Analysis of My YouTube Viewing Habits

Uncovering patterns, trends, and insights from personal watch history

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•Why Analyze My YouTube Watch History?

YouTube has become an integral part of my daily routine, offering both entertainment and learning opportunities. This project explores how I engage with the platform and uncovers meaningful patterns.

•Objectives:

- 1.Identify dominant content preferences (e.g., Music, Gaming, Education).
- 2.Understand viewing habits across time (hour of the day, day of the week).
- 3. Validate the hypothesis: "I watch more YouTube outside of regular 9–17 working hours, especially close to midnight."

•End Goal:

Gain actionable insights to optimize my digital consumption habits.



 Data Source: Google Takeout (YouTube watch history)

•Key Features:

- Video Titles
- Timestamps (Date and Time)
- Channel Names

Preprocessing Steps:

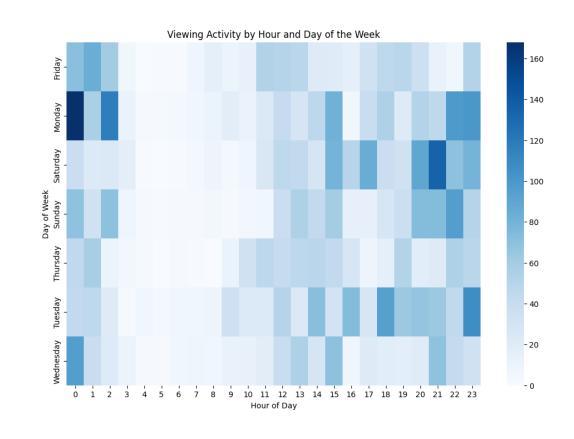
- •Parsed timestamps for hour and day analysis.
- Categorized videos based on content.
- Removed duplicates and irrelevant entries.



EDA

•EDA Goals:

- 1.Identify most-watched channels and categories.
- 2. Analyze viewing patterns by time and day.
- 3.Use visualizations (e.g., bar plots, heatmaps) for clear insights.
- •Visualizations:
- Category distribution (bar plot).
- •Hourly activity trends (bar plot).
- Weekly trends (day-of-week analysis).



Word Cloud of Video Titles



Key Insights

•Content Preferences:

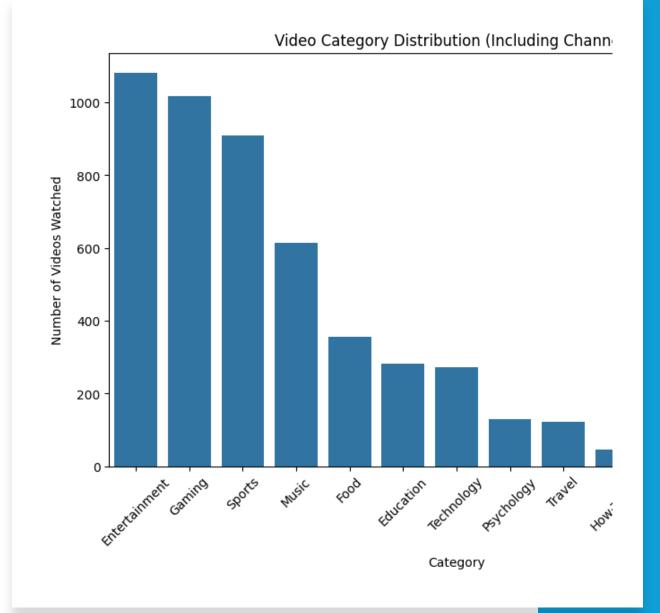
- •Dominant categories: Sports, Gaming, Entertainment.
- •Influential channels: *Crossover Talks*, *Boiler Room*.

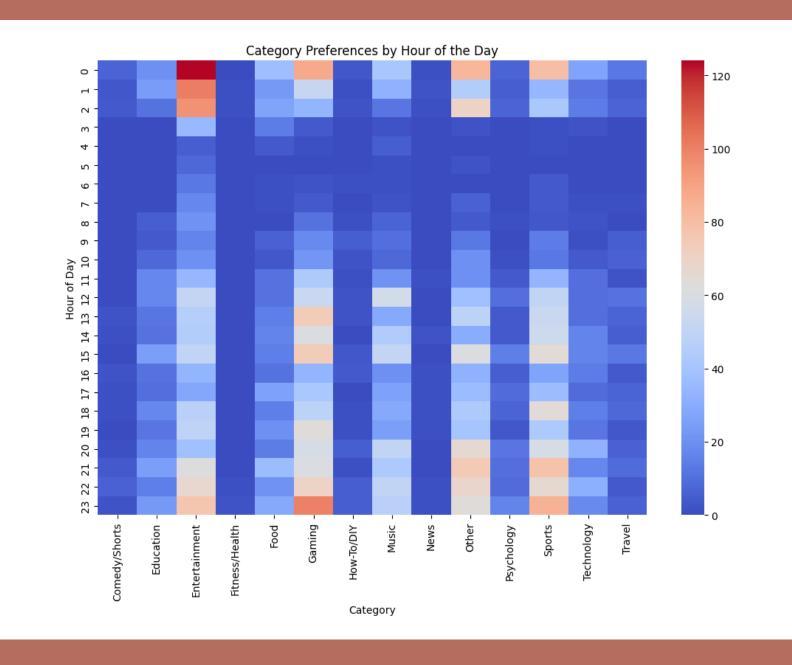
•Time Patterns:

- •Most activity occurs during non-working hours, particularly late at night.
- •Viewing peaks between 9 PM and midnight.

•Weekday vs. Weekend Trends:

•Higher activity on weekends, reflecting increased free time.





Hypothesis

Hypothesis:

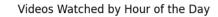
"I watch more YouTube outside of regular 9–17 working hours, especially close to midnight."

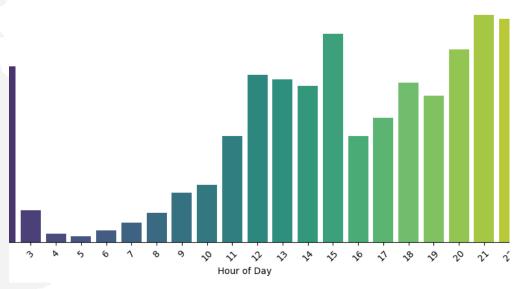
Key Findings:

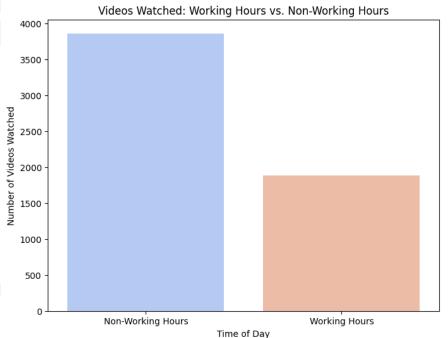
- Videos watched during working hours: 1882.
- Videos watched during non-working hours:
 3863.
- Significant peaks in viewing activity during late-night hours.

Conclusion:

The data strongly supports the hypothesis.







Future Work



Expand the Dataset:

Include data from platforms like Spotify and Netflix.



Category-Specific Analysis:

Investigate trends within categories (e.g., preferred Music genres).



Predictive Modeling:

Use machine learning to forecast future preferences.



Content Balance Analysis:

Explore the balance between educational and entertainment content.

Conclusion

•Key Findings:

- •Dominant content: Music, Gaming, Entertainment.
- •Peak activity during non-working hours, particularly close to midnight.
- •Weekends associated with higher activity.

•Takeaways:

- •This analysis validates my viewing patterns and provides insights for better content consumption choices.
- •Data-driven methods help uncover meaningful trends in everyday behavior.

