```
from google.colab import files
import pandas as pd

data = pd.read_csv('/content/new_predictions (1) (1).csv')
# Preview the dataset
data.head()
```

→ Property_Number specialNeeds roofDamage foundationDamage householdComposition destroyed residenceType repairAmount grossIncom 2 0 0 0 0 0 Apartment 0.00 5200 2 0 0 4 0 House/Duplex 3698.52 30000 1 2 3 0 0 0 2 0 House/Duplex 0.00 378 0 3 1804 3 4 0 0 0 House/Duplex 0.00 2 4 5 0 0 0 0 House/Duplex 0.00

Get an overview of the dataset
data.info()

<<class 'pandas.core.frame.DataFrame'>
 RangeIndex: 1048575 entries, 0 to 1048574
 Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype						
0	Property_Number	1048575 non-null	int64						
1	specialNeeds	1048575 non-null	int64						
2	roofDamage	1048575 non-null	int64						
3	foundationDamage	1048575 non-null	int64						
4	householdComposition	1048575 non-null	int64						
5	destroyed	1048575 non-null	int64						
6	residenceType	1048575 non-null	object						
7	repairAmount	1048575 non-null	float64						
8	grossIncome	1048575 non-null	int64						
9	waterLevel	1048575 non-null	int64						
10	tsa_eligible	1048575 non-null	float64						
<pre>dtypes: float64(2), int64(8), object(1)</pre>									
memory usage: 88.0+ MB									

Check for missing values

print("Missing values in each column:")
print(data.isnull().sum())

→ Missing values in each column: Property_Number specialNeeds roofDamage 0 ${\it foundation} {\it Damage}$ 0 householdComposition destroyed 0 residenceType0 repairAmount 0 grossIncome 0 waterLevel 0 tsa_eligible 0 dtype: int64

Display summary statistics for numerical columns
data.describe()



	Property_Number	specialNeeds	roofDamage	foundationDamage	$household {\tt Composition}$	destroyed	repairAmount	grossIncome	Wć
count	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.04
mean	5.242880e+05	3.386882e-02	3.667072e-02	5.400663e-03	2.253349e+00	1.754762e-03	3.609742e+02	4.903923e+04	8.41
std	3.026977e+05	1.808916e-01	1.879522e-01	7.329052e-02	1.459048e+00	4.185313e-02	2.189897e+03	5.572152e+05	5.50
min	1.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	1.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.00
25%	2.621445e+05	0.000000e+00	0.000000e+00	0.000000e+00	1.000000e+00	0.000000e+00	0.000000e+00	3.876000e+03	0.00
50%	5.242880e+05	0.000000e+00	0.000000e+00	0.000000e+00	2.000000e+00	0.000000e+00	0.000000e+00	2.000000e+04	0.00
75%	7.864315e+05	0.000000e+00	0.000000e+00	0.000000e+00	3.000000e+00	0.000000e+00	0.000000e+00	4.000000e+04	0.00
max	1.048575e+06	1.000000e+00	1.000000e+00	1.000000e+00	9.900000e+01	1.000000e+00	4.635872e+04	9.950000e+07	3.60
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```
# Count TSA-eligible vs. non-eligible families
tsa_counts = data['tsa_eligible'].value_counts()
print("TSA Eligibility Counts:\n", tsa_counts)

TSA Eligibility Counts:
```

tsa_eligible 0.0 925773 1.0 122802

Name: count, dtype: int64

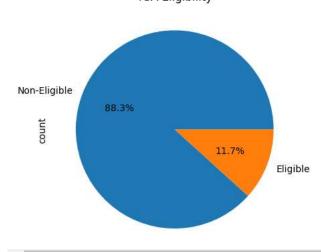
import matplotlib.pyplot as plt

```
# TSA eligibility counts
tsa_counts = data['tsa_eligible'].value_counts()
# Pie chart
tsa_counts.plot.pie(
    autopct='%1.1f%%',
```

autopct='%1.1f%%',
labels=['Non-Eligible', 'Eligible'],
title="TSA Eligibility"
)
plt.show()

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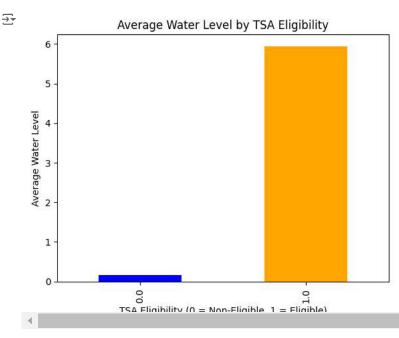
TSA Eligibility



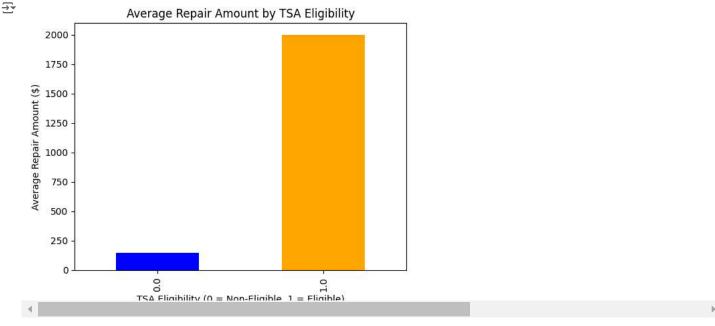
Average water level for TSA-eligible vs. non-eligible families
water_level_analysis = data.groupby('tsa_eligible')['waterLevel'].mean()
print("\nAverage Water Level by TSA Eligibility:")
print(water_level_analysis)

Average Water Level by TSA Eligibility: tsa_eligible 0.0 0.164329 1.0 5.942053
Name: waterLevel, dtype: float64

```
# Bar chart for average water level
water_level_analysis = data.groupby('tsa_eligible')['waterLevel'].mean()
water_level_analysis.plot.bar(
    color=['blue', 'orange'],
    title="Average Water Level by TSA Eligibility"
)
plt.xlabel('TSA Eligibility (0 = Non-Eligible, 1 = Eligible)')
plt.ylabel('Average Water Level')
plt.show()
```



```
# Average repair amount for TSA-eligible vs. non-eligible families
repair_amount_analysis = data.groupby('tsa_eligible')['repairAmount'].mean()
print("\nAverage Repair Amount by TSA Eligibility:")
print(repair_amount_analysis)
<del>_</del>
     Average Repair Amount by TSA Eligibility:
     {\tt tsa\_eligible}
     0.0
             143.594574
     1.0
            1999.744042
     Name: repairAmount, dtype: float64
# Bar chart for repair amount
repair_amount_analysis = data.groupby('tsa_eligible')['repairAmount'].mean()
repair_amount_analysis.plot.bar(
    color=['blue', 'orange'],
    title="Average Repair Amount by TSA Eligibility"
plt.xlabel('TSA Eligibility (0 = Non-Eligible, 1 = Eligible)')
plt.ylabel('Average Repair Amount ($)')
plt.show()
```

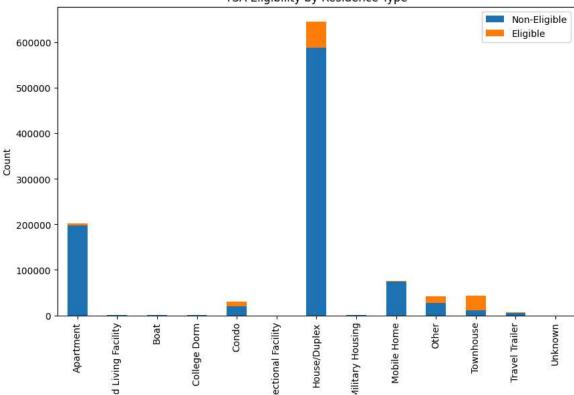


```
# Average gross income for TSA-eligible vs. non-eligible families
income_analysis = data.groupby('tsa_eligible')['grossIncome'].mean()
print("\nAverage Gross Income by TSA Eligibility:")
print(income_analysis)
\overline{2}
     Average Gross Income by TSA Eligibility:
     tsa eligible
           47156.511984
     0.0
     1.0
            63232.550577
     Name: grossIncome, dtype: float64
# Count of residence types by TSA eligibility
residence_tsa = data.groupby(['residenceType', 'tsa_eligible']).size().unstack(fill_value=0)
print("\nResidence Type by TSA Eligibility:")
print(residence_tsa)
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     Residence Type by TSA Eligibility:
     tsa_eligible
                                   0.0
                                          1.0
     residenceType
     Apartment
                               197921
                                         4516
     Assisted Living Facility
                                 1370
                                           29
     Boat
                                   679
                                           44
     College Dorm
                                   351
                                         9753
     Condo
                                19772
     Correctional Facility
                                  101
     House/Duplex
                                587662
                                        58439
     Military Housing
                                  518
     Mobile Home
                                74152
                                         1183
     Other
                                26736
                                       14871
     Townhouse
                                10692 33228
     Travel Trailer
                                 5768
                                          735
     Unknown
                                    51
                                            0
# Stacked bar chart for residence type
residence_tsa.plot(
    kind='bar',
    stacked=True,
    figsize=(10, 6),
    title="TSA Eligibility by Residence Type"
plt.xlabel('Residence Type')
plt.ylabel('Count')
plt.legend(['Non-Eligible', 'Eligible'])
plt.show()
```



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TSA Eligibility by Residence Type



```
# Count of special needs households by TSA eligibility
special_needs_tsa = data.groupby('tsa_eligible')['specialNeeds'].sum()
print("\nSpecial Needs by TSA Eligibility:")
print(special_needs_tsa)
```

```
Special Needs by TSA Eligibility:
tsa_eligible
0.0 31769
1.0 3745
Name: specialNeeds, dtype: int64

# Bar chart for special needs
special_needs_tsa.plot.bar(
    color=['blue', 'orange'],
    title="Special Needs Households by TSA Eligibility"
)
plt.xlabel('TSA Eligibility (0 = Non-Eligible, 1 = Eligible)')
plt.ylabel('Count')
```

