

Install libraries

```
In [12]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
import statsmodels.api as sm
from scipy.stats import pearsonr
```

Loading Data

```
In [3]: data = pd.read_csv('data.csv')
data.head()
```

```
Out[3]:
```

	Company	File No.	Opened	Closed	Coverage	SubCoverage	Reason	S
0	Anthem Health Plans, Inc	7045593	05/31/2022	06/02/2022	Group	Health Only	Claim Handling	
1	Anthem Health Plans, Inc	7043381	02/28/2022	06/02/2022	Group	Health Only	Claim Handling	
2	Anthem Health Plans, Inc	7044860	05/03/2022	06/02/2022	A & H	Health Only	Claim Handling	
3	Anthem Health Plans, Inc	7043381	02/28/2022	06/02/2022	Group	A & H	Claim Handling	
4	Anthem Health Plans, Inc	7052007	02/23/2023	03/17/2023	A & H	A & H	Marketing & Sales	

Descriptive Statistics & Summary Measures

Calculate mean, median, standard deviation, quartiles, and range for numerical variables

```
In [4]: data.info()

# File No is an id
# Recovery is useful

recovery = data['Recovery']
```

```

recovery_mean = np.mean(recovery)
recovery_median = np.median(recovery)
recovery_std = np.std(recovery)
recovery_min = np.min(recovery)
recovery_max = np.max(recovery)
recovery_range = recovery_max - recovery_min
recovery_quartiles = np.percentile(recovery, [25, 50, 75, 100])

print("\n")
print(f'Mean: {recovery_mean:.2f}')
print(f'Median: {recovery_median:.2f}')
print(f'Standard Deviation: {recovery_std:.2f}')
print(f'Range: {recovery_range:.2f}')
print(f'Quartiles: {recovery_quartiles}')

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 68469 entries, 0 to 68468
Data columns (total 12 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   Company               68469 non-null  object
 1   File No.              68469 non-null  int64
 2   Opened                68469 non-null  object
 3   Closed                66992 non-null  object
 4   Coverage              65130 non-null  object
 5   SubCoverage           55020 non-null  object
 6   Reason                65057 non-null  object
 7   SubReason             65057 non-null  object
 8   Disposition           41476 non-null  object
 9   Conclusion            42809 non-null  object
10   Recovery              68469 non-null  float64
11   Status                68469 non-null  object
dtypes: float64(1), int64(1), object(10)
memory usage: 6.3+ MB

```

```

Mean: 1723.51
Median: 0.00
Standard Deviation: 13989.96
Range: 843825.85
Quartiles: [    0.         0.         0.      843825.85]

```

Missing Value Analysis

Identify missing values, analyze their patterns (random or systematic), and calculate the percentage of missing entries for each column.

```

In [5]: # Missing values
missing_values = data.isnull().sum()
print("\nTotal Number of Missing Values:")
print(missing_values)

# Percentage of Missing Entries

```

```
total_entries = len(data)
missing_percentage = (missing_values / total_entries) * 100
print("\nPercentage of Missing Entries:")
print(missing_percentage)
```

Total Number of Missing Values:

```
Company      0
File No.     0
Opened       0
Closed      1477
Coverage     3339
SubCoverage  13449
Reason       3412
SubReason    3412
Disposition  26993
Conclusion   25660
Recovery     0
Status       0
dtype: int64
```

Percentage of Missing Entries:

```
Company      0.000000
File No.     0.000000
Opened       0.000000
Closed       2.157181
Coverage     4.876660
SubCoverage  19.642466
Reason       4.983277
SubReason    4.983277
Disposition  39.423681
Conclusion   37.476814
Recovery     0.000000
Status       0.000000
dtype: float64
```

Univariate Distribution Visualization

Create histograms, box plots, or density plots for individual variables.

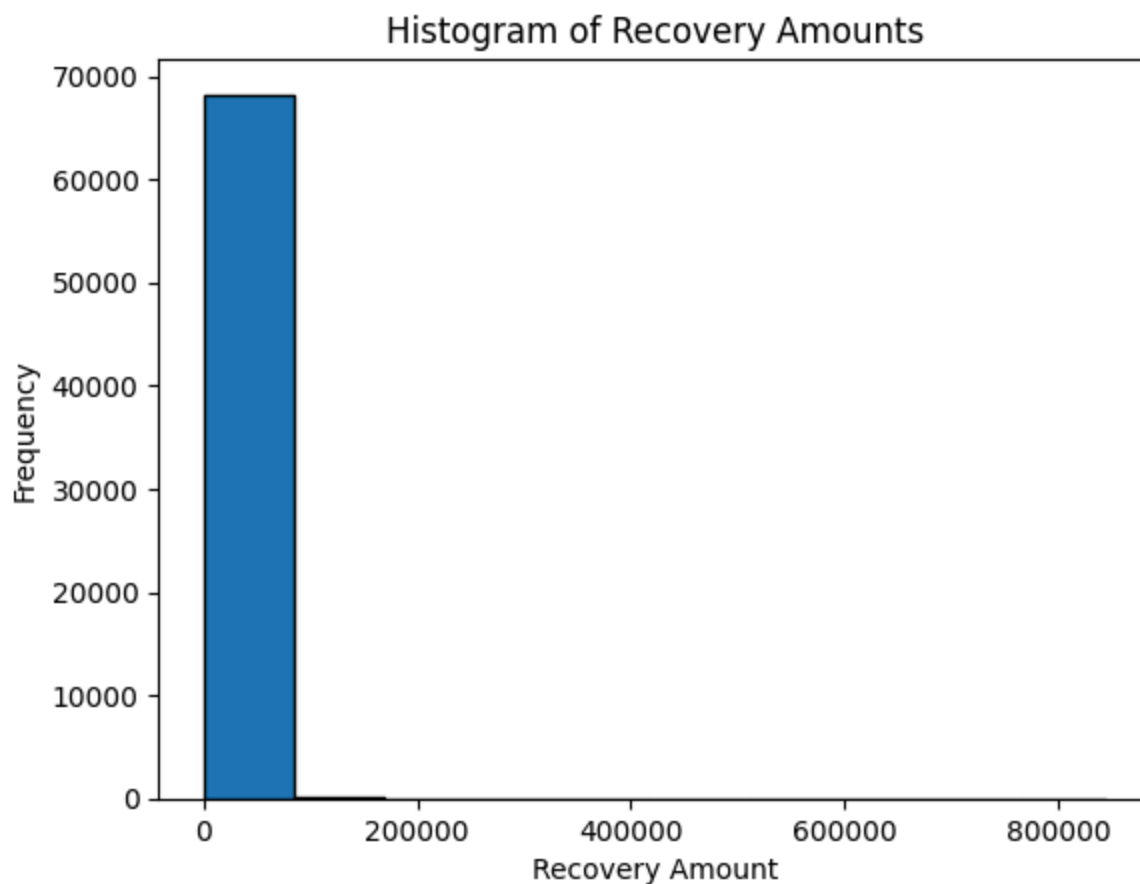
```
In [ ]: recovery_max = np.max(recovery)
print(recovery_max)
```

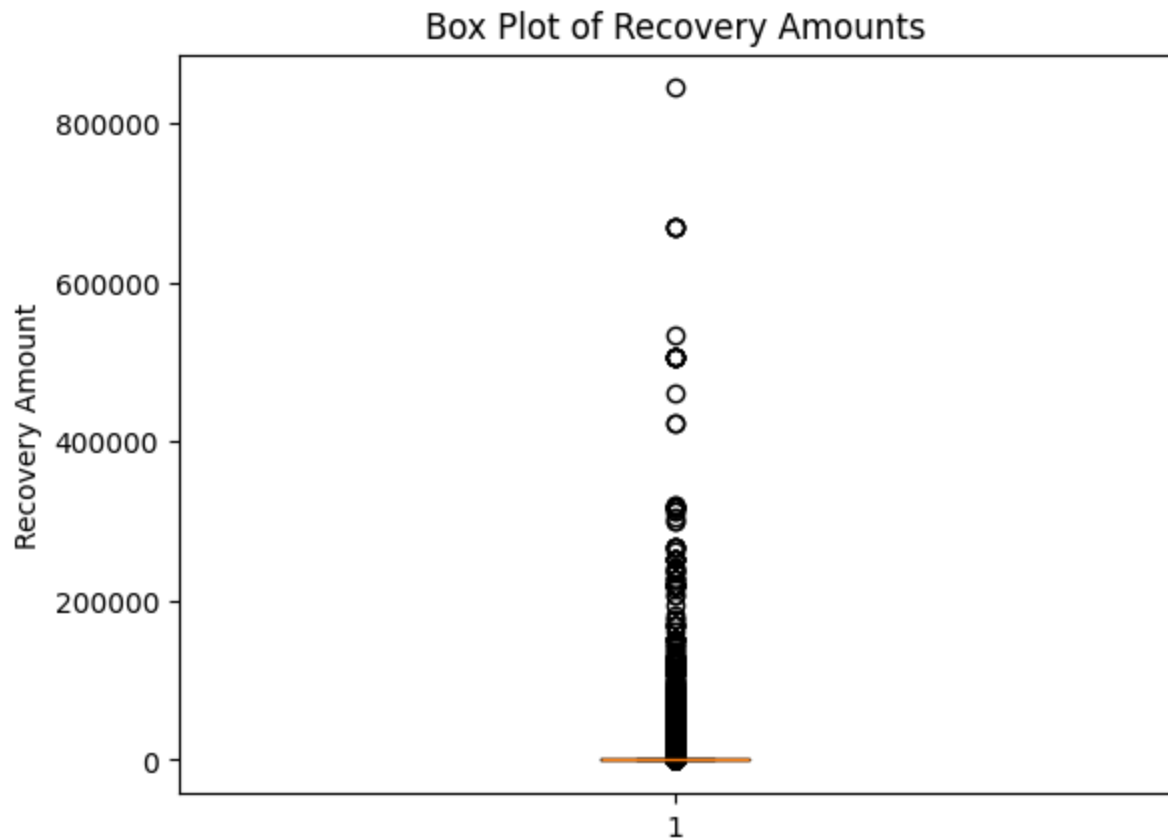
843825.85

```
In [6]: # histogram:
plt.hist(data['Recovery'], edgecolor='black')
plt.title('Histogram of Recovery Amounts')
plt.xlabel('Recovery Amount')
plt.ylabel('Frequency')
plt.show()

# box plot:
```

```
plt.boxplot(data['Recovery'])  
plt.title('Box Plot of Recovery Amounts')  
plt.ylabel('Recovery Amount')  
plt.show()  
  
# density plot:  
sns.kdeplot(data['Recovery'], shade=True)  
plt.title('Density Plot of Recovery Amounts')  
plt.xlabel('Recovery Amount')  
plt.ylabel('Density')  
plt.show()
```

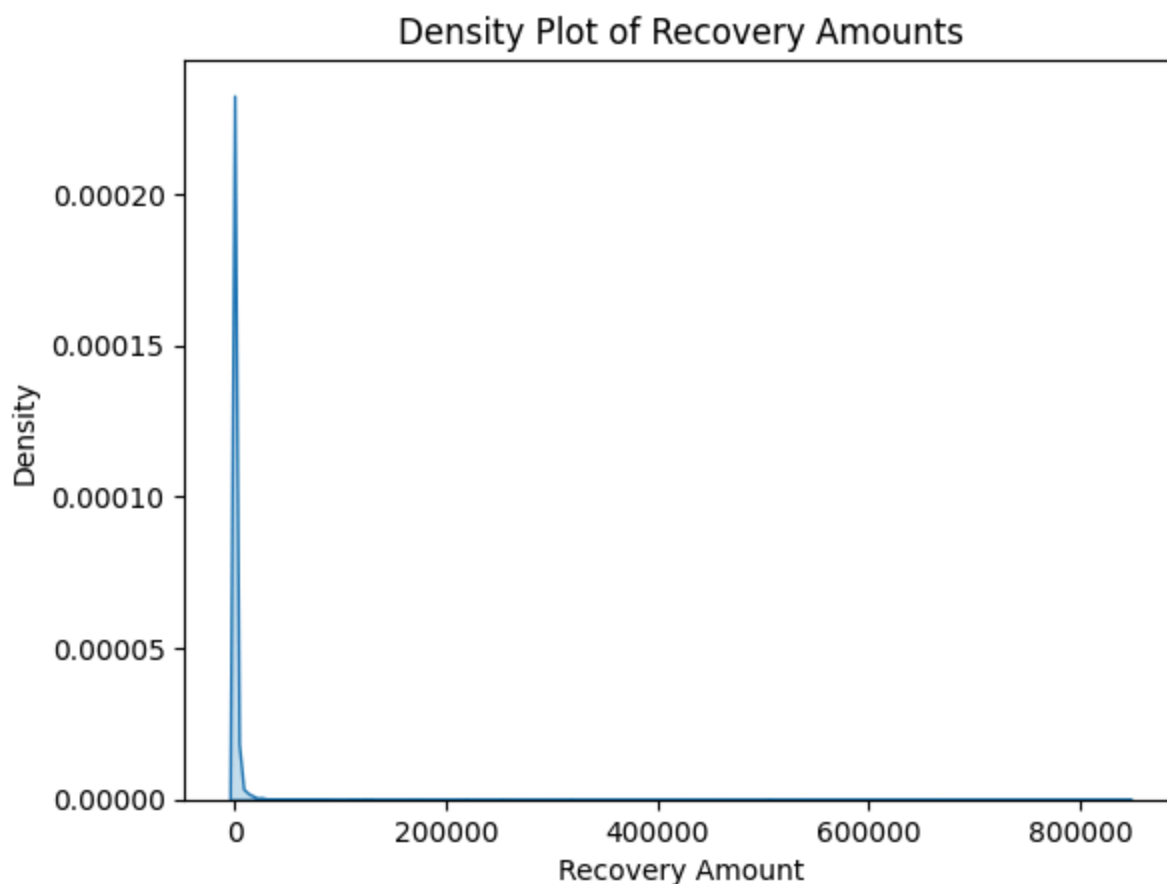




<ipython-input-6-0b29ca802637>:15: FutureWarning:

`shade` is now deprecated in favor of `fill`; setting `fill=True`.
This will become an error in seaborn v0.14.0; please update your code.

```
sns.kdeplot(data['Recovery'], shade=True)
```



Correlation and Relationship Analysis

```
In [13]: # Compute the correlation matrix (select only numeric columns)
numerical_data = data.select_dtypes(include=np.number)
corr_matrix = numerical_data.corr()

# Display the correlation matrix
print(corr_matrix)

# Set the figure size for better readability
plt.figure(figsize=(10, 8))
# Create a heatmap with correlation coefficients annotated
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', fmt=".2f")
# Add a title and display the plot
plt.title("Correlation Matrix Heatmap")
plt.show()

# Create a scatter plot with regression line using the actual column names
sns.regplot(x='Recovery', y='File No.', data=data, scatter_kws={'alpha': 0.5})
plt.title("Scatter Plot with Regression Line")
plt.xlabel("Recovery")
plt.ylabel("File No.")
plt.show()

# Compute Pearson correlation coefficient and p-value using 'Recovery' and 'File No.'
corr_coef, p_value = pearsonr(data['Recovery'], data['File No.'])
```

```

print(f"Pearson correlation coefficient: {corr_coef:.3f}")
print(f"P-value: {p_value:.3f}")

# Define the independent variable(s) and the dependent variable
X = data[['Recovery']] # Independent variable(s)
y = data['File No.']    # Dependent variable

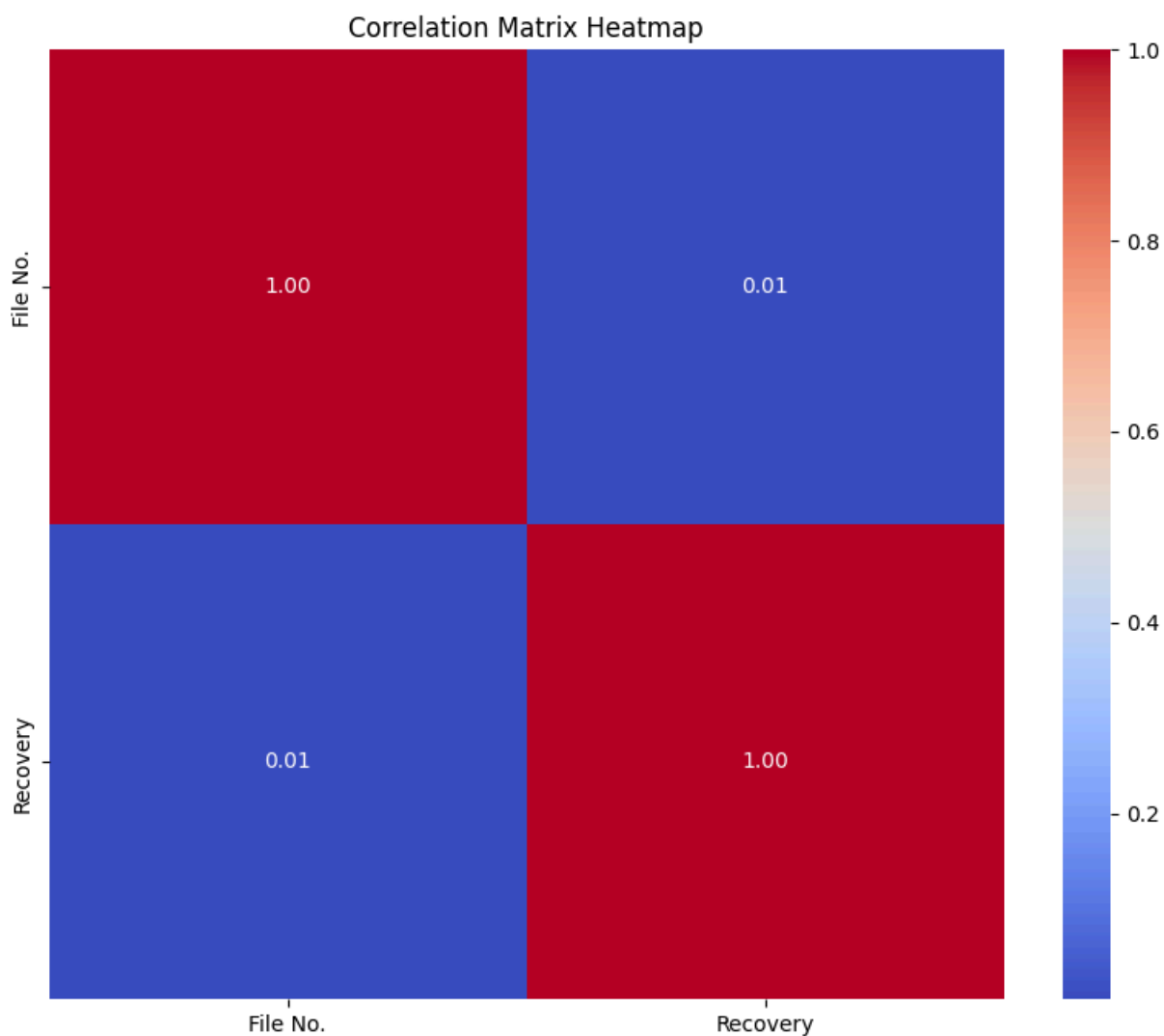
# Add a constant to the independent variables (the intercept)
X = sm.add_constant(X)

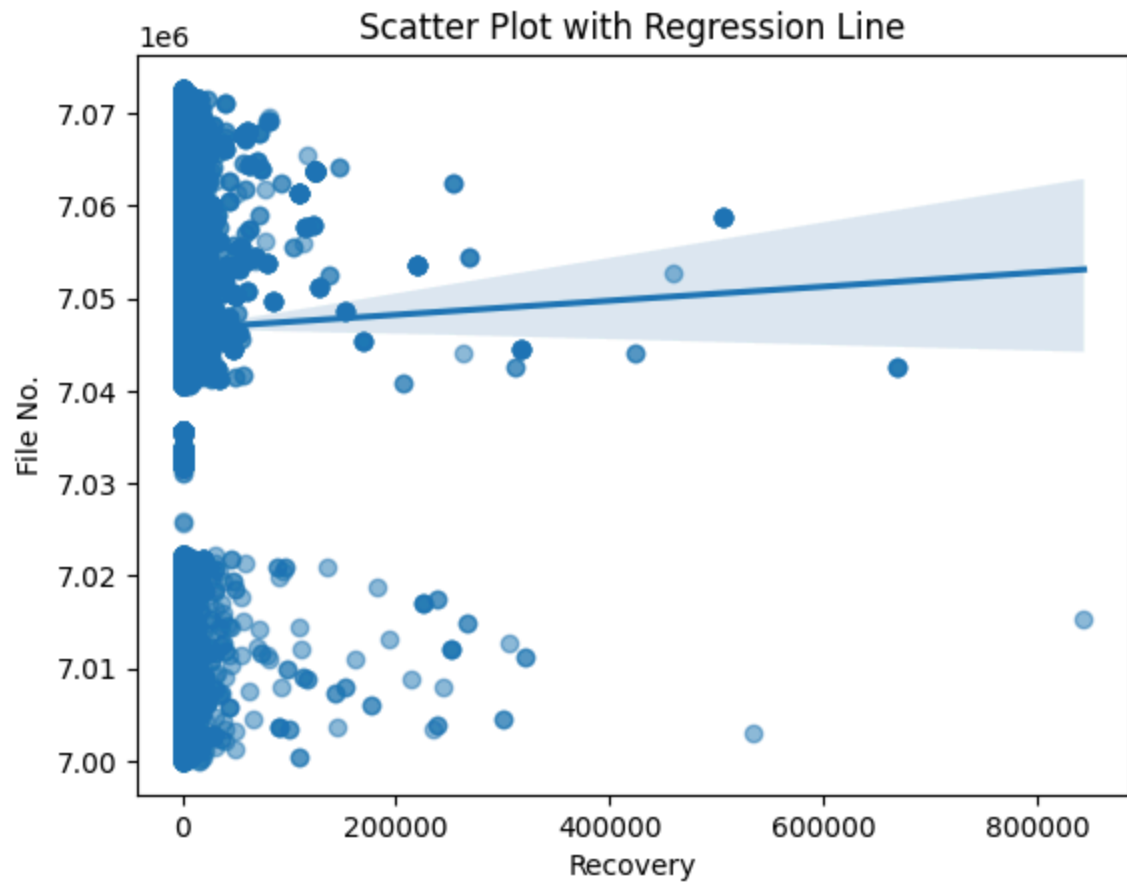
# Fit the Ordinary Least Squares (OLS) model
model = sm.OLS(y, X).fit()

# Print the regression summary
print(model.summary())

```

	File No.	Recovery
File No.	1.000000	0.005019
Recovery	0.005019	1.000000





Pearson correlation coefficient: 0.005
P-value: 0.189

OLS Regression Results						
=====						
==						
Dep. Variable:	File No.	R-squared:	0.000000			
Model:	OLS	Adj. R-squared:	0.000000			
Method:	Least Squares	F-statistic:	1.725000			
Date:	Tue, 04 Mar 2025	Prob (F-statistic):	0.189000			
Time:	22:38:47	Log-Likelihood:	-7.7929e+05			
No. Observations:	68469	AIC:	1.559e+06			
Df Residuals:	68467	BIC:	1.559e+06			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
==						
	coef	std err	t	P> t	[0.025	0.975]

const	7.047e+06	81.717	8.62e+04	0.000	7.05e+06	7.05e+06
Recovery	0.0076	0.006	1.313	0.189	-0.004	0.015
=====						
==						
Omnibus:	8470.043	Durbin-Watson:	1.000000			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	9471.000000			
Skew:	-0.864	Prob(JB):	0.000000			
Kurtosis:	2.420	Cond. No.	1.42e+04			
=====						
==						

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The condition number is large, 1.42e+04. This might indicate that there are strong multicollinearity or other numerical problems.

Analyzing High Cardinality and Grouping Rare Categories

```

In [8]: def analyze_high_cardinality(df, threshold=0.02):
        """
        Identifies high-cardinality categorical columns and groups rare categories.

        Parameters:
        df (pd.DataFrame): The input DataFrame.
        threshold (float): The minimum frequency a category must have to not be rare.

        Returns:
        pd.DataFrame: A modified DataFrame with rare categories grouped.
        """
        categorical_cols = df.select_dtypes(include=['object']).columns

        for col in categorical_cols:
            value_counts = df[col].value_counts(normalize=True)

            # Print unique value distribution before modification
            print(f"\nColumn: {col}")
            print(value_counts)

            rare_categories = value_counts[value_counts < threshold].index
            if len(rare_categories) > 0:
                print(f"\nGrouping {len(rare_categories)} rare categories in {col}")
                df[col] = df[col].replace(rare_categories, 'Other')

        return df

# Load CSV file
df = pd.read_csv("data.csv")

# Analyze and modify high-cardinality categorical columns
df_modified = analyze_high_cardinality(df, threshold=0.02)

# Save the modified DataFrame
df_modified.to_csv("modified_data.csv", index=False)

print("High-cardinality analysis complete. Modified data saved.")

```

Column: Company

Company

Anthem Health Plans, Inc	0.213308
ConnectiCare Benefits, Inc.	0.053572
UnitedHealthcare Insurance Company	0.049847
Cigna Health and Life Insurance Company	0.046532
ConnectiCare Insurance Company, Inc	0.033796

...

Plaza Insurance Company	0.000015
Trenwick America Reinsurance Corporation	0.000015
StarStone Specialty Insurance Company	0.000015
Continental Life Insurance Company of Brentwood, Tennessee	0.000015
ARAG Insurance Company	0.000015

Name: proportion, Length: 843, dtype: float64

Grouping 837 rare categories in Company under 'Other'

Column: Opened

Opened

01/03/2025	0.003023
09/18/2024	0.002746
02/29/2024	0.002731
02/28/2024	0.002658
02/07/2024	0.002527

...

11/02/2015	0.000015
02/18/2023	0.000015
09/08/2017	0.000015
11/06/2017	0.000015
12/26/2021	0.000015

Name: proportion, Length: 2139, dtype: float64

Grouping 2139 rare categories in Opened under 'Other'

Column: Closed

Closed

02/27/2024	0.003374
01/06/2025	0.003299
02/23/2023	0.003299
04/16/2024	0.003284
10/09/2024	0.003030

...

10/02/2020	0.000015
06/01/2024	0.000015
01/05/2018	0.000015
12/28/2018	0.000015
08/30/2021	0.000015

Name: proportion, Length: 1765, dtype: float64

Grouping 1765 rare categories in Closed under 'Other'

Column: Coverage

Coverage

A & H	0.233671
Individual Private Passenger	0.225672
Group	0.193382
Individual	0.133226
Homeowners	0.100077
Individual Life	0.027253
Commercial Multi-Peril	0.013619
Travel	0.010226
Commercial	0.010195
Individual Annuities	0.008875
Condo/Townhome	0.005988
General	0.005881
Group Life	0.004069
Workers' Compensation	0.003654
Renter/Tenants	0.003485
Pet Insurance	0.002948
Other [Enter Coverage]	0.002518
Extended Warranty & Service Contracts	0.002073
Dwelling Fire	0.001520
Credit Accident & Health	0.001336
Umbrella	0.001198
Inland Marine	0.001044
Fire, Allied Lines	0.000829
Title	0.000691
Watercraft	0.000645
Motorcycle	0.000614
Professional/E&O	0.000583
Unknown	0.000568
Motorhome	0.000522
Mobile Homeowner	0.000522
Portable Electronics Ins	0.000415
Rental	0.000384
Federal Flood	0.000322
Farm owner/Ranch owner	0.000230
Group Annuities	0.000230
Life and Annuity	0.000215
Fidelity & Surety	0.000215
Group Private Passenger	0.000138
Extended Warranty	0.000138
Portable Electronics	0.000092
Directors & Officers	0.000077
Ocean Marine	0.000061
Business Interruption	0.000061
Federal Programs	0.000061
Group Homeowners	0.000061
Credit Life	0.000061
Motorsport	0.000046
Credit Property	0.000046
Crop/Hail	0.000046
Aircraft	0.000031
GAP Ins	0.000031
IRA	0.000031
Surplus Lines	0.000031
Builder's Risk	0.000031
Auto Warranty	0.000031
Products	0.000015

Accelerated Benefits 0.000015

Name: proportion, dtype: float64

Grouping 51 rare categories in Coverage under 'Other'

Column: SubCoverage

SubCoverage

Health Only 0.267884

A & H 0.202163

Liability 0.135260

Homeowners 0.045674

Collision 0.044257

...

Residual Market/Joint Underwriting Assn 0.000018

COBRA 0.000018

Medicare Supplement Plan K 0.000018

Excess Loss 0.000018

Comprehensive Personal Liability 0.000018

Name: proportion, Length: 110, dtype: float64

Grouping 99 rare categories in SubCoverage under 'Other'

Column: Reason

Reason

Claim Handling 0.765313

PolicyHolder Service 0.107321

Underwriting 0.099036

Marketing & Sales 0.028329

Name: proportion, dtype: float64

Column: SubReason

SubReason

Claim Denial 0.113085

Claim Delay 0.097130

Unsatisfactory Settlement/Offer 0.087462

Medical Necessity Denial 0.057012

Premium & Rating 0.038658

...

Grace Period 0.000015

Wellness Program 0.000015

1035 Exchange 0.000015

Endorsement Rider 0.000015

Nonforfeiture 0.000015

Name: proportion, Length: 198, dtype: float64

Grouping 184 rare categories in SubReason under 'Other'

Column: Disposition

Disposition

Company Position Substantiated 0.440616

Claim Settled 0.244503

Question of Fact/Contract/Provision/Legal Issue 0.106158

Company Position Overturned 0.057841

Compromised Settlement/Resolution	0.046895
No Action Requested/Required	0.044411
No Jurisdiction	0.034261
Insufficient Information	0.012103
Complaint Withdrawn	0.005521
Referred to Outside Agency/Dept	0.002724
Referred to Another State's Dept of Insurance	0.001929
Claim Reopened	0.001423
Referred to Other Division for Possible Disciplinary Action	0.001230
Fine Assessed	0.000386
Name: proportion, dtype: float64	

Grouping 7 rare categories in Disposition under 'Other'

Column: Conclusion	
Conclusion	
Claim Paid	0.128968
Company Position Upheld	0.106753
Justified	0.085146
Furnished Information	0.083137
Contract Provision	0.081525
Corrective Action	0.073232
Provider Issue	0.072882
Coverage Granted	0.042701
Refer-Judicial/Attorney	0.036488
Claim Paid With Interest	0.033264
Coverage Denied	0.023336
Unjustified	0.020510
Voluntary Reconsideration	0.019295
Premium Refund	0.018734
Policy Restored/Reinstated	0.018407
Enter Arbitration	0.017052
Rate Increase Explained	0.014600
Satisfactory Explanation	0.013946
Additional Money Received	0.013081
Questionable	0.011983
Refer To Appraisal	0.010815
Satisfied	0.009951
External Review Info Sent	0.009834
No Action Necessary	0.006798
No Authority	0.005466
Record Only	0.004415
Non-Renewal Upheld	0.004345
No Cause For Action	0.003901
Non-Renewal Rescinded	0.003854
Cancellation Upheld	0.003784
Policy not written in CT	0.003738
Refer To Agency	0.002359
Federal	0.001962
Policy Issued	0.001939
Contract Violation	0.001822
Cancellation Withdrawn	0.001495
Rate Problem Solved	0.001402
Policy Not In Force	0.001308
Other [Enter Disposition]	0.001075

Coverage Extended	0.000888
Deductible Recovered	0.000584
Fees Returned	0.000561
Insured Retained Attorney	0.000420
Underwriting Guidelines	0.000374
Interest Paid	0.000350
Policy Offered	0.000280
Complaint Form Sent	0.000280
Class Revised	0.000280
Extl Rev Info Sent/SF	0.000210
Accident in Another State	0.000187
Underwriting Discretion	0.000140
Cease and Desist	0.000047
Cross Reference Only	0.000023
Med Jurisdiction Explained	0.000023
Filed Errors&Omission Clm	0.000023
Not Insurance Related	0.000023

Name: proportion, dtype: float64

Grouping 44 rare categories in Conclusion under 'Other'

Column: Status

Status	
Closed	0.975171
Sent to Company	0.008734
Supervisor Review	0.005258
Open	0.003184
Full Review – Standard	0.002308
Reopened	0.001636
Interim Letter Sent	0.001212
Incomplete Follow-up	0.000759
New Doc	0.000716
Awaiting Decision	0.000482
Missing Information	0.000161
Extension Granted	0.000146
Legal Review	0.000058
Preliminary Review – Standard	0.000029
Sent to Agent	0.000029
No Response Follow-up	0.000029
New	0.000029
In Progress	0.000029
Recovery Pending	0.000015
Verify Situs	0.000015

Name: proportion, dtype: float64

Grouping 19 rare categories in Status under 'Other'

High-cardinality analysis complete. Modified data saved.

Here's what happened:

-In the Company column, 837 rare companies were grouped as "Other" since they had very low proportions in the dataset. In the Opened, Closed, Coverage, SubCoverage,

SubReason, Disposition, Conclusion, and Status columns, categories that appeared infrequently were also grouped under "Other" to reduce complexity. This helps in:

- Reducing noise by merging very rare categories. Improving model performance by avoiding too many categories with little data. Making analysis and visualization clearer by focusing on the most relevant categories.

```
In [ ]: file_path = "modified_data.csv"
df = pd.read_csv(file_path)

# Ensure 'Recovery' is numeric
df['Recovery'] = pd.to_numeric(df['Recovery'], errors='coerce')

# Define outliers using the IQR method
Q1 = df['Recovery'].quantile(0.25)
Q3 = df['Recovery'].quantile(0.75)
IQR = Q3 - Q1
lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR

# Identify outliers
outliers = df[(df['Recovery'] < lower_bound) | (df['Recovery'] > upper_bound)]
non_outliers = df[(df['Recovery'] >= lower_bound) & (df['Recovery'] <= upper_bound)]

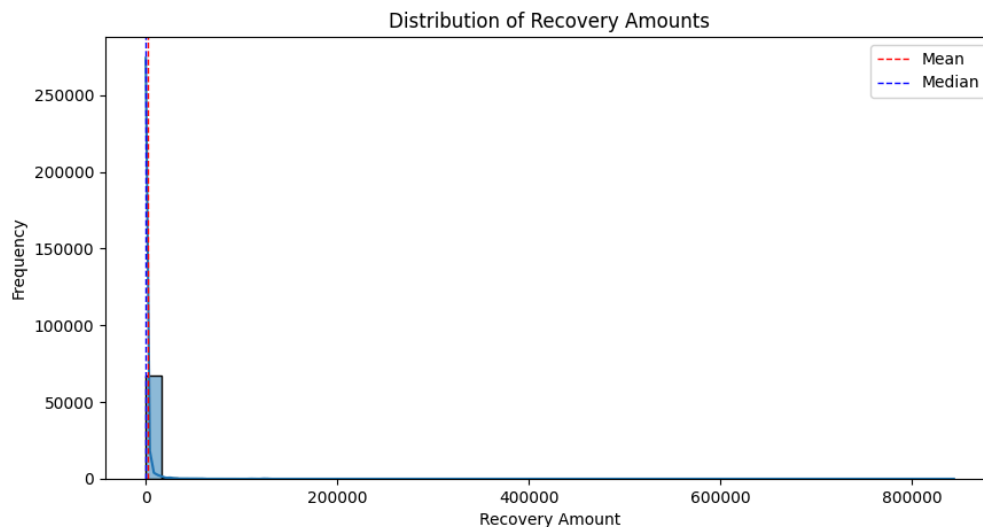
# Calculate statistics
mean_with_outliers = df['Recovery'].mean()
median_with_outliers = df['Recovery'].median()
mean_without_outliers = non_outliers['Recovery'].mean()
median_without_outliers = non_outliers['Recovery'].median()

# Print results
print(f"Mean with outliers: ${mean_with_outliers:.2f}")
print(f"Median with outliers: ${median_with_outliers:.2f}")
print(f"Mean without outliers: ${mean_without_outliers:.2f}")
print(f"Median without outliers: ${median_without_outliers:.2f}")

# Plot distribution
plt.figure(figsize=(10, 5))
sns.histplot(df['Recovery'], bins=50, kde=True)
plt.axvline(mean_with_outliers, color='red', linestyle='dashed', linewidth=1)
plt.axvline(median_with_outliers, color='blue', linestyle='dashed', linewidth=1)
plt.title('Distribution of Recovery Amounts')
plt.xlabel('Recovery Amount')
plt.ylabel('Frequency')
plt.legend()

# Save the plot
histogram_path = "recovery_histogram.png"
plt.savefig(histogram_path)
plt.close()

print(f"Histogram saved to: {histogram_path}")
```

Frequency and Counting Analysis

```
In [9]: df = pd.read_csv('data.csv')

# Remove exact duplicate rows
df = df.drop_duplicates()

# Remove duplicate complaints based on File No.
df = df.drop_duplicates(subset=['File No.'], keep='first')

# Show the first few rows and column names
print("Preview of the dataset:")
print(df.head())
print("\nColumn names:")
print(df.columns)

# Frequency Analysis: Complaints by Company
print("\nTop 10 Companies by Complaint Count:")
top_companies = df['Company'].value_counts().head(10)
print(top_companies)

# Frequency Analysis with Percentages
def print_count_and_percentage(column_name):
    counts = df[column_name].value_counts()
    percentages = df[column_name].value_counts(normalize=True) * 100
    summary = pd.DataFrame({
        'Count': counts,
        'Percentage': percentages
    })
    print(f"\n{column_name} Distribution (Counts and Percentages):")
    print(summary)

print_count_and_percentage('Company')
print_count_and_percentage('Reason')
```

```

print_count_and_percentage('Status')
print_count_and_percentage('Conclusion')

# Detect Rare Classes
def print_rare_classes(column_name, threshold=5):
    counts = df[column_name].value_counts()
    rare_classes = counts[counts <= threshold]
    if not rare_classes.empty:
        print(f"\nRare categories in {column_name} (<= {threshold} occurrence)
        print(rare_classes)
    else:
        print(f"\nNo rare categories found in {column_name} (<= {threshold}

print_rare_classes('Company')
print_rare_classes('Reason')
print_rare_classes('Status')
print_rare_classes('Conclusion')

# Visualization: Top 10 Companies by Complaints
plt.figure(figsize=(10, 6))
sns.barplot(x=top_companies.values, y=top_companies.index, palette='Blues_r'
plt.title('Top 10 Companies by Complaint Count')
plt.xlabel('Number of Complaints')
plt.ylabel('Company')
plt.tight_layout()
plt.savefig('top_companies_barplot.png')
plt.close()

# Visualization: Top Complaint Reasons
top_reasons = df['Reason'].value_counts().head(5)
plt.figure(figsize=(10, 6))
sns.barplot(x=top_reasons.values, y=top_reasons.index, palette='muted')
plt.title('Top Complaint Reasons')
plt.xlabel('Number of Complaints')
plt.ylabel('Reason')
plt.tight_layout()
plt.savefig('top_complaint_reasons_barplot.png')
plt.close()

# Visualization: Complaint Status Distribution (Pie Chart)
plt.figure(figsize=(8, 8))
plt.pie(df['Status'].value_counts().head(6), labels=df['Status'].value_count
plt.title('Complaint Status Distribution')
plt.tight_layout()
plt.savefig('complaint_status_pie_chart.png')
plt.close()

# Visualization: Complaint Conclusion Distribution (Pie Chart)
plt.figure(figsize=(8, 8))
plt.pie(df['Conclusion'].value_counts().head(6), labels=df['Conclusion'].va
plt.title('Complaint Conclusion Distribution')
plt.tight_layout()
plt.savefig('complaint_conclusion_pie_chart.png')
plt.close()

```

Preview of the dataset:

	Company	File No.	Opened	Closed	Coverage
\					
0	Anthem Health Plans, Inc	7045593	05/31/2022	06/02/2022	Group
1	Anthem Health Plans, Inc	7043381	02/28/2022	06/02/2022	Group
2	Anthem Health Plans, Inc	7044860	05/03/2022	06/02/2022	A & H
4	Anthem Health Plans, Inc	7052007	02/23/2023	03/17/2023	A & H
5	Oxford Health Plans (CT), Inc	7054762	06/01/2023	08/02/2023	A & H

	SubCoverage	Reason	SubReason	\
0	Health Only	Claim Handling	Medically Necessary	
1	Health Only	Claim Handling	Provider Contract Issue	
2	Health Only	Claim Handling	Denial	
4	A & H	Marketing & Sales	Duplicate Coverage	
5	Health Only	Claim Handling	External Review	

	Disposition	Conclusion	Recovery	\
0	Company Position Substantiated	Company Position Upheld	0.00	
1	Claim Settled	Satisfied	6467.30	
2	Claim Settled	Claim Paid	147.58	
4	Compromised Settlement/Resolution	Premium Refund	2179.32	
5	NaN	NaN	0.00	

	Status
0	Closed
1	Closed
2	Closed
4	Closed
5	Closed

Column names:

```
Index(['Company', 'File No.', 'Opened', 'Closed', 'Coverage', 'SubCoverage',
      'Reason', 'SubReason', 'Disposition', 'Conclusion', 'Recovery',
      'Status'],
      dtype='object')
```

Top 10 Companies by Complaint Count:

Company	
Anthem Health Plans, Inc	3657
UnitedHealthcare Insurance Company	1024
Progressive Direct Insurance Company	755
State Farm Mutual Automobile Insurance Company	732
Allstate Fire and Casualty Insurance Company	722
Cigna Health and Life Insurance Company	663
ConnectiCare Insurance Company, Inc	651
ConnectiCare Benefits, Inc.	643
Progressive Casualty Insurance Company	619
State Farm Fire & Casualty Company	611
Name: count, dtype: int64	

Company Distribution (Counts and Percentages):

Company	Count	Percentage
Anthem Health Plans, Inc	3657	12.798348
UnitedHealthcare Insurance Company	1024	3.583677
Progressive Direct Insurance Company	755	2.642262

State Farm Mutual Automobile Insurance Company	732	2.561769
Allstate Fire and Casualty Insurance Company	722	2.526773
...
Colony Specialty Insurance Company	1	0.003500
First Allmerica Financial Life Insurance Company	1	0.003500
Capitol Specialty Insurance Corporation	1	0.003500
21st Century Premier Insurance Company	1	0.003500
ARAG Insurance Company	1	0.003500

[825 rows x 2 columns]

Reason Distribution (Counts and Percentages):

	Count	Percentage
Reason		
Claim Handling	18183	71.724981
Underwriting	3655	14.417577
PolicyHolder Service	2576	10.161335
Marketing & Sales	937	3.696107

Status Distribution (Counts and Percentages):

	Count	Percentage
Status		
Closed	28061	98.204662
Sent to Company	187	0.654441
Open	185	0.647442
New Doc	33	0.115490
Supervisor Review	25	0.087492
Full Review - Standard	20	0.069994
Reopened	14	0.048996
Incomplete Follow-up	13	0.045496
Interim Letter Sent	11	0.038497
Awaiting Decision	6	0.020998
Missing Information	4	0.013999
Extension Granted	3	0.010499
Preliminary Review - Standard	2	0.006999
Sent to Agent	2	0.006999
In Progress	2	0.006999
New	2	0.006999
Recovery Pending	1	0.003500
Legal Review	1	0.003500
Verify Situs	1	0.003500
No Response Follow-up	1	0.003500

Conclusion Distribution (Counts and Percentages):

	Count	Percentage
Conclusion		
Justified	1430	12.459702
Company Position Upheld	1192	10.385989
Furnished Information	988	8.608521
Claim Paid	919	8.007319
Corrective Action	855	7.449682
Contract Provision	839	7.310273
Refer-Judicial/Attorney	826	7.197003
Unjustified	590	5.140716
Enter Arbitration	336	2.927594
Provider Issue	298	2.596497

No Action Necessary	258	2.247974
Coverage Granted	246	2.143417
Voluntary Reconsideration	233	2.030147
Claim Paid With Interest	211	1.838460
Questionable	207	1.803607
Refer To Appraisal	182	1.585780
Premium Refund	170	1.481223
Satisfactory Explanation	161	1.402806
Additional Money Received	159	1.385379
Coverage Denied	152	1.324388
Non-Renewal Upheld	148	1.289536
Rate Increase Explained	139	1.211118
Policy Restored/Reinstated	131	1.141413
No Cause For Action	102	0.888734
Policy not written in CT	99	0.862595
Cancellation Upheld	94	0.819029
No Authority	87	0.758038
Record Only	75	0.653481
Non-Renewal Rescinded	74	0.644768
External Review Info Sent	71	0.618629
Satisfied	41	0.357236
Cancellation Withdrawn	29	0.252679
Contract Violation	20	0.174262
Refer To Agency	16	0.139409
Other [Enter Disposition]	13	0.113270
Policy Issued	12	0.104557
Federal	12	0.104557
Policy Not In Force	11	0.095844
Underwriting Guidelines	9	0.078418
Coverage Extended	7	0.060992
Fees Returned	7	0.060992
Accident in Another State	5	0.043565
Underwriting Discretion	4	0.034852
Rate Problem Solved	3	0.026139
Interest Paid	3	0.026139
Extl Rev Info Sent/SF	2	0.017426
Deductible Recovered	2	0.017426
Policy Offered	2	0.017426
Class Revised	1	0.008713
Complaint Form Sent	1	0.008713
Cross Reference Only	1	0.008713
Cease and Desist	1	0.008713
Med Jurisdiction Explained	1	0.008713
Filed Errors&Omission Clm	1	0.008713
Not Insurance Related	1	0.008713

Rare categories in Company (<= 5 occurrences):

Company	
C.M. Life Insurance Company	5
Northfield Insurance Company	5
Markel Insurance Company	5
Lincoln Life Assurance Company of Boston	5
United Financial Casualty Company	5
..	
Colony Specialty Insurance Company	1
First Allmerica Financial Life Insurance Company	1

```

Capitol Specialty Insurance Corporation      1
21st Century Premier Insurance Company      1
ARAG Insurance Company                      1
Name: count, Length: 453, dtype: int64

```

No rare categories found in Reason (≤ 5 occurrences).

Rare categories in Status (≤ 5 occurrences):

```

Status
Missing Information      4
Extension Granted       3
Preliminary Review - Standard  2
Sent to Agent           2
In Progress             2
New                     2
Recovery Pending        1
Legal Review            1
Verify Situs            1
No Response Follow-up    1
Name: count, dtype: int64

```

Rare categories in Conclusion (≤ 5 occurrences):

```

Conclusion
Accident in Another State  5
Underwriting Discretion   4
Rate Problem Solved       3
Interest Paid             3
Extl Rev Info Sent/SF     2
Deductible Recovered      2
Policy Offered            2
Class Revised             1
Complaint Form Sent       1
Cross Reference Only      1
Cease and Desist          1
Med Jurisdiction Explained 1
Filed Errors&Omission Clm  1
Not Insurance Related      1
Name: count, dtype: int64

```

```
<ipython-input-9-3ff3d4ef68f3>:53: FutureWarning:
```

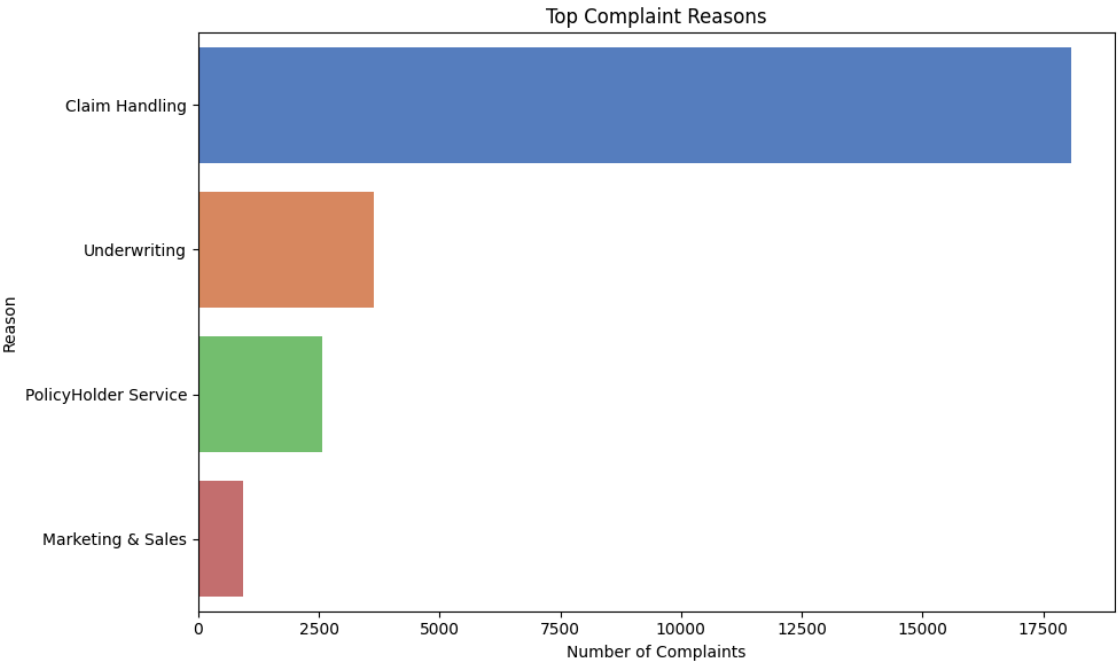
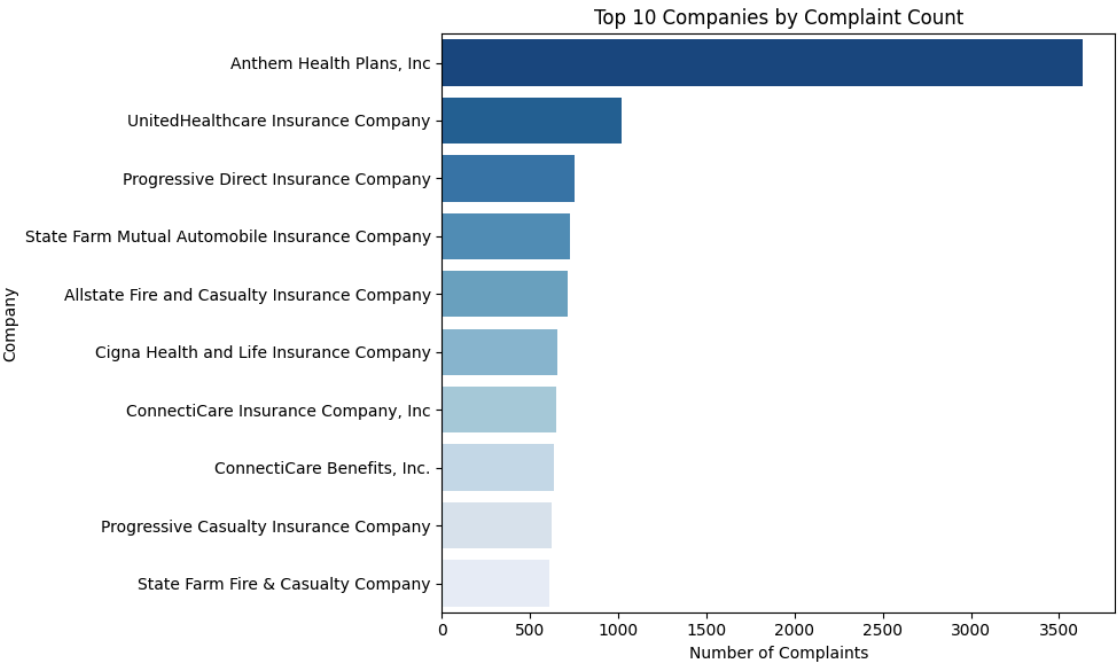
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

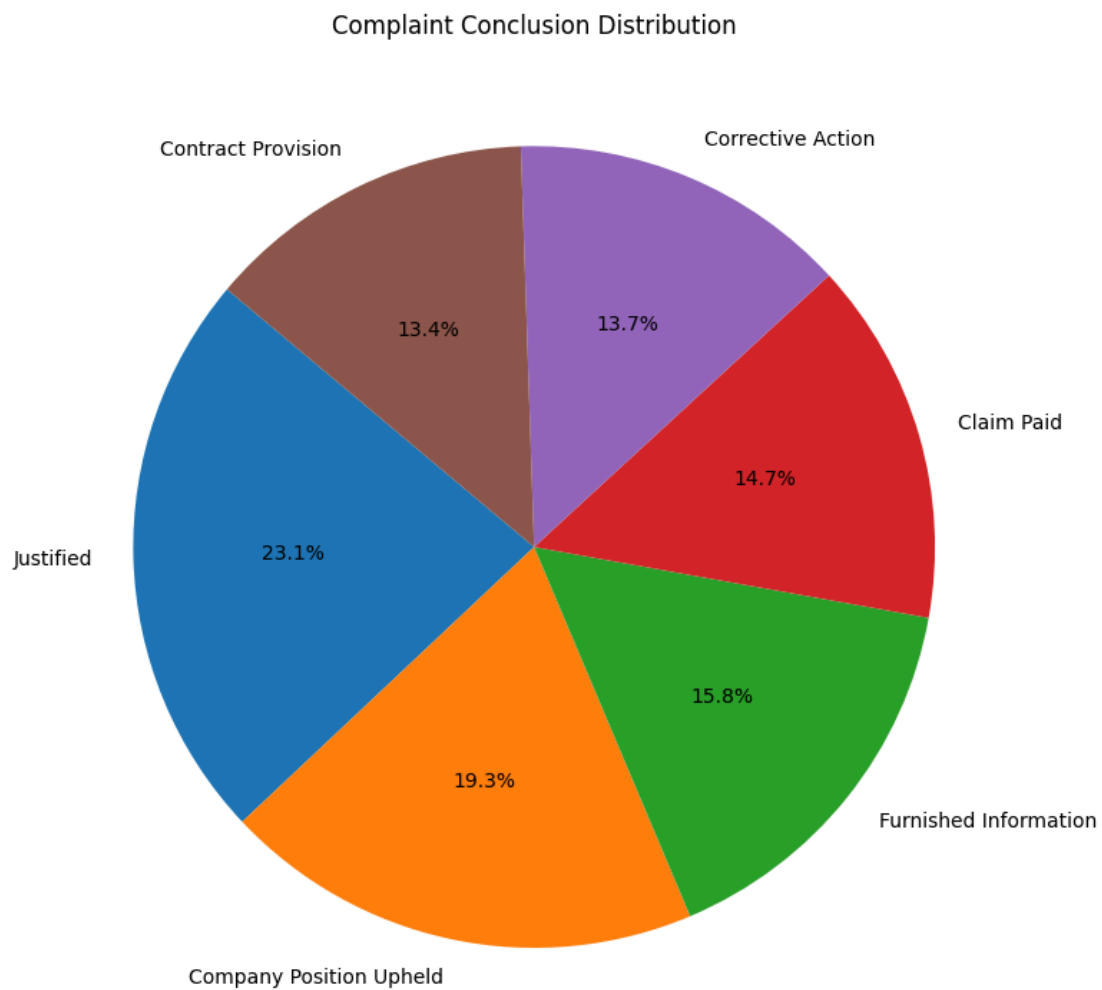
```
sns.barplot(x=top_companies.values, y=top_companies.index, palette='Blues_r')
```

```
<ipython-input-9-3ff3d4ef68f3>:64: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=top_reasons.values, y=top_reasons.index, palette='muted')
```





Extra

```
In [10]: df = pd.read_csv('data.csv')

if 'Company' in df.columns and 'Recovery' in df.columns:

    company_recovery_avg = df.groupby('Company')['Recovery'].mean().reset_index()

    company_recovery_count = df.groupby('Company').agg(
        recovered_count=('Recovery', lambda x: (x > 0).sum()), # Count non-
        not_recovered_count=('Recovery', lambda x: (x == 0).sum()) # Count
    ).reset_index()

    company_recovery_stats = pd.merge(company_recovery_avg, company_recovery_count)

    company_recovery_stats['recovery_percentage'] = (company_recovery_stats['recovered_count'] / (company_recovery_stats['recovered_count'] + company_recovery_stats['not_recovered_count'])) * 100
```



```
(company_recovery_stat
company_recovery_stat
```

```
top_companies = company_recovery_stats.nlargest(10, 'Recovery')

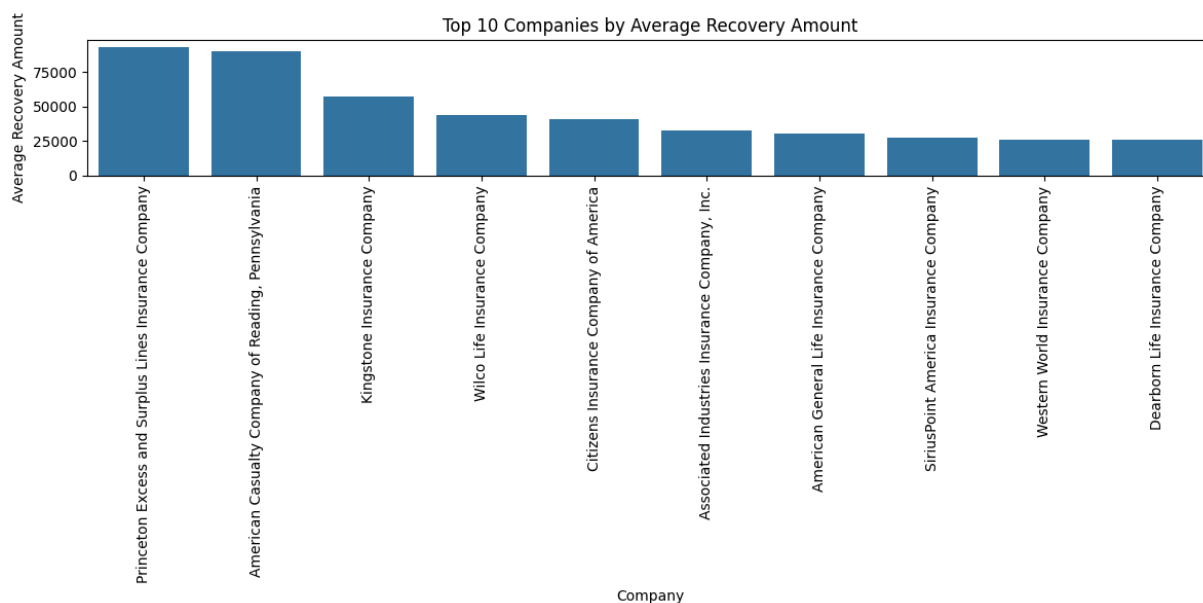
plt.figure(figsize=(12, 6))
sns.barplot(x='Company', y='Recovery', data=top_companies)
plt.xticks(rotation=90)
plt.title('Top 10 Companies by Average Recovery Amount')
plt.xlabel('Company')
plt.ylabel('Average Recovery Amount')
plt.tight_layout()

# Save the graph as a PNG file
plt.savefig('top_10_companies_vs_avg_recovery.png')

# Optionally, display the plot
plt.show()

# Print the recovery stats for the top 10 companies
print(top_companies[['Company', 'Recovery', 'recovered_count', 'not_recc

else:
    print("Necessary columns 'Company' and 'Recovery' are missing.")
```



	Company	Recovery \
615	Princeton Excess and Surplus Lines Insurance C...	93272.360000
57	American Casualty Company of Reading, Pennsylv...	89733.333333
426	Kingstone Insurance Company	57435.938261
834	Wilco Life Insurance Company	44071.456000
182	Citizens Insurance Company of America	40780.159200
110	Associated Industries Insurance Company, Inc.	32320.254000
67	American General Life Insurance Company	30556.362815
681	SiriusPoint America Insurance Company	27397.000000
831	Western World Insurance Company	26016.394167
232	Dearborn Life Insurance Company	25961.538462

	recovered_count	not_recovered_count	recovery_percentage
615	2	0	100.000000
57	5	4	55.555556
426	14	32	30.434783
834	2	3	40.000000
182	6	19	24.000000
110	1	4	20.000000
67	17	118	12.592593
681	1	1	50.000000
831	3	21	12.500000
232	3	10	23.076923

Recovery Analysis (with and without outliers)

```
In [ ]: file_path = "modified_data.csv"

# Ensure 'Recovery' is numeric
df['Recovery'] = pd.to_numeric(df['Recovery'], errors='coerce')

# Define outliers using the IQR method
Q1 = df['Recovery'].quantile(0.25)
Q3 = df['Recovery'].quantile(0.75)
IQR = Q3 - Q1
lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR

# Identify outliers
outliers = df[(df['Recovery'] < lower_bound) | (df['Recovery'] > upper_bound)]
non_outliers = df[(df['Recovery'] >= lower_bound) & (df['Recovery'] <= upper_bound)]

# Calculate statistics
mean_with_outliers = df['Recovery'].mean()
median_with_outliers = df['Recovery'].median()
mean_without_outliers = non_outliers['Recovery'].mean()
median_without_outliers = non_outliers['Recovery'].median()

# Print results
print(f"Mean with outliers: ${mean_with_outliers:.2f}")
print(f"Median with outliers: ${median_with_outliers:.2f}")
print(f"Mean without outliers: ${mean_without_outliers:.2f}")
print(f"Median without outliers: ${median_without_outliers:.2f}")
```

```
# Plot distribution
plt.figure(figsize=(10, 5))
sns.histplot(df['Recovery'], bins=50, kde=True)
plt.axvline(mean_with_outliers, color='red', linestyle='dashed', linewidth=1)
plt.axvline(median_with_outliers, color='blue', linestyle='dashed', linewidth=1)
plt.title('Distribution of Recovery Amounts')
plt.xlabel('Recovery Amount')
plt.ylabel('Frequency')
plt.legend()

# Save the plot
histogram_path = "recovery_histogram.png"
plt.savefig(histogram_path)
plt.close()

print(f"Histogram saved to: {histogram_path}")
```

Mean with outliers: \$1741.72

Median with outliers: \$0.00

Mean without outliers: \$0.00

Median without outliers: \$0.00

Histogram saved to: recovery_histogram.png

