

```
pip install pandas openpyxl matplotlib seaborn numpy
```

Requirement already satisfied: pandas in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (2.2.3)

Requirement already satisfied: openpyxl in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (3.1.5)

Requirement already satisfied: matplotlib in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (3.10.1)

Requirement already satisfied: seaborn in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (0.13.2)

Requirement already satisfied: numpy in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (2.2.3)

Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from pandas) (2.9.0.post0)

Requirement already satisfied: pytz>=2020.1 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from pandas) (2025.1)

Requirement already satisfied: tzdata>=2022.7 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from pandas) (2025.1)

Requirement already satisfied: et-xmlfile in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from openpyxl) (2.0.0)

Requirement already satisfied: contourpy>=1.0.1 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from matplotlib) (1.3.1)

Requirement already satisfied: cycler>=0.10 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from matplotlib) (0.12.1)

Requirement already satisfied: fonttools>=4.22.0 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from matplotlib) (4.56.0)

Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from matplotlib) (1.4.8)

Requirement already satisfied: packaging>=20.0 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from matplotlib) (24.2)

Requirement already satisfied: pillow>=8 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from matplotlib) (11.1.0)

Requirement already satisfied: pyparsing>=2.3.1 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from matplotlib) (3.2.1)

Requirement already satisfied: six>=1.5 in c:\users\91952\appdata\local\programs\python\python311\lib\site-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)

Note: you may need to restart the kernel to use updated packages.

```
[notice] A new release of pip available: 22.3.1 -> 25.1.1
[notice] To update, run: python.exe -m pip install --upgrade pip
```

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np

print("Libraries loaded successfully.")
```

Libraries loaded successfully.

```
df = pd.read_excel(r"C:\Users\91952\Documents\ACADEMICS\Programming
1x\Projects\Axion Assessment\Data for Task 1.xlsx")
```

Basic overview of the data

```
df.head()
```

	VIN	TRANSACTION_ID	\
0	3HCFDDE89SH220903	13021	
1	1HRFFEE8XSZ230636	13028	
2	1HYKSMRK6SZ000990	13035	
3	3HCFDFEL3SH241701	13021	
4	1HRFFHEL1RZ181474	13021	

	CORRECTION_VERBATIM	\
0	REPLACED STEERING WHEEL NOW OKAY	
1	CHECKED - FOUND DTC'S U0229 - U1530 SET IN BCM...	
2	APPROVED 4.9(OLH) FOR ADDED DIAGNOSTICS WITH T...	
3	STEERING WHEEL REPLACEMENT	
4	REPLACED STEERING MESSAGE NO LONGER DISPLAYED	

	CUSTOMER_VERBATIM	REPAIR_DATE	\
0	STEERING WHEEL COMING APART	2024-01-02	
1	CUSTOMER STATES HEATED STEERING WHEEL INOP	2024-01-03	
2	OWNER REPORTS: THE SUPER CRUISE BAR ON THE STE...	2024-01-04	
3	CUSTOMER STATES THE LETTERING AND FINISH ON TH...	2024-01-04	
4	C/S: CUSTOMER STATES THE SERVICE DRIVER ASSIST...	2024-01-05	

	CAUSAL_PART_NM	
GLOBAL_LABOR_CODE_DESCRIPTION	\	
0	WHEEL ASM-STRG *JET BLACK	Steering Wheel Replacement
1	MODULE ASM-STRG WHL HT CONT	Heated Steering Wheel Module Replacement
2	WHEEL ASM-STRG *BACKEN BLACKK	Steering Wheel Replacement
3	WHEEL ASM-STRG *JET BLACK	Steering Wheel Replacement

4 WHEEL ASM-STRG *JET BLACK Steering Wheel Replacement

	PLATFORM	BODY_STYLE	VPPC	...
TRANSMISSION_TRACE_NBR \				
0 Full-Size Trucks		Crew Cab	T1CCF	...
S2210121CNJX0941				
1 Full-Size Trucks		Crew Cab	T1CGF	...
R2210881CNJX0287				
2	BEV	4 Door Utility	L233-LS0P	...
NaN				
3 Full-Size Trucks		Crew Cab	T1CCF	...
S1210822CKJX0291				
4 Full-Size Trucks		Crew Cab	T1CGF	...
R2212982CKJX0282				

	SRC_TXN_ID	SRC_VER_NBR	TRANSACTION_CNTR	MEDIA_FLAG	VIN_MODL_DESGTR
\					
0	2808908219	6	1	N	CF10543
1	2808841910	6	1	Y	TF10543
2	2809979441	4	1	Y	6MB26
3	2808892288	6	1	Y	CF10543
4	2808901882	8	1	N	TF10543

	LINE_SERIES	LAST_KNOWN_DELVRY_TYPE_CD	NON_CAUSAL_PART_QTY
SALES_REGION_CODE			
0	1500	21.0	0
1			
1	1500	10.0	0
1			
2	Lux-1	10.0	0
1			
3	1500	10.0	0
1			
4	1500	10.0	0
1			

[5 rows x 52 columns]

Shape, Column Names and Types

```
print("Shape of the dataset:", df.shape)
print("\nColumn names:")
print(df.columns.tolist())
```

```
df.info()
```

```
Shape of the dataset: (100, 52)
```

```
Column names:
```

```
['VIN', 'TRANSACTION_ID', 'CORRECTION_VERBATIM', 'CUSTOMER_VERBATIM',  
'REPAIR_DATE', 'CAUSAL_PART_NM', 'GLOBAL_LABOR_CODE_DESCRIPTION',  
'PLATFORM', 'BODY_STYLE', 'VPPC', 'PLANT', 'BUILD_COUNTRY',  
'LAST_KNOWN_DLR_NAME', 'LAST_KNOWN_DLR_CITY', 'REPAIRING_DEALER_CODE',  
'DEALER_NAME', 'REPAIR_DLR_CITY', 'STATE', 'DEALER_REGION',  
'REPAIR_DLR_POSTAL_CD', 'REPAIR_AGE', 'KM', 'COMPLAINT_CD_CSI',  
'COMPLAINT_CD', 'VEH_TEST_GRP', 'COUNTRY_SALE_ISO',  
'ORD_SELLING_SRC_CD', 'OPTN_FAMILY_CERTIFICATION',  
'OPTF_FAMILY_EMISSION_SYSTEM', 'GLOBAL_LABOR_CODE',  
'TRANSACTION_CATEGORY', 'CAMPAIGN_NBR', 'REPORTING_COST', 'TOTALCOST',  
'LBR_COST', 'ENGINE', 'ENGINE_DESC', 'TRANSMISSION',  
'TRANSMISSION_DESC', 'ENGINE_SOURCE_PLANT', 'ENGINE_TRACE_NBR',  
'TRANSMISSION_SOURCE_PLANT', 'TRANSMISSION_TRACE_NBR', 'SRC_TXN_ID',  
'SRC_VER_NBR', 'TRANSACTION_CNTR', 'MEDIA_FLAG', 'VIN_MODL_DESGTR',  
'LINE_SERIES', 'LAST_KNOWN_DELVRY_TYPE_CD', 'NON_CAUSAL_PART_QTY',  
'SALES_REGION_CODE']
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 100 entries, 0 to 99
```

```
Data columns (total 52 columns):
```

#	Column	Non-Null Count	Dtype
0	VIN	100 non-null	object
1	TRANSACTION_ID	100 non-null	int64
2	CORRECTION_VERBATIM	100 non-null	object
3	CUSTOMER_VERBATIM	100 non-null	object
4	REPAIR_DATE	100 non-null	datetime64[ns]
5	CAUSAL_PART_NM	95 non-null	object
6	GLOBAL_LABOR_CODE_DESCRIPTION	100 non-null	object
7	PLATFORM	100 non-null	object
8	BODY_STYLE	100 non-null	object
9	VPPC	100 non-null	object
10	PLANT	99 non-null	object
11	BUILD_COUNTRY	100 non-null	object
12	LAST_KNOWN_DLR_NAME	100 non-null	object
13	LAST_KNOWN_DLR_CITY	100 non-null	object
14	REPAIRING_DEALER_CODE	100 non-null	object
15	DEALER_NAME	100 non-null	object
16	REPAIR_DLR_CITY	100 non-null	object
17	STATE	98 non-null	object
18	DEALER_REGION	100 non-null	int64
19	REPAIR_DLR_POSTAL_CD	98 non-null	object
20	REPAIR_AGE	100 non-null	int64
21	KM	100 non-null	int64
22	COMPLAINT_CD_CSI	100 non-null	int64

23	COMPLAINT_CD	100	non-null	object
24	VEH_TEST_GRP	98	non-null	object
25	COUNTRY_SALE_ISO	100	non-null	object
26	ORD_SELLING_SRC_CD	100	non-null	int64
27	OPTN_FAMLY_CERTIFICATION	90	non-null	object
28	OPTF_FAMLY_EMISSION_SYSTEM	95	non-null	object
29	GLOBAL_LABOR_CODE	100	non-null	int64
30	TRANSACTION_CATEGORY	100	non-null	object
31	CAMPAIGN_NBR	0	non-null	float64
32	REPORTING_COST	100	non-null	float64
33	TOTALCOST	100	non-null	float64
34	LBRCOST	100	non-null	float64
35	ENGINE	100	non-null	object
36	ENGINE_DESC	100	non-null	object
37	TRANSMISSION	100	non-null	object
38	TRANSMISSION_DESC	100	non-null	object
39	ENGINE_SOURCE_PLANT	88	non-null	object
40	ENGINE_TRACE_NBR	88	non-null	object
41	TRANSMISSION_SOURCE_PLANT	88	non-null	float64
42	TRANSMISSION_TRACE_NBR	88	non-null	object
43	SRC_TXN_ID	100	non-null	int64
44	SRC_VER_NBR	100	non-null	int64
45	TRANSACTION_CNTR	100	non-null	int64
46	MEDIA_FLAG	100	non-null	object
47	VIN_MODL_DESGTR	100	non-null	object
48	LINE_SERIES	99	non-null	object
49	LAST_KNOWN_DELVRY_TYPE_CD	98	non-null	float64
50	NON_CAUSAL_PART_QTY	100	non-null	int64
51	SALES_REGION_CODE	100	non-null	int64
dtypes: datetime64[ns](1), float64(6), int64(12), object(33)				
memory usage: 40.8+ KB				
df.describe()				
	TRANSACTION_ID	REPAIR_DATE	DEALER_REGION	REPAIR_AGE
\count	100.000000	100	100.000000	100.000000
mean	13036.900000	2024-01-22 07:26:24	1.090000	14.940000
min	13021.000000	2024-01-02 00:00:00	1.000000	0.000000
25%	13027.750000	2024-01-12 00:00:00	1.000000	5.000000
50%	13036.000000	2024-01-24 12:00:00	1.000000	12.000000
75%	13041.250000	2024-02-01 00:00:00	1.000000	21.000000
max	13081.000000	2024-02-07 00:00:00	4.000000	50.000000

std	12.028166	NaN	0.51434	12.367945
-----	-----------	-----	---------	-----------

	KM	COMPLAINT_CD_CSI	ORD_SELLING_SRC_CD
GLOBAL_LABOR_CODE \			
count	100.000000	100.0	100.000000
100.000000			
mean	24914.230000	0.0	24.590000
251.900000			
min	3.000000	0.0	11.000000
20.000000			
25%	8883.250000	0.0	13.000000
130.000000			
50%	21962.000000	0.0	13.000000
130.000000			
75%	35493.250000	0.0	48.000000
130.000000			
max	107905.000000	0.0	72.000000
2400.000000			
std	20747.078206	0.0	17.822976
546.451722			

	CAMPAIGN_NBR	REPORTING_COST	TOTALCOST	LBRCOST \
count	0.0	100.000000	100.000000	100.000000
mean	NaN	531.193200	563.321400	106.344900
min	NaN	27.690000	27.690000	20.000000
25%	NaN	305.432500	320.930000	61.855000
50%	NaN	433.970000	455.805000	78.560000
75%	NaN	554.062500	604.695000	108.055000
max	NaN	2457.450000	3205.450000	1012.670000
std	NaN	411.161608	456.075552	113.223074

	TRANSMISSION_SOURCE_PLANT	SRC_TXN_ID	SRC_VER_NBR
TRANSACTION_CNTR \			
count	8.800000e+01	1.000000e+02	100.000000
100.0			
mean	1.676383e+08	2.815767e+09	5.720000
1.0			
min	2.878270e+05	2.808842e+09	2.000000
1.0			
25%	2.878270e+05	2.809436e+09	4.000000
1.0			
50%	8.042172e+06	2.820097e+09	4.000000
1.0			
75%	1.774929e+07	2.820880e+09	6.000000
1.0			
max	8.282984e+08	2.823000e+09	26.000000
1.0			
std	3.237538e+08	5.790727e+06	4.040402
0.0			

SALES_REGION_CODE	LAST_KNOWN_DELVRY_TYPE_CD	NON_CAUSAL_PART_QTY
count	98.000000	100.000000
100.00000		
mean	14.132653	0.070000
1.09000		
min	10.000000	0.000000
1.00000		
25%	10.000000	0.000000
1.00000		
50%	10.000000	0.000000
1.00000		
75%	16.000000	0.000000
1.00000		
max	37.000000	1.000000
4.00000		
std	6.694570	0.256432
0.51434		

Check Missing Values and Unique Entries

```
for col in df.columns:
    print(f"--- {col} ---")
    print(f"Missing: {df[col].isnull().sum()}")
    print(f"Unique: {df[col].nunique()}")
    print()

--- VIN ---
Missing: 0
Unique: 98

--- TRANSACTION_ID ---
Missing: 0
Unique: 32

--- CORRECTION_VERBATIM ---
Missing: 0
Unique: 93

--- CUSTOMER_VERBATIM ---
Missing: 0
Unique: 100

--- REPAIR_DATE ---
Missing: 0
Unique: 29

--- CAUSAL_PART_NM ---
Missing: 5
```

Unique: 18

--- GLOBAL_LABOR_CODE_DESCRIPTION ---

Missing: 0

Unique: 4

--- PLATFORM ---

Missing: 0

Unique: 11

--- BODY_STYLE ---

Missing: 0

Unique: 6

--- VPPC ---

Missing: 0

Unique: 26

--- PLANT ---

Missing: 1

Unique: 11

--- BUILD_COUNTRY ---

Missing: 0

Unique: 3

--- LAST_KNOWN_DLR_NAME ---

Missing: 0

Unique: 100

--- LAST_KNOWN_DLR_CITY ---

Missing: 0

Unique: 94

--- REPAIRING_DEALER_CODE ---

Missing: 0

Unique: 95

--- DEALER_NAME ---

Missing: 0

Unique: 100

--- REPAIR_DLR_CITY ---

Missing: 0

Unique: 93

--- STATE ---

Missing: 2

Unique: 39

--- DEALER_REGION ---

Missing: 0

Unique: 2

--- REPAIR_DLR_POSTAL_CD ---

Missing: 2

Unique: 92

--- REPAIR_AGE ---

Missing: 0

Unique: 35

--- KM ---

Missing: 0

Unique: 100

--- COMPLAINT_CD_CSI ---

Missing: 0

Unique: 1

--- COMPLAINT_CD ---

Missing: 0

Unique: 7

--- VEH_TEST_GRP ---

Missing: 2

Unique: 23

--- COUNTRY_SALE_ISO ---

Missing: 0

Unique: 6

--- ORD_SELLING_SRC_CD ---

Missing: 0

Unique: 7

--- OPTN_FAMLY_CERTIFICATION ---

Missing: 10

Unique: 3

--- OPTF_FAMLY_EMISSION_SYSTEM ---

Missing: 5

Unique: 8

--- GLOBAL_LABOR_CODE ---

Missing: 0

Unique: 4

--- TRANSACTION_CATEGORY ---

Missing: 0

Unique: 2

--- CAMPAIGN_NBR ---

Missing: 100

Unique: 0

--- REPORTING_COST ---

Missing: 0

Unique: 100

--- TOTALCOST ---

Missing: 0

Unique: 100

--- LBRCOST ---

Missing: 0

Unique: 99

--- ENGINE ---

Missing: 0

Unique: 12

--- ENGINE_DESC ---

Missing: 0

Unique: 12

--- TRANSMISSION ---

Missing: 0

Unique: 19

--- TRANSMISSION_DESC ---

Missing: 0

Unique: 20

--- ENGINE_SOURCE_PLANT ---

Missing: 12

Unique: 9

--- ENGINE_TRACE_NBR ---

Missing: 12

Unique: 88

--- TRANSMISSION_SOURCE_PLANT ---

Missing: 12

Unique: 6

--- TRANSMISSION_TRACE_NBR ---

Missing: 12

Unique: 88

```

--- SRC_TXN_ID ---
Missing: 0
Unique: 100

--- SRC_VER_NBR ---
Missing: 0
Unique: 10

--- TRANSACTION_CNTR ---
Missing: 0
Unique: 1

--- MEDIA_FLAG ---
Missing: 0
Unique: 2

--- VIN_MODL_DESGTR ---
Missing: 0
Unique: 41

--- LINE_SERIES ---
Missing: 1
Unique: 22

--- LAST_KNOWN_DELVRY_TYPE_CD ---
Missing: 2
Unique: 11

--- NON_CAUSAL_PART_QTY ---
Missing: 0
Unique: 2

--- SALES_REGION_CODE ---
Missing: 0
Unique: 2

```

View all column names with nulls only

```

null_cols = df.columns[df.isnull().any()]
df[null_cols].isnull().sum()

```

CAUSAL_PART_NM	5
PLANT	1
STATE	2
REPAIR_DLR_POSTAL_CD	2
VEH_TEST_GRP	2
OPTN_FAMLY_CERTIFICATION	10
OPTF_FAMLY_EMISSION_SYSTEM	5
CAMPAIGN_NBR	100

```
ENGINE_SOURCE_PLANT      12
ENGINE_TRACE_NBR         12
TRANSMISSION_SOURCE_PLANT 12
TRANSMISSION_TRACE_NBR   12
LINE_SERIES              1
LAST_KNOWN_DELVRY_TYPE_CD 2
dtype: int64
```

Drop rows where Transaction ID values (key identifier) is missing

```
df = df.dropna(subset=['TRANSACTION_ID'])
```

Convert All Columns to Lowercase

```
df.columns = df.columns.str.strip().str.lower()
```

Fill Nulls Only for Columns That Exist

```
df.columns.tolist()

['vin',
 'transaction_id',
 'correction_verbatim',
 'customer_verbatim',
 'repair_date',
 'causal_part_nm',
 'global_labor_code_description',
 'platform',
 'body_style',
 'vppc',
 'plant',
 'build_country',
 'last_known_dlr_name',
 'last_known_dlr_city',
 'repairing_dealer_code',
 'dealer_name',
 'repair_dlr_city',
 'state',
 'dealer_region',
 'repair_dlr_postal_cd',
 'repair_age',
 'km',
 'complaint_cd_csi',
 'complaint_cd',
 'veh_test_grp',
 'country_sale_iso',
 'ord_selling_src_cd',
 'optn_famly_certification',
 'optf_famly_emissiof_system',
 'global_labor_code',
```

```

'transaction_category',
'campaign_nbr',
'reporting_cost',
'totalcost',
'lbrcost',
'engine',
'engine_desc',
'transmission',
'transmission_desc',
'engine_source_plant',
'engine_trace_nbr',
'transmission_source_plant',
'transmission_trace_nbr',
'src_txn_id',
'src_ver_nbr',
'transaction_cntr',
'media_flag',
'vin_modl_desgtr',
'line_series',
'last_known_delvry_type_cd',
'non_causal_part_qty',
'sales_region_code']

minor_null_cols = [
    'causal_part_nm', 'plant', 'state', 'repair_dlr_postal_cd',
    'veh_test_grp', 'optn_famly_certification',
    'optf_famly_emissiof_system', 'engine_source_plant',
    'engine_trace_nbr', 'transmission_source_plant',
    'transmission_trace_nbr', 'line_series',
    'last_known_delvry_type_cd'
]

# Fill "unknown" only if column exists
for col in minor_null_cols:
    if col in df.columns:
        df[col] = df[col].fillna("unknown")
    else:
        print(f" Column not found: {col}")

df.rename(columns={'causal_part_nm': 'component'}, inplace=True)

```

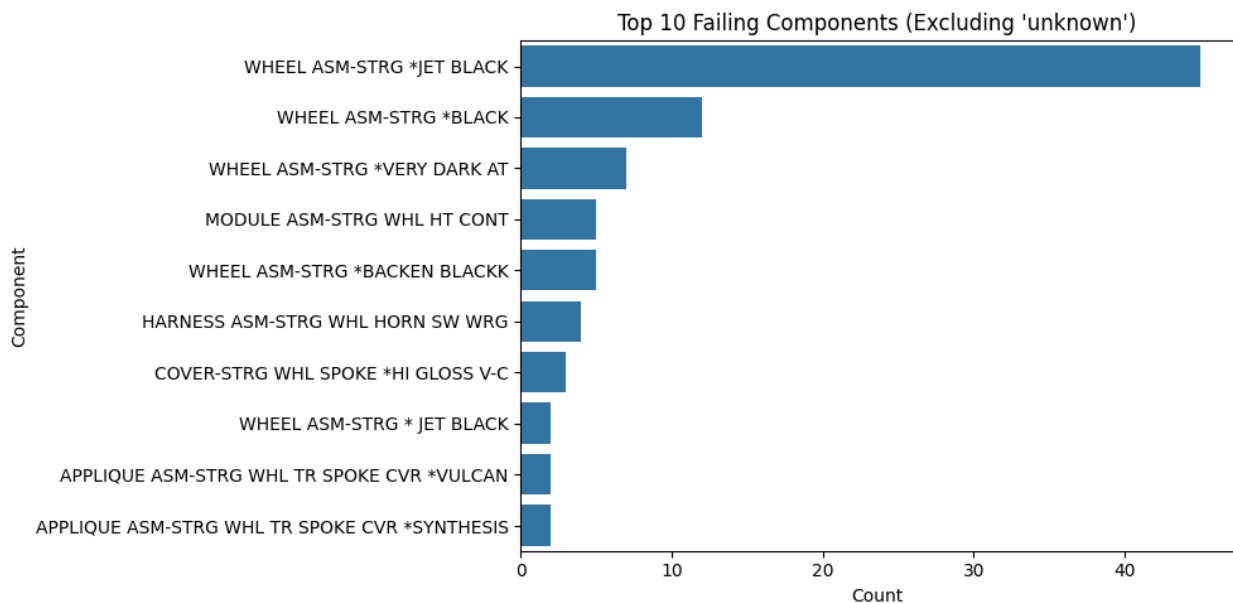
Identify Critical Columns & Create Visualizations

Column	Reason
component	Which part/component is failing
repair_date	Timeline of repairs
totalcost or reporting_cost	Cost of failure (financial impact)
actual_hrs or repair_age	Time impact of the failure
correction_verbatim	Root cause or fix summary (for tag extraction)

```
import seaborn as sns
import matplotlib.pyplot as plt
```

Top 10 failing components

```
plt.figure(figsize=(10, 5))
sns.countplot(
    data=df[df['component'] != 'unknown'],
    y='component',
    order=df[df['component'] != 'unknown']
['component'].value_counts().head(10).index
)
plt.title("Top 10 Failing Components (Excluding 'unknown')")
plt.xlabel("Count")
plt.ylabel("Component")
plt.tight_layout()
plt.show()
```



Records labeled as 'unknown' represent missing or unrecorded component data and were excluded from core visual analysis to maintain the accuracy and actionability of insights.

Investigate 'unknown' Root Causes

```
df[df['component'] == 'unknown'][['vin', 'repair_date',
'correction_verbatim']].head(10)
```

	vin	repair_date	\
8	1HRF9CED6NZ221061	2024-01-08	
19	1HR49WEY2NF322460	2024-01-11	
36	3HCFDFED5SH364828	2024-01-16	
47	1HCRYDED2NZ208869	2024-01-23	

```
98 1HC4WLE78RF260518 2024-02-07
```

```
correction_verbatim
8  TECHNICIAN FOUND THE STEERING COLUMN PLASTIC T...
19 Reconnected horn checked operation, ok.
36 REMOVED AND REPLACED HEATED STEERING WHEEL MOD...
47 TEST DROVE AND CONFIRMED CUSTOMER COMPLAINT BR...
98 REMOVED STEERING WHEEL AND DISASSEMBLED AND FO...
```

Group by Key Identifiers ('plant', 'dealer_name')

```
df[df['component'] ==
'unknown'].groupby('plant').size().sort_values(ascending=False)

plant
FLT      2
FTW      2
SIL      1
dtype: int64

df[df['component'] ==
'unknown'].groupby('dealer_name').size().sort_values(ascending=False)

dealer_name
BlueSky Auto Sales    1
IronClad Wheels      1
Legacy Car Sales      1
NovaCar Sales         1
WestPoint Motors      1
dtype: int64
```

Missing Component Analysis:

A small portion of records 6% had missing values in the `component` field, labeled as `'unknown'`. Further investigation revealed the following:

Key Findings:

- All missing entries originated from just **three plants**: FLT, FTW, and SIL (2 records each).
- Only **five dealers** were associated with these cases — each contributing just **1 record**.
- The free-text field `correction_verbatim` still provided clues about the actual component involved (e.g., "steering wheel", "horn").

Possible Reasons:

- System-level data mapping failure at the plant level.
- Repair submission forms allowing component fields to be skipped.
- Lack of validation in structured data capture.

Suggestions:

- Add mandatory field validation for `component` during data entry.

- Implement simple keyword-based auto-tagging from `correction_verbatim`.
- Review repair workflows or systems at plants `FLT`, `FTW`, and `SIL` to close the data gap.

These steps will help reduce data quality issues and improve insight reliability for future analyses.