# **Customer Shopping Trends Analytics**



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# **Building an Interactive Customer Shopping Trends Analytics Dashboard**

A comprehensive journey through designing and implementing a data visualization platform for retail analytics

#### 1. Introduction

In today's data-driven retail landscape, understanding customer shopping behaviors is critical for business success. This blog post documents my recent project: an interactive web-based dashboard for visualizing and analyzing customer shopping trends. The platform integrates multiple Python visualization libraries, a Flask backend, and a responsive frontend to provide comprehensive insights into customer preferences, purchase patterns, and demographic information.

# 2. Project Overview

This customer shopping trends analytics dashboard processes retail transaction data to generate dynamic visualizations that help identify patterns in consumer behavior. The platform offers multiple visualization approaches through four popular Python libraries - Matplotlib, Seaborn, Pandas, and Plotly - allowing users to explore the same dataset through different analytical lenses.

# 3. Key Features

- Multi-library visualization with Matplotlib, Seaborn, Pandas, and Plotly
- Interactive dashboard with theme switching and responsive design
- **Dynamic data loading** with real-time visualization generation
- Dataset information panel for contextual understanding
- Downloadable visualizations for reports and presentations
- Dark/light theme support for different viewing preferences
- Comprehensive retail metrics including demographics, purchases, and reviews

#### 4. The Data Set

The project uses a comprehensive retail dataset containing customer shopping information including:

- **Demographics:** Age, gender, location
- Purchase details: Item, category, price
- Payment methods: Credit/debit cards, PayPal, etc.

- Shipping types: Standard, express, next-day
- Review ratings: Customer satisfaction levels
- Subscription status: Active subscriptions

This rich dataset enables deep analysis of shopping patterns across different customer segments and product categories

#### 5. Technical Architecture:

## **Backend Components**

The backend is powered by Flask and incorporates several Python libraries for data processing and visualization:

```
from flask import Flask, render_template, jsonify, request, send_file
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import base64
import io
```

The Flask server handles visualization requests through dedicated endpoints:

```
deapp.route('/get_visualizations', methods=['POST'])
def get_visualizations():
    library = request.form.get('library')
df = load_data()

if library == 'matplotlib':
    results = generate_matplotlib_visualizations(df)
elif library == 'seaborn':
    results = generate_seaborn_visualizations(df)
elif library == 'pandas':
    results = generate_pandas_visualizations(df)
elif library == 'plotly':
    results = generate_plotly_visualizations(df)
else:
    return jsonify({'error': 'Invalid library selected'})
return jsonify(results)
```

#### **Visualization Generation**

Each visualization library has its own module for generating specific insights:

## **Matplotlib Visualizations**

Matplotlib provides foundational visualizations with direct control over plot elements:

```
def generate_matplotlib_visualizations(df):
    visualizations = []

# Age Distribution
plt.figure(figsize=(10, 6))
plt.hist(df_clean['Age'].values, bins=20, edgecolor='black', color='skyblue')
plt.title('Age Distribution of Customers')
plt.xlabel('Age')
plt.ylabel('Age')
plt.ylabel('Count')
visualizations.append(save_plot_to_base64(plt, 'Age Distribution'))

# Other visualizations...

return {
    'library': 'matplotlib',
    'visualizations': visualizations
}
```

#### Seaborn Visualizations

Seaborn enhances statistical visualizations with minimal code:

```
def generate_seaborn_visualizations(df):
    visualizations = []

# Age Distribution by Gender
plt.figure(figsize=(10, 6))
sns.histplot(data=df, x='Age', hue='Gender', kde=True, bins=20, alpha=0.6)
plt.title('Age Distribution by Gender')
visualizations.append(save_plot_to_base64(plt, 'Age Distribution by Gender'))

# Additional visualizations...

return {
    'library': 'seaborn',
    'visualizations': visualizations
}
```

#### **Pandas Visualizations**

Pandas provides quick, integrated visualizations directly from DataFrames:

```
def generate_pandas_visualizations(df):
    visualizations = []

# Top 10 Items Purchased

plt.figure(figsize=(10, 6))
    item_counts = df['Item Purchased'].value_counts()

top_10_items = item_counts.head(10)

sorted_top_items = top_10_items.sort_values()

sorted_top_items.plot(kind='barh', color='teal')

visualizations.append(save_plot_to_base64(plt, 'Top 10 Items Purchased'))

# More visualizations...

return {
    'library': 'pandas',
    'visualizations': visualizations
}
```

#### **NumPy Visualizations**

NumPy visualizations offer low-level, efficient control over data representation, enabling optimized, custom-built charts using array operations and manual plotting.

```
def generate_numpy_visualizations(df):
   # Visualization 6: Average Rating by Season
   unique_seasons = np.unique(seasons)
   avg_ratings_by_season = np.zeros(len(unique_seasons))
   for i, season in enumerate(unique_seasons):
       avg_ratings_by_season[i] = np.mean(review_ratings[seasons == season])
   plt.figure(figsize=(10, 6))
   plt.plot(unique_seasons, avg_ratings_by_season, marker='o', color='purple')
   plt.title('Average Rating by Season (NumPy)')
   plt.xlabel('Season')
   plt.ylabel('Average Rating')
   plt.ylim(0, 5)
   plt.grid(True, linestyle='--', alpha=0.3)
   visualizations.append(save_plot_to_base64(plt, 'Average Rating by Season'))
   plt.close()
   return {
       'library': 'numpy',
        'visualizations': visualizations
```

#### **Frontend Components**

The frontend combines HTML, CSS, and JavaScript to create an interactive user experience:

**Responsive UI Structure** The HTML structure creates a responsive dashboard with a sidebar for library selection.

```
cdiv class="app-container">
cnav class="sidebar">
cliv class="sidebar">
cliv class="logo">
class="fas fa-chart-line"></i>
cspan>Data Viz</span>
cliv
class="fas fa-bars"></i>
cspan>Data Viz</span>
class="fas fa-bars"></i>
class="fas fa-chart-bar"></i>
class="fas fa-chart-bar"></ii>
class="fas fa-bars"></ii>
class="fas fa-bars"></ii>
class="fas fa-chart-bar"></ii>
class="fas fa-ch
```

## **Dynamic Visualization Loading**

JavaScript handles the loading and display of visualizations:

```
function loadVisualizations(library) {
       // Store selected library
       localStorage.setItem('selectedLibrary', library);
       // Hide welcome message and show loading
       welcomeMessage.style.display = 'none';
        loadingIndicator.style.display = 'flex';
        visualizationContainer.innerHTML = '';
       // Send request to backend
        const formData = new FormData();
        formData.append('library', library);
        fetch('/get visualizations', {
            method: 'POST',
           body: formData
        })
        .then(response => response.json())
        .then(data => {
            loadingIndicator.style.display = 'none';
            if (data.error) {
                showErrorMessage(data.error);
               return;
            displayVisualizations(data);
        })
        .catch(error => {
            loadingIndicator.style.display = 'none';
            showErrorMessage('Failed to load visualizations. Please try again.');
            console.error('Error:', error);
       });
```

# 6. Key Visualizations and Insights:

#### **Customer Demographics**

The age distribution visualizations reveal that most customers fall in the 30-60 age range, with a significant concentration around 35-45 and 50-60 years old. This bimodal distribution suggests two key customer segments to target with different marketing strategies.

#### **Purchase Patterns**

The "Purchase Amount by Category" visualization shows that:

- Footwear has the highest average purchase amount (\$80+)
- Accessories show the most variability in spending
- Clothing items maintain consistent mid-range purchases

#### **Payment Preferences**

The payment method distribution reveals:

- Credit cards are the most popular payment method (40%)
- Digital payment options (PayPal, Venmo) are gaining traction (35% combined)
- Cash payments represent a smaller but significant portion (15%)

#### **Seasonal Trends**

The seasonal purchase pattern visualization demonstrates:

- Spring has the highest purchase activity
- Winter shows the highest average purchase amounts
- Summer purchases have the most diversity in product categories

# 7. Implementation Challenges and Solutions:

#### **Challenge 1: Responsive Visualization Display**

**Problem:** Different visualization libraries produce different image sizes and formats that didn't consistently scale well on various devices.

**Solution:** Implemented standardized card-based layouts with CSS Grid and responsive design techniques:

```
.visualization-container {
    display: grid;
    grid-template-columns: repeat(auto-fit, minmax(500px, 1fr));
    gap: 30px;
}

.visualization-card {
    background-color: var(--card-bg);
    border-radius: var(--border-radius-md);
    box-shadow: var(--shadow-md);
    overflow: hidden;
    transition: transform 0.3s ease, box-shadow 0.3s ease;
}
```

## **Challenge 2: Integrating Plotly's Interactive Visualizations**

**Problem:** Plotly's interactive visualizations required special handling compared to static image-based charts.

**Solution:** Created a dual-approach rendering system in JavaScript:

**User Experience Features** 

## **Theme Switching**

The application supports both light and dark themes, toggled with a simple button click. The theme preference is stored in localStorage for a consistent experience across visits:

```
1 // Check for saved theme preference
2 const savedTheme = localStorage.getItem('darkTheme');
3 if (savedTheme === 'true') {
4    document.body.classList.add('dark-theme');
5    themeToggle.innerHTML = '<i class="fas fa-sun"></i>';
6 }
```

#### **Information Panel**

A slide-out information panel provides dataset context and explanation:

```
datasetInfoToggle.addEventListener('click', function() {
    datasetInfoPanel.classList.add('active');
});

closePanel.addEventListener('click', function() {
    datasetInfoPanel.classList.remove('active');
});
```

## **Download Functionality**

Each visualization can be downloaded for use in presentations or reports:

```
function downloadImage(imageData, title) {
        const formData = new FormData();
        formData.append('image_data', imageData);
        formData.append('chart title', title);
        fetch('/download image', {
            method: 'POST',
            body: formData
        1)
        .then(response => response.blob())
        .then(blob => {
            const url = window.URL.createObjectURL(blob);
            const a = document.createElement('a');
            a.href = url;
            a.download = `${title}.png`;
            a.click();
            // Show download success notification
            showNotification('Chart downloaded successfully!', 'success');
        })
        .catch(error => {
            showNotification('Failed to download image', 'error');
        });
24 }
```

# 8. Business Insights:

The visualizations reveal several actionable business insights:

- 1. **Target demographic:** Marketing efforts should focus on the 35-45 and 50-60 age groups, which represent the core customer base.
- 2. **Product pricing strategy:** Footwear items command higher prices and should be positioned as premium offerings.

- 3. **Seasonal marketing:** Spring campaigns should emphasize variety, while winter campaigns should focus on high-value items.
- 4. **Payment optimization:** Digital payment methods should be expanded and promoted, as they show growing adoption rates among customers.
- 5. **Review correlation:** Higher-priced items don't necessarily correlate with higher review ratings, suggesting quality improvement opportunities in premium product lines.

## 9. Future Enhancements:

This project has several potential avenues for expansion:

- 1. **Machine Learning Integration:** Add predictive analytics for sales forecasting and customer segmentation.
- 2. **Real-time Data Updates:** Connect to live databases or APIs for continuously updated visualizations.
- 3. **Advanced Filtering:** Add UI controls to filter visualizations by demographic or product attributes.
- 4. **Comparison Tools:** Implement features to compare time periods or customer segments side-by-side.
- 5. **Export Options:** Add functionality to export insights as PDF reports or data in various formats.

#### 10. Conclusion:

This Customer Shopping Trends Analysis dashboard demonstrates the power of combining multiple visualization libraries with a responsive, user-friendly interface. By providing multiple perspectives on the same dataset, it enables deeper insights into customer behavior and preferences.

The project showcases how modern web technologies and data visualization techniques can transform raw retail data into actionable business intelligence. Through careful attention to user experience, performance, and visual communication, the dashboard makes complex data accessible and meaningful to business stakeholders.

Whether you're analyzing customer demographics, purchase patterns, or seasonal trends, this visualization platform provides the tools needed to drive data-informed business decisions in the retail sector.