

# **Titanic - Machine Learning from Disaster**

Machine Learning Project – Kaggle Titanic Competition

# The Challenge

- The sinking of the Titanic is one of the most infamous shipwrecks in history.
- On April 15, 1912, during her maiden voyage, the widely considered “unsinkable” RMS Titanic sank after colliding with an iceberg. Unfortunately, there weren’t enough lifeboats for everyone onboard, resulting in the death of 1502 out of 2224 passengers and crew.
- While there was some element of luck involved in surviving, it seems some groups of people were more likely to survive than others.
- In this challenge, we ask you to build a predictive model that answers the question: “what sorts of people were more likely to survive?” using passenger data (i.e name, age, gender, socio-economic class, etc).

# Dataset Overview (Kaggle Titanic Competition)

The dataset consists of **three CSV files**:

## 1. **train.csv**

- Contains data for **891 passengers**.
- Includes the "**Survived**" column:
  - 1 → Passenger survived
  - 0 → Passenger did not survive
- Used to **train** and understand survival patterns.

# Dataset Overview (Kaggle Titanic Competition)

## 2. test.csv

- Contains data for 418 passengers.
- Does not include the "Survived" column.
- My task is to predict survival for these passengers.

## 3. gender\_submission.csv

A sample submission file:

- Assumes all females survived and all males did not.

Shows the correct format for your submission.csv:

- Columns: PassengerId, Survived

# Data Preprocessing on training dataset

**Import the pandas library**

```
[43]: import pandas as pd
```

**Read the data**

```
[44]: titanic_train = pd.read_csv("/kaggle/input/train-data/train.csv")
```

# Data Preprocessing on training dataset

```
[45]: titanic_train.head()
```

```
[45]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...)	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

# Data Preprocessing on training dataset

```
titanic_train.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 891 entries, 0 to 890  
Data columns (total 12 columns):  
#   Column      Non-Null Count  Dtype  
---  -  
0   PassengerId  891 non-null    int64  
1   Survived     891 non-null    int64  
2   Pclass       891 non-null    int64  
3   Name         891 non-null    object  
4   Sex          891 non-null    object  
5   Age          714 non-null    float64  
6   SibSp        891 non-null    int64  
7   Parch        891 non-null    int64  
8   Ticket       891 non-null    object  
9   Fare         891 non-null    float64  
10  Cabin        204 non-null    object  
11  Embarked     889 non-null    object  
dtypes: float64(2), int64(5), object(5)  
memory usage: 83.7+ KB
```

# Data Preprocessing on training dataset

## ➤ Fill the missing values

```
[47]: titanic_train['Age'] = titanic_train['Age'].fillna(titanic_train['Age'].median())
```

```
[48]: titanic_train['Embarked'] = titanic_train['Embarked'].fillna(titanic_train['Embarked'].mode()[0], inplace=True)
```



# Data Preprocessing on training dataset



```
titanic_train.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 891 entries, 0 to 890  
Data columns (total 12 columns):  
#   Column      Non-Null Count  Dtype  
---  -  
0   PassengerId  891 non-null    int64  
1   Survived     891 non-null    int64  
2   Pclass       891 non-null    int64  
3   Name         891 non-null    object  
4   Sex          891 non-null    object  
5   Age          891 non-null    float64  
6   SibSp        891 non-null    int64  
7   Parch        891 non-null    int64  
8   Ticket       891 non-null    object  
9   Fare         891 non-null    float64  
10  Cabin        204 non-null    object  
11  Embarked     891 non-null    object  
dtypes: float64(2), int64(5), object(5)  
memory usage: 83.7+ KB
```

# Data Preprocessing on training dataset

## ➤ Encoding Categorical Variables

```
[50]: from sklearn.preprocessing import LabelEncoder  
  
le = LabelEncoder()  
titanic_train['Embarked'] = titanic_train['Embarked'].map({'S': 0, 'C': 1, 'Q': 2}).astype(int)
```

```
[51]: titanic_train['Sex'] = titanic_train['Sex'].map({'male': 0, 'female': 1}).astype(int)
```

# Data Preprocessing on training dataset

- See the data after encoding

```
▶ titanic_train.head()
```

```
[52]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	0	22.0	1	0	A/5 21171	7.2500	NaN	0
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	1	38.0	1	0	PC 17599	71.2833	C85	1
2	3	1	3	Heikkinen, Miss. Laina	1	26.0	0	0	STON/O2. 3101282	7.9250	NaN	0
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	1	35.0	1	0	113803	53.1000	C123	0
4	5	0	3	Allen, Mr. William Henry	0	35.0	0	0	373450	8.0500	NaN	0

# Data Preprocessing on training dataset

## ➤ Drop the unnecessary columns

```
[53]: titanic_train.drop(['Ticket', 'Cabin', 'Name'], axis=1, inplace=True)
```

# Model Training using Random Forest

## ➤ Feature and Target Separation

```
[54]: X = titanic_train.drop('Survived',axis=1)
      y = titanic_train['Survived']
```

This separated the dataset into:

- X containing the input features (all columns except 'Survived')
- y containing the target variable 'Survived' for model training.

# Model Training using Random Forest

X								
	PassengerId	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	1	3	0	22.0	1	0	7.2500	0
1	2	1	1	38.0	1	0	71.2833	1
2	3	3	1	26.0	0	0	7.9250	0
3	4	1	1	35.0	1	0	53.1000	0
4	5	3	0	35.0	0	0	8.0500	0
...	...	...	...	...	...	...	...	...
886	887	2	0	27.0	0	0	13.0000	0
887	888	1	1	19.0	0	0	30.0000	0
888	889	3	1	28.0	1	2	23.4500	0
889	890	1	0	26.0	0	0	30.0000	1
890	891	3	0	32.0	0	0	7.7500	2

891 rows × 8 columns

```
[57]: y
```

0	0
1	1
2	1
3	1
4	0
...	...
886	0
887	1
888	0
889	1
890	0

Name: Survived, Length: 891, dtype: int64

+ Code + Markdown

# Model Training using Random Forest

## ➤ Splitting the data into training and testing

```
: from sklearn.model_selection import train_test_split
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 42)
```

```
[60]: print(X_train.shape)
      print(X_test.shape)
      print(y_train.shape)
      print(y_test.shape)
```

```
(712, 8)
(179, 8)
(712,)
(179,)
```

# Model Training using Random Forest

## ➤ Import the Randomforest Model

```
[61]: from sklearn.ensemble import RandomForestClassifier  
      from sklearn.metrics import accuracy_score
```

```
[62]: model = RandomForestClassifier(random_state = 42)
```



# Model Training using Random Forest

- Train the model on training data

```
▶ model.fit(X_train,y_train)
```

[63]: RandomForestClassifier

RandomForestClassifier(random\_state=42)

# Model Training using Random Forest

## ➤ Predictions of the model on testing data

```
[65]: y_pred = model.predict(X_test)
```

```
[47]: comparison_df = pd.DataFrame({
        'Actual': y_test.values,
        'Predicted': y_pred
    })

print(comparison_df.head(10))
```

	Actual	Predicted
0	1	0
1	0	0
2	0	0
3	1	1
4	1	0
5	1	1
6	1	1
7	0	0
8	1	1
9	1	1

# Model Training using Random Forest

## ➤ Accuracy score of this model

```
[48]: accuracy_score(y_pred, y_test)
```

```
[48]: 0.8324022346368715
```

# Data Preprocessing on testing data

## ➤ Read the testing data

```
[68]: titanic_test = pd.read_csv("/kaggle/input/test-dataset/test.csv")
```

```
[69]: titanic_test.head()
```

```
[69]:
```

	PassengerId	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	892	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	NaN	Q
1	893	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	NaN	S
2	894	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	NaN	Q
3	895	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	NaN	S
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	NaN	S

# Data Preprocessing on testing data

```
[70]: titanic_test.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 11 columns):
 #   Column      Non-Null Count  Dtype  
---  -
 0   PassengerId 418 non-null   int64  
 1   Pclass      418 non-null   int64  
 2   Name        418 non-null   object  
 3   Sex         418 non-null   object  
 4   Age         332 non-null   float64 
 5   SibSp       418 non-null   int64  
 6   Parch       418 non-null   int64  
 7   Ticket      418 non-null   object  
 8   Fare        417 non-null   float64 
 9   Cabin       91 non-null    object  
10   Embarked    418 non-null   object  
dtypes: float64(2), int64(4), object(5)
memory usage: 36.1+ KB
```

# Data Preprocessing on testing data

## ➤ Filling the missing values

```
[71]: titanic_test['Age'] = titanic_test['Age'].fillna(titanic_test['Age'].median())
```

```
[78]: titanic_test['Fare'] = titanic_test['Fare'].fillna(titanic_test['Fare'].median())
```

# Data Preprocessing on testing data

## ➤ Encoding Categorical Variables

```
[72]: from sklearn.preprocessing import LabelEncoder  
  
le = LabelEncoder()  
titanic_test['Embarked'] = titanic_test['Embarked'].map({'S': 0, 'C': 1, 'Q': 2}).astype(int)
```

```
[73]: titanic_test['Sex'] = titanic_test['Sex'].map({'male': 0, 'female': 1}).astype(int)
```

# Data Preprocessing on testing data

## ➤ Preview of data after encoding

```
[74]: titanic_test.head()
```

```
[74]:
```

	PassengerId	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	892	3	Kelly, Mr. James	0	34.5	0	0	330911	7.8292	NaN	2
1	893	3	Wilkes, Mrs. James (Ellen Needs)	1	47.0	1	0	363272	7.0000	NaN	0
2	894	2	Myles, Mr. Thomas Francis	0	62.0	0	0	240276	9.6875	NaN	2
3	895	3	Wirz, Mr. Albert	0	27.0	0	0	315154	8.6625	NaN	0
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	1	22.0	1	1	3101298	12.2875	NaN	0



# Data Preprocessing on testing data

## ➤ Removing the unnecessary columns

```
[76]: titanic_test.drop(['Ticket', 'Cabin', 'Name'], axis=1, inplace=True)
```

```
[79]: titanic_test.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 8 columns):
 #   Column      Non-Null Count  Dtype  
---  -
 0   PassengerId 418 non-null   int64  
 1   Pclass      418 non-null   int64  
 2   Sex         418 non-null   object  
 3   Age         418 non-null   float64 
 4   SibSp       418 non-null   int64  
 5   Parch       418 non-null   int64  
 6   Fare        418 non-null   float64 
 7   Embarked    418 non-null   object  
dtypes: float64(2), int64(6)
memory usage: 26.3 KB
```

# Predicting the testing data

## ➤ Predicting the testing data

```
[82]: y_test_pred = model.predict(titanic_test)
```

# Predicting the testing data

## ➤ Downloading the submission file

```
[83]: submission = pd.DataFrame({  
      'PassengerId': titanic_test['PassengerId'],  
      'Survived': y_test_pred  
    })  
  
      submission.to_csv('submission.csv', index=False)
```

# Predicting the testing data

- See the predictions of the model on testing data

[65]:

```
print(submission.tail(10))
```

	PassengerId	Survived
408	1300	1
409	1301	1
410	1302	1
411	1303	1
412	1304	0
413	1305	0
414	1306	1
415	1307	0
416	1308	0
417	1309	0

# My submission score on the Kaggle leaderboard

## Submissions

All

Successful

Errors

Recent ▼

Submission and Description

Public Score ⓘ



**submission.csv**

Complete · 3h ago · I used a Random Forest Classifier to predict Titanic passenger survival based on features like Pclass, Sex, A...

**0.78708**