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import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
import matplotlib.pyplot as plt
import seaborn as sns

data = pd.read_csv("/content/drive/MyDrive/nearest-earth-objects(1910-2024).csv")

data.head()

{"type": "dataframe", "variable_name": "data"}

data.describe(include='all')

{"summary": "{\n  \"name\": \"data\", \n  \"rows\": 11, \n  \"fields\": [\n    {\n      \"column\": \"neo_id\", \n      \"properties\": {\n        \"dtype\": \"number\", \n        \"std\": 22716782.698697973, \n        \"min\": 338199.0, \n        \"max\": 54462807.0, \n        \"num_unique_values\": 8, \n        \"samples\": [\n          17599388.682018574, \n          3742127.0, \n          338199.0\n        ], \n        \"semantic_type\": \"\", \n        \"description\": \"\"\n      }\n    }, \n    {\n      \"column\": \"name\", \n      \"properties\": {\n        \"dtype\": \"category\", \n        \"num_unique_values\": 4, \n        \"samples\": [\n          33514, \n          \"211\", \n          \"338199\"\n        ], \n        \"semantic_type\": \"\", \n        \"description\": \"\"\n      }\n    }, \n    {\n      \"column\": \"absolute_magnitude\", \n      \"properties\": {\n        \"dtype\": \"number\", \n        \"std\": 119554.56863782684, \n        \"min\": 2.911216390292293, \n        \"max\": 338171.0, \n        \"num_unique_values\": 8, \n        \"samples\": [\n          22.93252495926617, \n          22.8, \n          338171.0\n        ], \n        \"semantic_type\": \"\", \n        \"description\": \"\"\n      }\n    }, \n    {\n      \"column\": \"estimated_diameter_min\", \n      \"properties\": {\n        \"dtype\": \"number\", \n        \"std\": 119559.56966825192, \n        \"min\": 0.0005111578, \n        \"max\": 338171.0, \n        \"num_unique_values\": 8, \n        \"samples\": [\n          0.1578120466605549, \n          0.0732073989, \n          338171.0\n        ], \n        \"semantic_type\": \"\", \n        \"description\": \"\"\n      }\n    }, \n    {\n      \"column\": \"estimated_diameter_max\", \n      \"properties\": {\n        \"dtype\": \"number\", \n        \"std\": 119557.18111686174, \n        \"min\": 0.0011429835, \n        \"max\": 338171.0, \n        \"num_unique_values\": 8, \n        \"samples\": [\n          0.3528784640005492, \n          0.1636967205, \n          338171.0\n        ], \n        \"semantic_type\": \"\", \n        \"description\": \"\"\n      }\n    }, \n    {\n      \"column\": \"orbiting_body\", \n      \"properties\": {\n        \"dtype\": \"category\", \n        \"num_unique_values\": 3, \n        \"samples\": [\n          \"338199\", \n          1, \n          \"Earth\"\n        ], \n        \"semantic_type\": \"\", \n        \"description\": \"\"\n      }\n    }\n  ]\n}"

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\"semantic_type\": \"\", \n          \"description\": \"\" \n      } \n  }, \n  { \n    \"column\": \"relative_velocity\", \n    \"properties\": { \n      \"dtype\": \"number\", \n      \"std\": 130715.5645630245, \n      \"min\": 203.34643253, \n      \"max\": 338199.0, \n      \"num_unique_values\": 8, \n      \"samples\": [ \n        51060.662907595586, \n        47560.4654744848, \n        338199.0 \n      ], \n      \"semantic_type\": \"\", \n      \"description\": \"\" \n    } \n  }, \n  { \n    \"column\": \"miss_distance\", \n    \"properties\": { \n      \"dtype\": \"number\", \n      \"std\": 26699262.28644739, \n      \"min\": 6745.532515957, \n      \"max\": 74798651.4521972, \n      \"num_unique_values\": 8, \n      \"samples\": [ \n        41535350.93219019, \n        43326743.82834823, \n        338199.0 \n      ], \n      \"semantic_type\": \"\", \n      \"description\": \"\" \n    } \n  }, \n  { \n    \"column\": \"is_hazardous\", \n    \"properties\": { \n      \"dtype\": \"category\", \n      \"num_unique_values\": 4, \n      \"samples\": [ \n        295037, \n        338199 \n      ], \n      \"semantic_type\": \"\", \n      \"description\": \"\" \n    } \n  } \n ] \n }\", \"type\": \"dataframe\"}

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data.info()
data.isnull().sum()
data = data.drop(['neo_id', 'name', 'orbiting_body'], axis = 1)
data.head()

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<class 'pandas.core.frame.DataFrame'>
RangeIndex: 338199 entries, 0 to 338198
Data columns (total 9 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   neo_id                                338199 non-null  int64
1   name                                  338199 non-null  object
2   absolute_magnitude                    338171 non-null  float64
3   estimated_diameter_min                 338171 non-null  float64
4   estimated_diameter_max                 338171 non-null  float64
5   orbiting_body                          338199 non-null  object
6   relative_velocity                     338199 non-null  float64
7   miss_distance                         338199 non-null  float64
8   is_hazardous                          338199 non-null  bool
dtypes: bool(1), float64(5), int64(1), object(2)
memory usage: 21.0+ MB

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{"type": "dataframe", "variable_name": "data"}

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data['is_hazardous'] = data['is_hazardous'].map({True:1 , False:0})
data.head()

```

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{"type": "dataframe", "variable_name": "data"}

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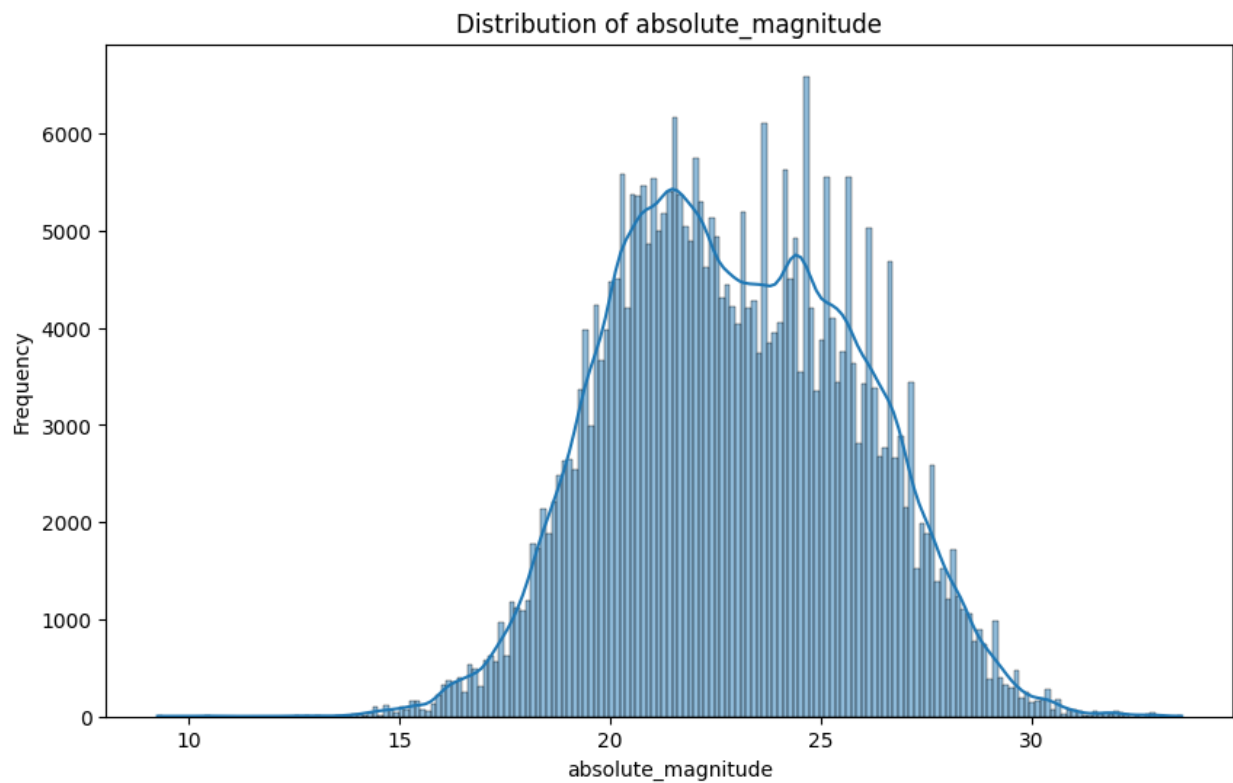
```

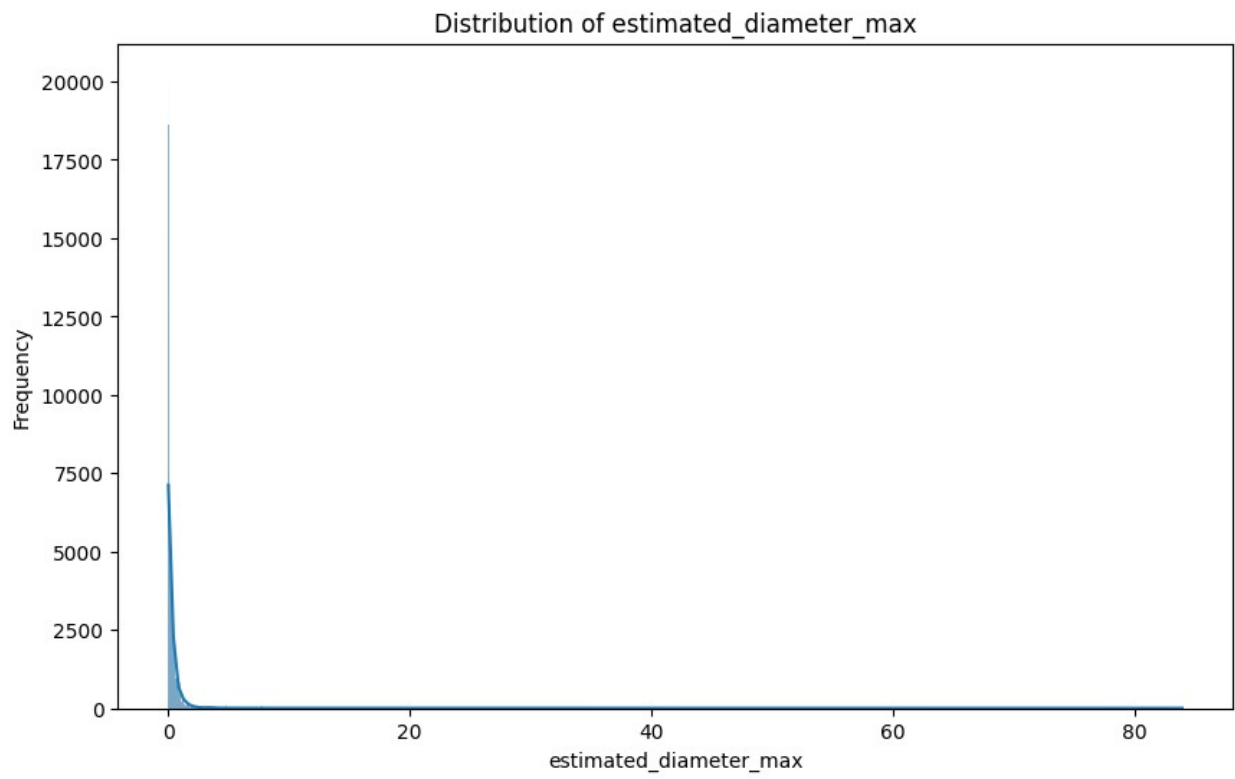
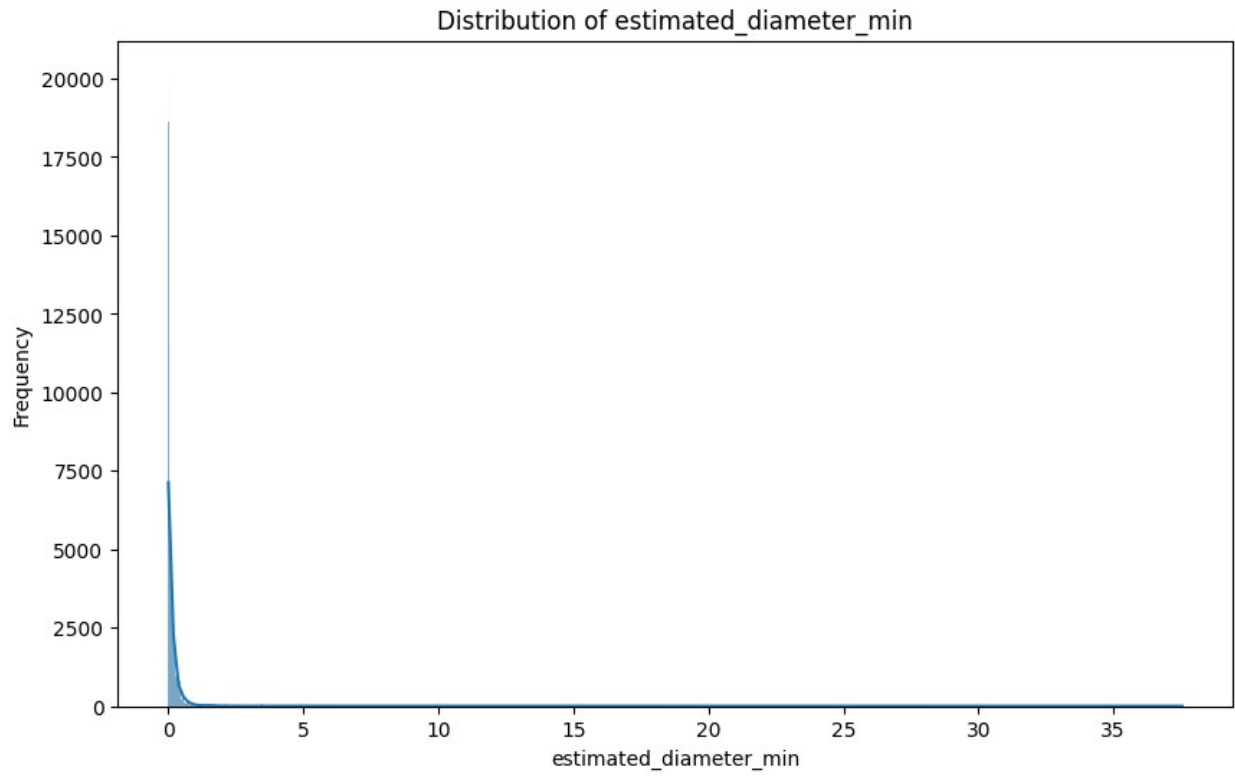
columns = data.columns

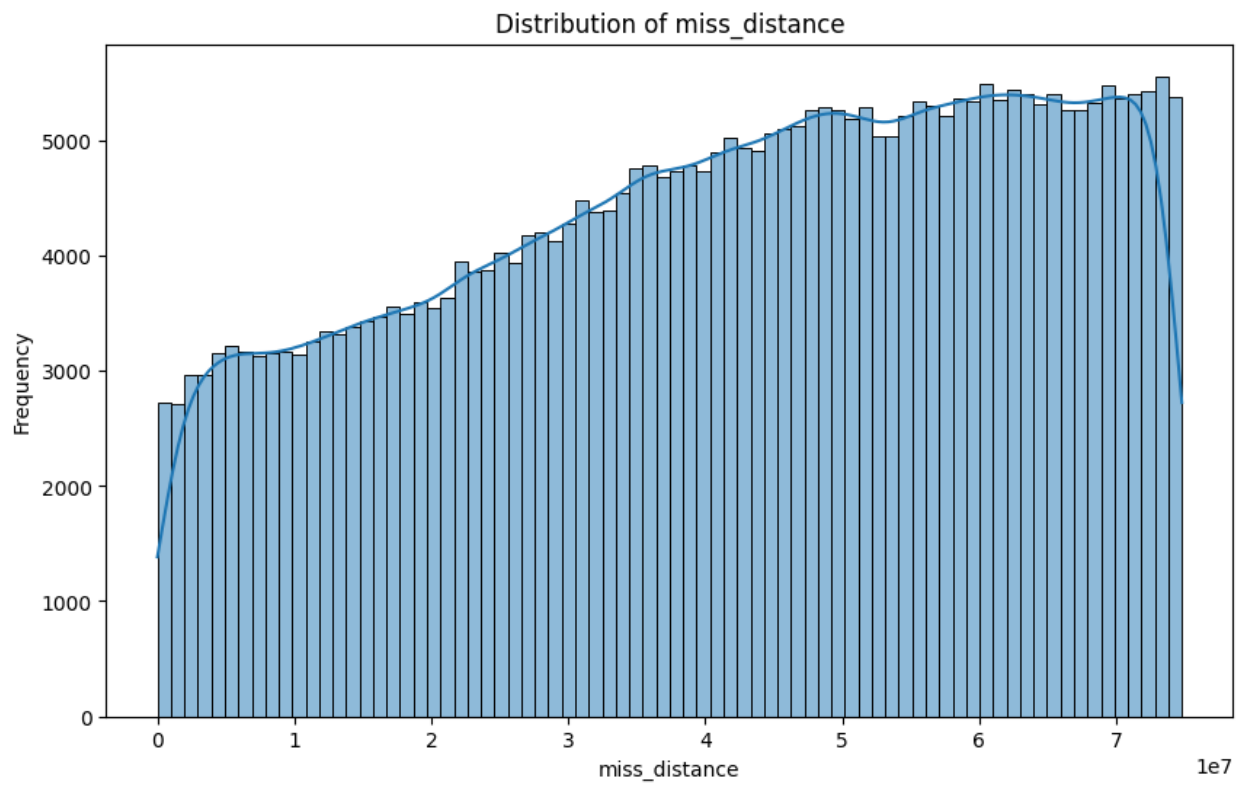
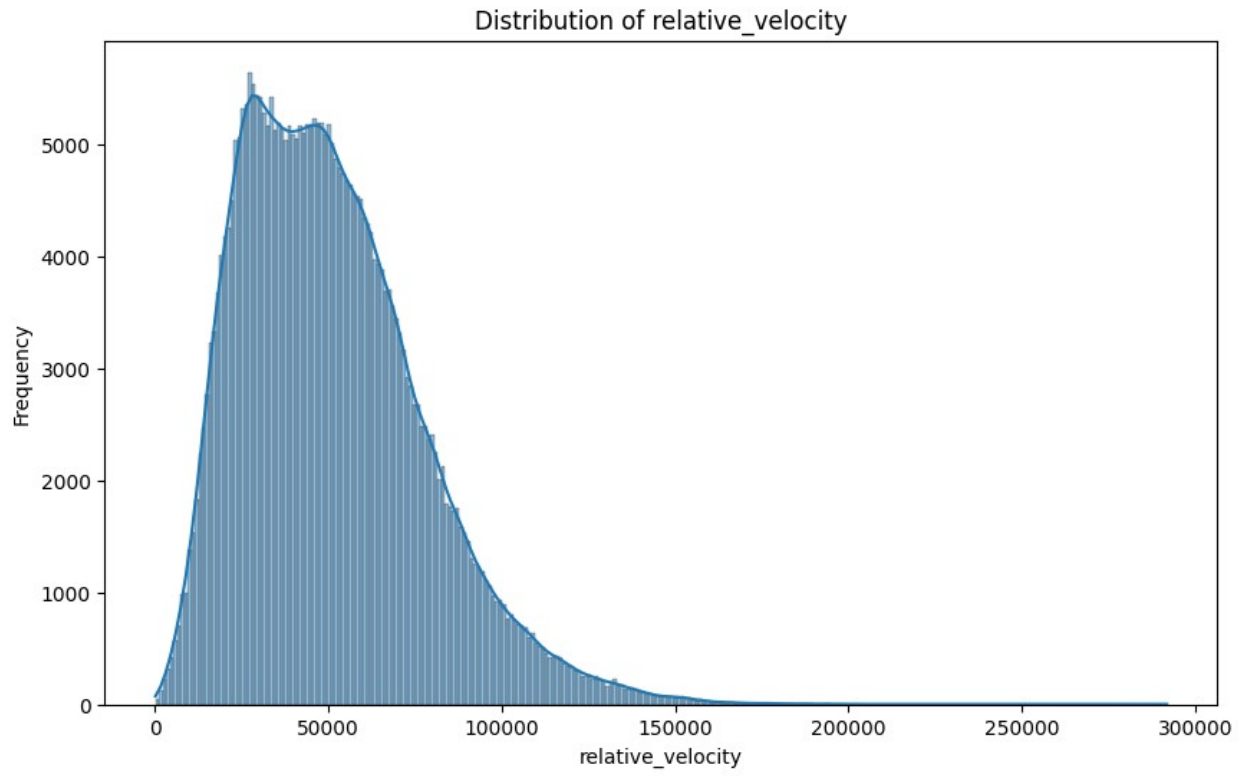
# Plot each column
for col in columns:
    plt.figure(figsize=(10, 6))
    sns.histplot(data[col], kde=True)
    plt.title(f'Distribution of {col}')
    plt.xlabel(col)
    plt.ylabel('Frequency')
    plt.show()

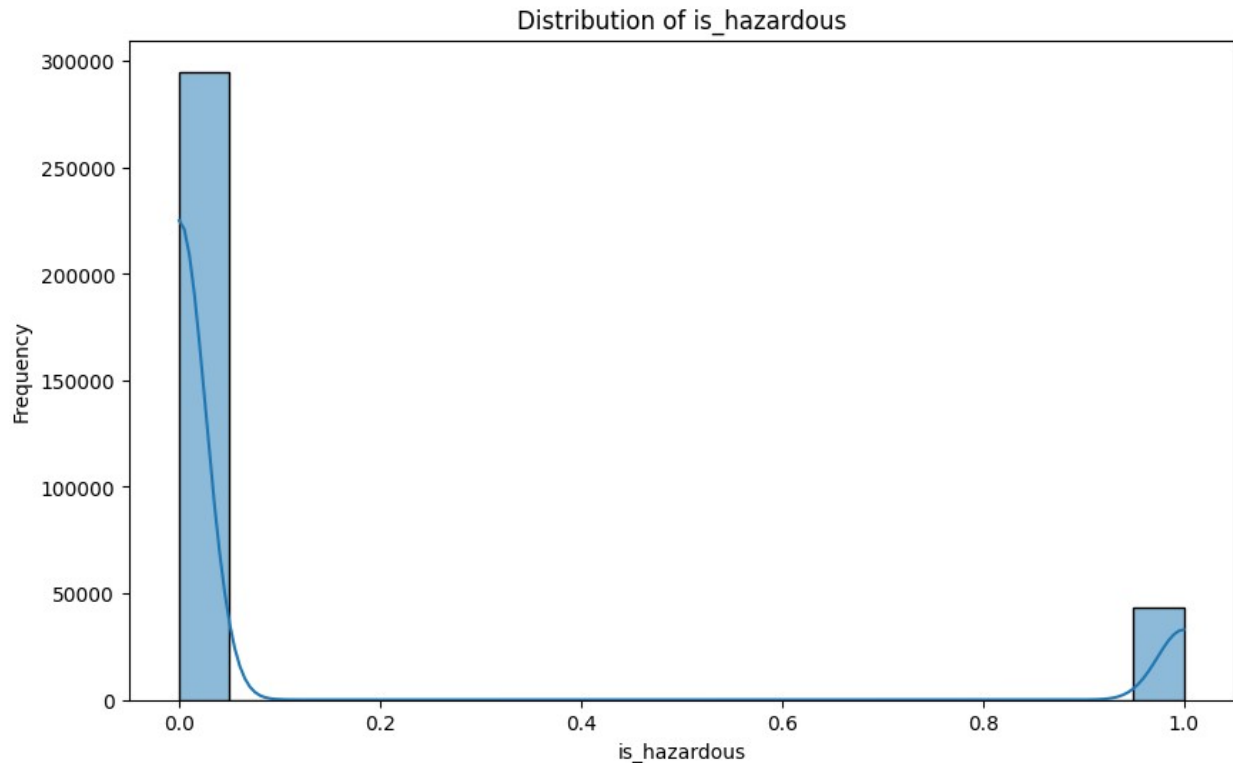
data = data.dropna()

```









```
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
x = data.drop('is_hazardous',axis = 1)
y = data['is_hazardous']
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x_scaled = scaler.fit_transform(x)
print(x_scaled)
```

```
print(x)
print(y)
```

```
[[-1.30273056  0.75553115  0.75553115  0.78354069  0.79946485]
 [-1.52257027  1.18683296  1.18683296  2.23068176  0.68670052]
 [-0.50924664 -0.06847576 -0.06847576 -0.9922207  1.2357423 ]
 ...
 [-0.34814535 -0.15283782 -0.15283782 -1.48591476  0.5740475 ]
 [ 0.32786173 -0.36139181 -0.36139181  0.19463588 -1.7498115 ]
 [ 0.00634617 -0.28520788 -0.28520788 -0.34089772 -1.65633252]]
```

```
absolute_magnitude  estimated_diameter_min
estimated_diameter_max \
0                      19.140                0.394962
0.883161
1                      18.500                0.530341
1.185878
2                      21.450                0.136319
0.304818
```

3	20.630	0.198863
0.444672		
4	22.700	0.076658
0.171412		
...
...		
338194	28.580	0.005112
0.011430		
338195	28.690	0.004859
0.010865		
338196	21.919	0.109839
0.245607		
338197	23.887	0.044377
0.099229		
338198	22.951	0.068290
0.152700		

	relative_velocity	miss_distance
0	71745.401048	5.814362e+07
1	109949.757148	5.580105e+07
2	24865.506798	6.720689e+07
3	78890.076805	3.039644e+07
4	56036.519484	6.311863e+07
...
338194	56646.985988	6.406548e+07
338195	21130.768947	2.948883e+07
338196	11832.041031	5.346078e+07
338197	56198.382733	5.184742e+06
338198	42060.357830	7.126682e+06

[338171 rows x 5 columns]

0	0
1	1
2	0
3	0
4	0
...	...
338194	0
338195	0
338196	0
338197	0
338198	0

Name: is_hazardous, Length: 338171, dtype: int64

```
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test = train_test_split(x_scaled,y,test_size
= 0.2 , random_state = 42)
print(X_train)
```

```
[[-1.45387036  1.03816614  1.03816614  2.94555792  0.99482004]
 [-0.06269724 -0.26410792 -0.26410792 -0.46793005 -1.1647807 ]
 [ 1.34908585 -0.46681368 -0.46681368 -0.82639138  0.31651595]
 ...
 [-0.49894165 -0.07443451 -0.07443451 -1.06662046  1.26623063]
 [ 1.56892556 -0.47599238 -0.47599238 -1.10754607  0.68708169]
 [ 1.39717579 -0.46905877 -0.46905877  0.32777902  1.30049185]]
```

```
from sklearn.neighbors import KNeighborsClassifier
Knn = KNeighborsClassifier(n_neighbors=10)
Knn.fit(X_train,y_train)
```

```
KNeighborsClassifier(n_neighbors=10)
```

```
y_pred = Knn.predict(X_test)
```

```
from sklearn.metrics import accuracy_score
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
```

```
Accuracy: 0.8851630073186959
```