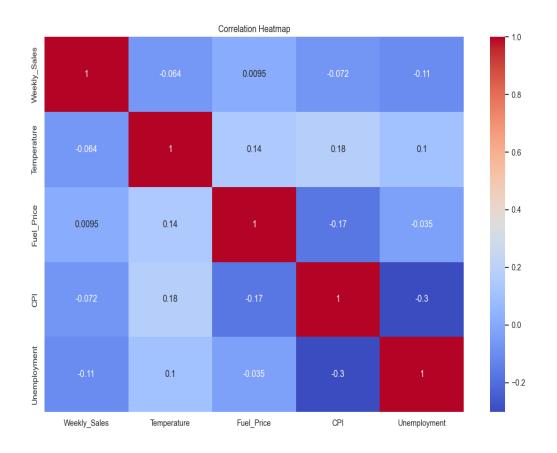
"WALMART SALES ANALYSIS"

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GRAPH DESCRIPTION

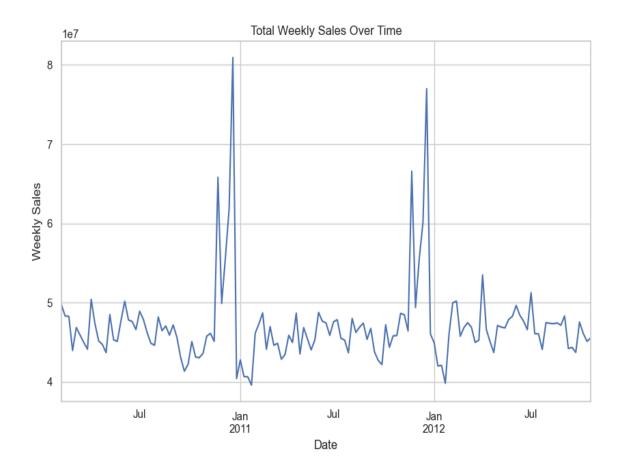
1. CORRELATION HEATMAP



• **Explanation:** This heatmap shows the **correlation** between different variables in the dataset. Correlation is a statistical measure that indicates how closely two variables move together. The values range from -1 to +1. A value close to +1 (dark red) indicates a strong positive correlation (when one variable increases, the other tends to increase). A value close to -1 (dark blue) indicates a strong negative correlation (when one increases, the other decreases). A value close to 0 means there is no significant linear relationship.

- **Insights:** The most important insight from this map is the relationship between **Weekly_Sales** and other variables.
 - Temperature has a weak positive correlation with sales, meaning warmer weather might be associated with slightly higher sales.
 - Fuel_Price and Unemployment have a weak to moderate negative correlation with sales, suggesting that as fuel prices and unemployment rise, sales tend to fall. This makes intuitive sense as these factors reflect a weaker economy and reduced consumer spending.
 - CPI (Consumer Price Index) has a moderate negative correlation with sales, indicating that as the cost-of-living increases, sales might decrease.
- **Future Business Use:** The insights from this heatmap are crucial for building a strong predictive model. They confirm which external factors are most relevant to sales. In the future, this information can be used to:
 - Develop Strategic Pricing: Walmart could adjust prices or promotions in response to changes in fuel prices or unemployment rates.
 - Refine the Predictive Model: The features with the strongest correlations (like Unemployment and CPI) should be prioritized in more complex models to improve forecasting accuracy.
 - Monitor Economic Indicators: The company can keep a close eye on economic data to anticipate shifts in consumer behavior and adjust its business strategy accordingly.

2. TOTAL WEEKLY SALES OVER TIME



Insights

1. Overall Stability with Spikes

- Most weeks, sales stay between \$40M-\$50M.
- However, there are significant spikes around late December and early January in both 2011 and 2012, where sales peak above \$80M.

2. Seasonality

- The spikes align with holiday shopping seasons (Christmas/New Year).
- This indicates highly seasonal demand, with sales doubling during these periods compared to the baseline.

3. Dips in Sales

 Noticeable dips just before and after the spikes, suggesting post-holiday slowdowns.

4. No Strong Upward/Downward Trend

 Apart from seasonal effects, the baseline sales remain relatively flat over time, showing stable but not strongly growing sales.

What It Tells for the Future

1. Inventory & Supply Chain Planning

- Ensure extra stock, logistics, and staff during December

 January to capture peak demand.
- o Avoid overstocking in **February–March** when sales drop back down.

2. Promotions & Marketing

- Invest more in holiday campaigns (Black Friday, Christmas, New Year) since these are proven high-revenue periods.
- Run clearance or loyalty offers post-holiday to smooth out the demand dip.

3. Forecasting

- Use time-series forecasting models (ARIMA, Prophet, or LSTM) that account for seasonality.
- o Predict spikes more accurately for better financial planning.

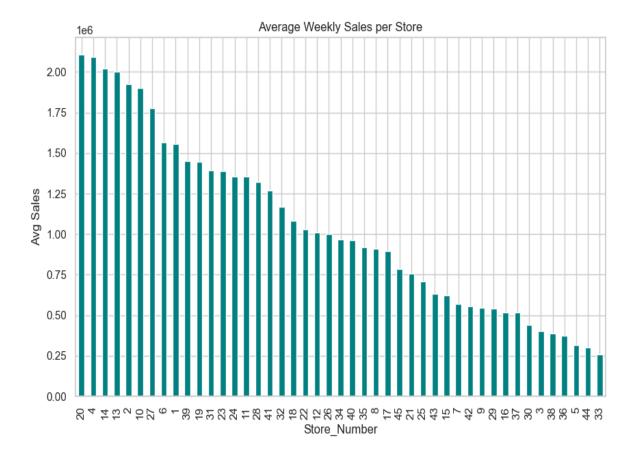
4. Diversification of Revenue

Explore strategies to boost sales in off-peak months (April–September),
 e.g., summer promotions, product bundling, or targeting different customer needs.

5. Operational Efficiency

 Since sales are stable most of the year, the company can optimize resources in normal months and scale up during seasonal demand.

3. AVERAGE WEEKLY SALES PER STORE



Insights

1. High Variability Across Stores

- Some stores (like Store 20, Store 4, Store 14) achieve over \$2M average weekly sales,
- While others (like Store 33, Store 5, Store 44) average only around \$250K-\$400K.
- This indicates a large performance gap between top and bottom stores.

2. Pareto Effect (80/20 Rule)

- A small number of stores contribute disproportionately to overall sales.
- Top ~10 stores generate significantly more revenue compared to the bottom 20+ stores.

3. Potential Store Characteristics

- High-performing stores may be in urban, high-traffic, or economically stronger regions.
- Low-performing stores may face location disadvantages, competition, or operational inefficiencies.

4. Room for Optimization

 If low-performing stores can even partially match the performance of midlevel ones, overall revenue would increase substantially.

What It Tells for the Future

1. Best Practice Replication

- Study what top-performing stores (20, 4, 14, etc.) are doing right location, promotions, product mix, staffing, or customer engagement.
- Apply these strategies to mid- and low-performing stores.

2. Targeted Interventions for Low-Performers

- Diagnose reasons why certain stores (33, 5, 44) underperform.
- o Interventions may include:
 - Better inventory management
 - Regionalized promotions
 - Store layout improvements
 - Training for staff/customer service

3. Resource Allocation

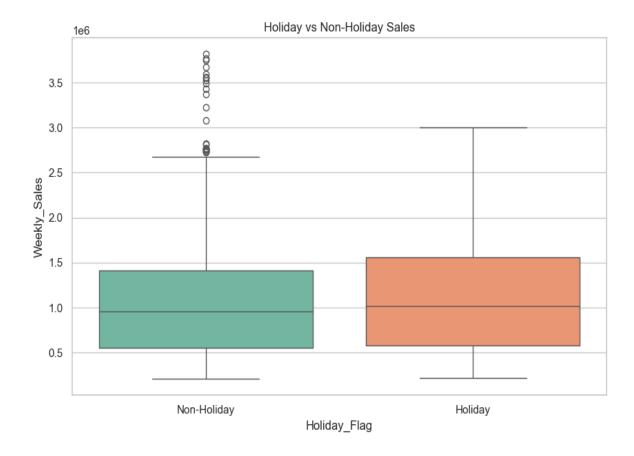
- Invest more marketing and logistics support into top-performing stores, since they deliver higher returns.
- For underperformers, either improve efficiency or evaluate if resources should be reduced/reallocated.

4. Customer Segmentation & Regional Insights

- Conduct regional analysis to see if demographics drive sales differences.
- o Tailor product offerings (e.g., local preferences, seasonal items) by store.

5. Long-Term Strategy

- Consider closing or rebranding persistently underperforming stores if improvements don't yield results.
- Explore new store openings in areas resembling top performers.



Insights

1. Median Sales Comparison

- Holiday weeks have a slightly higher median sales than non-holiday weeks (~\$1M vs ~\$0.9M).
- This confirms holidays do boost sales, but not drastically for the "average" week.

2. Spread (Variation in Sales)

- Holiday weeks show a wider spread (greater variability) in sales, meaning some holidays perform extremely well while others are closer to normal weeks.
- Non-holiday weeks have a **tighter spread**, showing more consistent performance.

3. Outliers in non-Holiday Weeks

- Interestingly, many extreme outliers (above \$3M) appear in the non-holiday weeks.
- These likely align with special events (e.g., Black Friday, promotions) that may not be tagged as "holiday" in the dataset.

4. Upper Bound of Sales

- Holiday weeks reach up to ~\$3M, while some non-holiday weeks reach even higher due to outliers.
- This suggests holidays guarantee some boost, but exceptional spikes can also happen outside official holiday flags.

What It Tells for the Future

1. Holiday Planning

 Holidays bring higher median sales, so stores should prepare with inventory, staffing, and promotions during holiday periods.

2. Event-Based Strategy

 Since large outliers occur in non-holiday weeks, the company should also capitalize on promotional events (Black Friday, back-to-school, clearance sales) that may not fall under "official holidays."

3. Forecasting

- Models should account for holiday flags but also consider other eventdriven spikes.
- A combined seasonality + event calendar model will give better accuracy.

4. Marketing Investments

- Strong marketing campaigns during holidays and special events can maximize the already higher demand.
- Focus particularly on major shopping events that outperform even holiday weeks.

5. ACTUAL VS PREDICTED SALES

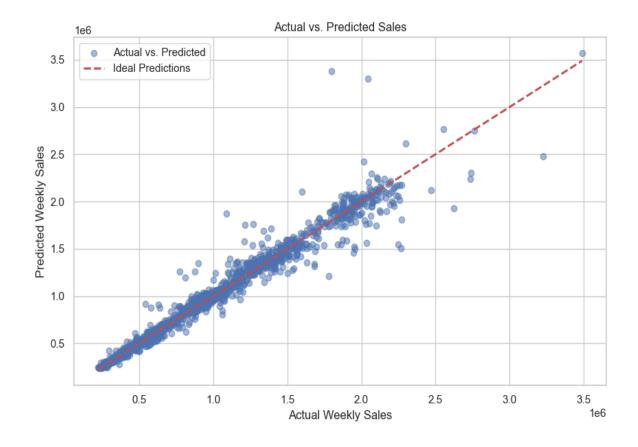
Insights

1. Strong Model Fit

- Most points lie close to the red diagonal line (ideal predictions).
- This means the model predicts sales with high accuracy across most weeks.

2. Good Performance for Majority Cases

- For low-to-medium sales ranges (up to ~\$2M), the predictions align very well with actuals.
- The model captures general sales behavior effectively.



3. Deviation at High Sales

- At higher sales values (> \$2M), predictions start to underestimate actuals.
- The model struggles slightly to capture extreme spikes (e.g., holiday peaks, Black Friday).

4. Few Outliers

 Some scattered points show larger prediction errors. These are likely special event weeks not fully captured by the model.

What It Tells for the Future

1. Model Reliability

- The model can be trusted for baseline and regular forecasting.
- Useful for inventory planning, staffing, and financial projections in non-extreme weeks.

2. Improvement for High Sales Weeks

- Need to enhance the model's ability to capture holiday/event-driven surges.
- Consider adding:
 - Holiday/event flags
 - Promotional campaign data
 - External factors (weather, economic indicators)

3. Business Use Case

- Reliable forecasting allows proactive decision-making:
 - Avoids understocking/overstocking
 - Optimizes workforce allocation
 - Supports accurate revenue projections

4. Continuous Monitoring

 Forecasting should be iteratively retrained with new data to adapt to changing consumer behavior.

CONLUSION:

Walmart's sales data shows a **stable baseline throughout the year**, with **sharp spikes during holidays and special events**, highlighting strong seasonality. A few **top-performing stores drive a major share of revenue**, while many others underperform, suggesting room for targeted improvements. Holiday weeks deliver higher sales, but some of the **biggest spikes occur outside official holidays**, likely due to promotions or events like Black Friday. Forecasting models perform well overall but **underestimate peak sales**, meaning external factors like holidays, promotions, and regional demand should be incorporated.

What to Do Next:

- Prepare inventory, staffing, and logistics well ahead of holiday and event weeks.
- Replicate best practices from top stores and intervene in underperforming ones.
- Use **event-driven forecasting** to better capture peak sales.
- Explore off-season promotions to stabilize sales during slower months.