

5.1 This is a list of over 34,000 consumer reviews for Amazon products like the Kindle, Fire TV Stick, and more provided by Datafiniti's Product Database. The dataset includes basic product information, rating, review text, and more for each product. This is a sample of a large dataset. The full dataset is available through Datafiniti.

I've used one of the downloaded csv files. It contains 28,332 rows and 24 columns. There are no empty cells or data for the review.text column.

5.2 The code is performing some preprocessing steps on the Consumer Reviews of Amazon Products dataset from Kaggle. The preprocessing steps are:

- **Cleaning:** Removing any errors or missing values in the data. The code uses the `dropna()` method on the `reviews_data` Series to remove any rows that have empty text reviews in the `reviews.text` column. This can improve the quality and accuracy of the data and avoid misleading or biased results.
- **Formatting:** Converting the data into a standard or uniform format that can be easily processed by the computer. The code uses the `pd.read_csv()` function to read the csv file and convert it into a pandas DataFrame, which is a two-dimensional tabular data structure.
- **Sentiment analysis:** Applying the `textblob` component for `spaCy` to calculate the polarity and subjectivity of the text reviews. The code defines a function that takes a text review and returns its sentiment score and label based on the polarity value. The code also tests the function with three sample reviews and prints the results.

5.3 The model is based on the `textblob` component for `spaCy`, which uses the `TextBlob` library to calculate the polarity and subjectivity of texts. The polarity is a float value between -1 and 1, where -1 indicates a negative sentiment, 0 indicates a neutral sentiment, and 1 indicates a positive sentiment. The subjectivity is a float value between 0 and 1, where 0 indicates an objective statement and 1 indicates a subjective statement.

We can see that the function returns the polarity, (if needed can subjectivity), and sentiment label for each review. We can use these results to perform further analysis, such as:

- Calculate the average polarity and subjectivity for each product or category, and compare them to see which products or categories have the highest or lowest customer satisfaction.
- Plot the distribution of polarity and subjectivity scores for each product or category, and identify any outliers or trends that may indicate customer preferences or issues.
- Group the reviews by sentiment label, and perform text analysis to extract the most frequent or relevant words, phrases, or topics that describe the positive, negative, or neutral sentiments.

5.4 Some of the strengths of the model are:

- It is easy to implement and use, as you only need to add the textblob component to your spaCy pipeline and access the doc.\_.polarity and doc.\_.subjectivity attributes.
- It can handle different types of texts, such as reviews, tweets, comments, etc., and provide a quick and simple way to measure their sentiment.
- It can also provide the subjectivity score, which can be useful to identify how factual or opinionated the text is.

Some of the limitations of the model are:

- It relies on a predefined lexicon of words and phrases that have assigned polarity and subjectivity values, which may not capture the nuances and contexts of natural language.
- It does not account for the intensity or degree of the sentiment, such as very positive or slightly negative, which may affect the accuracy and usefulness of the polarity score.
- It does not consider the structure or syntax of the text, such as negations, modifiers, or conjunctions, which may alter the meaning and sentiment of the text.

Resources Used for This Task:

<https://realpython.com/sentiment-analysis-python/>

<https://www.kaggle.com/datasets/datafiniti/consumer-reviews-of-amazon-products>

[https://en.wikipedia.org/wiki/Data\\_Preprocessing](https://en.wikipedia.org/wiki/Data_Preprocessing)

<https://learn.g2.com/data-preprocessing>

<https://monkeylearn.com/blog/data-preprocessing/>

[https://www.w3schools.com/python/pandas/ref\\_df\\_dropna.asp#:~:text=The%20dropna\(\)%20method%20removes,in%20the%20original%20DataFrame%20instead.](https://www.w3schools.com/python/pandas/ref_df_dropna.asp#:~:text=The%20dropna()%20method%20removes,in%20the%20original%20DataFrame%20instead.)