

Notebook 2: EDA & Visualization Final Project - Ordinal vs Nominal Sentiment Analysis

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Purpose: Exploratory Data Analysis and create visualizations for report.

Input: amazon_electronics_cleaned.csv

Output: class_distribution.png

```
1 from google.colab import drive
2 drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount('/content/drive', force_remount=True)


```
1 # Import libraries
2 import pandas as pd
3 import numpy as np
4 import matplotlib.pyplot as plt
5 import seaborn as sns
6 import warnings
7 warnings.filterwarnings('ignore')
8
9 # Plot settings
10 plt.style.use('seaborn-v0_8-whitegrid')
11 sns.set_palette("husl")
12 plt.rcParams['figure.dpi'] = 150
13
14 print("✅ Libraries imported")
```

✅ Libraries imported

Step 1: Load Cleaned Data

```
1 # Load cleaned data from Notebook 1
2 df = pd.read_csv(r'/content/drive/MyDrive/Fall 2025/Foundations of Artificial Intelligence/EDA/Amazon Electronics Reviews/Amazon Electronics Reviews.csv')
3
4 print(f"✅ Loaded {len(df):,} reviews")
5 print(f"Columns: {list(df.columns)}")
6 df.head()
```

✓ Loaded 49,960 reviews
Columns: ['text', 'rating']

	text	rating	
0	We got this GPS for my husband who is an (OTR)...	5	
1	I'm a professional OTR truck driver, and I bou...	1	
2	Well, what can I say. I've had this unit in m...	3	
3	Not going to write a long review, even thought...	2	
4	I've had mine for a year and here's what we go...	1	

Next steps:

[Generate code with df](#)

[New interactive sheet](#)

Step 2: Class Distribution Visualization

```

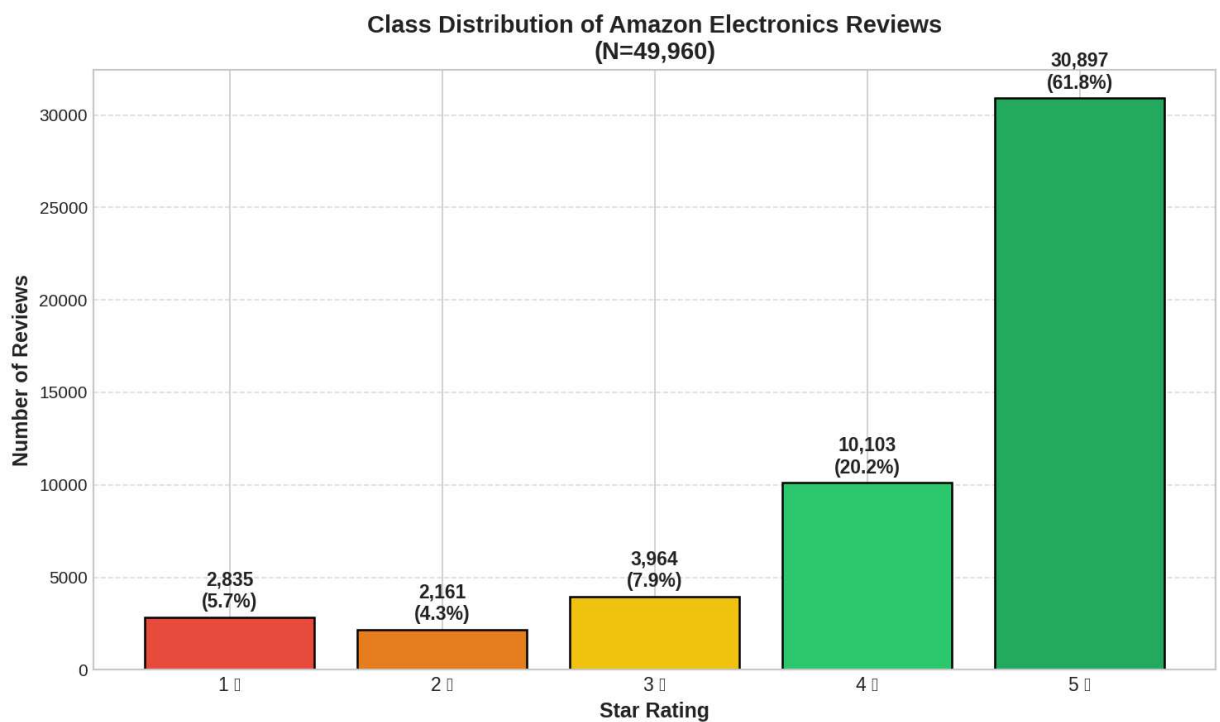
1 # =====
2 # CLASS DISTRIBUTION PLOT (For Report)
3 # =====
4
5 fig, ax = plt.subplots(figsize=(10, 6))
6
7 rating_counts = df['rating'].value_counts().sort_index()
8 colors = ['#e74c3c', '#e67e22', '#f1c40f', '#2ecc71', '#27ae60']
9
10 bars = ax.bar(rating_counts.index, rating_counts.values, color=colors,
11               edgecolor='black', linewidth=1.2)
12
13 # Add value labels on bars
14 for bar, count in zip(bars, rating_counts.values):
15     height = bar.get_height()
16     ax.text(bar.get_x() + bar.get_width()/2, height + max(rating_counts)*
17             f'{count:}, }\n({count/len(df)*100:.1f}%',
18             ha='center', va='bottom', fontsize=11, fontweight='bold')
19
20 ax.set_xlabel('Star Rating', fontsize=12, fontweight='bold')
21 ax.set_ylabel('Number of Reviews', fontsize=12, fontweight='bold')
22 ax.set_title(f'Class Distribution of Amazon Electronics Reviews\n(N={len(
23             fontsize=14, fontweight='bold')
24 ax.set_xticks([1, 2, 3, 4, 5])
25 ax.set_xticklabels(['1 ★', '2 ★', '3 ★', '4 ★', '5 ★'], fontsize=1
26
27 # Add grid
28 ax.yaxis.grid(True, linestyle='--', alpha=0.7)
29 ax.set_axisbelow(True)
30
31 plt.tight_layout()
32 plt.savefig('/content/drive/MyDrive/Fall 2025/Foundations of Artificial I

```

```

33         facecolor='white', edgecolor='none')
34 plt.show()
35
36 print("\n✅ Saved: /content/drive/MyDrive/Fall 2025/Foundations of Artifi

```



✅ Saved: /content/drive/MyDrive/Fall 2025/Foundations of Artificial Intelligence

Step 3: Review Length Analysis

```

1 # Calculate text statistics
2 df['text_length'] = df['text'].str.len()
3 df['word_count'] = df['text'].str.split().str.len()
4
5 print("🔍 Review Length Statistics:")
6 print(df[['text_length', 'word_count']].describe())

```

🔍 Review Length Statistics:

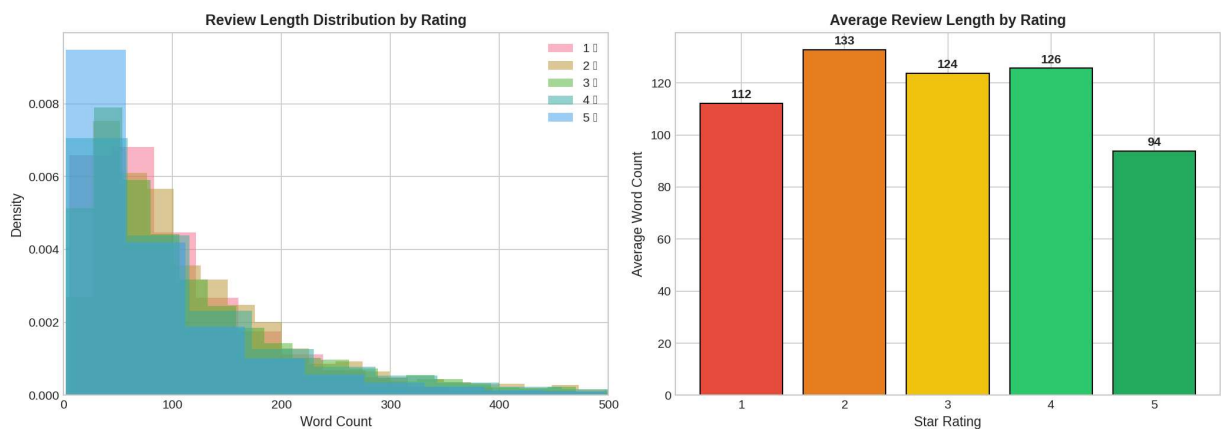
	text_length	word_count
count	49960.000000	49960.000000
mean	577.486930	105.335028

std	698.808926	124.742919
min	10.000000	2.000000
25%	173.000000	32.000000
50%	337.000000	62.500000
75%	698.000000	128.000000
max	15567.000000	2845.000000

```

1 # =====
2 # REVIEW LENGTH BY RATING
3 # =====
4
5 # Ensure 'word_count' is available in case it was dropped prematurely
6 if 'word_count' not in df.columns:
7     df['text_length'] = df['text'].str.len()
8     df['word_count'] = df['text'].str.split().str.len()
9
10 fig, axes = plt.subplots(1, 2, figsize=(14, 5))
11
12 # Word count distribution
13 for rating in [1, 2, 3, 4, 5]:
14     subset = df[df['rating'] == rating]['word_count']
15     axes[0].hist(subset, bins=50, alpha=0.5, label=f'{rating} ★', density
16
17 axes[0].set_xlabel('Word Count', fontsize=11)
18 axes[0].set_ylabel('Density', fontsize=11)
19 axes[0].set_title('Review Length Distribution by Rating', fontsize=12, font
20 axes[0].legend()
21 axes[0].set_xlim([0, 500])
22
23 # Average word count by rating
24 avg_words = df.groupby('rating')['word_count'].mean()
25 bars = axes[1].bar(avg_words.index, avg_words.values, color=colors, edgecol
26 axes[1].set_xlabel('Star Rating', fontsize=11)
27 axes[1].set_ylabel('Average Word Count', fontsize=11)
28 axes[1].set_title('Average Review Length by Rating', fontsize=12, fontweigh
29 axes[1].set_xticks([1, 2, 3, 4, 5])
30
31 for bar, val in zip(bars, avg_words.values):
32     axes[1].text(bar.get_x() + bar.get_width()/2, val + 2, f'{val:.0f}',
33                 ha='center', fontsize=10, fontweight='bold')
34
35 plt.tight_layout()
36 plt.savefig('/content/drive/MyDrive/Fall 2025/Foundations of Artificial Int
37 plt.show()
38
39 print("\n✅ Saved: /content/drive/MyDrive/Fall 2025/Foundations of Artific

```



✓ Saved: /content/drive/MyDrive/Fall 2025/Foundations of Artificial Intelligence

Step 4: Sample Reviews

```
1 # Show sample reviews for each rating
2 print("=" * 70)
3 print("SAMPLE REVIEWS BY RATING")
4 print("=" * 70)
5
6 for rating in [1, 2, 3, 4, 5]:
7     sample = df[df['rating'] == rating]['text'].iloc[0][:200]
8     print(f"\n{'★' * rating} ({rating}-star):")
9     print(f"    {sample}...")
```

```
=====
SAMPLE REVIEWS BY RATING
=====
```

★ (1-star):

I'm a professional OTR truck driver, and I bought a TND 700 at a truck stop h

★★ (2-star):

Not going to write a long review, even though this unit deserves one. I've d

★★★ (3-star):

Well, what can I say. I've had this unit in my truck for about four days now

★★★★ (4-star):

This adapter easily connects my Nook HD 7" to my HDTV through the HDMI ca

★★★★★ (5-star):

We got this GPS for my husband who is an (OTR) over the road trucker. Very I

Step 5: Key Insights

```

1 # =====
2 # KEY INSIGHTS FOR REPORT
3 # =====
4
5 print("=" * 70)
6 print("📋 KEY INSIGHTS FOR REPORT")
7 print("=" * 70)
8
9 rating_counts = df['rating'].value_counts().sort_index()
10
11 print(f"""
12 1. DATASET SIZE:
13     - Total reviews: {len(df):,}
14     - Source: Amazon Electronics Reviews (McAuley Lab, UCSD)
15
16 2. CLASS IMBALANCE:
17     - 5-star reviews: {rating_counts[5]:,} ({rating_counts[5]/len(df)*100:.1
18     - 1-star reviews: {rating_counts[1]:,} ({rating_counts[1]/len(df)*100:.1
19     - Imbalance ratio: {rating_counts[5]/rating_counts.min():.1f}:1
20
21 3. REVIEW LENGTH:
22     - Average words: {df['word_count'].mean():.0f}
23     - Median words: {df['word_count'].median():.0f}
24     - Negative reviews tend to be longer (more detail about complaints)
25
26 4. CHALLENGE FOR CLASSIFICATION:
27     - Adjacent ratings (4★ vs 5★) use similar vocabulary
28     - Class imbalance affects minority class performance
29     - Ordinal structure: 1 < 2 < 3 < 4 < 5
30 """)

```

```

=====
📋 KEY INSIGHTS FOR REPORT
=====

```

1. DATASET SIZE:
 - Total reviews: 49,960
 - Source: Amazon Electronics Reviews (McAuley Lab, UCSD)
2. CLASS IMBALANCE:

- 5-star reviews: 30,897 (61.8%)
- 1-star reviews: 2,835 (5.7%)
- Imbalance ratio: 14.3:1

3. REVIEW LENGTH:

- Average words: 105
- Median words: 62
- Negative reviews tend to be longer (more detail about complaints)

4. CHALLENGE FOR CLASSIFICATION:

- Adjacent ratings (4★ vs 5★) use similar vocabulary
- Class imbalance affects minority class performance
- Ordinal structure: 1 < 2 < 3 < 4 < 5

```
1 # Clean up helper columns
2 df = df.drop(columns=['text_length', 'word_count'], errors='ignore')
3
4 # Images are now saved directly to Google Drive, no need for separate download
5 print("Images have been saved directly to Google Drive.")
```

Images have been saved directly to Google Drive.

```
1 # This cell is no longer needed, as images are now saved directly to
  Google Drive by the plt.savefig commands.
2 print("Images have been saved directly to Google Drive, no further move
  operation is required.")
```

Images have been saved directly to Google Drive, no further move operation is required.

✓ Summary

Visualizations created:

- `class_distribution.png` - For Dataset section of report
- `review_length_analysis.png` - Additional analysis

Key findings:

- Severe class imbalance (5-star dominant)
- Negative reviews are longer on average
- Ordinal structure should be leveraged

Next: Run `3_Models_Nominal.ipynb`

