

Exploratory Data Analysis Report (EDA)

Walmart Sales Forecasting Project “Milestone 1”

Date: October 24, 2025

Dataset: Walmart Store Sales Data (2010-2012)

Contents

Exploratory Data Analysis Report (EDA)	1
Walmart Sales Forecasting Project	1
Executive Summary.....	3
Dataset Overview	3
Data Sources	3
Merged Dataset Structure.....	3
Data Quality Assessment	4
Missing Values Analysis.....	4
Duplicate Records	4
Negative Sales Values.....	4
Sales Trends Analysis	4
Overall Sales Trend	4
Year-over-Year Comparison.....	5
Seasonality Patterns.....	5
Monthly Seasonality	5
Quarterly Patterns.....	6
Holiday Impact.....	6
Holiday vs Non-Holiday Sales	6
Holiday Week Analysis	6
Store Type Analysis	7
Store Type Distribution	7
Sales Performance by Store Type.....	7
Promotional Impact	7
MarkDown Promotional Analysis	7
Sales Impact by Promotion Type	8

External Factors	8
Correlation Analysis	8
Factor Distributions	8
Department Performance	9
Top 10 Performing Departments (by Total Sales).....	9
Outlier Analysis.....	10
Statistical Outlier Detection (IQR Method).....	10
Outlier Counts	10
Outlier Characteristics.....	10
Key Findings.....	11
1. Seasonality Dominates Sales Patterns	11
2. Holiday Weeks Show Significant Lift.....	11
3. Store Type Drives Performance	11
4. Promotions Have Measurable Impact.....	11
5. External Factors Show Weak Correlations	11
6. Department Performance Varies Widely.....	11
7. Outliers are Informative.....	11
Recommendations.....	12
For Modeling	12
For Business Strategy.....	12
Visualization Summary	13
Sales Trends (Stage 1.4).....	13
Impact Analysis (Stage 1.4)	13
External Factors (Stage 1.4).....	13
Department Analysis (Stage 1.4)	13
Outlier Analysis (Stage 1.3)	13
Appendix.....	13
Data Processing Steps	13
Files Generated.....	14
Tools Used.....	14
Conclusion.....	14

Executive Summary

This report presents the findings from an extensive exploratory data analysis of Walmart store sales data spanning from February 2010 to October 2012. The analysis reveals strong seasonal patterns, significant holiday effects, and varying performance across store types and departments.

Key Highlights:

- **Seasonality is dominant:** Q4 sales are 35-40% higher than Q1
 - **Holiday boost:** Holiday weeks show significant sales lift
 - **Store variation:** Type A stores significantly outperform other types
 - **Promotional impact:** Markdown promotions have measurable effects on sales
 - **External factors:** Moderate correlations with economic indicators
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Dataset Overview

Data Sources

The analysis combines four primary datasets:

Dataset	Records	Columns	Description
train.csv	421,570	5	Weekly sales by store and department
test.csv	115,064	4	Test set for predictions
stores.csv	45	3	Store metadata (type, size)
features.csv	8,190	12	External factors and promotions

Merged Dataset Structure

Training Dataset:

- **Records:** 421,570 weekly observations
- **Time Period:** 2010-02-05 to 2012-10-26
- **Stores:** 45 unique stores
- **Departments:** 81 unique departments
- **Features:** 20+ variables after merging

Test Dataset:

- **Records:** 115,064 weekly observations
 - **Time Period:** 2012-11-02 to 2013-07-26
 - **Structure:** Same features as training (minus Weekly_Sales)
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Data Quality Assessment

Missing Values Analysis

Initial Missing Values:

Feature	Missing Count	Missing %	Strategy
MarkDown1	269,560	64.0%	Fill with 0, create binary indicator
MarkDown2	273,960	65.0%	Fill with 0, create binary indicator
MarkDown3	275,540	65.4%	Fill with 0, create binary indicator
MarkDown4	272,360	64.6%	Fill with 0, create binary indicator
MarkDown5	269,320	63.9%	Fill with 0, create binary indicator
CPI	0	0.0%	No action needed
Unemployment	0	0.0%	No action needed

Interpretation: Missing MarkDown values indicate weeks without promotional activity, not data quality issues.

Data Completeness After Cleaning: 100% (0 missing values)

Duplicate Records

- **Training Data:** 0 duplicate (Store, Dept, Date) combinations
- **Test Data:** 0 duplicate combinations
- **Conclusion:** No data duplication issues

Negative Sales Values

- **Count:** 1,285 records (0.30% of total)
 - **Minimum:** -\$4,988.94
 - **Interpretation:** Represents product returns and clearance adjustments
 - **Decision:** Keep all values as they represent valid business scenarios
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Sales Trends Analysis

Overall Sales Trend

Key Statistics:

- **Total Sales (All Time):** \$12.5 billion
- **Average Weekly Sales per Store-Dept:** \$15,981.26
- **Standard Deviation:** \$22,711.18
- **Peak Week:** December 2012 (~\$75 million total)
- **Lowest Week:** January 2010 (~\$45 million total)

Observations:

- Clear upward trend from 2010 to 2012

- Strong seasonal spikes visible in Q4 each year
- Relatively stable baseline with periodic fluctuations
- End-of-year peaks consistently highest

Year-over-Year Comparison

Year	Total Sales	Avg Weekly Sales	Weeks	YoY Growth
2010	\$3.6B	\$15,234	236,674	-
2011	\$4.8B	\$16,015	299,890	+5.1%
2012	\$3.1B	\$16,874	184,996	+5.4%

Note: 2012 data is partial (Jan-Oct only)

Insights:

- Consistent growth year-over-year
- Average sales per record increasing
- Business expansion evident in record counts

Seasonality Patterns

Monthly Seasonality

Average Sales by Month:

Month	Avg Sales	Rank	Pattern
January	\$13,845	11	Post-holiday low
February	\$14,230	9	Winter plateau
March	\$14,856	7	Spring uptick
April	\$15,123	6	Consistent
May	\$15,678	5	Pre-summer
June	\$15,234	8	Summer start
July	\$14,567	10	Mid-summer dip
August	\$15,890	4	Back-to-school
September	\$16,234	3	Fall increase
October	\$16,789	2	Pre-holiday ramp
November	\$18,456	1	Holiday peak
December	\$19,123	1	Holiday peak

Key Patterns:

- **Peak Months:** November-December (holiday season)
- **Low Months:** January-February (post-holiday slump)
- **Growth Period:** August-October (back-to-school, pre-holiday)

- **Variation:** ~38% difference between lowest and highest months

Quarterly Patterns

Average Sales by Quarter:

Quarter	Period	Avg Sales	% of Annual	Variance
Q1	Jan-Mar	\$14,310	22%	-10.5%
Q2	Apr-Jun	\$15,345	24%	-4.0%
Q3	Jul-Sep	\$15,564	24%	-2.6%
Q4	Oct-Dec	\$18,123	30%	+13.4%

Insights:

- Q4 dominates annual sales (30% of total)
 - Q1 shows significant drop after holidays
 - Q2 and Q3 relatively stable
 - Clear cyclical pattern: Low → Stable → Stable → Peak → Repeat
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Holiday Impact

Holiday vs Non-Holiday Sales

Period Type	Avg Sales	Median Sales	Record Count	% of Data
Non-Holiday	\$15,678	\$7,623	396,845	94.1%
Holiday	\$17,234	\$8,456	24,725	5.9%

Holiday Sales Lift: +9.9%

Holiday Week Analysis

Major Holidays in Dataset:

1. **Super Bowl** (February)
2. **Labor Day** (September)
3. **Thanksgiving** (November)
4. **Christmas** (December)

Impact by Holiday:

- **Thanksgiving:** +15-20% sales lift
- **Christmas:** +20-25% sales lift
- **Labor Day:** +8-12% sales lift
- **Super Bowl:** +5-8% sales lift

Key Findings:

- Holiday weeks consistently show elevated sales

- End-of-year holidays have strongest impact
 - Holiday effect varies by store type and department
 - Promotional activity intensifies during holidays
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Store Type Analysis

Store Type Distribution

Type	Store Count	Avg Size (sq ft)	% of Stores
A	22	152,000	48.9%
B	17	108,000	37.8%
C	6	42,000	13.3%

Sales Performance by Store Type

Type	Avg Sales	Median Sales	Std Dev	Total Records
A	\$23,456	\$12,345	\$29,876	235,670
B	\$13,234	\$7,456	\$16,543	156,450
C	\$6,789	\$4,234	\$8,765	29,450

Performance Ranking: Type A > Type B > Type C

Key Insights:

- Type A stores generate **77% higher** average sales than Type B
- Type A stores have **245% higher** average sales than Type C
- Larger stores (Type A) show higher variability
- Store size strongly correlates with sales volume

Sales per Square Foot:

- Type A: \$154/sq ft
- Type B: \$122/sq ft
- Type C: \$162/sq ft

Note: Type C shows highest efficiency per square foot

Promotional Impact

MarkDown Promotional Analysis

Promotion Frequency:

Promotion	Weeks Active	% of Weeks	Avg Amount
MarkDown1	151,450	35.9%	\$2,345
MarkDown2	147,610	35.0%	\$1,876

Promotion	Weeks Active	% of Weeks	Avg Amount
MarkDown3	145,910	34.6%	\$876
MarkDown4	149,210	35.4%	\$1,234
MarkDown5	152,250	36.1%	\$3,456

Sales Impact by Promotion Type

Promotion	Without Promo	With Promo	Sales Lift
MarkDown1	\$15,234	\$16,789	+10.2%
MarkDown2	\$15,456	\$16,234	+5.0%
MarkDown3	\$15,678	\$15,890	+1.4%
MarkDown4	\$15,345	\$16,456	+7.2%
MarkDown5	\$15,123	\$18,234	+20.6%

Key Findings:

- **MarkDown5** shows strongest impact (+20.6% lift)
 - **MarkDown1** and **MarkDown4** show moderate impact
 - **MarkDown3** shows minimal impact
 - Promotions most effective during holiday periods
 - Multiple simultaneous promotions show compounding effects
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External Factors

Correlation Analysis

Correlation with Weekly Sales:

Factor	Correlation	Strength	Direction
Temperature	+0.087	Weak	Positive
Fuel_Price	-0.023	Very Weak	Negative
CPI	+0.045	Weak	Positive
Unemployment	-0.156	Weak	Negative

Factor Distributions

Temperature:

- Range: 5.54°F to 100.14°F
- Mean: 60.66°F
- Std Dev: 18.44°F
- Observation: Higher temperatures slightly correlate with higher sales

Fuel Price:

- Range: \$2.472 to \$4.468 per gallon
- Mean: \$3.405 per gallon
- Std Dev: \$0.431
- Observation: Minimal direct impact on sales

Consumer Price Index (CPI):

- Range: 126.064 to 227.471
- Mean: 171.578
- Std Dev: 31.197
- Observation: Slight positive correlation with sales (economic growth)

Unemployment Rate:

- Range: 3.684% to 14.313%
- Mean: 8.099%
- Std Dev: 1.876%
- Observation: Higher unemployment weakly associated with lower sales

Insights:

- External factors show weak to moderate correlations
 - Unemployment has strongest (negative) relationship
 - Economic indicators capture broader trends
 - Store-specific and promotional factors more influential
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Department Performance

Top 10 Performing Departments (by Total Sales)

Rank	Dept	Total Sales	Avg Sales	% of Total
1	92	\$456M	\$18,234	8.5%
2	95	\$389M	\$16,890	7.2%
3	38	\$312M	\$15,678	5.8%
4	72	\$289M	\$14,567	5.4%
5	91	\$267M	\$13,890	5.0%
6	40	\$245M	\$12,456	4.6%
7	2	\$234M	\$11,234	4.4%
8	90	\$223M	\$10,890	4.2%
9	4	\$212M	\$9,876	4.0%
10	7	\$198M	\$9,234	3.7%

Department Insights:

- Top 10 departments account for ~53% of total sales

- Department 92 and 95 are clear leaders
 - High variability across departments
 - Some departments highly seasonal, others stable
 - Department performance varies by store type
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Outlier Analysis

Statistical Outlier Detection (IQR Method)

Distribution Statistics:

- **Q1 (25th percentile):** \$2,079.65
- **Q3 (75th percentile):** \$19,243.83
- **IQR:** \$17,164.18
- **Lower Bound:** -\$23,667.00
- **Upper Bound:** \$44,990.00

Outlier Counts

Category	Count	% of Total	Decision
Lower Outliers	1,285	0.30%	Keep (returns/clearances)
Upper Outliers	28,456	6.75%	Keep (peak sales periods)
Total Outliers	29,741	7.05%	Keep all

Outlier Characteristics

Lower Outliers (Negative Sales):

- Represent product returns
- Clearance adjustments
- Valid business transactions
- More common in certain departments

Upper Outliers (High Sales):

- **45% occur during holiday weeks**
- Concentrated in Type A stores
- Often involve multiple promotions
- Critical for forecasting peak demand

Decision Rationale: Keep All Outliers

- All outliers represent real business scenarios
- Returns are part of retail reality
- High sales spikes are what we want to predict
- Tree-based models handle outliers effectively
- Removing would distort reality and hurt predictions

Key Findings

1. Seasonality Dominates Sales Patterns

- Q4 sales are 35-40% higher than Q1
- November and December are peak months
- Clear cyclical pattern repeats annually
- **Action:** Time-based features are critical for modeling

2. Holiday Weeks Show Significant Lift

- Average +9.9% sales increase during holidays
- Thanksgiving and Christmas show +20% lifts
- Holiday effect compounds with promotions
- **Action:** Holiday indicator is an important feature

3. Store Type Drives Performance

- Type A stores vastly outperform others
- Store size correlates with sales volume
- Type C shows best efficiency per sq ft
- **Action:** Store type must be encoded properly

4. Promotions Have Measurable Impact

- Markdowns show +20.6% sales lift
- Multiple promotions compound effects
- Timing matters (holidays vs regular weeks)
- **Action:** Both amount and presence features needed

5. External Factors Show Weak Correlations

- Unemployment has strongest effect (-0.156)
- Temperature, CPI, fuel price weakly correlated
- Store-level factors more influential
- **Action:** Include but don't over-weight in models

6. Department Performance Varies Widely

- Top 10 departments drive 53% of sales
- High variability in averages and trends
- Department-specific seasonality exists
- **Action:** Department as feature or separate models

7. Outliers are Informative

- 7% of data classified as statistical outliers
- Most represent valid business scenarios
- High sales periods critical to predict
- **Action:** Keep all data, use robust algorithms

Recommendations

For Modeling

1. Feature Engineering Priorities:

- Time-based features (month, quarter, week)
- Lag features (previous weeks' sales)
- Rolling statistics (moving averages)
- Holiday indicators and interactions
- Promotion presence + amount features
- Store type one-hot encoding

2. Model Selection:

- Recommend tree-based models (Random Forest, XGBoost, LightGBM)
- These handle:
 - Non-linear relationships
 - Outliers naturally
 - Feature interactions
 - Seasonal patterns

3. Validation Strategy:

- Time-based split (not random)
- Preserve temporal ordering
- Test on future periods
- Cross-validation by store/department

For Business Strategy

1. Inventory Management:

- Plan for 35-40% Q4 increase
- Pre-position inventory for holiday peaks
- Account for post-holiday returns
- Department-specific strategies needed

2. Promotional Planning:

- Markdowns most effective → prioritize
- Combine promotions during holidays
- Time promotions strategically
- Monitor promotion fatigue

3. Store Operations:

- Type A stores need more resources
- Scale staffing with predicted demand
- Focus on high-performing departments

- Optimize space allocation

4. Risk Management:

- Expect high variability in Q4
 - Plan for negative sales (returns)
 - Monitor external factors (unemployment)
 - Department-level contingencies
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Visualization Summary

The following visualizations were generated during EDA:

Sales Trends (Stage 1.4)

1. 01_overall_sales_trend.png - Time series of total weekly sales
2. 02_sales_by_year.png - Annual sales comparison
3. 03_monthly_seasonality.png - Monthly sales patterns
4. 04_quarterly_pattern.png - Quarterly sales breakdown

Impact Analysis (Stage 1.4)

5. 05_holiday_impact.png - Holiday vs non-holiday comparison
6. 06_store_type_comparison.png - Sales by store type
7. 07_promotion_impact.png - Promotional effectiveness

External Factors (Stage 1.4)

8. 08_external_factors_correlation.png - Correlation heatmap
9. 09_external_factors_scatter.png - Scatter plots vs sales

Department Analysis (Stage 1.4)

10. 10_top_departments.png - Top 10 departments by sales

Outlier Analysis (Stage 1.3)

11. boxplot_sales_by_type.png - Distribution by store type
 12. histogram_sales_distribution.png - Overall distribution with bounds
 13. boxplot_holiday_impact.png - Holiday impact boxplot
 14. scatter_sales_over_time.png - Sales over time with outlier bounds
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Appendix

Data Processing Steps

1. Data loading and merging (Step 1.1)
2. Missing value handling (Step 1.2)
3. Outlier analysis (Step 1.3)
4. Exploratory data analysis (Step 1.4)
5. Feature engineering (Steps 1.3.1-1.3.4)

Files Generated

- **Processed Data:** processed_data/Stage1.3.4_Final/
- **Visualizations:** visualizations/Stage1.3/ and visualizations/Stage1.4/
- **Reports:** EDA_REPORT.md (this file)

Tools Used

- **Python 3.x**
 - **pandas** - Data manipulation
 - **numpy** - Numerical operations
 - **matplotlib** - Plotting
 - **seaborn** - Statistical visualizations
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Conclusion

The exploratory data analysis reveals a rich dataset with strong seasonal patterns, clear business drivers, and substantial predictive potential. The dominant factors influencing sales are:

1. **Seasonality** (Q4 peak)
2. **Holiday effects** (+10-20% lift)
3. **Store characteristics** (type, size)
4. **Promotional activity** (markdowns)
5. **Department variations**

The data is clean, complete, and ready for advanced modeling. Tree-based ensemble methods are recommended given the non-linear relationships, outlier distribution, and feature interactions observed.