

```
from google.colab import files
uploaded = files.upload()
```



Choose Files Titanic-Dataset.csv

- **Titanic-Dataset.csv**(text/csv) - 61194 bytes, last modified: 6/9/2025 - 100% done  
Saving Titanic-Dataset.csv to Titanic-Dataset.csv

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# Set plot styles
sns.set(style="whitegrid")
plt.rcParams["figure.figsize"] = (10, 6)
```

```
# Load the dataset
df = pd.read_csv('Titanic-Dataset.csv')
```

```
# 1. Basic Data Overview
print("First 5 rows:")
print(df.head())
```

```
print("\nData Info:")
print(df.info())
```

```
print("\nStatistical Summary:")
print(df.describe(include='all'))
```

```
print("\nMissing Values:")
print(df.isnull().sum())
```

```
print("\nDuplicate Rows:", df.duplicated().sum())
```



mean	440.000000	0.500000	2.500000	NaN	NaN
std	257.353842	0.486592	0.836071	NaN	NaN
min	1.000000	0.000000	1.000000	NaN	NaN
25%	223.500000	0.000000	2.000000	NaN	NaN
50%	446.000000	0.000000	3.000000	NaN	NaN
75%	668.500000	1.000000	3.000000	NaN	NaN
max	891.000000	1.000000	3.000000	NaN	NaN

	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
count	714.000000	891.000000	891.000000	891	891.000000	204	889
unique	NaN	NaN	NaN	681	NaN	147	3
top	NaN	NaN	NaN	347082	NaN	G6	S
freq	NaN	NaN	NaN	7	NaN	4	644
mean	29.699118	0.523008	0.381594	NaN	32.204208	NaN	NaN
std	14.526497	1.102743	0.806057	NaN	49.693429	NaN	NaN
min	0.420000	0.000000	0.000000	NaN	0.000000	NaN	NaN
25%	20.125000	0.000000	0.000000	NaN	7.910400	NaN	NaN
50%	28.000000	0.000000	0.000000	NaN	14.454200	NaN	NaN
75%	38.000000	1.000000	0.000000	NaN	31.000000	NaN	NaN
max	80.000000	8.000000	6.000000	NaN	512.329200	NaN	NaN

Missing Values:

PassengerId	0
Survived	0
Pclass	0
Name	0
Sex	0
Age	177
SibSp	0
Parch	0
Ticket	0
Fare	0
Cabin	687
Embarked	2

dtype: int64

Duplicate Rows: 0

```
print("\nValue Counts for Survived:")
print(df['Survived'].value_counts())
```

```
sns.countplot(x='Survived', data=df)
plt.title("Survival Count")
plt.show()
```

```
# Numerical: Age
df['Age'].hist(bins=30)
plt.title("Age Distribution")
plt.xlabel("Age")
plt.ylabel("Frequency")
plt.show()
```

```
sns.boxplot(x='Age', data=df)
plt.title("Boxplot of Age")
plt.show()
```

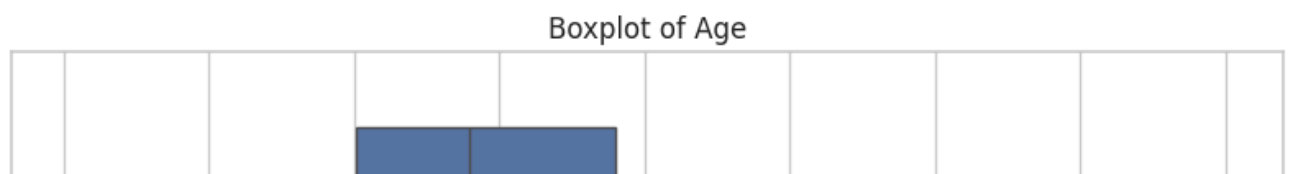
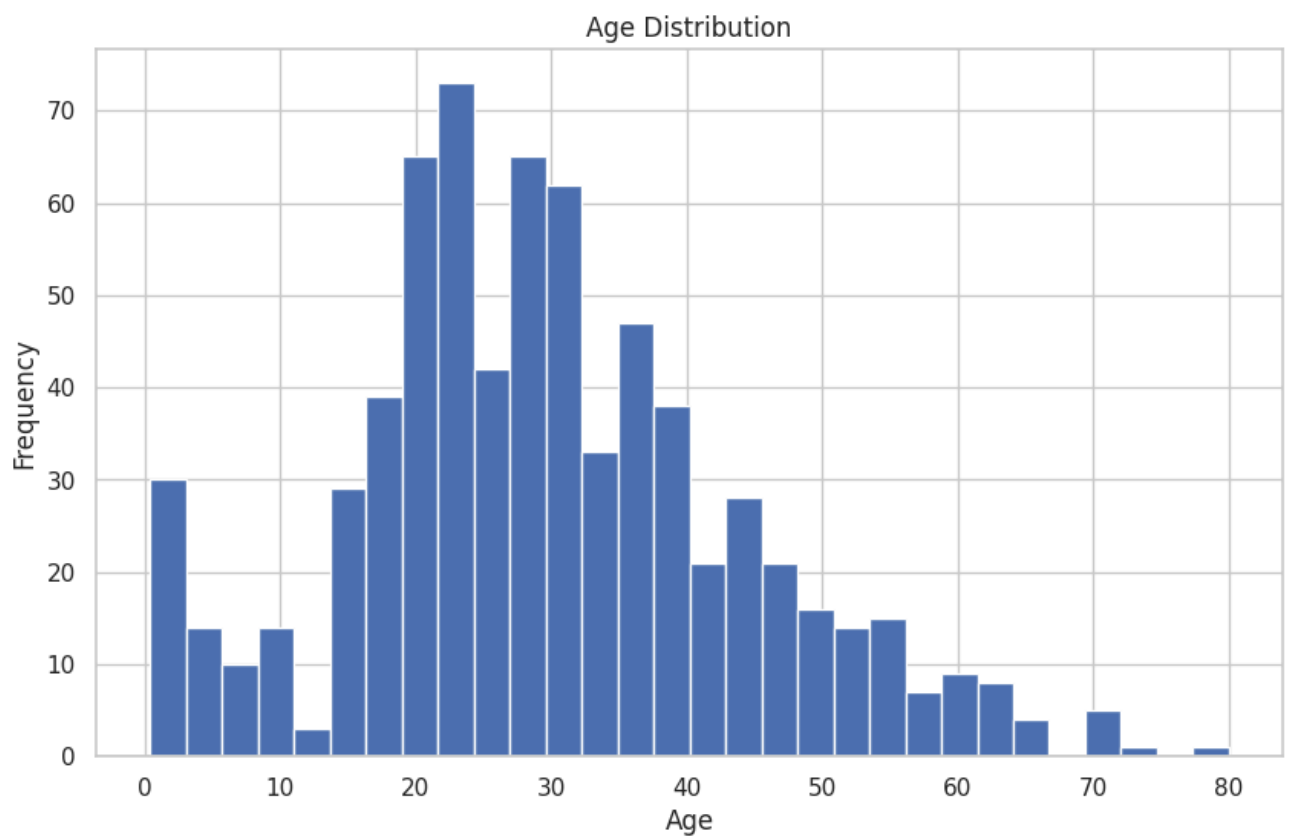
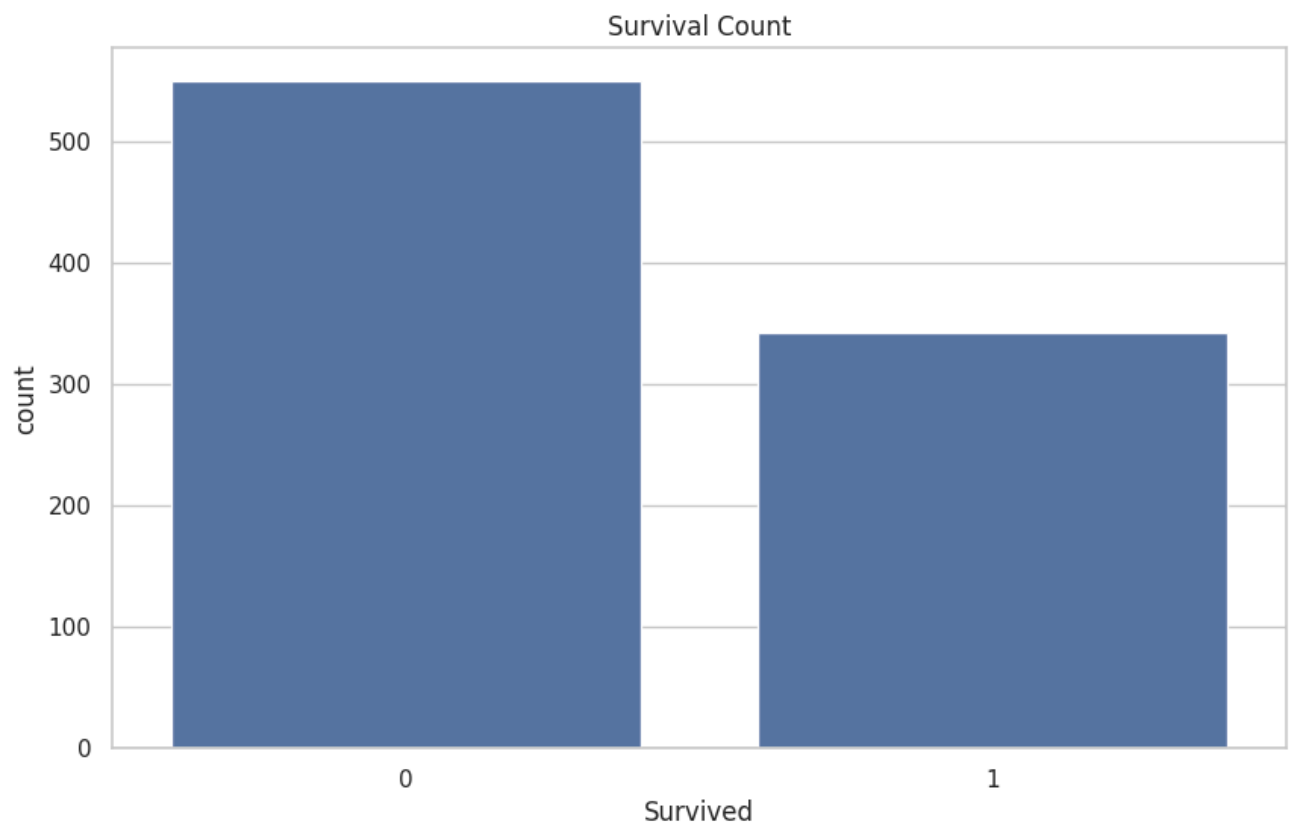


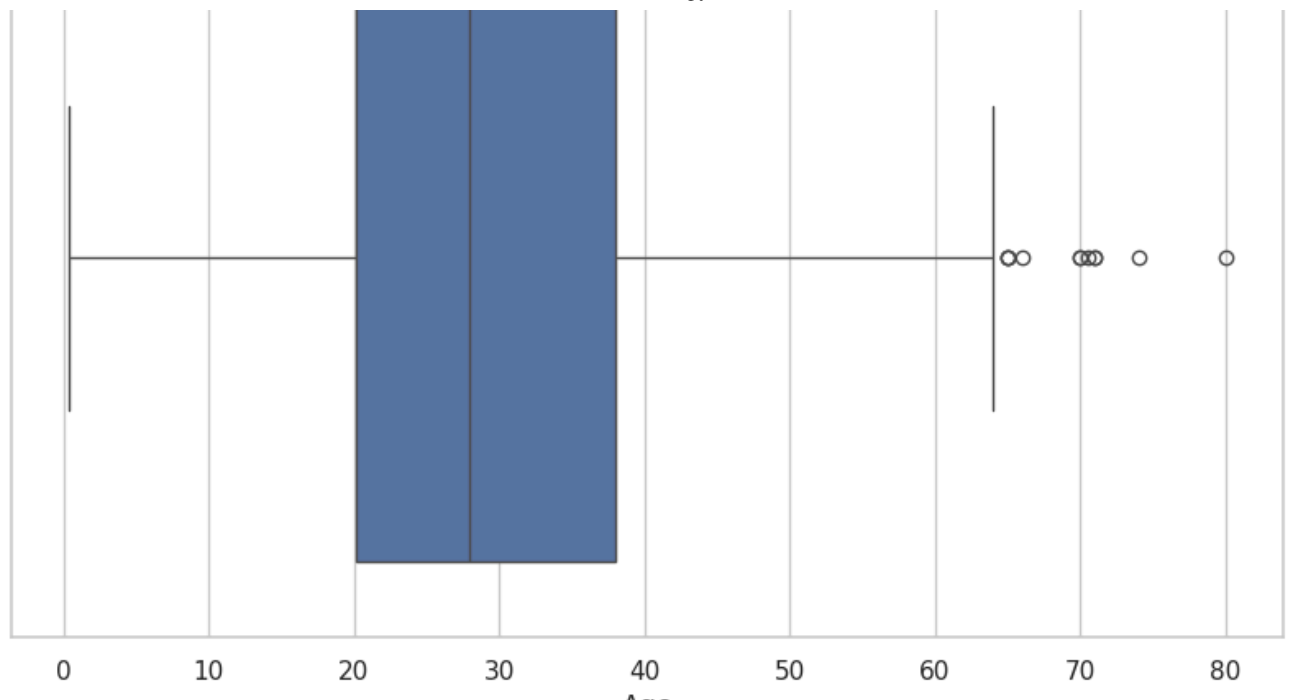
Survived

0 549

1 342

Name: count, dtype: int64

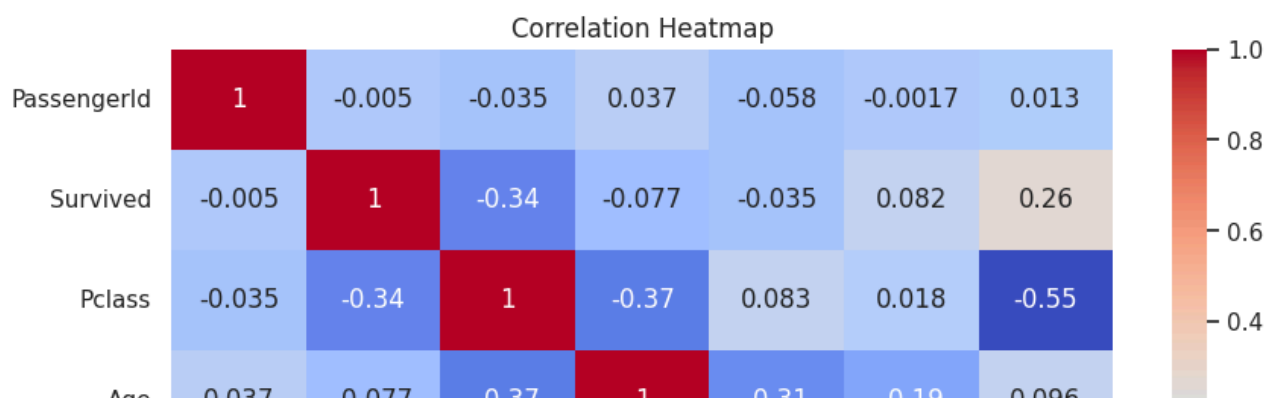
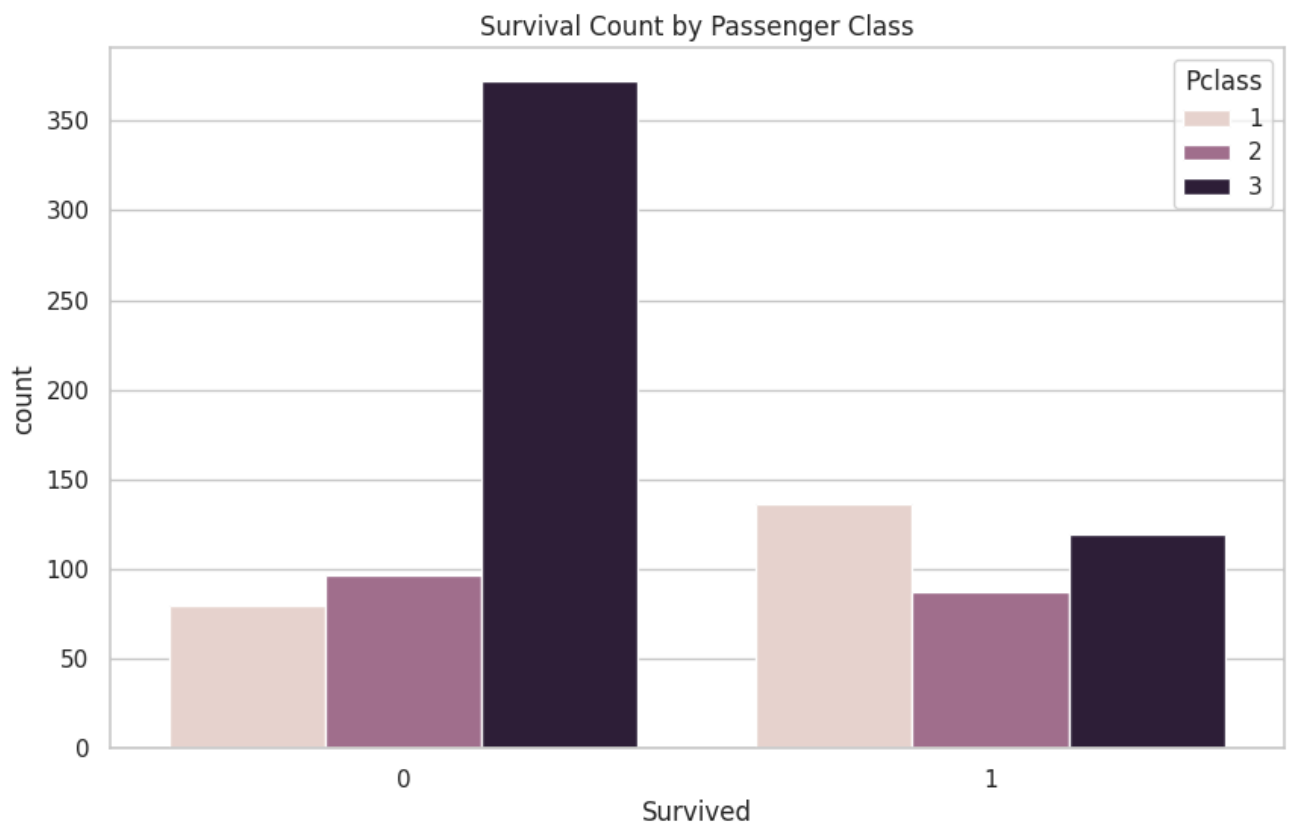
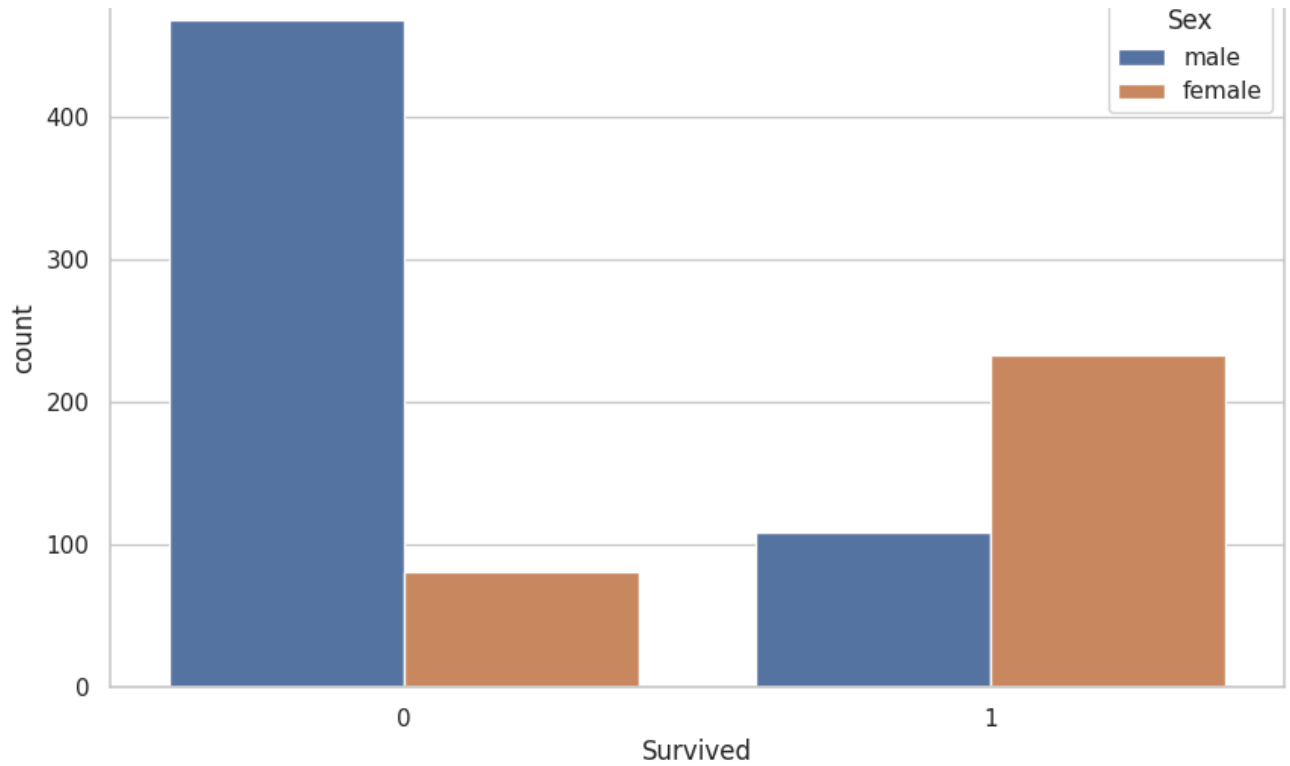


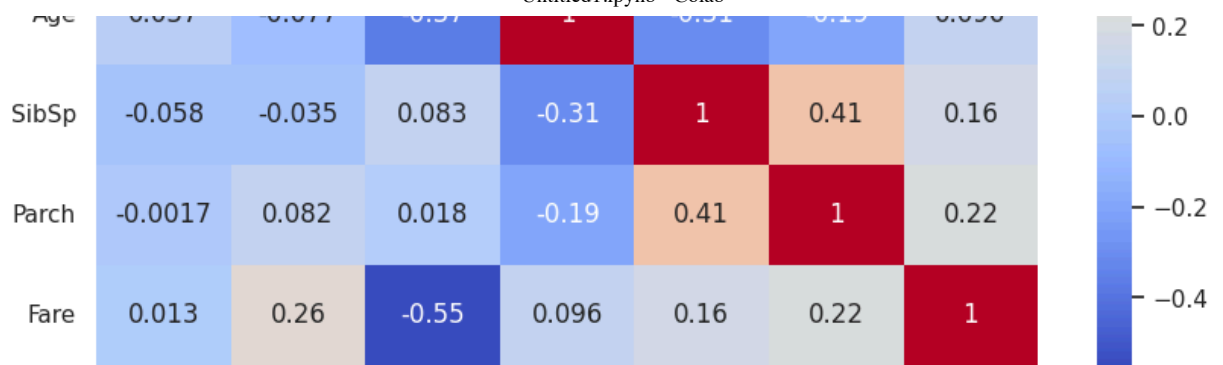


```
sns.countplot(x='Survived', hue='Sex', data=df)
plt.title("Survival Count by Gender")
plt.show()
```

```
# Survived vs Pclass
sns.countplot(x='Survived', hue='Pclass', data=df)
plt.title("Survival Count by Passenger Class")
plt.show()
```

```
# Heatmap of correlations (numeric features)
numeric_df = df.select_dtypes(include=['int64', 'float64'])
sns.heatmap(numeric_df.corr(), annot=True, cmap='coolwarm')
plt.title("Correlation Heatmap")
plt.show()
```





```
print("\nAverage Age by Survival Status:")
print(df.groupby('Survived')['Age'].mean())
```

```
print("\nSurvival Rate by Passenger Class:")
print(df.groupby('Pclass')['Survived'].mean())
```



Average Age by Survival Status:

Survived

0 30.626179

1 28.343690

Name: Age, dtype: float64

Survival Rate by Passenger Class:

Pclass

1 0.629630

2 0.472826

3 0.242363

Name: Survived, dtype: float64

```
print("\nSummary:")
print("- More passengers did not survive (0) than survived (1).")
print("- Females had a higher survival rate than males.")
print("- Passengers in higher classes (Pclass=1) had better chances of survival.")
print("- Age distribution is right-skewed, with many young passengers.")
```



Summary:

- More passengers did not survive (0) than survived (1).
- Females had a higher survival rate than males.
- Passengers in higher classes (Pclass=1) had better chances of survival.
- Age distribution is right-skewed, with many young passengers.

```
!apt-get install -y texlive-xetex texlive-fonts-recommended texlive-gene
!jupyter nbconvert --to pdf --output "Titanic_EDA_Report.pdf" "Titanic_E
```



```

the notebook version to write.
    Use this to downgrade notebooks.
Choices: any of [1, 2, 3, 4]
Default: 4
Equivalent to: [--NotebookExporter.nbformat_version]

```

## Examples

-----

The simplest way to use nbconvert is

```
> jupyter nbconvert mynotebook.ipynb --to html
```

Options include ['asciidoc', 'custom', 'html', 'latex', 'markdown']

```
> jupyter nbconvert --to latex mynotebook.ipynb
```

Both HTML and LaTeX support multiple output templates. LaTeX includes 'base', 'article' and 'report'. HTML includes 'basic', 'lab' and 'classic'. You can specify the flavor of the format used.

```
> jupyter nbconvert --to html --template lab mynotebook.ipynb
```

You can also pipe the output to stdout, rather than a file

```
> jupyter nbconvert mynotebook.ipynb --stdout
```

PDF is generated via latex

```
> jupyter nbconvert mynotebook.ipynb --to pdf
```

You can get (and serve) a Reveal.js-powered slideshow

```
> jupyter nbconvert myslides.ipynb --to slides --post serve
```

Multiple notebooks can be given at the command line in a couple of different ways:

```
> jupyter nbconvert notebook*.ipynb
```

```
> jupyter nbconvert notebook1.ipynb notebook2.ipynb
```

or you can specify the notebooks list in a config file, containing

```
c.NbConvertApp.notebooks = ["my_notebook.ipynb"]
```

```
> jupyter nbconvert --config mycfg.py
```

To see all available configurables, use `--help-all`.

```

from google.colab import files
files.download("Titanic_EDA_Report.pdf")

```