have the highest depression rate at 70.77%, indicating that pre-university pressure and uncertainty may be a major factor.
- Other degrees like **B.Arch**, **BSc**, and **BBA** also show high depression rates (around 58–59%), possibly due to academic pressure or career concerns.
- Advanced degrees like **M.Tech**, **M.Ed**, and **MHM** show relatively lower depression rates (around 50–52%), suggesting that students at this stage may be better at handling academic stress or have more clarity in their goals.

Level 4: Ranking & Comparison

print(head(ranked students, 10))

11. Rank students by academic performance and depression levels.

```
# 11: Rank students based on CGPA and Depression levels ranked_data <- data %>%

# Rank based on CGPA (higher CGPA = better rank)
mutate(CGPA_Rank = dense_rank(desc(CGPA))) %>%

# Arrange by Depression (1 first) and CGPA_Rank
arrange(desc(Depression), CGPA_Rank) %>%

# Assign final ranking
mutate(Final_Rank = row_number())

# Select relevant columns
ranked_students <- ranked_data %>%
select(id, CGPA, CGPA_Rank, Depression, Final_Rank)

# Print top 10 ranked students
```

```
Console Terminal × Background Jobs ×
R 4.4.2 · C:/Users/Aditya Yadav/Desktop/Student Depression/
> ranked_data <- data %>%
    "anked_data <- data %>%

# Rank based on CGPA (higher CGPA = better rank)
mutate(CGPA_Rank = dense_rank(desc(CGPA))) %>%

# Arrange by Depression (1 first) and CGPA_Rank
arrange(desc(Depression), CGPA_Rank) %>%
     # Assign final ranking
     mutate(Final_Rank = row_number())
> # Select relevant columns
> ranked_students <- ranked_data %>%
+ select(id, CGPA, CGPA_Rank, Depression, Final_Rank)
> # Print top 10 ranked students
> print(head(ranked_students, 10))
# A tibble: 10 \times 5
         id CGPA CGPA_Rank Depression Final_Rank
                           <int>
                                          <db7>
     <db1> <db1>
                                               1
 1 13170
 2 <u>15</u>800
                                   1
                                                    1
                  10
                                                 1
 3 <u>22</u>499
                  10
    <u>24</u>975
                  10
                                  1
                                                   1
    <u>25</u>353
                  10
 6 <u>25</u>482
                  10
                                                                    6
7
 7 26892
                  10
 8 32697
    34831
```

- Top-ranked students (CGPA = 10) are depressed, proving that academic excellence ≠ mental well-being.
- CGPA alone does not determine depression; other factors like pressure and stress contribute.
- High achievers might face extreme expectations, leading to mental health struggles.
- Universities should focus on student well-being, not just grades.

12. What is the count and percentage of depression cases among males and females?

```
# Count of depression cases by gender table(data$Gender, data$Depression)
```

Percentage of depression in each gender prop.table(table(data\$Gender, data\$Depression)) * 100

Output

Interpretation:

- Total Cases: More males (9115) suffer from depression compared to females (7220).
- Percentage:
 - **Females:** 25.88% have depression, while 18.39% do not.
 - Males: 32.67% have depression, while 23.05% do not.
- **Insight:** Males have a higher percentage of depression cases than females in this dataset.

13. How many students from each dietary habit category (Healthy, Moderate, Unhealthy) have depression, and what is their percentage?

```
# 13: Count and percentage of depressed students by dietary habit dietary_depression <- data %>%
group_by('Dietary Habits', Depression) %>%
summarise(count = n(), .groups = "drop") %>%
mutate(Percentage = (count / sum(count)) * 100)

# Print result
print(dietary_depression)
```

Output

```
> print(dietary_depression)
  A tibble: 8 \times 4
   `Dietary Habits` Depression count Percentage
   <chr.
                                  <db1> <int>
                                                          <db7>
                                                       15.0
                                       0 4177
1 Healthy
                                                      12.4
                                       1 <u>3</u>472
0 <u>4</u>363
2 Healthy
3 Moderate
                                                       15.6
                                       1 5558 19.9
0 4 0.014<u>3</u>
1 8 0.028<u>7</u>
0 <u>3</u>019 10.8
1 <u>7</u>297 26.2
4 Moderate
5 Others
6 Others
  Unhealthy
8 Unhealthy
```

Interpretation:

- Unhealthy diet: Highest depression cases (26.2%).
- Moderate diet: Significant depression cases (19.9%).
- Healthy diet: Lowest depression cases (12.4%).
- **Key insight**: Poor dietary habits may be linked to higher depression risks.
- Suggestion: Encouraging a healthy diet could help improve mental well-being.

14. How many students who are depressed have also had suicidal thoughts?

```
suicidal_depressed <- data %>%
  filter(Depression == 1, `Have you ever had suicidal thoughts ?` == "Yes") %>%
  summarise(count = n())
print(suicidal depressed)
```

```
Console Terminal x Background Jobs x

R - R 4.4.2 · C:/Users/Aditya Yadav/Desktop/Student Depression/ 
> suicidal_depressed <- data %>%
+ filter(Depression == 1, `Have you ever had suicidal thoughts ?` == "Yes") %>%
+ summarise(count = n())
> print(suicidal_depressed)
# A tibble: 1 × 1
count
<int>
13957

13957
```

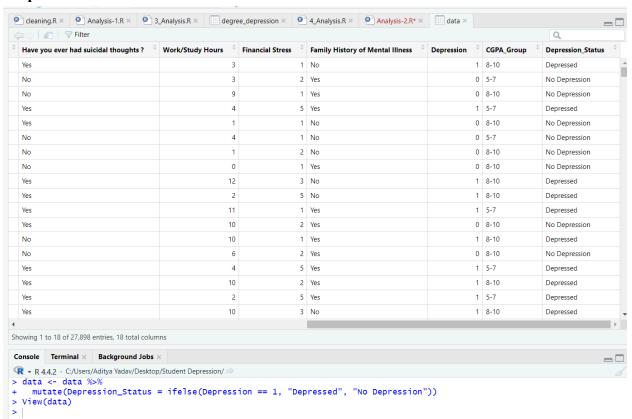
- Total 13,957 students who are depressed have also reported having suicidal thoughts.
- This suggests a strong link between depression and suicidal thoughts among students.
- Identifying and supporting these students through mental health initiatives is crucial.
- Out of all students, 16,335 are depressed and 17,656 have had suicidal thoughts.
 Among them, 13,957 students are both depressed and have experienced suicidal thoughts.
 This means around 85.44% of depressed students also reported suicidal thoughts, showing a strong link between depression and suicidal thinking in students.

Level 5: Creating New Insights

15. Add a new column "Depression Status" based on the depression column.

- "No Depression" (0)
- "Depressed" (1)

data <- data %>% mutate(Depression_Status = ifelse(Depression == 1, "Depressed", "No Depression")) head(data)



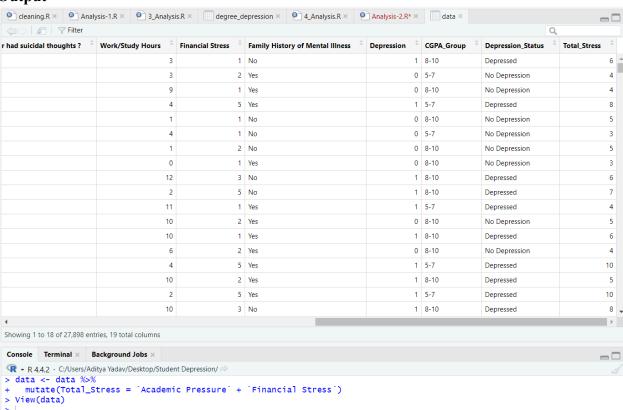
- A new column named Depression Status has been added to the dataset.
- It converts the numeric values in the Depression column (0 or 1) into more readable labels:
 - If Depression == 1, it shows "Depressed"
 - If Depression == 0, it shows "No Depression"
- This makes the data easier to understand and analyze, especially for reporting or visualization purposes.

16. What is the total stress level of each student by combining academic and financial stress?

```
data <- data %>%

mutate(Total_Stress = `Academic Pressure` + `Financial Stress`)

View(data)
```



- A new column called **Total Stress** has been created in the dataset.
- It is calculated by adding the values from two existing columns:
 - Academic Pressure
 - o Financial Stress
- This helps in understanding the combined effect of both academic and financial pressures on a student.
- You can now analyze how total stress relates to depression, CGPA, sleep, etc., more effectively.

Conclusion

The project successfully analyzed student depression data using R. It included data cleaning, transformation, and filtering steps to make the dataset analysis-ready. Using group-wise summaries and logical filtering, key insights were generated like:

- Percentage of students with depression
- Gender-wise and CGPA-wise depression trends
- Identification of students with suicidal thoughts
- Relationship between financial and academic pressure
- Programs with the highest depression levels

These findings show that data analysis can help in understanding student mental health issues more deeply and suggest focus areas for intervention.

Result

- Dataset cleaned and irrelevant columns removed.
- Depression % among students was calculated.
- Depression was compared with CGPA, age, and gender.
- Suicidal thoughts were analyzed among depressed students.
- "Depression_Status" and "Total_Stress" columns were created for deeper insights.
- A final ranked list of students based on CGPA and depression was generated.