

ISYE Homework 13

Problem: Optimizing Shelf Space Allocation to Maximize Profit and Validate Effectiveness

To tackle this problem, we can combine data analysis, predictive modeling, optimization, and real-world testing. This way, we can make sure our recommendations are practical, effective, and backed by solid data.

Data Requirements

A robust dataset is essential for effective analysis and modeling. Required data:

1. Product-Level Data

- Product ID: Unique identifier for each product
- Category: Classification of products into groups
- Price and Cost: Retail price and supplier cost for calculating profit margins
- Package Size: Dimensions to calculate shelf space requirements
- Historical Shelf Space: Allocation in previous periods
- Cross-Selling Potential: Groupings of frequently purchased products

2. Sales Data

- Transaction Data: Includes basket-level purchases, quantity sold, and timestamps
- Seasonality: Historical sales trends to identify seasonal patterns and demand spikes
- Store ID: Links sales to specific store layouts and characteristics

3. Store-Level Data

- Shelf Space Constraints: Minimum and maximum space requirements per product type
 - Total Space: Store-wide capacity for shelf space
 - Store Layout: Aisle arrangements, shelf configurations, and product placement impacts
 - Customer Demographics: Buyer behavior linked to age, income, and other characteristics
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Analytical Models

This phase involves using advanced analytics to understand relationships, predict outcomes, and optimize shelf space.

A. Time-Series Analysis

- Given: Historical sales data, including seasonality and trends
- Use:
 - Apply exponential smoothing or ARIMA models to adjust for seasonality and isolate long-term trends
 - Identify high-demand periods for specific products
- To:
 - Ensure sales data used in further models accurately reflects product performance without seasonal bias

B. Regression Analysis

- Given: Seasonally adjusted sales data, shelf space allocation, and price data.
- Use:
 - Fit linear or polynomial regression models to quantify the relationship between shelf space and sales or profit
 - Use advanced regression methods (lasso, ridge, elastic net) to identify key drivers of sales
- To:
 - Test the hypothesis: Does increasing shelf space lead to higher sales or profit?
 - Determine elasticity: How much sales increase for each additional unit of shelf space

C. k-Means Clustering

- Given: Transaction data showing products frequently purchased together and product attributes
- Use:
 - Cluster products based on co-occurrence and shared attributes
 - Identify cross-selling opportunities by grouping complementary items
- To:
 - Inform shelf layouts by placing related products close together to encourage cross-selling

D. Optimization

- Given:
 - Regression outputs, clustering results, and store constraints
- Use:
 - Develop an optimization model with the objective function to maximize profit:
 - Constraints:
 - Total space available in the store.
 - Minimum and maximum shelf space limits for each product.
 - Placement of clustered products for cross-selling opportunities.
- To:
 - Allocate shelf space optimally to maximize profit while adhering to physical and business constraints.

Validating Results

To ensure the effectiveness of the recommendations, employ experimental validation techniques:

A. A/B Testing

- Given:
 - Current sales data as a baseline
 - Optimized shelf layouts versus legacy layouts
- Use:
 - Test the optimized layout in selected stores (treatment group) and retain the legacy layout in others (control group)
 - Measure metrics like total sales, profit, and cross-selling increases
- To:
 - Validate whether the optimized layout leads to statistically significant improvements.
 - Ensure the model's predictions align with real-world outcomes before scaling company-wide

B. Statistical Testing

- Given: A/B testing results and regression outputs.
- Use:
 - Perform hypothesis testing (e.g., t-tests) to determine if shelf space changes significantly impact sales or profit
- To:
 - Verify that observed improvements are not due to chance

Practical Considerations

1. Scalability:
 - Ensure models and recommendations can be scaled across multiple stores with varying layouts and constraints
2. Data Quality:

- Ensure all data is accurate, complete, and up-to-date for reliable insights
- 3. Customer Behavior:
 - Incorporate customer demographics and local preferences to fine-tune recommendations for each store
- 4. Diminishing Returns:
 - Account for diminishing returns in sales as shelf space for a product increases