# AMAZON BEAUTY REVIEW

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### INTRODUCTION

Analzying customer data on Amazon beauty products and products restrictly review textual sentiment, star-based ratings, and the effects on overall brand choice and product popularity.

## Main Question: Is there a correlation between reviews, sentiment, and star ratings?

- Purpose: To analyze customer feedback on Amazon beauty products and uncover patterns between textual sentiment and star-based ratings, time period trends, and overall user's reasoning behind buying certain products and their popularity.
- 2. Beauty product sellers and marketers on Amazon,
  Data analysts studying consumer behavior, and
  Product development teams

## METHODOLOGY

#### Dataset:

- Source: <u>Amazon Reviews 2018</u>
- Each row = One product review

#### **Key Features Used:**

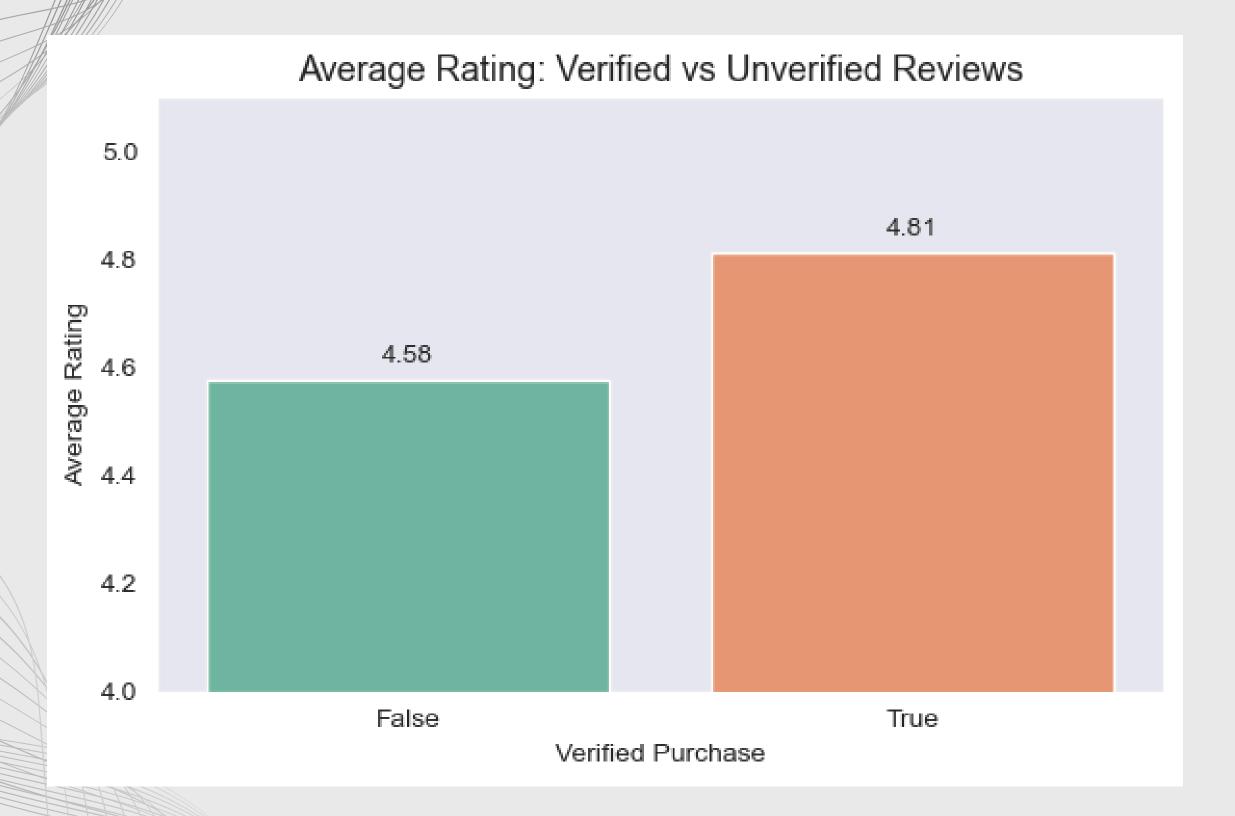
- Star rating (overall)
- Verified purchase (verified)
- Review text (reviewtext)
- Review length

#### Tools:

- Jupyter Notebook
- Python libraries: pandas, seaborn, matplotlib, TextBlob

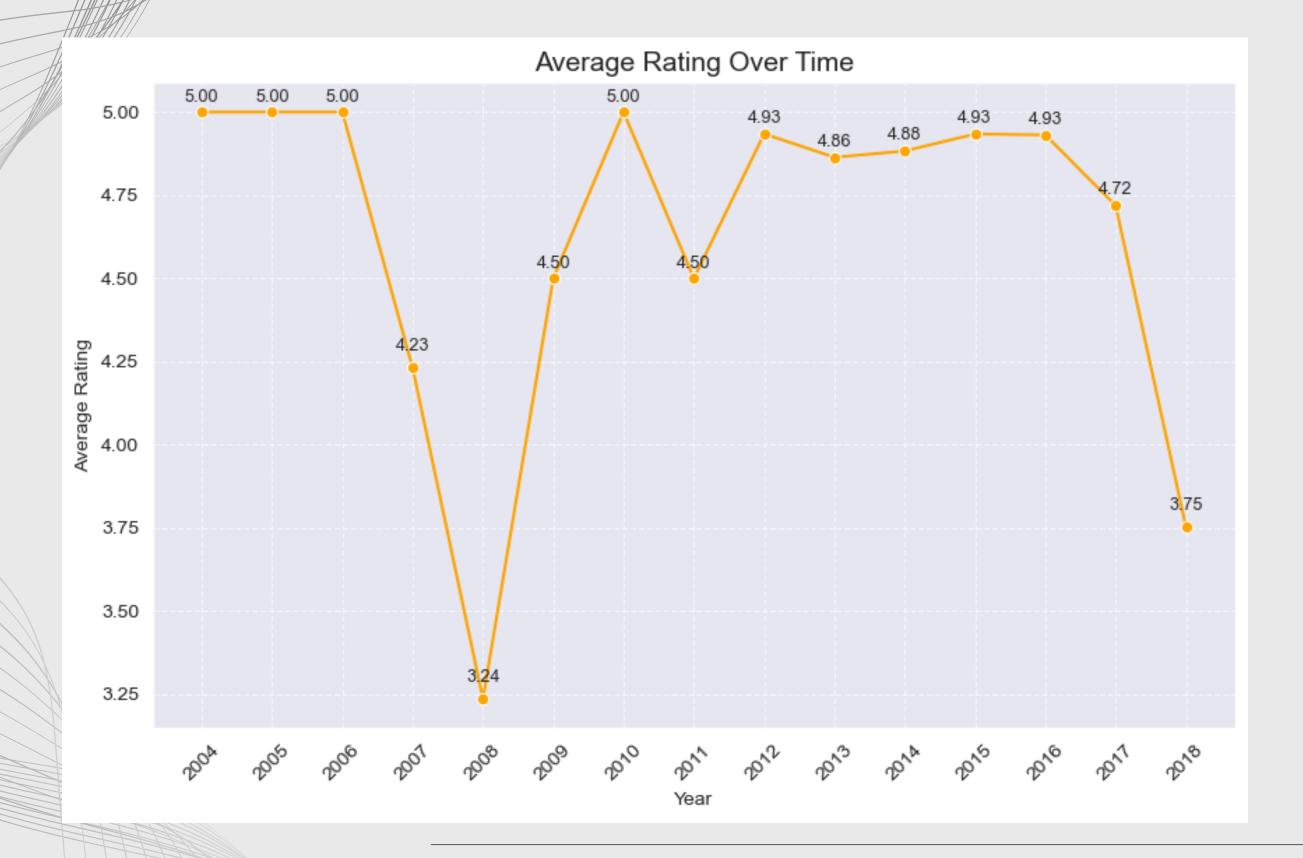
## RESULTS

#### Do verified purchase reviews differ in rating compared to unverified ones?



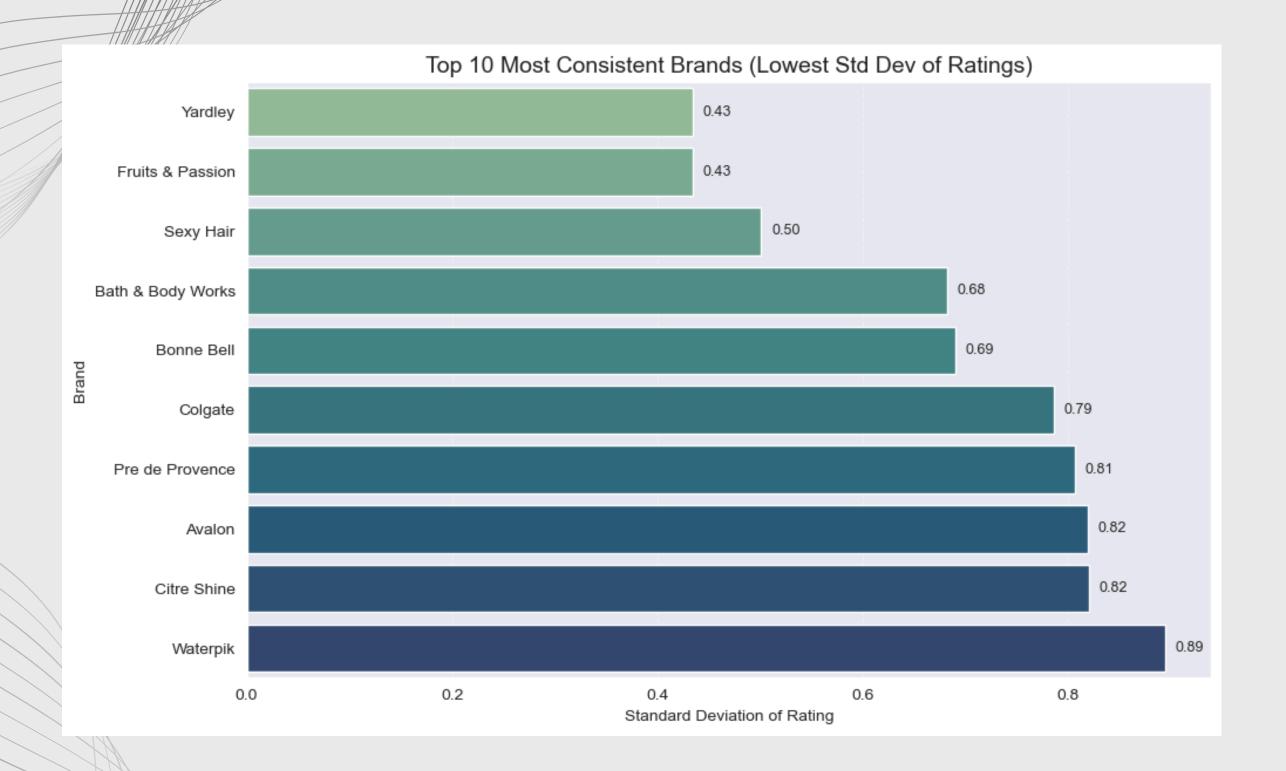
- Verified purchases have a higher average rating than unverified ones.
- Unverified reviews show slightly more extreme (low or high) ratings.
- Verified reviews may reflect more genuine customer experiences.

#### Are there temporal trends in reviews – do ratings trend higher or lower over time?



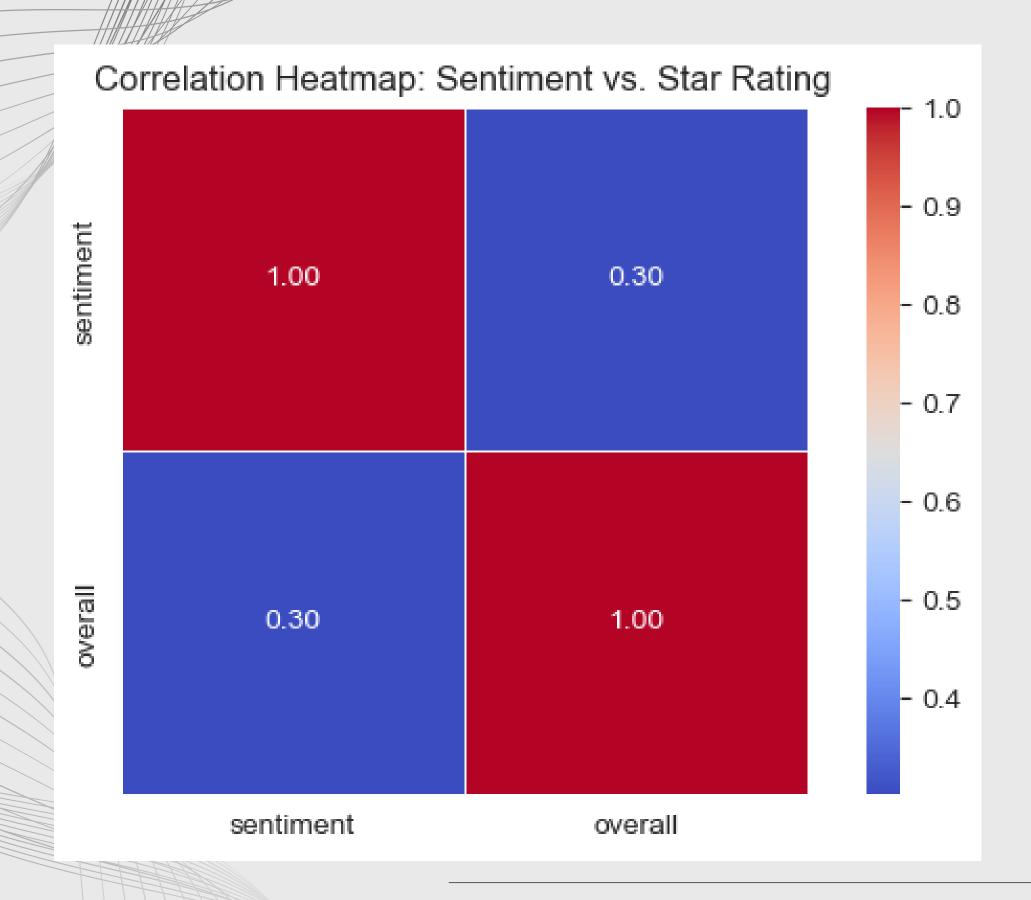
- Ratings fluctuated significantly between 2007–2009, reaching a low in 2008 (3.24)
- More recent years show consistently high ratings, except for a decline in 2018.
- External factors (policy changes, product updates, platform adjustments) may influence rating trends.

#### Which brands receive the most consistent ratings?



- Yardley and Fruits & Passion had the lowest standard deviation (0.43), suggesting consistent customer satisfaction.
- Brands like Waterpik showed higher rating variability.
- Lower standard deviation = more predictable customer satisfaction
   = brand reliability.

#### Is there a correlation between the sentiment of review text and the star rating?



- Sentiment polarity shows a positive correlation (0.30) with star rating.
- Higher sentiment scores generally align with higher star ratings.
- Textual analysis is a useful early signal for product satisfaction before ratings shift.

#### What Do These Insights Mean?

- Focus on Verified Purchase Programs
  - Encourage verified reviews via post-purchase emails or incentives they trend more positive and build trust.
- Monitor Sentiment in Real Time
  - A 0.30 correlation shows sentiment can forecast rating drops. Use automated tools to flag rising negative tone.
- Leverage Consistency as a Brand Asset
  - Brands with stable ratings (like Yardley) should highlight this in marketing copy –
     "Trusted by thousands, consistently."
- Investigate Temporal Spikes
  - Analyze time periods with lower ratings for potential operational or product issues (e.g., changes in formulation, packaging, or logistics).
- Use Standard Deviation to Assess Risk
  - High variability in ratings could point to quality control issues or niche customer satisfaction.

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#### **Future Work**

What new questions emerged from this project?

- \*How do specific product types (e.g., skincare vs. haircare) compare in sentiment and rating consistency?
- Are certain keywords or phrases in reviews strong predictors of low ratings?
- How do ratings correlate with price, brand size, or product rank?
- Are there time-based patterns within specific brands (e.g., did a formula change affect ratings)?

What still needs to be answered?

- How does review helpfulness influence sentiment or rating trends?
- Is there bias in reviews based on product visibility (e.g., top-ranked products getting inflated ratings)?
- Can verified reviews be trusted as proxies for genuine sentiment, or are they more susceptible to incentivization?

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## THANKYOU