

Titanic

July 30, 2022

1 Titanic Kaggle Challenge

1.1 Downloading the Dataset

```
[438]: import os
import opendatasets as od
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
%matplotlib inline
sns.set_style('darkgrid')
```

```
[439]: os.listdir('.')
```

```
[439]: ['gender_submission.csv', 'test.csv', 'Titanic.ipynb', 'train.csv']
```

```
[440]: data_df = pd.read_csv('./train.csv')
test_df = pd.read_csv('./test.csv')
```

```
[441]: data_df
```

```
[441]:
```

	PassengerId	Survived	Pclass	\
0	1	0	3	
1	2	1	1	
2	3	1	3	
3	4	1	1	
4	5	0	3	
..	
886	887	0	2	
887	888	1	1	
888	889	0	3	
889	890	1	1	
890	891	0	3	

	Name	Sex	Age	SibSp	\
0	Braund, Mr. Owen Harris	male	22.0	1	

1	Cummings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1
2	Heikkinen, Miss. Laina	female	26.0	0
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1
4	Allen, Mr. William Henry	male	35.0	0
..
886	Montvila, Rev. Juozas	male	27.0	0
887	Graham, Miss. Margaret Edith	female	19.0	0
888	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1
889	Behr, Mr. Karl Howell	male	26.0	0
890	Dooley, Mr. Patrick	male	32.0	0

	Parch	Ticket	Fare	Cabin	Embarked
0	0	A/5 21171	7.2500	NaN	S
1	0	PC 17599	71.2833	C85	C
2	0	STON/O2. 3101282	7.9250	NaN	S
3	0	113803	53.1000	C123	S
4	0	373450	8.0500	NaN	S
..
886	0	211536	13.0000	NaN	S
887	0	112053	30.0000	B42	S
888	2	W./C. 6607	23.4500	NaN	S
889	0	111369	30.0000	C148	C
890	0	370376	7.7500	NaN	Q

[891 rows x 12 columns]

[442]: test_df

[442]:

	PassengerId	Pclass	Name \
0	892	3	Kelly, Mr. James
1	893	3	Wilkes, Mrs. James (Ellen Needs)
2	894	2	Myles, Mr. Thomas Francis
3	895	3	Wirz, Mr. Albert
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)
..
413	1305	3	Spector, Mr. Woolf
414	1306	1	Oliva y Ocana, Dona. Fermina
415	1307	3	Saether, Mr. Simon Sivertsen
416	1308	3	Ware, Mr. Frederick
417	1309	3	Peter, Master. Michael J

	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	male	34.5	0	0	330911	7.8292	NaN	Q
1	female	47.0	1	0	363272	7.0000	NaN	S
2	male	62.0	0	0	240276	9.6875	NaN	Q
3	male	27.0	0	0	315154	8.6625	NaN	S
4	female	22.0	1	1	3101298	12.2875	NaN	S

```

..      ...      ...      ...      ...      ...      ...      ...      ...
413    male    NaN      0      0      A.5. 3236      8.0500    NaN      S
414    female  39.0      0      0      PC 17758    108.9000    C105     C
415    male    38.5      0      0    SOTON/O.Q. 3101262    7.2500    NaN      S
416    male    NaN      0      0      359309      8.0500    NaN      S
417    male    NaN      1      1      2668      22.3583    NaN      C

```

[418 rows x 11 columns]

1.2 Analyzing Data

[443]: `data_df.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

```

[444]: `data_df.describe()`

```

[444]:
      PassengerId  Survived  Pclass    Age  SibSp  \
count    891.000000   891.000000   891.000000  714.000000  891.000000
mean     446.000000     0.383838     2.308642   29.699118    0.523008
std      257.353842     0.486592     0.836071   14.526497    1.102743
min         1.000000     0.000000     1.000000     0.420000    0.000000
25%      223.500000     0.000000     2.000000    20.125000    0.000000
50%      446.000000     0.000000     3.000000    28.000000    0.000000
75%      668.500000     1.000000     3.000000    38.000000    1.000000
max      891.000000     1.000000     3.000000    80.000000    8.000000

      Parch    Fare
count    891.000000  891.000000
mean       0.381594   32.204208

```

```
std      0.806057  49.693429
min      0.000000  0.000000
25%      0.000000  7.910400
50%      0.000000  14.454200
75%      0.000000  31.000000
max      6.000000  512.329200
```

```
[445]: data_df.corr()
```

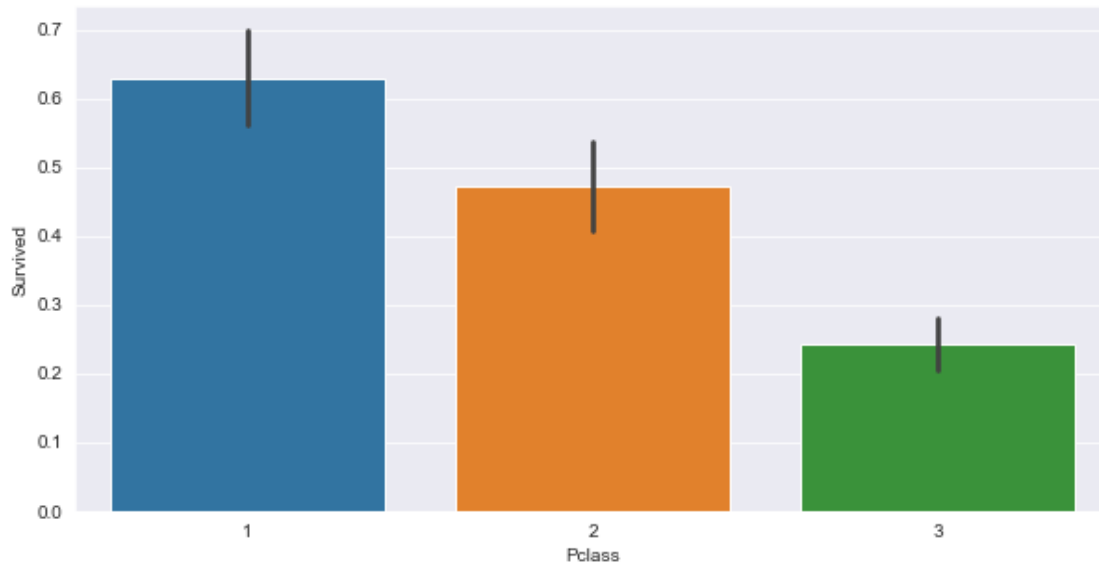
```
[445]:      PassengerId  Survived  Pclass     Age  SibSp  Parch  \
PassengerId    1.000000 -0.005007 -0.035144  0.036847 -0.057527 -0.001652
Survived       -0.005007  1.000000 -0.338481 -0.077221 -0.035322  0.081629
Pclass         -0.035144 -0.338481  1.000000 -0.369226  0.083081  0.018443
Age            0.036847 -0.077221 -0.369226  1.000000 -0.308247 -0.189119
SibSp         -0.057527 -0.035322  0.083081 -0.308247  1.000000  0.414838
Parch         -0.001652  0.081629  0.018443 -0.189119  0.414838  1.000000
Fare           0.012658  0.257307 -0.549500  0.096067  0.159651  0.216225
```

```
      Fare
PassengerId  0.012658
Survived     0.257307
Pclass       -0.549500
Age          0.096067
SibSp        0.159651
Parch        0.216225
Fare         1.000000
```

```
[446]: plt.figure(figsize=(15,5))
sns.heatmap(data_df.corr(), annot=True);
```



```
[447]: plt.figure(figsize=(10,5))
sns.barplot(data=data_df, x='Pclass', y='Survived');
```



The higher the class, the higher the possibility to survive.

```
[448]: data_df.Fare.describe()
```

```
[448]: count      891.000000
      mean       32.204208
      std       49.693429
      min        0.000000
      25%       7.910400
      50%      14.454200
      75%      31.000000
      max      512.329200
      Name: Fare, dtype: float64
```

```
[449]: data_df.Fare.value_counts()
```

```
[449]: 8.0500      43
      13.0000     42
      7.8958     38
      7.7500     34
      26.0000     31
      ..
      35.0000      1
      28.5000      1
      6.2375       1
      14.0000       1
      10.5167       1
      Name: Fare, Length: 248, dtype: int64
```

```
[450]: fig = px.histogram(data_df, x='Fare', y='Survived', nbins=100, color='Pclass')

fig.update_layout(bargap=0.1)
fig.show();
```

The higher the class, the higher the price paid for the fee.

1.3 Dataframe Adjustments

```
[451]: data_df
```

```
[451]:
```

	PassengerId	Survived	Pclass	\
0	1	0	3	
1	2	1	1	
2	3	1	3	
3	4	1	1	
4	5	0	3	
..	
886	887	0	2	
887	888	1	1	
888	889	0	3	
889	890	1	1	
890	891	0	3	

	Name	Sex	Age	SibSp	\
0	Braund, Mr. Owen Harris	male	22.0	1	
1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	
2	Heikkinen, Miss. Laina	female	26.0	0	
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	
4	Allen, Mr. William Henry	male	35.0	0	
..	
886	Montvila, Rev. Juozas	male	27.0	0	
887	Graham, Miss. Margaret Edith	female	19.0	0	
888	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	
889	Behr, Mr. Karl Howell	male	26.0	0	
890	Dooley, Mr. Patrick	male	32.0	0	

	Parch	Ticket	Fare	Cabin	Embarked
0	0	A/5 21171	7.2500	NaN	S
1	0	PC 17599	71.2833	C85	C
2	0	STON/O2. 3101282	7.9250	NaN	S
3	0	113803	53.1000	C123	S
4	0	373450	8.0500	NaN	S
..	
886	0	211536	13.0000	NaN	S
887	0	112053	30.0000	B42	S
888	2	W./C. 6607	23.4500	NaN	S

```

889      0      111369  30.0000  C148      C
890      0      370376   7.7500   NaN      Q

```

[891 rows x 12 columns]

```
[452]: test_df
```

```

[452]:      PassengerId  Pclass      Name \
0           892         3      Kelly, Mr. James
1           893         3  Wilkes, Mrs. James (Ellen Needs)
2           894         2    Myles, Mr. Thomas Francis
3           895         3      Wirz, Mr. Albert
4           896         3  Hirvonen, Mrs. Alexander (Helga E Lindqvist)
..      ...      ...      ...
413        1305         3      Spector, Mr. Woolf
414        1306         1  Oliva y Ocana, Dona. Fermina
415        1307         3  Saether, Mr. Simon Sivertsen
416        1308         3      Ware, Mr. Frederick
417        1309         3  Peter, Master. Michael J

      Sex  Age  SibSp  Parch      Ticket      Fare Cabin Embarked
0    male  34.5     0     0      330911   7.8292   NaN      Q
1  female  47.0     1     0      363272   7.0000   NaN      S
2    male  62.0     0     0      240276   9.6875   NaN      Q
3    male  27.0     0     0      315154   8.6625   NaN      S
4  female  22.0     1     1      3101298  12.2875   NaN      S
..      ...  ...  ...  ...      ...      ...  ...
413   male   NaN     0     0      A.5. 3236   8.0500   NaN      S
414  female  39.0     0     0      PC 17758 108.9000  C105      C
415   male  38.5     0     0  SOTON/O.Q. 3101262   7.2500   NaN      S
416   male   NaN     0     0      359309   8.0500   NaN      S
417   male   NaN     1     1        2668  22.3583   NaN      C

```

[418 rows x 11 columns]

```

[453]: survived_binary = {1: 'yes', 0: 'no'}

survived_binary_col = data_df['Survived'].map(survived_binary)

```

```
[454]: data_df['SurvivedBinary'] = survived_binary_col
```

```
[455]: data_df
```

```

[455]:      PassengerId  Survived  Pclass \
0           1         0         3
1           2         1         1
2           3         1         2

```

```

3          4          1          1
4          5          0          3
..          ...          ...          ...
886        887          0          2
887        888          1          1
888        889          0          3
889        890          1          1
890        891          0          3

```

```

                                Name      Sex  Age  SibSp  \
0                Braund, Mr. Owen Harris   male  22.0    1
1  Cumings, Mrs. John Bradley (Florence Briggs Th... female  38.0    1
2                Heikkinen, Miss. Laina   female  26.0    0
3      Futrelle, Mrs. Jacques Heath (Lily May Peel) female  35.0    1
4                Allen, Mr. William Henry   male  35.0    0
..                ...                ...  ...  ...
886                Montvila, Rev. Juozas   male  27.0    0
887                Graham, Miss. Margaret Edith female  19.0    0
888      Johnston, Miss. Catherine Helen "Carrie" female   NaN    1
889                Behr, Mr. Karl Howell   male  26.0    0
890                Dooley, Mr. Patrick   male  32.0    0

```

```

      Parch      Ticket    Fare Cabin Embarked SurvivedBinary
0         0      A/5 21171   7.2500   NaN      S          no
1         0      PC 17599  71.2833   C85      C          yes
2         0  STON/O2. 3101282   7.9250   NaN      S          yes
3         0      113803  53.1000  C123      S          yes
4         0      373450   8.0500   NaN      S          no
..        ...          ...      ...      ...      ...
886        0      211536  13.0000   NaN      S          no
887        0      112053  30.0000   B42      S          yes
888        2      W./C. 6607  23.4500   NaN      S          no
889        0      111369  30.0000  C148      C          yes
890        0      370376   7.7500   NaN      Q          no

```

[891 rows x 13 columns]

```
[456]: name_col = data_df.pop('Name')
```

```
[457]: data_df.insert(2, 'Name', name_col)
```

```
[458]: data_df
```

```

[458]:   PassengerId  Survived      Name  \
0           1         0      Braund, Mr. Owen Harris
1           2         1  Cumings, Mrs. John Bradley (Florence Briggs Th...
2           3         1      Heikkinen, Miss. Laina

```


3	4	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)
4	5	0	Allen, Mr. William Henry
..
886	887	0	Montvila, Rev. Juozas
887	888	1	Graham, Miss. Margaret Edith
888	889	0	Johnston, Miss. Catherine Helen "Carrie"
889	890	1	Behr, Mr. Karl Howell
890	891	0	Dooley, Mr. Patrick

	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	\
0	3	male	22.0	1	0	A/5 21171	7.2500	NaN	
1	1	female	38.0	1	0	PC 17599	71.2833	C85	
2	3	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	
3	1	female	35.0	1	0	113803	53.1000	C123	
4	3	male	35.0	0	0	373450	8.0500	NaN	
..	
886	2	male	27.0	0	0	211536	13.0000	NaN	
887	1	female	19.0	0	0	112053	30.0000	B42	
888	3	female	NaN	1	2	W./C. 6607	23.4500	NaN	
889	1	male	26.0	0	0	111369	30.0000	C148	
890	3	male	32.0	0	0	370376	7.7500	NaN	

	Embarked	Survived	Binary
0	S	no	
1	C	yes	
2	S	yes	
3	S	yes	
4	S	no	
..	
886	S	no	
887	S	yes	
888	S	no	
889	C	yes	
890	Q	no	

[891 rows x 13 columns]

```
[459]: ticket = data_df.pop('Ticket')
       cabin = data_df.pop('Cabin')
```

```
[460]: data_df.insert(3, 'Ticket', ticket)
       data_df.insert(4, 'Cabin', cabin)
```

```
[461]: data_df
```

```
[461]: PassengerId  Survived  Name \
0          1          0  Braund, Mr. Owen Harris
```

1	2	1	Cummings, Mrs. John Bradley (Florence Briggs Th...
2	3	1	Heikkinen, Miss. Laina
3	4	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)
4	5	0	Allen, Mr. William Henry
..
886	887	0	Montvila, Rev. Juozas
887	888	1	Graham, Miss. Margaret Edith
888	889	0	Johnston, Miss. Catherine Helen "Carrie"
889	890	1	Behr, Mr. Karl Howell
890	891	0	Dooley, Mr. Patrick

	Ticket	Cabin	Pclass	Sex	Age	SibSp	Parch	Fare	\
0	A/5 21171	NaN	3	male	22.0	1	0	7.2500	
1	PC 17599	C85	1	female	38.0	1	0	71.2833	
2	STON/O2. 3101282	NaN	3	female	26.0	0	0	7.9250	
3	113803	C123	1	female	35.0	1	0	53.1000	
4	373450	NaN	3	male	35.0	0	0	8.0500	
..	
886	211536	NaN	2	male	27.0	0	0	13.0000	
887	112053	B42	1	female	19.0	0	0	30.0000	
888	W./C. 6607	NaN	3	female	NaN	1	2	23.4500	
889	111369	C148	1	male	26.0	0	0	30.0000	
890	370376	NaN	3	male	32.0	0	0	7.7500	

	Embarked	Survived	Binary
0	S		no
1	C		yes
2	S		yes
3	S		yes
4	S		no
..
886	S		no
887	S		yes
888	S		no
889	C		yes
890	Q		no

[891 rows x 13 columns]

```
[462]: sex_binary = {'male':0, 'female':1}
```

```
sex_binary_col=data_df.Sex.map(sex_binary)
```

```
[463]: data_df.insert(10, 'SexBinary', sex_binary_col)
```

```
[464]: data_df.pop('Sex')
```

```
[464]: 0      male
        1      female
        2      female
        3      female
        4      male
        ...
        886     male
        887     female
        888     female
        889     male
        890     male
Name: Sex, Length: 891, dtype: object
```

```
[465]: data_df
```

```
[465]: PassengerId  Survived  Name \
0          1         0      Braund, Mr. Owen Harris
1          2         1  Cumings, Mrs. John Bradley (Florence Briggs Th...
2          3         1      Heikkinen, Miss. Laina
3          4         1  Futrelle, Mrs. Jacques Heath (Lily May Peel)
4          5         0      Allen, Mr. William Henry
..         ...         ...
886        887         0      Montvila, Rev. Juozas
887        888         1      Graham, Miss. Margaret Edith
888        889         0  Johnston, Miss. Catherine Helen "Carrie"
889        890         1      Behr, Mr. Karl Howell
890        891         0      Dooley, Mr. Patrick

Ticket Cabin  Pclass  Age  SibSp  Parch  SexBinary  Fare \
0      A/5 21171   NaN    3  22.0    1    0         0  7.2500
1      PC 17599   C85    1  38.0    1    0         1 71.2833
2  STON/O2. 3101282   NaN    3  26.0    0    0         1  7.9250
3      113803  C123    1  35.0    1    0         1 53.1000
4      373450   NaN    3  35.0    0    0         0  8.0500
..         ...         ...
886      211536   NaN    2  27.0    0    0         0 13.0000
887      112053  B42    1  19.0    0    0         1 30.0000
888  W./C. 6607   NaN    3   NaN    1    2         1 23.4500
889      111369  C148    1  26.0    0    0         0 30.0000
890      370376   NaN    3  32.0    0    0         0  7.7500

Embarked SurvivedBinary
0      S             no
1      C             yes
2      S             yes
3      S             yes
4      S             no
```

```

..      ...
886      S      no
887      S      yes
888      S      no
889      C      yes
890      Q      no

```

[891 rows x 13 columns]

```
[466]: test_df
```

```

[466]:      PassengerId  Pclass                                Name \
0           892         3                                Kelly, Mr. James
1           893         3          Wilkes, Mrs. James (Ellen Needs)
2           894         2          Myles, Mr. Thomas Francis
3           895         3              Wirz, Mr. Albert
4           896         3  Hirvonen, Mrs. Alexander (Helga E Lindqvist)
..      ...      ...
413        1305         3              Spector, Mr. Woolf
414        1306         1      Oliva y Ocana, Dona. Fermina
415        1307         3      Saether, Mr. Simon Sivertsen
416        1308         3          Ware, Mr. Frederick
417        1309         3      Peter, Master. Michael J

      Sex  Age  SibSp  Parch            Ticket     Fare Cabin Embarked
0    male  34.5    0.0    0.0        330911     7.8292   NaN      Q
1  female  47.0    1.0    0.0        363272     7.0000   NaN      S
2    male  62.0    0.0    0.0        240276     9.6875   NaN      Q
3    male  27.0    0.0    0.0        315154     8.6625   NaN      S
4  female  22.0    1.0    1.0       3101298    12.2875   NaN      S
..      ...  ...  ...  ...
413   male   NaN    0.0    0.0          A.5. 3236     8.0500   NaN      S
414  female  39.0    0.0    0.0          PC 17758   108.9000  C105      C
415   male  38.5    0.0    0.0  SOTON/O.Q. 3101262     7.2500   NaN      S
416   male   NaN    0.0    0.0          359309     8.0500   NaN      S
417   male   NaN    1.0    1.0           2668     22.3583   NaN      C

```

[418 rows x 11 columns]

```
[467]: test_df.insert(2, 'Name', test_df.pop('Name'))
```

```
[468]: test_df
```

```

[468]:      PassengerId  Pclass                                Name \
0           892         3                                Kelly, Mr. James
1           893         3          Wilkes, Mrs. James (Ellen Needs)
2           894         2          Myles, Mr. Thomas Francis

```

3	895	3	Wirz, Mr. Albert
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)
..
413	1305	3	Spector, Mr. Woolf
414	1306	1	Oliva y Ocana, Dona. Fermina
415	1307	3	Saether, Mr. Simon Sivertsen
416	1308	3	Ware, Mr. Frederick
417	1309	3	Peter, Master. Michael J

	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	male	34.5	0	0	330911	7.8292	NaN	Q
1	female	47.0	1	0	363272	7.0000	NaN	S
2	male	62.0	0	0	240276	9.6875	NaN	Q
3	male	27.0	0	0	315154	8.6625	NaN	S
4	female	22.0	1	1	3101298	12.2875	NaN	S
..
413	male	NaN	0	0	A.5. 3236	8.0500	NaN	S
414	female	39.0	0	0	PC 17758	108.9000	C105	C
415	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	NaN	S
416	male	NaN	0	0	359309	8.0500	NaN	S
417	male	NaN	1	1	2668	22.3583	NaN	C

[418 rows x 11 columns]

```
[469]: test_df.insert(3, 'Pclass', test_df.pop('Pclass'))
```

```
[470]: test_df
```

```
[470]:
```

	PassengerId	Name	Sex	\
0	892	Kelly, Mr. James	male	
1	893	Wilkes, Mrs. James (Ellen Needs)	female	
2	894	Myles, Mr. Thomas Francis	male	
3	895	Wirz, Mr. Albert	male	
4	896	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	
..	
413	1305	Spector, Mr. Woolf	male	
414	1306	Oliva y Ocana, Dona. Fermina	female	
415	1307	Saether, Mr. Simon Sivertsen	male	
416	1308	Ware, Mr. Frederick	male	
417	1309	Peter, Master. Michael J	male	

	Pclass	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	3	34.5	0	0	330911	7.8292	NaN	Q
1	3	47.0	1	0	363272	7.0000	NaN	S
2	2	62.0	0	0	240276	9.6875	NaN	Q
3	3	27.0	0	0	315154	8.6625	NaN	S
4	3	22.0	1	1	3101298	12.2875	NaN	S

..
413	3	NaN	0	0	A.5.	3236	8.0500	NaN	S	
414	1	39.0	0	0	PC	17758	108.9000	C105	C	
415	3	38.5	0	0	SOTON/O.Q.	3101262	7.2500	NaN	S	
416	3	NaN	0	0		359309	8.0500	NaN	S	
417	3	NaN	1	1		2668	22.3583	NaN	C	

[418 rows x 11 columns]

```
[471]: test_df.insert(2, 'Ticket', test_df.pop('Ticket'))
test_df.insert(3, 'Cabin', test_df.pop('Cabin'))
test_df.insert(8, 'SexBinary', sex_binary_col)
```

```
[472]: test_df
```

```
[472]: PassengerId      Name \
0      892      Kelly, Mr. James
1      893  Wilkes, Mrs. James (Ellen Needs)
2      894  Myles, Mr. Thomas Francis
3      895  Wirz, Mr. Albert
4      896  Hirvonen, Mrs. Alexander (Helga E Lindqvist)
..      ...
413     1305      Spector, Mr. Woolf
414     1306  Oliva y Ocana, Dona. Fermina
415     1307  Saether, Mr. Simon Sivertsen
416     1308  Ware, Mr. Frederick
417     1309  Peter, Master. Michael J

      Ticket Cabin  Sex  Pclass  Age  SibSp  SexBinary  Parch \
0      330911  NaN  male      3  34.5      0          0      0
1      363272  NaN  female    3  47.0      1          1      0
2      240276  NaN  male      2  62.0      0          1      0
3      315154  NaN  male      3  27.0      0          1      0
4      3101298  NaN  female    3  22.0      1          0      1
..      ...
413      A.5. 3236  NaN  male      3  NaN      0          0      0
414      PC 17758  C105  female    1  39.0      0          0      0
415  SOTON/O.Q. 3101262  NaN  male      3  38.5      0          1      0
416      359309  NaN  male      3  NaN      0          1      0
417      2668  NaN  male      3  NaN      1          1      1

      Fare Embarked
0      7.8292      Q
1      7.0000      S
2      9.6875      Q
3      8.6625      S
4     12.2875      S
```

```

..      ...      ...
413      8.0500      S
414     108.9000      C
415      7.2500      S
416      8.0500      S
417     22.3583      C

```

[418 rows x 12 columns]

[473]: data_df

```

[473]:      PassengerId  Survived  Name \
0              1         0    Braund, Mr. Owen Harris
1              2         1  Cumings, Mrs. John Bradley (Florence Briggs Th...
2              3         1    Heikkinen, Miss. Laina
3              4         1  Futrelle, Mrs. Jacques Heath (Lily May Peel)
4              5         0    Allen, Mr. William Henry
..      ...      ...      ...
886          887         0    Montvila, Rev. Juozas
887          888         1    Graham, Miss. Margaret Edith
888          889         0  Johnston, Miss. Catherine Helen "Carrie"
889          890         1    Behr, Mr. Karl Howell
890          891         0    Dooley, Mr. Patrick

```

```

      Ticket Cabin  Pclass  Age  SibSp  Parch  SexBinary  Fare \
0      A/5 21171   NaN      3  22.0      1      0          0  7.2500
1      PC 17599   C85      1  38.0      1      0          1  71.2833
2  STON/O2. 3101282   NaN      3  26.0      0      0          1   7.9250
3      113803  C123      1  35.0      1      0          1  53.1000
4      373450   NaN      3  35.0      0      0          0   8.0500
..      ...      ...      ...      ...      ...      ...      ...
886      211536   NaN      2  27.0      0      0          0  13.0000
887      112053   B42      1  19.0      0      0          1  30.0000
888  W./C. 6607   NaN      3   NaN      1      2          1  23.4500
889      111369  C148      1  26.0      0      0          0  30.0000
890      370376   NaN      3  32.0      0      0          0   7.7500

```

```

      Embarked SurvivedBinary
0      S      no
1      C      yes
2      S      yes
3      S      yes
4      S      no
..      ...      ...
886      S      no
887      S      yes
888      S      no

```

```
889      C      yes
890      Q      no
```

```
[891 rows x 13 columns]
```

1.4 Input and Target Columns

```
[474]: input_cols = list(data_df.columns[5:12])
       target_col = 'SurvivedBinary'
```

```
[475]: input_cols
```

```
[475]: ['Pclass', 'Age', 'SibSp', 'Parch', 'SexBinary', 'Fare', 'Embarked']
```

```
[476]: target_col
```

```
[476]: 'SurvivedBinary'
```

```
[477]: input_df = data_df[input_cols].copy()
       target_df = data_df[target_col].copy()
```

```
[478]: input_df
```

```
[478]:
```

	Pclass	Age	SibSp	Parch	SexBinary	Fare	Embarked
0	3	22.0	1	0	0	7.2500	S
1	1	38.0	1	0	1	71.2833	C
2	3	26.0	0	0	1	7.9250	S
3	1	35.0	1	0	1	53.1000	S
4	3	35.0	0	0	0	8.0500	S
..
886	2	27.0	0	0	0	13.0000	S
887	1	19.0	0	0	1	30.0000	S
888	3	NaN	1	2	1	23.4500	S
889	1	26.0	0	0	0	30.0000	C
890	3	32.0	0	0	0	7.7500	Q

```
[891 rows x 7 columns]
```

```
[479]: target_df
```

```
[479]:
```

0	no
1	yes
2	yes
3	yes
4	no
..	...
886	no
887	yes


```

888     no
889     yes
890     no
Name: SurvivedBinary, Length: 891, dtype: object

```

1.5 Numerical and Categorical Columns

```

[480]: numerical_cols = input_df.select_dtypes(include=np.number).columns.tolist()
       categorical_cols = input_df.select_dtypes(include='object').columns.tolist()

```

```

[481]: numerical_cols

```

```

[481]: ['Pclass', 'Age', 'SibSp', 'Parch', 'SexBinary', 'Fare']

```

```

[482]: categorical_cols

```

```

[482]: ['Embarked']

```

1.6 Feature Engineering

Scaling Numerical Columns

```

[483]: from sklearn.preprocessing import MinMaxScaler

```

```

[484]: scaler = MinMaxScaler()
       scaler.fit(input_df[numerical_cols])

```

```

[484]: MinMaxScaler()

```

```

[485]: input_df[numerical_cols] = scaler.transform(input_df[numerical_cols])
       test_df[numerical_cols] = scaler.transform(test_df[numerical_cols])

```

```

[486]: input_df

```

```

[486]:
      Pclass      Age  SibSp  Parch  SexBinary      Fare  Embarked
0         1.0  0.271174  0.125  0.000000         0.0  0.014151         S
1         0.0  0.472229  0.125  0.000000         1.0  0.139136         C
2         1.0  0.321438  0.000  0.000000         1.0  0.015469         S
3         0.0  0.434531  0.125  0.000000         1.0  0.103644         S
4         1.0  0.434531  0.000  0.000000         0.0  0.015713         S
..      ...      ...      ...      ...      ...      ...
886        0.5  0.334004  0.000  0.000000         0.0  0.025374         S
887        0.0  0.233476  0.000  0.000000         1.0  0.058556         S
888        1.0         NaN  0.125  0.333333         1.0  0.045771         S
889        0.0  0.321438  0.000  0.000000         0.0  0.058556         C
890        1.0  0.396833  0.000  0.000000         0.0  0.015127         Q

```

```

[891 rows x 7 columns]

```

```
[487]: test_df[numerical_cols]
```

```
[487]:
```

	Pclass	Age	SibSp	Parch	SexBinary	Fare
0	1.0	0.428248	0.000	0.000000	0.0	0.015282
1	1.0	0.585323	0.125	0.000000	1.0	0.013663
2	0.5	0.773813	0.000	0.000000	1.0	0.018909
3	1.0	0.334004	0.000	0.000000	1.0	0.016908
4	1.0	0.271174	0.125	0.166667	0.0	0.023984
..
413	1.0	NaN	0.000	0.000000	0.0	0.015713
414	0.0	0.484795	0.000	0.000000	0.0	0.212559
415	1.0	0.478512	0.000	0.000000	1.0	0.014151
416	1.0	NaN	0.000	0.000000	1.0	0.015713
417	1.0	NaN	0.125	0.166667	1.0	0.043640

[418 rows x 6 columns]

Imputing Missing Values

```
[488]: from sklearn.impute import SimpleImputer
```

```
[489]: input_df.isna().sum()
```

```
[489]: Pclass      0
Age          177
SibSp        0
Parch        0
SexBinary    0
Fare         0
Embarked     2
dtype: int64
```

```
[490]: input_df.Embarked.value_counts()
```

```
[490]: S      644
C      168
Q       77
Name: Embarked, dtype: int64
```

```
[491]: input_df.Embarked.fillna('Q', inplace=True)
```

```
[492]: input_df.Embarked.value_counts()
```

```
[492]: S      644
C      168
Q       79
Name: Embarked, dtype: int64
```

```
[493]: test_df.isna().sum()
```

```
[493]: PassengerId      0
      Name            0
      Ticket          0
      Cabin          327
      Sex             0
      Pclass          0
      Age            86
      SibSp           0
      SexBinary       0
      Parch           0
      Fare            1
      Embarked        0
      dtype: int64
```

```
[494]: imputer = SimpleImputer(strategy='mean')
      imputer.fit(input_df[numerical_cols])
```

```
[494]: SimpleImputer()
```

```
[495]: input_df[numerical_cols] = imputer.transform(input_df[numerical_cols])
      test_df[numerical_cols] = imputer.transform(test_df[numerical_cols])
```

```
[496]: input_df.isna().sum()
```

```
[496]: Pclass      0
      Age       0
      SibSp     0
      Parch     0
      SexBinary 0
      Fare      0
      Embarked  0
      dtype: int64
```

```
[497]: test_df.isna().sum()
```

```
[497]: PassengerId      0
      Name            0
      Ticket          0
      Cabin          327
      Sex             0
      Pclass          0
      Age            0
      SibSp           0
      SexBinary       0
      Parch           0
      Fare            0
```

```
Embarked      0
dtype: int64
```

One Hot Encoding

```
[498]: from sklearn.preprocessing import OneHotEncoder
```

```
[499]: encoder = OneHotEncoder(sparse=False, handle_unknown='ignore')
encoder.fit(input_df[categorical_cols])
```

```
[499]: OneHotEncoder(handle_unknown='ignore', sparse=False)
```

```
[500]: encoded_cols = list(encoder.get_feature_names_out(categorical_cols))
encoded_cols
```

```
[500]: ['Embarked_C', 'Embarked_Q', 'Embarked_S']
```

```
[501]: input_df[encoded_cols] = encoder.transform(input_df[categorical_cols])
test_df[encoded_cols] = encoder.transform(test_df[categorical_cols])
```

```
[502]: input_df
```

```
[502]:
```

	Pclass	Age	SibSp	Parch	SexBinary	Fare	Embarked \
0	1.0	0.271174	0.125	0.000000	0.0	0.014151	S
1	0.0	0.472229	0.125	0.000000	1.0	0.139136	C
2	1.0	0.321438	0.000	0.000000	1.0	0.015469	S
3	0.0	0.434531	0.125	0.000000	1.0	0.103644	S
4	1.0	0.434531	0.000	0.000000	0.0	0.015713	S
..	
886	0.5	0.334004	0.000	0.000000	0.0	0.025374	S
887	0.0	0.233476	0.000	0.000000	1.0	0.058556	S
888	1.0	0.367921	0.125	0.333333	1.0	0.045771	S
889	0.0	0.321438	0.000	0.000000	0.0	0.058556	C
890	1.0	0.396833	0.000	0.000000	0.0	0.015127	Q

	Embarked_C	Embarked_Q	Embarked_S
0	0.0	0.0	1.0
1	1.0	0.0	0.0
2	0.0	0.0	1.0
3	0.0	0.0	1.0
4	0.0	0.0	1.0
..
886	0.0	0.0	1.0
887	0.0	0.0	1.0
888	0.0	0.0	1.0
889	1.0	0.0	0.0
890	0.0	1.0	0.0

[891 rows x 10 columns]

[503]: test_df

```
[503]:
```

	PassengerId	Name \
0	892	Kelly, Mr. James
1	893	Wilkes, Mrs. James (Ellen Needs)
2	894	Myles, Mr. Thomas Francis
3	895	Wirz, Mr. Albert
4	896	Hirvonen, Mrs. Alexander (Helga E Lindqvist)
..
413	1305	Spector, Mr. Woolf
414	1306	Oliva y Ocana, Dona. Fermina
415	1307	Saether, Mr. Simon Sivertsen
416	1308	Ware, Mr. Frederick
417	1309	Peter, Master. Michael J

		Ticket	Cabin	Sex	Pclass	Age	SibSp	SexBinary \
0		330911	NaN	male	1.0	0.428248	0.000	0.0
1		363272	NaN	female	1.0	0.585323	0.125	1.0
2		240276	NaN	male	0.5	0.773813	0.000	1.0
3		315154	NaN	male	1.0	0.334004	0.000	1.0
4		3101298	NaN	female	1.0	0.271174	0.125	0.0
..	
413		A.5. 3236	NaN	male	1.0	0.367921	0.000	0.0
414		PC 17758	C105	female	0.0	0.484795	0.000	0.0
415	SOTON/O.Q.	3101262	NaN	male	1.0	0.478512	0.000	1.0
416		359309	NaN	male	1.0	0.367921	0.000	1.0
417		2668	NaN	male	1.0	0.367921	0.125	1.0

	Parch	Fare	Embarked	Embarked_C	Embarked_Q	Embarked_S
0	0.000000	0.015282	Q	0.0	1.0	0.0
1	0.000000	0.013663	S	0.0	0.0	1.0
2	0.000000	0.018909	Q	0.0	1.0	0.0
3	0.000000	0.016908	S	0.0	0.0	1.0
4	0.166667	0.023984	S	0.0	0.0	1.0
..
413	0.000000	0.015713	S	0.0	0.0	1.0
414	0.000000	0.212559	C	1.0	0.0	0.0
415	0.000000	0.014151	S	0.0	0.0	1.0
416	0.000000	0.015713	S	0.0	0.0	1.0
417	0.166667	0.043640	C	1.0	0.0	0.0

[418 rows x 15 columns]

2 Creating a Training and a Validation Set

```
[504]: from sklearn.model_selection import train_test_split
```

```
[505]: train_inputs, val_inputs, train_targets, val_targets =  
↳ train_test_split(input_df[numerical_cols+encoded_cols], target_df,  
↳ test_size=0.20, random_state=42)
```

```
[506]: train_inputs
```

```
[506]:      Pclass      Age  SibSp   Parch  SexBinary      Fare  Embarked_C  \  
331      0.0  0.566474  0.000  0.000000      0.0  0.055628      0.0  
733      0.5  0.283740  0.000  0.000000      0.0  0.025374      0.0  
382      1.0  0.396833  0.000  0.000000      0.0  0.015469      0.0  
704      1.0  0.321438  0.125  0.000000      0.0  0.015330      0.0  
813      1.0  0.070118  0.500  0.333333      1.0  0.061045      0.0  
..  
106      1.0  0.258608  0.000  0.000000      1.0  0.014932      0.0  
270      0.0  0.367921  0.000  0.000000      0.0  0.060508      0.0  
860      1.0  0.509927  0.250  0.000000      0.0  0.027538      0.0  
435      0.0  0.170646  0.125  0.333333      1.0  0.234224      0.0  
102      0.0  0.258608  0.000  0.166667      0.0  0.150855      0.0
```

```
      Embarked_Q  Embarked_S  
331           0.0           1.0  
733           0.0           1.0  
382           0.0           1.0  
704           0.0           1.0  
813           0.0           1.0  
..  
106           0.0           1.0  
270           0.0           1.0  
860           0.0           1.0  
435           0.0           1.0  
102           0.0           1.0
```

```
[712 rows x 9 columns]
```

```
[507]: val_inputs
```

```
[507]:      Pclass      Age  SibSp   Parch  SexBinary      Fare  Embarked_C  \  
709      1.0  0.367921  0.125  0.166667      0.0  0.029758      1.0  
439      0.5  0.384267  0.000  0.000000      0.0  0.020495      0.0  
840      1.0  0.246042  0.000  0.000000      0.0  0.015469      0.0  
720      0.5  0.070118  0.000  0.166667      1.0  0.064412      0.0  
39       1.0  0.170646  0.125  0.000000      1.0  0.021942      1.0  
..  
      ...      ...      ...      ...      ...      ...
```

433	1.0	0.208344	0.000	0.000000	0.0	0.013907	0.0
773	1.0	0.367921	0.000	0.000000	0.0	0.014102	1.0
25	1.0	0.472229	0.125	0.833333	1.0	0.061264	0.0
84	0.5	0.208344	0.000	0.000000	1.0	0.020495	0.0
10	1.0	0.044986	0.125	0.166667	1.0	0.032596	0.0

	Embarked_Q	Embarked_S
709	0.0	0.0
439	0.0	1.0
840	0.0	1.0
720	0.0	1.0
39	0.0	0.0
..
433	0.0	1.0
773	0.0	0.0
25	0.0	1.0
84	0.0	1.0
10	0.0	1.0

[179 rows x 9 columns]

```
[508]: train_targets
```

```
[508]: 331    no
       733    no
       382    no
       704    no
       813    no
       ...
       106   yes
       270    no
       860    no
       435   yes
       102    no
Name: SurvivedBinary, Length: 712, dtype: object
```

```
[509]: val_targets
```

```
[509]: 709   yes
       439   no
       840   no
       720   yes
       39    yes
       ...
       433   no
       773   no
       25    yes
```

```
84     yes
10     yes
Name: SurvivedBinary, Length: 179, dtype: object
```

```
[510]: X_train = train_inputs[numerical_cols+encoded_cols]
X_val = val_inputs[numerical_cols+encoded_cols]
X_test = test_df[numerical_cols+encoded_cols]
```

3 Creating the Model

The chosen model is Random Forest.

```
[511]: from sklearn.ensemble import RandomForestClassifier
```

```
[512]: %%time
ran_for_model = RandomForestClassifier(n_jobs=-1, random_state=42)
ran_for_model.fit(X_train, train_targets)
```

CPU times: total: 281 ms

Wall time: 161 ms

```
[512]: RandomForestClassifier(n_jobs=-1, random_state=42)
```

Accuracies

```
[513]: ran_for_model_base_accs = ran_for_model.score(X_train, train_targets),
↳ ran_for_model.score(X_val, val_targets)
```

```
[514]: ran_for_model_base_accs
```

```
[514]: (0.9803370786516854, 0.8100558659217877)
```

4 Hypertuning

max_depth

```
[515]: def max_depth_accuracy(md):
    model = RandomForestClassifier(n_jobs=-1, random_state=42, max_depth=md)
    model.fit(X_train, train_targets)
    train_acc = model.score(X_train, train_targets)
    val_acc = model.score(X_val, val_targets)
    return {'MaxDepth': md, 'TrainAccuracy': train_acc, 'ValidationAccuracy':
↳ val_acc}
```

```
[516]: max_depth_acc_df = pd.DataFrame([max_depth_accuracy(md) for md in range(1,20)])
```

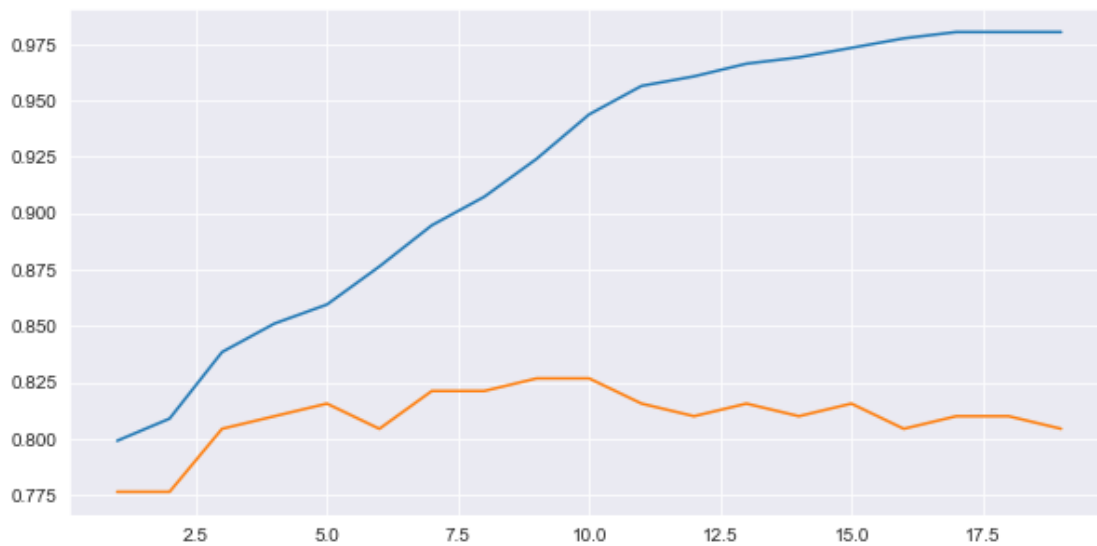
```
[517]: max_depth_acc_df.head(10)
```



```
[517]:
```

	MaxDepth	TrainAccuracy	ValidationAccuracy
0	1	0.799157	0.776536
1	2	0.808989	0.776536
2	3	0.838483	0.804469
3	4	0.851124	0.810056
4	5	0.859551	0.815642
5	6	0.876404	0.804469
6	7	0.894663	0.821229
7	8	0.907303	0.821229
8	9	0.924157	0.826816
9	10	0.943820	0.826816

```
[518]: plt.figure(figsize=(10,5))
plt.plot(max_depth_acc_df['MaxDepth'], max_depth_acc_df['TrainAccuracy'])
plt.plot(max_depth_acc_df['MaxDepth'], max_depth_acc_df['ValidationAccuracy']);
```



```
[519]: ran_for_model = RandomForestClassifier(n_jobs=-1, random_state=42, max_depth=10)
ran_for_model.fit(X_train, train_targets)
```

```
[519]: RandomForestClassifier(max_depth=10, n_jobs=-1, random_state=42)
```

```
[520]: ran_for_model_base_accs
```

```
[520]: (0.9803370786516854, 0.8100558659217877)
```

```
[521]: ran_for_model.score(X_train, train_targets), ran_for_model.score(X_val,
↪ val_targets)
```

```
[521]: (0.9438202247191011, 0.8268156424581006)
```

n_estimators

```
[522]: ran_for_model_estim = RandomForestClassifier(n_jobs=-1, random_state=42,
↳ n_estimators=57, max_depth=9)
ran_for_model_estim.fit(X_train, train_targets)
ran_for_model_estim.score(X_train, train_targets), ran_for_model_estim.
↳ score(X_val, val_targets)
```

[522]: (0.9199438202247191, 0.8379888268156425)

min_impurity_decrease

```
[523]: ran_for_model_imp = RandomForestClassifier(n_jobs=-1, random_state=42,
↳ n_estimators=57, max_depth=9, min_impurity_decrease=1e-6)
ran_for_model_imp.fit(X_train, train_targets)
ran_for_model_imp.score(X_train, train_targets), ran_for_model_imp.score(X_val,
↳ val_targets)
```

[523]: (0.9199438202247191, 0.8435754189944135)

4.0.1 class_weight

```
[524]: saved=list(ran_for_model_imp.predict(X_train))
```

```
[525]: len(saved)
```

[525]: 712

```
[526]: count = 0
for x in saved:
    if x == 'yes':
        count += 1
```

```
[527]: count
```

[527]: 219

```
[539]: ran_for_model_imp = RandomForestClassifier(n_jobs=-1, random_state=42,
↳ n_estimators=57, max_depth=9, min_impurity_decrease=1e-7,
↳ class_weight={'yes':2, 'no':1})
ran_for_model_imp.fit(X_train, train_targets)
ran_for_model_imp.score(X_train, train_targets), ran_for_model_imp.score(X_val,
↳ val_targets)
```

[539]: (0.9353932584269663, 0.8435754189944135)

5 Final Model

```
[529]: ran_for_model_final = RandomForestClassifier(n_jobs=-1, random_state=42,  
    ↪n_estimators=57, max_depth=9, min_impurity_decrease=1e-7,  
    ↪class_weight={'yes':2, 'no':1})  
ran_for_model_final.fit(X_train, train_targets)
```

```
[529]: RandomForestClassifier(class_weight={'no': 1, 'yes': 2}, max_depth=9,  
    min_impurity_decrease=1e-07, n_estimators=57, n_jobs=-1,  
    random_state=42)
```

```
[530]: preds=ran_for_model_final.predict(X_test)
```