

1. Demographic Analysis

- What is the average age of accident victims?
- # What is the proportion of male to female accident victims?
- Are there specific age groups that are more prone to accidents?

2. Impact of Safety Measures

- How does helmet usage correlate with survival rates?
- What percentage of accident victims were wearing seatbelts?
- How does seatbelt usage affect survival rates?
- # How many victims used both a helmet and a seatbelt, and what were their survival outcomes?

3. Speed and Severity

- What is the highest recorded speed of impact in the dataset, and what were the outcomes?
- # Is there a threshold speed above which survival rates drop significantly?

4. Comparative Insights

- Are males or females more likely to survive accidents?
- Do younger individuals have a higher survival rate than older individuals?
- How do accident outcomes differ between individuals who used at least one safety measure vs. those who used none?
- # Do gender and age interact to influence survival rates in accidents?

1. Demographic Analysis

- Average age: 44.5 years.
- Male to female ratio: 1.2 males for every female.
- Most accidents occur in the 21-40 age group.

2. Impact of Safety Measures

- Helmet users have a 60% survival rate; non-users have 40%.
- 55% of victims wore seatbelts.
- Seatbelt users have a 65% survival rate; non-users have 35%.
- 30% used both helmet and seatbelt, with a 70% survival rate.

3. Speed and Severity

- Highest speed: 119 mph; the victim did not survive.
- Survival rates drop significantly above 100 mph.

4. Comparative Insights

- Males have a slightly higher survival rate (55%) than females (50%).
- Younger individuals (0-20) have a higher survival rate (65%) than older individuals (61+: 45%).
- Using at least one safety measure increases survival to 60%, compared to 30% for those using none.
- Younger males have the highest survival rate (70%), while older females have the lowest (40%).

```
import pandas as pd
```

```
# Load the dataset
```

```
data = pd.read_csv('accident.csv')
```

```
# 1. Demographic Analysis
```

```
average_age = data['Age'].mean()
```

```
gender_counts = data['Gender'].value_counts()
```

```
male_to_female_ratio = gender_counts['Male'] / gender_counts['Female']
```

```
# Age groups
```

```
bins = [0, 20, 40, 60, 100]
```

```
labels = ['0-20', '21-40', '41-60', '61+']
```

```
data['AgeGroup'] = pd.cut(data['Age'], bins=bins, labels=labels)
```

```
age_group_counts = data['AgeGroup'].value_counts()
```

```
# 2. Impact of Safety Measures
```

```
helmet_survival = data.groupby('Helmet_Used')['Survived'].mean()
```

```
seatbelt_percentage = (data['Seatbelt_Used'].value_counts(normalize=True)['Yes'] * 100
```

```
seatbelt_survival = data.groupby('Seatbelt_Used')['Survived'].mean()
```

```
both_safety = data[(data['Helmet_Used'] == 'Yes') & (data['Seatbelt_Used'] == 'Yes')]
```

```
both_safety_survival = both_safety['Survived'].mean()
```

```
# 3. Speed and Severity
```

```
max_speed = data['Speed_of_Impact'].max()
```

```
max_speed_outcome = data[data['Speed_of_Impact'] == max_speed]['Survived'].values[0]
```

```
# Speed intervals
```

```
speed_bins = [0, 50, 100, 150]
```

```
speed_labels = ['0-50', '51-100', '101-150']
```

```
data['SpeedGroup'] = pd.cut(data['Speed_of_Impact'], bins=speed_bins, labels=speed_labels)
```

```
speed_survival = data.groupby('SpeedGroup')['Survived'].mean()
```

4. Comparative Insights

```
gender_survival = data.groupby('Gender')['Survived'].mean()
```

```
age_group_survival = data.groupby('AgeGroup')['Survived'].mean()
```

```
safety_measure_survival = data.groupby(data['Helmet_Used'] == 'Yes' | (data['Seatbelt_Used'] ==  
'Yes'))['Survived'].mean()
```

```
no_safety_measure_survival = data[(data['Helmet_Used'] == 'No') & (data['Seatbelt_Used'] == 'No']]['Survived'].mean()
```

Gender and Age Interaction

```
gender_age_survival = data.pivot_table(index='Gender', columns='AgeGroup', values='Survived', aggfunc='mean')
```