

## ✓ SECTION 6: Packaging Preferences

### 1. Processed Dataset

- Recoded **Q24 How would you rate the quality of the packaging?** & **Q27 If this shampoo changed only its packaging, how likely are you to try it again?**
- Sentiment scores for **Q25 Why do you prefer this type of packaging?/Q26 How Can This Packaging Be Improved**

	Q24 How would you rate the quality of the packaging?	Q24_Label	Q27 If this shampoo changed only its packaging, how likely are you to try it again?	Q27_Label
0		3 Average	5	Extremely likely
1		5 Excellent	4	Likely
2		5 Excellent	5	Extremely likely
3		5 Excellent	3	Neutral
4		5 Excellent	4	Likely

[https://docs.google.com/spreadsheets/d/1ejnuioXFQa2MBZ4P95fUspL6CNPP3AqmDqG\\_eoeqtSc/edit?gid=1073419341#gid=1073419341](https://docs.google.com/spreadsheets/d/1ejnuioXFQa2MBZ4P95fUspL6CNPP3AqmDqG_eoeqtSc/edit?gid=1073419341#gid=1073419341)

### Statistical Reports

#### 1. ANOVA Summary Tables

Used to evaluate whether differences in **packaging quality (Q24)** ratings and **likelihood to repurchase (Q27)** differ across **brands** or other groups (e.g., demographics or ingredients).

**Example ANOVA Output Summary:**

Variable	F-value	p-value	Partial Eta Squared ( $\eta^2$ )
Brand → Packaging Quality	4.35	0.015	0.06
Brand → Likelihood (Q27)	3.89	0.021	0.05

## 2. Mann–Whitney U Test

Used when comparing **two independent groups** (e.g., Male vs. Female) on an **ordinal outcome** like Q27 (likelihood to repurchase).

**Example Result:**

- **U = 5213.0, p = 0.047**

**Interpretation:**

- There's a **significant difference** in repurchase likelihood based on the grouping variable (e.g., gender).
- Since Mann–Whitney is non-parametric, it doesn't assume normal distribution—ideal for ordinal or skewed data.

## 3. Levene's Test – Homogeneity of Variance

Tested whether **brand groups** had equal variance in packaging evaluations.

**Result:**

- **Statistic = 0.70, p = 0.7767**

**Interpretation:**

- The high p-value ( $> 0.05$ ) means **no significant difference in variances** across groups.
- Therefore, **ANOVA assumptions are met** regarding equal variance—supporting the validity of the ANOVA conclusions above.

### Interpretation:

1. There is a statistically significant difference between brands on both variables.
2. **Effect sizes ( $\eta^2 \sim 0.05\text{--}0.06$ )** suggest a **small to medium practical impact**, indicating brand perception influences packaging evaluations.
  - ANOVA summary tables with F, p-values, effect sizes
  - Mann–Whitney U results (U, p)

Levene's test for Brand: Stat=0.70, p=0.7767

Source	Sum of Squares (SS)	df	F-value	p-value (PR > F)
C(Brand)	15.212	14	0.841	0.624
C(ScalpSegment)	2.142	2	0.829	0.437
Residual	752.983	583	—	—

### Interpretation:

- **Brand and scalp segment have no statistically significant effect** on the dependent variable ( $p > 0.05$ ).
- The **model explains very little variance**, as indicated by the low F-values.
- High residual variance (753) suggests the dependent variable may be influenced more by other factors or individual differences.

[https://docs.google.com/spreadsheets/d/1ejnuioXFQa2MBZ4P95fUspL6CNPP3AqmDqG\\_eoeqtSc/edit?gid=1203334407#gid=1203334407](https://docs.google.com/spreadsheets/d/1ejnuioXFQa2MBZ4P95fUspL6CNPP3AqmDqG_eoeqtSc/edit?gid=1203334407#gid=1203334407)

Group 1	Group 2	Mean Difference	p-value (adj)	Lower Bound	Upper Bound	Significant (Reject H <sub>0</sub> )
Mild	Moderate	-0.1512	0.3749	-0.4168	0.1144	No
Mild	Severe	-0.0011	1.0000	-0.2700	0.2678	No
Moderate	Severe	0.1502	0.3791	-0.1151	0.4154	No

### Interpretation:

- No pairwise comparison among the scalp condition groups (Mild, Moderate, Severe) shows a **statistically significant** difference in the mean values.
- All **p-values > 0.05**, and the **confidence intervals include 0**, indicating a lack of effect.

'QualityRating ~ C(Brand)'

Source	Sum of Squares (SS)	df	F-value	p-value (PR > F)
C(Brand)	16.148	14	0.894	0.566
Residual	755.125	585	—	—

### Interpretation:

- The **effect of shampoo brand is not statistically significant** ( $p = 0.566$ ).
- The **F-value (0.894)** is well below the typical cutoff for significance.
- This suggests **no meaningful variation** in the dependent variable can be attributed to brand differences in this context.

### 'QualityRating ~ C(ScalpCondition)'

Source	Sum of Squares (SS)	df	F-value	p-value (PR > F)
C(ScalpCondition)	3.079	2	1.196	0.303
Residual	768.195	597	—	—

### Interpretation:

- The **effect of scalp condition is not statistically significant** ( $p = 0.303$ ).
- The **F-value of 1.196** indicates low between-group variance compared to within-group variance.
- This suggests that **scalp condition (Mild, Moderate, Severe)** does not significantly influence the dependent variable in this model.

**Levene's test:  $W = 0.699$ ,  $p = 0.777$**

**Table 1: Levene's Test for Homogeneity of Variance**

Test	W-value	p-value
Levene's Test	0.699	0.777

**Interpretation:**

- Since  $p = 0.777 > 0.05$ , the assumption of **equal variances** across groups (brands) is met.
- This validates the use of ANOVA for comparing group means.

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**Table 2: One-Way ANOVA Summary – Brand**

Source	F-value	p-value	Partial Eta Squared ( $\eta^2$ )
Brand	0.894	0.566	0.021

**Interpretation:**

- The effect of **Brand** is **not statistically significant** ( $p = 0.566$ ).
  - The **effect size** ( $\eta^2 = 0.021$ ) is **small**, indicating that only about 2.1% of the variance in the outcome can be attributed to differences between brands.
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# Description:

## Task Description

Extract only adjectives from qualitative responses (e.g., product feedback) to analyze descriptive sentiment, texture, or user perception.

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## Why This Matters

Adjectives are key indicators of sentiment and perception in open-ended responses. Filtering for adjectives allows you to:

- Isolate qualitative descriptors like “*smooth*,” “*sticky*,” “*natural*”
  - Conduct targeted sentiment or brand attribute analysis
  - Visualize frequently used descriptive terms in word clouds or bar charts
  - Enhance emotional profiling or psychographic segmentation
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## How to Do It

1. **Import necessary NLTK functions** for tokenization and part-of-speech (POS) tagging.
2. **Download NLTK resources** (only once per environment).
3. **Define a function** that:
  - Tokenizes the response text
  - Tags each word with a part of speech
  - Filters out adjectives (**JJ**, **JJR**, **JJS**)
  - Joins and returns them as a single string

4. **Apply this function** to the column with open-ended responses using `.apply()`
5. **Save the extracted adjectives** in a new DataFrame column for further analysis.

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### Tools/Modules

- `nltk.tokenize.word_tokenize` – for breaking text into words
- `nltk.pos_tag` – for tagging each word with its part of speech
- `nltk.download('punkt')` – tokenizer model
- `nltk.download('averaged_perceptron_tagger')` – POS tagger model
- `pandas` – to apply the function across rows of the DataFrame

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### Output

- A new column (e.g., `Q15_Adjectives`) in your DataFrame that contains adjectives extracted from the original open-ended text.
- Example:

Q15_OpenEnd	Q15_Adjectives
"The cream is very smooth and light"	smooth light
"Feels sticky but smells natural"	sticky natural
"Packaging is bold, but texture is harsh"	bold harsh