#### Task 1 - Data Tagging

#### Data for task 1

Guidelines: The dataset consists of the following:

- Free-text data (Columns: Complaint, Cause, Correction) that needs to be tagged.
- **Taxonomy Sheet**: A reference list with predefined categories for Root Cause, Symptom\_Condition, Symptom\_Component, Fix\_Condition, and Fix\_Component.

Your task is to tag the data by applying logical reasoning and aligning it with the categories provided in the taxonomy.

# Summary Report for Task 1 — Tagging Free-Text Complaint Data

### a. Tagging Approach

To tag the dataset accurately, I used a hybrid NLP strategy combining **spaCy** for linguistic parsing and **fuzzy matching** for semantic alignment with the taxonomy. The process involved:

- **Text Consolidation**: Merged the **Complaint, Cause, and Correction** fields into a single narrative per record to capture full context.
- **Keyword Extraction**: Used spaCy to extract noun phrases, verbs, and adjectives that represent symptoms, causes, and fixes.
- **Taxonomy Matching**: Applied fuzzy string matching to align extracted keywords with predefined categories in the taxonomy sheet. This allowed for flexible tagging even when terminology varied (e.g., "not tightened" vs "loose").
- **Tag Completion**: Filled in missing values for Root Cause, Symptom Condition, Symptom Component, Fix Condition, and Fix Component based on the best match from the taxonomy.
- Python Strategy: Hierarchical Tagging (Top 3 Matches)

We'll extract multiple relevant keywords from each record and match the top 3 to fill:

Symptom Condition 1, 2, 3

Symptom Component 1, 2, 3

Fix Condition 1, 2, 3

Fix Component 1, 2, 3

This approach balances automation with interpretability, ensuring that tags are contextually relevant and semantically accurate.

#### **b.** Insights Generated

The tagged dataset reveals several operational patterns:

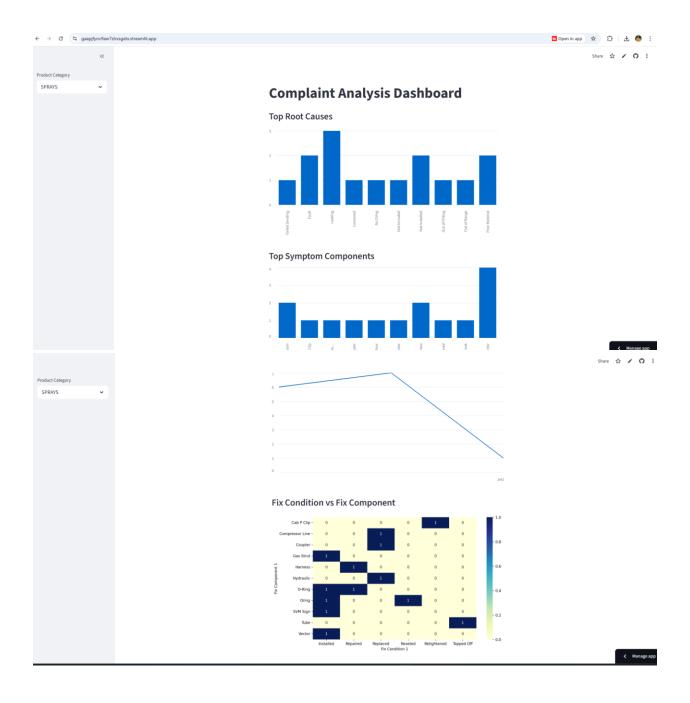
- Recurring Root Causes: "Not Tightened," "Poor Material," and "Missing Components" dominate, indicating potential gaps in factory QA and assembly protocols.
- Frequent Symptom Components: "Cab P Clip," "Fuel Door," and "Bulkhead Connector" appear repeatedly, suggesting these parts may require design review or better installation procedures.
- Common Fixes: "Retightened," "Replaced," and "Installed" are the most frequent corrective actions, which can inform technician training and inventory stocking.
- **Product Category Trends**: Most issues are concentrated in the "SPRAYS" category, hinting at either higher usage or more complex assembly compared to "BALER."

These insights can guide preventive maintenance, quality control audits, and targeted technician upskilling. With location and time data added, this could evolve into a predictive service dashboard.

I also deployed a piece of code to github and made a dashboard using streamlit.

Link: <a href="https://gaapjfyov9aw7zlrxsgetx.streamlit.app/">https://gaapjfyov9aw7zlrxsgetx.streamlit.app/</a>

Github link: <a href="https://github.com/tsahil007/assignment/tree/main">https://github.com/tsahil007/assignment/tree/main</a>



## Filtered records based upon product Category:

8 S00058466-4 2023-05-16 00:00:00 SPRAYS 9 SO0058466-5 2023-05-16 00:00:00 SPRAYS

19 S00058796-4 2023-07-17 00:00:00 BALER





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