

Dynamic Contextualization for Starbucks Orders

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Problem Statement

Starbucks recently deployed a new product recommendation system on their drive-thru screens at 4,000 Starbucks locations. This recommendation system has been in use for the last year. Now the company would like to do a redesign of the interface and has tasked a UX team to refresh the screens. The ultimate goal of this project is to both optimize the conversion rate of Starbucks customers and to better communicate why the customer should purchase the items of interest to increase the likelihood of the product being purchased. The new design recommendations include more dynamic content to increase screen engagement and sales. The hypothesis is that if better communication is provided through headlines that show the customer why they are seeing particular products, there will be a higher revenue.

DATA USED:

> ACTION REWARDS

Maps impressions to successful conversions

> WEATHER: store-hour-level

Normalized weather features per seasonality

> STORE

Store level features such as number and location

> PRODUCT

Product features such as product type, name, ingredients and flavor profile

Project Goals

- Determine context labels to requesting store
- Predict optimal headline choice
- Ensure product recommendation alignment with chosen headline

Feature Vectors

INPUTS
requesting store, time, weather, 4 products

CONTEXT ELEMENTS
{preferred mode}
{weather state}
{store city}
{daypart}

STRING TEMPLATES
“{weather state} in {store city}”
“{daypart} {preferred mode}”



Context Element Mappings

The four context element keys are determined based on a set of rules displayed below. The “preferred customer mode” key is assigned to the value if the product is above the 50th percentile of the distribution ie. a product is a “treat” if it surpasses the 50th percentile of calories of all the products. This same logic applies to products that are a “Treat” or “Boost” when looking at the sugar and caffeine distributions respectively. Cognitive dissonance is taken into consideration to ensure that the headlines align with the products ie. model will only recommend hot drinks when the weather is chilly and iced drinks when it is sunny.

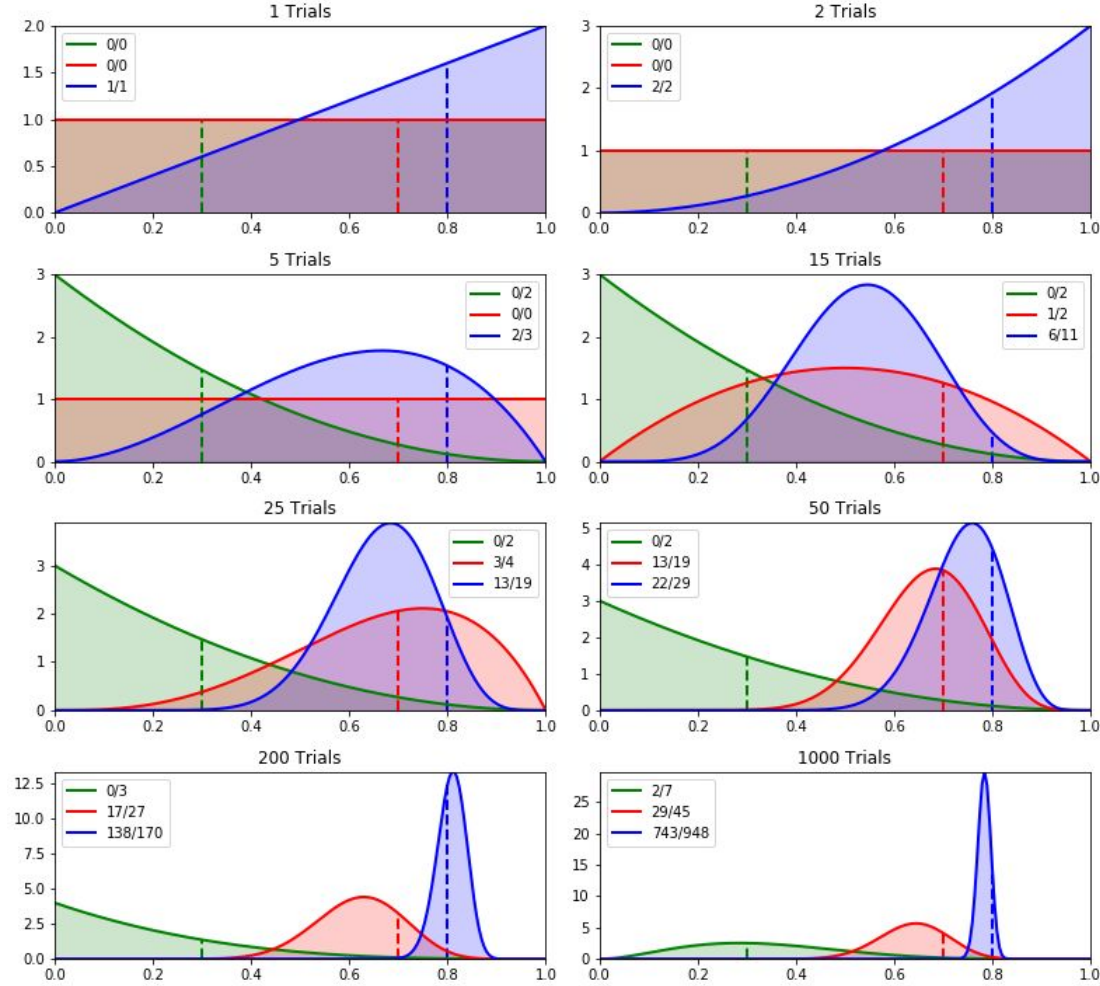
| Context Element Key | Context Element Value | Product Discordancy Notes |
|--------------------------|-----------------------|-------------------------------------|
| Hour store_city | {store num mapping} | |
| weather_state | Sunny | No hot drinks |
| weather_state | Chilly | No iced drinks |
| weather_state | Snowy | No iced drinks |
| weather_state | Rainy | No iced drinks |
| weather_state | Pleasant | |
| daypart | Morning | |
| daypart | Lunch | |
| daypart | Afternoon | |
| daypart | Evening | |
| preferred_customer_mode | Light Pick-Me-Up | Calorie threshold/weight potential |
| preferred_customer_mode | Treat | Sugar threshold/weight potential |
| preferred_customer_mode | Boost | Caffeine threshold/weight potential |
| preferred_customber_mode | Flavor | |

Addressing Cold Start

There is no historical data on the performance of these dynamic contextualized headlines since they haven’t yet been pushed to production yet. We used Thompson Sampling which utilizes a random uniform distribution to create random probabilities for each headline. The prior is set to an optimistic prior and we observe how quickly we can converge to a headline that produces the most incremental ticket.

Multi-Armed Bandits

The multi-armed bandit problem models an agent that wants to simultaneously acquire new knowledge (exploration) and optimize decisions based on existing knowledge (exploitation).



The information (customers purchasing or not) gained from the data gathered will shift the distribution in different ways to reflect an accurate reward model.

Headlines

We utilized the product features to create 11 potential headlines and then employed our implementation of multi-armed bandits to choose the best headline for conversion rate.

| |
|---|
| Headline 0 was displayed 49.0 times |
| Headline 1 was displayed 95.0 times |
| Headline 2 was displayed 34.0 times |
| Headline 3 was displayed 57.0 times |
| Headline 4 was displayed 650.0 times |
| Headline 5 was displayed 33.0 times |
| Headline 6 was displayed 57.0 times |
| Headline 7 was displayed 1605.0 times |
| Headline 8 was displayed 119.0 times |
| Headline 9 was displayed 125.0 times |
| Headline 10 was displayed 176.0 times |
| Overall Conclusion: The best headline to display is headline “Sunny Treat”! |