DSA210 Final Report

The Impact of Smartphone Notifications on Cognitive Control

Student: Metin Gündüz

Course: DSA 210 - Introduction to Data Science

Instructor: Özgür Asar **Term:** Spring 2025

1. Introduction

Smartphones generate hundreds of notifications per day, fragmenting attention and potentially degrading executive control. This project investigates how daily notification load correlates with performance on a Go/No-Go task, a standard test of response inhibition. We additionally assess the roles of mood and caffeine intake.

2. Data Collection & Description

Period: 12 March – 25 April 2025 (45 days)

• **Device**: iPhone (notification tracking via Screen Time)

Task: Browser-based Go/No-Go task (PsyToolkit)

Variables:

- Notifications (daily count)
- Commission Error Rate (%)
- Reaction Time (ms)
- Mood (1–5 scale)
- Caffeine intake (mg)

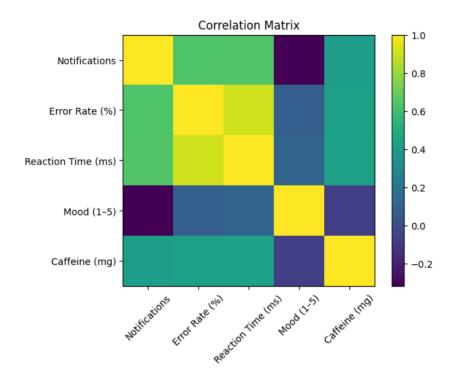
Data were collected manually and semi-automatically, converted from JSON to CSV using Google Colab, and merged on date. Cleaning steps included numeric conversions, character stripping, and null removal.

3. Exploratory Data Analysis (EDA)

3.1 Distributions

- Notifications mostly ranged between 250–400.
- Error Rate between 8–18%.
- Reaction Time 450-600ms.
- Mood typically 3–4 with notable dips.
- Caffeine peaked at 0, 60, 120, and 180mg.

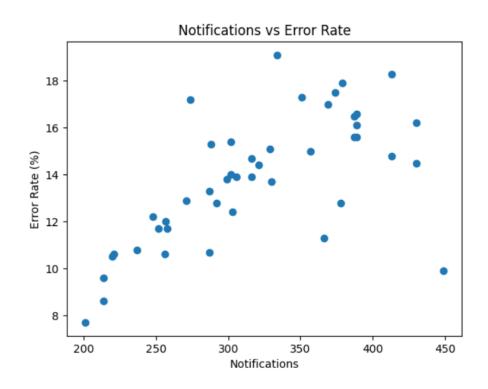
3.2 Correlation Matrix



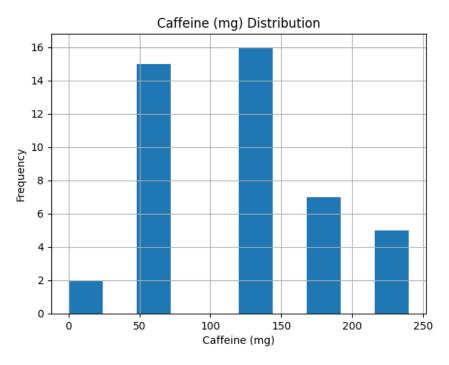
- Strong correlation between Error Rate and Notifications (r ≈ 0.64).
- Error Rate and Reaction Time also strongly correlated (r ≈ 0.90).

3.3 Key Scatter Plots

Notifications vs. Error Rate



Caffeine vs. Reaction Time



4. Hypothesis Testing

- **H0**: There is no relationship between notification count and error rate.
- **H1**: There is a positive linear relationship.

Result: Pearson r = 0.637, p < 0.001 \rightarrow Reject H0.

Conclusion: High notification days are associated with more commission errors.

Additional:

- Mood vs. Error Rate: r = 0.08 (p = 0.58) → No correlation.
- Caffeine vs. Reaction Time: r = 0.436 (p = 0.003) → Moderate correlation.

5. Machine Learning Modeling

5.1 Regression (Predicting Error Rate)

Features: Notifications, Mood, Caffeine

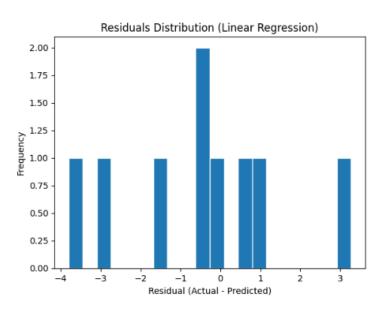
Models:

Model MSE R²

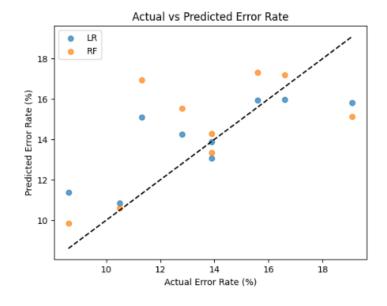
Linear Regression 4.04 0.57

Random Forest 6.72 0.28

Residual Histogram (LR)



Actual vs. Predicted (LR vs. RF)



5.2 Classification (High vs Low Error)

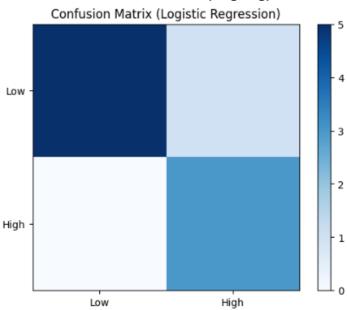
Labeling: Error Rate > median = High Error Day **Models**: Logistic Regression, Decision Tree

Model Accuracy AUC

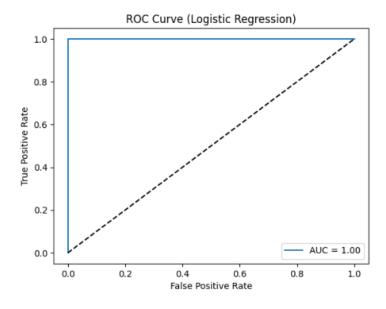
Logistic Regression 0.89 1.00

Decision Tree 0.78 0.88

Confusion Matrix (LogReg)



ROC Curve (LogReg)



6. Interpretation & Discussion

- **Notifications** was consistently the most predictive feature across tasks.
- Caffeine had a moderate but unexpected positive correlation with slower reaction times.
- Mood had weak or inconsistent predictive value.
- ML models performed well despite the small dataset; Logistic Regression classified high-error days with 0.89 accuracy and perfect AUC.

7. Limitations & Future Work

- Sample size limited to 45 days.
- Mood and caffeine self-reported; sleep and stress were not tracked.
- Future work could:
 - 1. Expand to 90+ days
 - 2. Incorporate sleep metrics and screen time types
 - 3. Use cross-validation and hyperparameter tuning
 - 4. Apply time-lagged models

8. Conclusion

This project supports the hypothesis that frequent daily smartphone notifications negatively affect cognitive control. Notifications predicted increased Go/No-Go commission errors and slower response times. Caffeine and mood had secondary roles.

9. References

- Apple Screen Time Documentation
- PsyToolkit.org Go/No-Go Task
- Breiman, L. (2001). Random Forests. Machine Learning
- Field, A. (2013). Discovering Statistics Using IBM SPSS Statistics
- scikit-learn Documentation