import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

import seaborn as sns
%matplotlib inline

train = pd.read_csv('/content/titanic_train.csv')

train.head()

| \Rightarrow | P | assengerId | 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 | С |
| | 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 + Code | female + T | 35.0 ext | 1 | 0 | 113803 | 53.1000 | C123 | S |

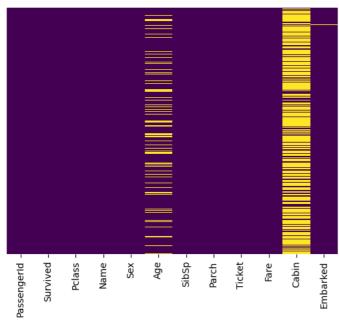
train.isnull()

| | PassengerId | Survived | Pclass | Name | Sex | Age | SibSp | Parch | Ticket | Fare | Cabin | Embarked |
|-----|-------------|----------|--------|-------|-------|-------|-------|-------|--------|-------|-------|----------|
| 0 | False | False | False | False | False | False | False | False | False | False | True | False |
| 1 | False | False | False | False | False | False | False | False | False | False | False | False |
| 2 | False | False | False | False | False | False | False | False | False | False | True | False |
| 3 | False | False | False | False | False | False | False | False | False | False | False | False |
| 4 | False | False | False | False | False | False | False | False | False | False | True | False |
| | | ••• | | | | | | | | | | |
| 886 | False | False | False | False | False | False | False | False | False | False | True | False |
| 887 | False | False | False | False | False | False | False | False | False | False | False | False |
| 888 | False | False | False | False | False | True | False | False | False | False | True | False |
| 889 | False | False | False | False | False | False | False | False | False | False | False | False |
| 890 | False | False | False | False | False | False | False | False | False | False | True | False |

sns.heatmap(train.isnull(),yticklabels=False,cbar=False,cmap='viridis')

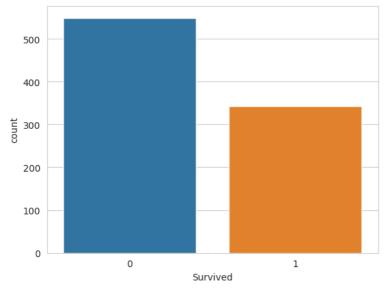
<Axes: >

891 rows x 12 columns



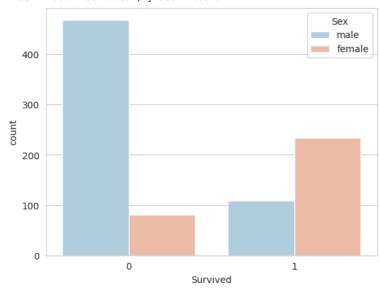
sns.set_style('whitegrid')
sns.countplot(x='Survived',data=train)

<Axes: xlabel='Survived', ylabel='count'>



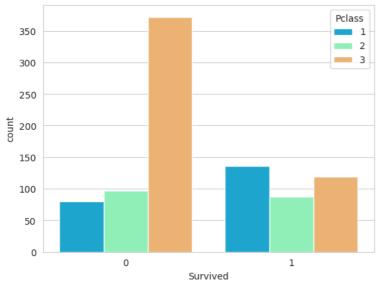
sns.set_style('whitegrid')
sns.countplot(x='Survived',hue='Sex',data=train,palette='RdBu_r')

<Axes: xlabel='Survived', ylabel='count'>

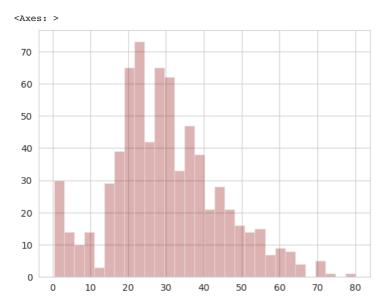


sns.set_style('whitegrid')
sns.countplot(x='Survived',hue='Pclass',data=train,palette='rainbow')

<Axes: xlabel='Survived', ylabel='count'>

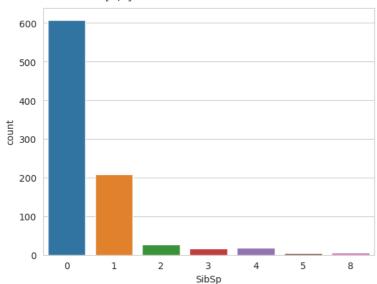


train['Age'].hist(bins=30,color='darkred',alpha=0.3)

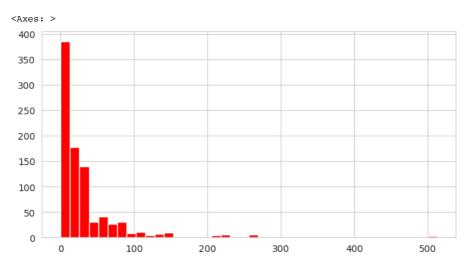


sns.countplot(x='SibSp',data=train)

<Axes: xlabel='SibSp', ylabel='count'>



train['Fare'].hist(color='red',bins=40,figsize=(8,4))



plt.figure(figsize=(12, 7))
sns.boxplot(x='Pclass',y='Age',data=train,palette='winter')

<Axes: xlabel='Pclass', ylabel='Age'>
80
60
50
30
20

```
def impute_age(cols):
    Age = cols[0]
    Pclass = cols[1]

if pd.isnull(Age):
    if Pclass == 1:
        return 37

    elif Pclass == 2:
        return 29

    else:
        return 24

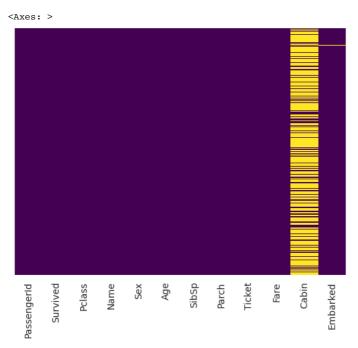
else:
    return Age
```

10

0

train['Age'] = train[['Age','Pclass']].apply(impute_age,axis=1)

 $\verb|sns.heatmap(train.isnull(),yticklabels=False,cbar=False,cmap='viridis')|$



train.drop('Cabin',axis=1,inplace=True)

train.head()

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

train.dropna(inplace=True)

train.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 889 entries, 0 to 890
Data columns (total 11 columns):

| # | Column | Non-Null Count | Dtype |
|------|---------------|-----------------|---------|
| | | | |
| 0 | PassengerId | 889 non-null | int64 |
| 1 | Survived | 889 non-null | int64 |
| 2 | Pclass | 889 non-null | int64 |
| 3 | Name | 889 non-null | object |
| 4 | Sex | 889 non-null | object |
| 5 | Age | 889 non-null | float64 |
| 6 | SibSp | 889 non-null | int64 |
| 7 | Parch | 889 non-null | int64 |
| 8 | Ticket | 889 non-null | object |
| 9 | Fare | 889 non-null | float64 |
| 10 | Embarked | 889 non-null | object |
| dtyp | es: float64(2 |), int64(5), ob | ject(4) |
| memo | ry usage: 83. | 3+ KB | |

pd.get_dummies(train['Embarked'],drop_first=True).head()

- **Q S 0** 0 1
- **1** 0 0
- **2** 0 1
- **3** 0 1
- 4 0 1

sex = pd.get_dummies(train['Sex'],drop_first=True)
embark = pd.get_dummies(train['Embarked'],drop_first=True)

train.drop(['Sex','Embarked','Name','Ticket'],axis=1,inplace=True)

train.head()

| | PassengerId | Survived | Pclass | Age | SibSp | Parch | Fare |
|---|-------------|----------|--------|------|-------|-------|---------|
| 0 | 1 | 0 | 3 | 22.0 | 1 | 0 | 7.2500 |
| 1 | 2 | 1 | 1 | 38.0 | 1 | 0 | 71.2833 |
| 2 | 3 | 1 | 3 | 26.0 | 0 | 0 | 7.9250 |
| 3 | 4 | 1 | 1 | 35.0 | 1 | 0 | 53.1000 |
| 4 | 5 | 0 | 3 | 35.0 | 0 | 0 | 8.0500 |

train = pd.concat([train,sex,embark],axis=1)

train.head()

| | PassengerId | Survived | Pclass | Age | SibSp | Parch | Fare | male | Q | s |
|---|-------------|----------|--------|------|-------|-------|---------|------|---|---|
| 0 | 1 | 0 | 3 | 22.0 | 1 | 0 | 7.2500 | 1 | 0 | 1 |
| 1 | 2 | 1 | 1 | 38.0 | 1 | 0 | 71.2833 | 0 | 0 | 0 |
| 2 | 3 | 1 | 3 | 26.0 | 0 | 0 | 7.9250 | 0 | 0 | 1 |
| _ | | | | | | | | _ | _ | |

train.drop('Survived',axis=1).head()

| | PassengerId | Pclass | Age | SibSp | Parch | Fare | male | Q | s |
|---|-------------|--------|------|-------|-------|---------|------|---|---|
| 0 | 1 | 3 | 22.0 | 1 | 0 | 7.2500 | 1 | 0 | 1 |
| 1 | 2 | 1 | 38.0 | 1 | 0 | 71.2833 | 0 | 0 | 0 |
| 2 | 3 | 3 | 26.0 | 0 | 0 | 7.9250 | 0 | 0 | 1 |
| 3 | 4 | 1 | 35.0 | 1 | 0 | 53.1000 | 0 | 0 | 1 |
| 4 | 5 | 3 | 35.0 | 0 | 0 | 8.0500 | 1 | 0 | 1 |

train['Survived'].head()

```
0 0
1 1
2 1
3 1
```

0

4

Name: Survived, dtype: int64

from sklearn.model_selection import train_test_split

from sklearn.linear_model import LogisticRegression

```
logmodel = LogisticRegression()
logmodel.fit(X_train,y_train)
```

/usr/local/lib/python3.10/dist-packages/sklearn/linear_model/_logistic.py:458: ConvergenceWarning: lbfgs failed to conver STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

```
Increase the number of iterations (max_iter) or scale the data as shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
    https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
    n_iter_i = _check_optimize_result(
    v LogisticRegression |
    LogisticRegression()
```

predictions = logmodel.predict(X_test)

from sklearn.metrics import confusion_matrix
accuracy=confusion_matrix(y_test,predictions)

accuracy

```
array([[149, 14],
[ 39, 65]])
```

from sklearn.metrics import accuracy_score

accuracy=accuracy_score(y_test,predictions)
accuracy

0.8014981273408239

predictions

from sklearn.metrics import classification_report

print(classification_report(y_test,predictions))

| support | f1-score | recall | precision | |
|---------|----------|--------|-----------|--------------|
| 163 | 0.85 | 0.91 | 0.79 | 0 |
| 104 | 0.71 | 0.62 | 0.82 | 1 |
| 267 | 0.80 | | | accuracy |
| 267 | 0.78 | 0.77 | 0.81 | macro avg |
| 267 | 0.80 | 0.80 | 0.80 | weighted avg |