

# Data Collection and Preprocessing Phase

Date	June 2024
Team ID	Team-739856
Project Title	Auto Insurance Fraud Detection Using Machine Learning
Maximum Marks	6 Marks

## Preparation Template

The images will be preprocessed by resizing, normalizing, augmenting, de noising, adjusting contrast, detecting edges, converting color space, cropping, batch normalizing, and whitening data. These steps will enhance data quality, promote model generalization, and improve convergence during neural network training, ensuring robust and efficient performance across various computer vision tasks.

Section	Description
Data Overview	There are many popular open sources for collecting the data. E.g.: kaggle.com, UCI repository, etc. In this project we have used .csv data.
Data Preparation	These are the general steps of pre-processing the data before using it for machine learning
Handling missing values	We use Handling missing values For checking the null values
Handling categorical data	As we can see our dataset has categorical data we must convert the categorical data to integer encoding or binary encoding
Handling Outliers in Data	With the help of boxplot, outliers are visualized. And here we are going to find upper bound and lower bound of numerical features with some mathematical formula.

# Data Preparation

Collect the dataset	<p>Please refer to the link given below to download the dataset.</p> <p>Link: <a href="https://www.kaggle.com/datasets/bunttyshah/auto-insurance-claims-data">https://www.kaggle.com/datasets/bunttyshah/auto-insurance-claims-data</a></p>
Importing the libraries	<pre> import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns from sklearn.model_selection import train_test_split from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier from sklearn.tree import DecisionTreeClassifier from sklearn.neighbors import KNeighborsClassifier from sklearn.metrics import f1_score from sklearn.metrics import classification_report, confusion_matrix import warnings import pickle from scipy import stats warnings.filterwarnings('ignore') </pre>
Loading Data	<p>We use the code</p> <pre>df=pd.read_csv("/content/Train.csv")</pre> <p>For reading the dataset</p>

Handling missing values

▶ `df.isna().any()`

```

→ months_as_customer      False
   age                    False
   policy_number           False
   policy_bind_date        False
   policy_state            False
   policy_csl              False
   policy_deductable       False
   policy_annual_premium   False
   umbrella_limit          False
   insured_zip             False
   insured_sex             False
   insured_education_level False
   insured_occupation      False
   insured_hobbies          False
   insured_relationship     False
   capital-gains            False
   capital-loss             False
   incident_date            False
   incident_type            False
   collision_type           False
   incident_severity        False
   authorities_contacted    True
   incident_state           False
   incident_city            False
   incident_location        False
   incident_hour_of_the_day False
   number_of_vehicles_involved False
   property_damage          False
   bodily_injuries          False
   witnesses                False
   police_report_available  False
   total_claim_amount       False
   injury_claim             False
   property_claim           False
   vehicle_claim            False
   auto_make                False
   auto_model               False
   auto_year                False
   fraud_reported           False
   _c39                     True
dtype: bool
  
```

## Handling Categorical values

```
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd # Import pandas for handling dataframes

for k in df.columns:
    if df[k].dtype in ['int64', 'float64']: # Check if column is numeric
        print(k)
        # Handle potential missing values
        sns.boxplot(df[k].dropna()) # Drop missing values before plotting
        plt.show()
    else:
        print(f"Skipping column '{k}' as it is non-numeric.") # Inform user about skipped columns
```

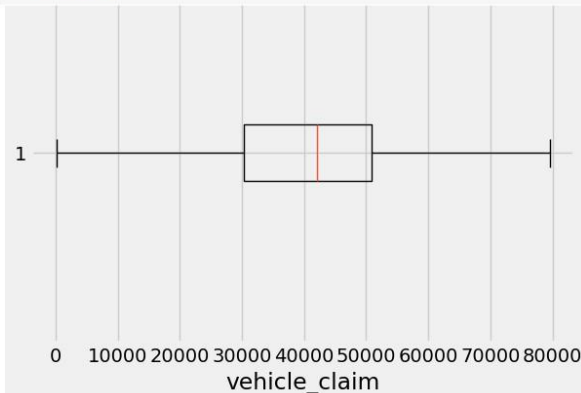


Skipping column 'incident\_date' as it is non-numeric.  
 Skipping column 'incident\_type' as it is non-numeric.  
 Skipping column 'collision\_type' as it is non-numeric.  
 Skipping column 'incident\_severity' as it is non-numeric.  
 Skipping column 'authorities\_contacted' as it is non-numeric.  
 Skipping column 'incident\_state' as it is non-numeric.  
 Skipping column 'incident\_city' as it is non-numeric.  
 Skipping column 'incident\_location' as it is non-numeric.  
 incident\_hour\_of\_the\_day

## Handling Outliers

```
#outlier handling

import matplotlib.pyplot as plt
df=pd.read_csv('/content/insurance_claims.csv')
for i in dict(df.dtypes):
    if dict(df.dtypes)[i] == 'int64' or dict(df.dtypes)[i] == 'float64':
        plt.boxplot(df[i], vert = False)
        plt.title("Detecting outlier using Boxplot")
        plt.xlabel(i)
        plt.show()
```



Detecting outlier using Boxplot