# Predicting the #1 Song on Spotify

An Analytics Approach for Kalshi Market Insights

Presented by Kenneth Lent

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Presented by Juliana Silva

# The Challenge

An Analytics Approach for Kalshi Market Insights

- Kalshi offers daily prediction markets: "Which song will be #1 on Spotify USA tomorrow?"
- Kalshi

- Market settles at 11 a.m. ET based on official Spotify U.S. Top 50 chart.
- Prices (1¢-99¢) reflect market's collective belief in a song's probability of hitting #1.
- Opportunity: External popularity signals (streaming, search, lyric engagement) often arrive before market settlement, creating potential for a data-driven trading edge.

# Project Goal

Streaming-Sentiment Signals

- Primary Objective: Develop an analytical tool to predict the probability of a song reaching #1 on Spotify USA, providing actionable insights for Kalshi market traders.
- Core Idea:
  - Ingest and process timely data signals (Spotify charts, API metadata).
  - Engineer features capturing song momentum and characteristics.
  - Train a predictive model to generate calibrated probabilities.
    Identify mispricings in the Kalshi market.

# Fueling the Model

Data & Features

**01**Spotify Charts CSVs

Daily rank, streams, track/artist IDs.

02 Spotify Web API

Track popularity, duration, explicitness, release date.



Rate limiting challenges. Didn't end up needing these features for a wellperforming model.





# Fueling the Model

Data & Features

# Key Engineered Features (Data Mining): days\_since\_release rank\_change (daily)

- stream\_momentum (vs. 3-day rolling average)





# Forecasting #1

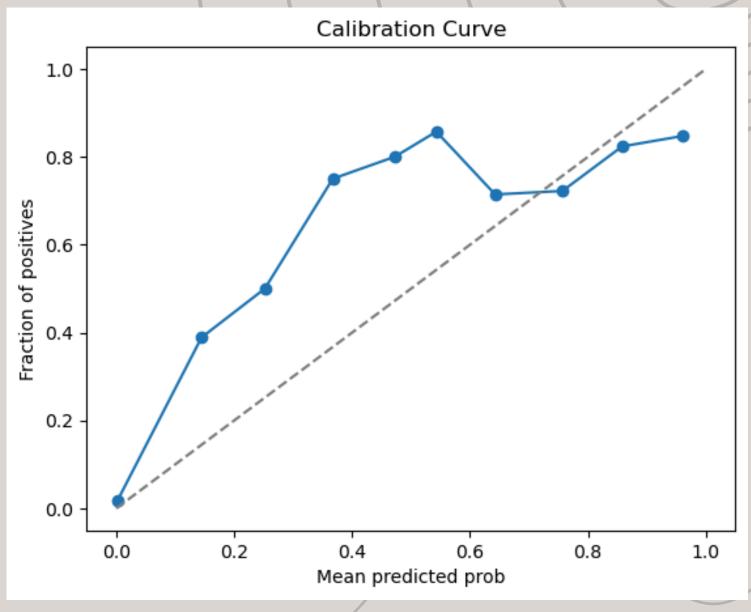
The Model

- Model: LightGBM Classifier
  - Chosen for efficiency and performance.
  - o class\_weight='balanced' to handle #1 song rarity.
- Training & Testing Strategy:
  - Time-Based Rolling Window:
    - Evaluated daily from Jan 1, 2025, to May 11, 2025.
    - Train on all prior data, predict for the next day.
       Simulates real-world application.
- Key Features Used: rank, streams, rank\_change, stream\_momentum, popularity, duration\_ms, explicit, days\_since\_release.



## How Well Does It Predict?

- Overall Predictive Power (Aggregated Multi-Day Results):
  - ROC AUC: 0.97 (Excellent discrimination)
  - o Average Precision (PR Curve): 0.81
- Daily Prediction Averages (for #1 song):
  - Precision: 0.68
  - Recall: 0.68
- Calibration: Model probabilities reasonably well-calibrated.



"Predicted probabilities are generally reliable."

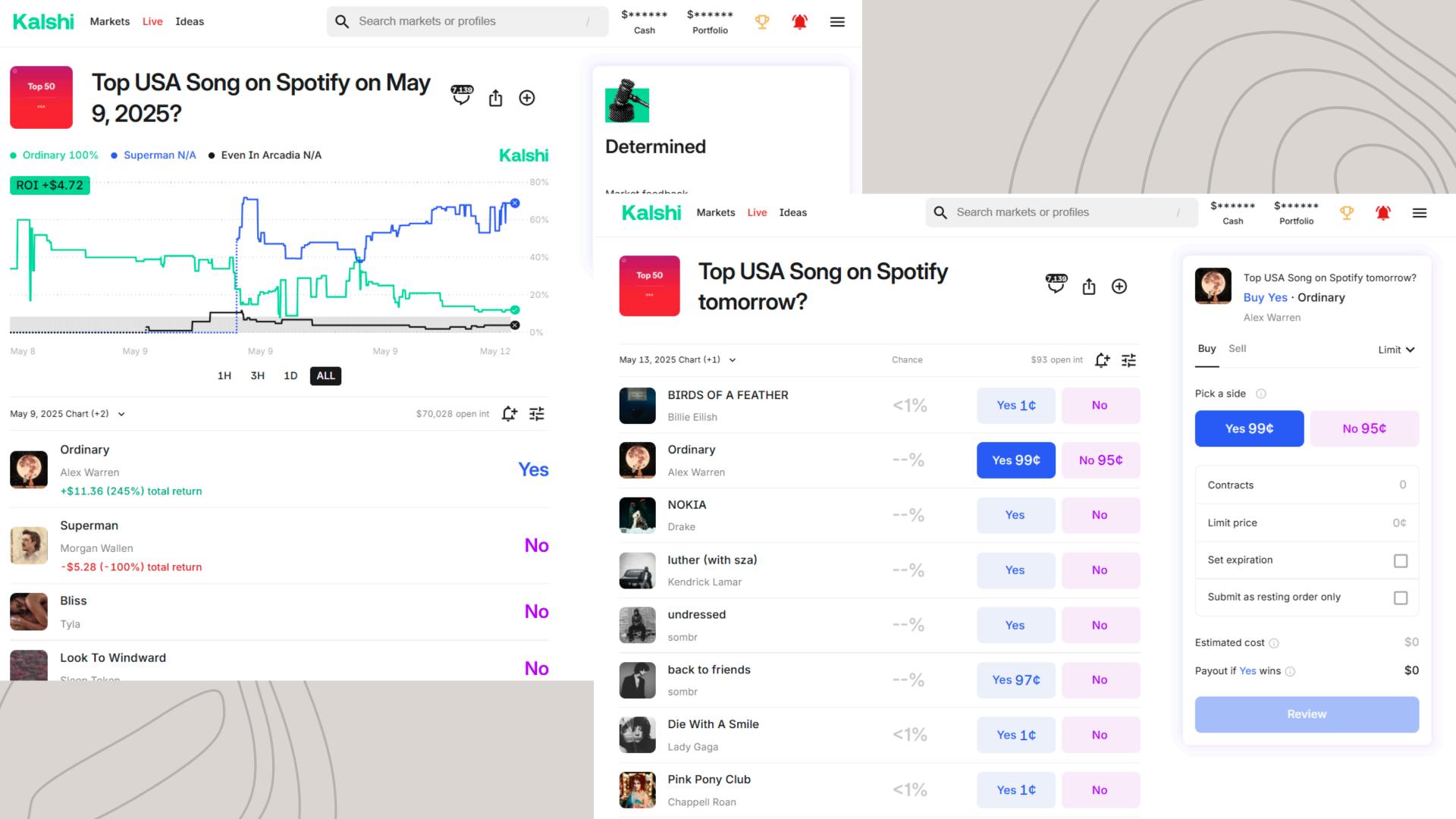
## From Prediction to Potential Profit

### Kalshi Market Edge

- Model generates probabilities before market settlement.
- Allows identification of potential over/undervalued contracts by comparing model probability to market-implied odds.
- Important Note: Predicting #1 is the first step. Actual profitability requires backtesting against Kalshi market conditions (liquidity, spreads).

#### **Key Feature Drivers**

- streams and current rank are highly influential.
- stream\_momentum and general Spotify popularity also important.



## Conclusion & Path Forward

#### • Conclusion:

- Successfully developed an analytical tool predicting Spotify's #1 song with strong performance (ROC AUC 0.97).
- Demonstrates value of API usage, feature engineering, and predictive modeling for market insights.

#### • Key Recommendations:

- Develop Backtesting Framework: Crucial for assessing real-world trading viability against Kalshi odds.
- Expand Data Sources: Incorporate web scraping (social media, news) and other APIs (Genius) to bolster the model.
- Monetization Potential: Data pipeline and model can be basis for a subscription insights service.

## Acknowledging Limits & Next Steps

#### Limitations

- Google Trends unfeasibility (due to rate limits, but model still performed well).
- Relies on data availability; no live trading implemented.

#### **Future Work**

- Implement financial backtesting.
- Integrate diverse data (social media, news).
- Explore advanced models (ensembles, neural networks).
- Develop a user-friendly dashboard.
- Automate the daily pipeline.

## Thank You & Questions



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