# Planogram Optimization

Finding optimal product placement on shelf

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## **Problem Statement**

In the retail industry, placement of products plays an important role in affecting the profit. The products at the center of modular, for instance, are more likely to be purchased by customers.



# Steps

Location-sales Analysis

Store Effect Location Effect Optimization

# **Store Effect**

- 1. Sales of a product could be strongly influenced by which store it is at
- Need to account for store effect before studying location-sales relationship

Store Category Sales

#### **Location Effect\***

Store-related Features

# Results

Results have been generated for 10 different sizes of planograms in Dog Treats category. Here is an example.

Walmar



NATURAL MEAT PREMIUM BISCUITS SOFT & CHEWY

#### **Current placement**



#### **Project Goal**

- Study the location-sales relationship
- Build an optimization algorithm 2. that maximizes sales

# Data Description

Scope

**1 Product Category: Dog Treats** 

7 Product Groups: Biscuits, Meat...

#### Sales Data

- 4000+ stores & 300+ products
- Time span: 12/01/19 02/15/20
- Aggregated over time per store

- 1. Build one model per product group per modular size
- 2. Use the predicted values for dog treats
- category sales as a predictor to account for the store effect

Store Effect







Figure 3 3-D illustration of location effect

\*Possible qualification: location data not evenly distributed.

# Optimization

Based on the insights from location-sales analysis, an optimization formulation has been created to generate the product placement on planogram that maximizes sales.

#### Idea

The planogram is broken down into small blocks, and assign each block to some product group.

#### **Our recommendation**





In this example, by changing from current placement to our recommendation, we expect an overall uplift of 6% in sales

# Conclusion

Different planograms with different sizes have different expected sales uplift. In general, the numbers range from 2% to 10%. While the numbers are just rough

estimations, they are good signs showing that there is a great potential that the produced recommendation will lead to increase in sales.

In the case of Walmart especially, the world's largest retailer, the numbers already indicate huge real world impacts.



*Figure 1 Distribution for store category sales* 

## **Location Data**

- (x, y) location on modular
- Bottom left point is (0, 0)
- Normalized by modular size



Figure 2 Distribution of historical locations for one group

# **Additional Data**

Store-related:

- Size
- Location
- Demographic...

**Product-related:** 

- Price
- Flavor
- Height...

# **Objective**

maximize expected sales

rewards if blocks of the same group stay close

# **Key Decision Variables**

1 if block i is assigned to product  $z_{i,k} \in \{0,1\}$ group k



1 if block i and block j are near each other and assigned to the same group

# **Constraints**

- Each block assigned to exactly one group
- Each group gets enough space
- Groups are in contiguous shape
- Exclude assignment to extremely uncommon locations

# Future Works

#### Add additional constraints



Generate results for more product categories



Improve modeling accuracy through experimentation and learning