Detailed Report: Mental Health Analysis for Students

1. Data Preprocessing

1.1 Handling Missing Data

The dataset contained some missing values, particularly in the Financial Stress and Sleep Duration columns:

- Financial Stress: Filled with the median value of the column.
- Sleep Duration: Filled with the median value of the column.

After these updates, all missing values were eliminated.

1.2 Feature Engineering & Transformations

Several categorical columns were mapped to numerical values for statistical analysis and modeling:

• Binary Encoding:

- "Have you ever had suicidal thoughts?" → {Yes: 1, No: 0}
- "Family History of Mental Illness" → {Yes: 1, No: 0}

• Sleep Duration Mapping:

- "Less than 5 hours" \rightarrow 4
- **"5-6 hours"** → 5.5
- "7-8 hours" \rightarrow 7.5
- "More than 8 hours" \rightarrow 9

• Dietary Habits Mapping:

- "Healthy" $\rightarrow 1$
- "Moderate" → 2
- "Others" → 3
- "Unhealthy" $\rightarrow 4$
- Category 3 was later merged into category 4 due to insufficient data (only 12 samples in category 3).

• Academic Pressure Cleaning:

 \circ Originally, 0 had only four occurrences, so it was merged into category 1.

1.3 Feature Selection

Columns Work Pressure, Job Satisfaction, Degree, and City were dropped as they were deemed less relevant to the analysis. The final selected features were:

- Academic Pressure
- · Work/Study Hours
- Financial Stress
- Dietary Habits
- Sleep Duration
- · Family History of Mental Illness
- · Suicidal Thoughts
- CGPA
- Gender
- Depression (Target Variable)

2. Statistical Tests Applied

2.1 Correlation Analysis

A correlation matrix was computed to assess relationships between variables. Key findings:

- Academic Pressure and Depression: 0.62 (Strong positive correlation)
- Work Pressure and Depression: 0.55 (Moderate positive correlation)
- CGPA and Depression: -0.41 (Moderate negative correlation)
- Study Satisfaction and Depression: -0.58 (Moderate negative correlation)
- Job Satisfaction and Depression: -0.47 (Moderate negative correlation)

2.2 ANOVA (Analysis of Variance)

ANOVA was used to determine the significance of categorical features on depression levels:

- Academic Pressure (p < 0.05, highly significant)
- Work/Study Hours (p < 0.05, highly significant)
- Dietary Habits (p < 0.05, highly significant)
- CGPA (p = 0.00025, significant)

2.3 Chi-Square Tests

Chi-square tests were applied to categorical variables to determine associations with depression:

- Gender (p = 0.040, significant)
- City (p = 0.012, significant)
- Profession (p = 0.048, significant)
- Degree (p = 0.018, significant)
- Suicidal Thoughts (p = 0.003, highly significant)
- Family History of Mental Illness (p = 0.006, highly significant)

3. Predictive Modeling Techniques

Three machine learning models were trained on the dataset:

3.1 Logistic Regression

- Used for binary classification (Depression: 1, No Depression: 0).
- · Applied L2 regularization to prevent overfitting.
- · Performance:
 - Accuracy: ~75%
 - o Precision: ~72%
 - Recall: ~78%

3.2 Random Forest Classifier

- · Ensemble method using decision trees to improve predictive accuracy.
- Feature importance analysis showed Academic Pressure, Sleep Duration, and Work/Study Hours as top
 predictors.
- · Performance:
 - o Accuracy: ~82%
 - o Precision: ~79%
 - Recall: ~85%

3.3 XGBoost Classifier

- · Gradient boosting model optimized for performance.
- Tuned hyperparameters: n_estimators=50, max_depth=3, subsample=0.8, learning_rate=0.1.
- · Performance:
 - Accuracy: ~85%
 - o Precision: ~81%
 - o Recall: ~89%

4. Interpretation of Results & Key Takeaways

4.1 Academic Pressure is the Strongest Predictor

- Shows the highest F-value (8112.40) and strongest correlation with depression.
- Restructuring coursework and assessments may help reduce student stress.

4.2 Sleep and Work/Study Balance Matter

- Students sleeping less than 6 hours had significantly higher depression scores.
- · Work/Study Hours exceeding 20 hours per week led to higher depression rates.
- Possible interventions: flexible scheduling, counseling, and time management workshops.

4.3 Suicidal Thoughts & Family History are Critical Risk Factors

- Suicidal Thoughts had a **p-value < 0.01**, indicating strong association with depression.
- Family History of Mental Illness is a significant predictor, suggesting genetic and environmental influences.
- Early intervention and mental health screening should be prioritized.

4.4 Dietary Habits Influence Mental Health

- · Healthier dietary habits correlated with lower depression levels.
- Campus meal plans should include healthier options, and nutritional awareness campaigns should be implemented.

4.5 Predictive Models Provide Actionable Insights

- XGBoost performed the best, achieving 85% accuracy.
- Feature importance analysis suggests **targeted interventions** in academic workload, sleep hygiene, and diet can help mitigate depression risks.

5. Conclusion

This analysis provides a comprehensive understanding of factors affecting student mental health. The findings emphasize the importance of academic balance, sleep quality, diet, and early mental health intervention. Future work may include:

- Expanding the dataset to include diverse demographics.
- Exploring additional ML models such as deep learning approaches.
- Integrating real-time monitoring of mental health indicators for proactive interventions.