#data analysis libraries
import numpy as np
import pandas as pd

#visualization libraries
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline

#ignore warnings
import warnings
warnings.filterwarnings('ignore')

Start coding or generate with AI.

Double-click (or enter) to edit

data=pd.read_csv('/content/Titanic-Dataset.csv')
data

													
ے ک		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	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 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
8	386	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN	S
8	387	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42	S
8	888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN	S
8	389	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148	С

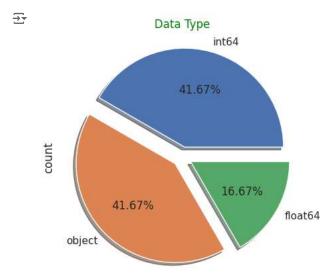
data.describe()

₹		PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
	count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
	mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
	std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
	min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
	25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400
	50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
	75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
	max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

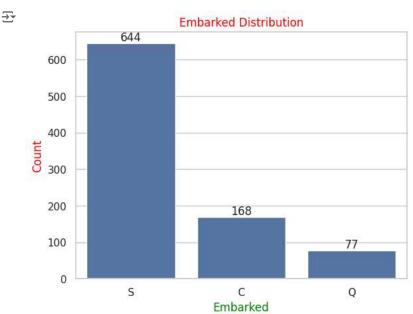
data.info()

vata	columns (tot	al 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	Δσρ	714 non-null	float64

```
SibSp
                    891 non-null
                                     int64
 6
     Parch
                    891 non-null
                                     int64
     Ticket
                    891 non-null
                                     object
                                     float64
     Fare
                    891 non-null
                                     object
 10 Cabin
                    204 non-null
 11 Embarked
                    889 non-null
                                     object
dtypes: float64(2), int64(5), object(5) memory usage: 83.7+ KB
```

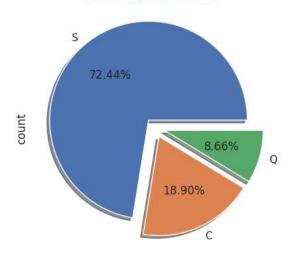


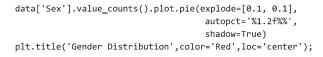
```
ax = sns.set(style="whitegrid")
ax = sns.countplot(data=data,x='Embarked');
ax.bar_label(ax.containers[0])
plt.title('Embarked Distribution',color='Red',loc='center');
plt.xlabel('Embarked',color='Green',loc='center')
plt.ylabel('Count',color='Red',loc='center');
```





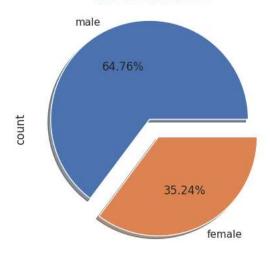
Embarked Distribution

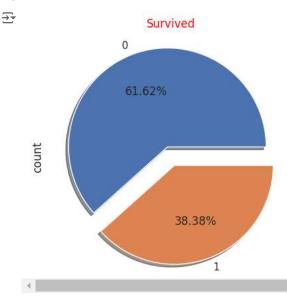




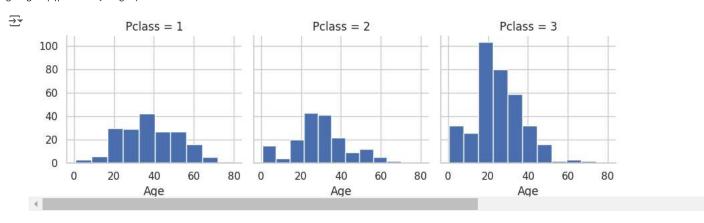


Gender Distribution





g = sns.FacetGrid(data, col="Pclass")
g = g.map(plt.hist, "Age")



Start coding or $\underline{\text{generate}}$ with AI.

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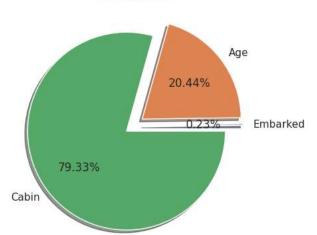
```
def missing_value (df):
    missing_Number = df.isnull().sum().sort_values(ascending=False)[df.isnull().sum().sort_values(ascending=False) !=0]
    missing_percent=round((df.isnull().sum()/df.isnull().count())*100,2)[round((df.isnull().sum()/df.isnull().count())*100,2) !=0]
    missing = pd.concat([missing_Number,missing_percent],axis=1,keys=['Missing Number','Missing Percentage'])
    return missing
```

Apply styling with precision setting
styled_df = missing_value(data).style.background_gradient(cmap='coolwarm')
styled_df.format("{:.2f}", subset=['Missing Percentage']) # Format 'Missing Percentage' column with 2 decimal places
styled_df

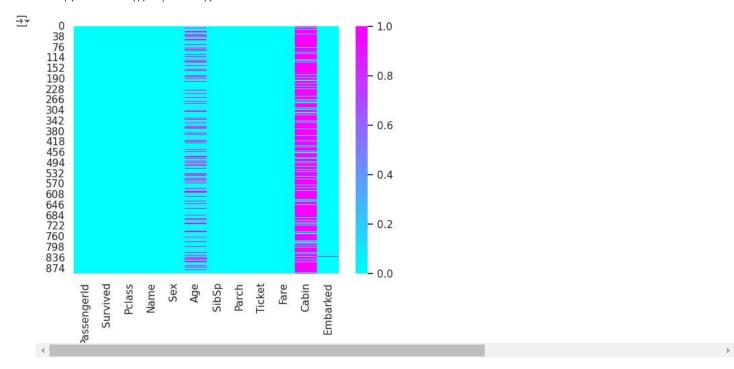
	Missing Number	Missing Percentage
Cabin	687	77.10
Age	177	19.87
Embarked	2	0.22
4		

4

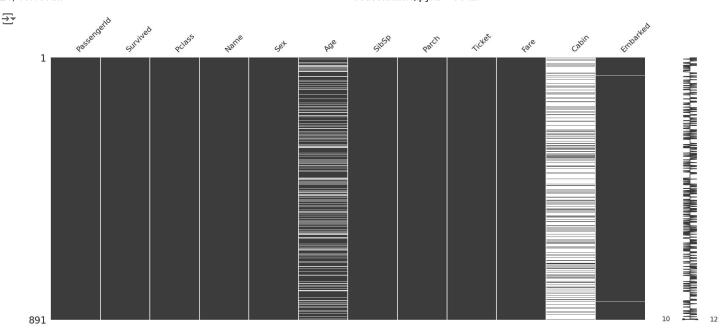




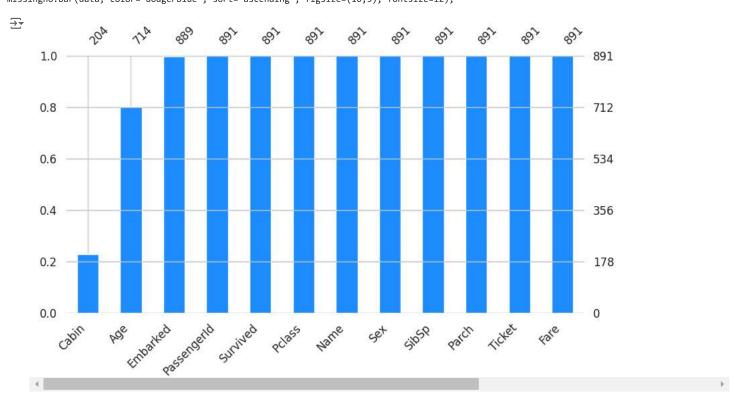
sns.heatmap(data.isnull(),cmap='cool');



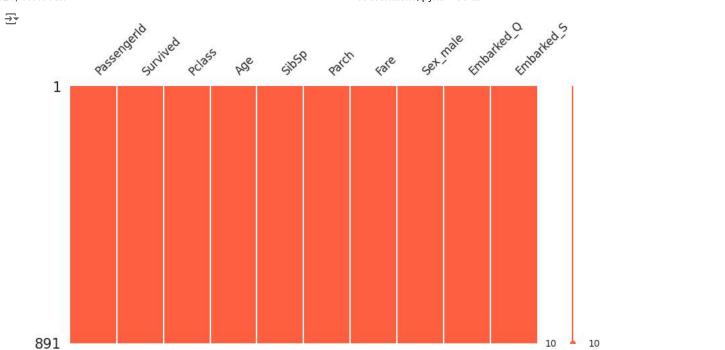
import missingno as msno
msno.matrix(data)
plt.show()



import missingno
missingno.bar(data, color="dodgerblue", sort="ascending", figsize=(10,5), fontsize=12);



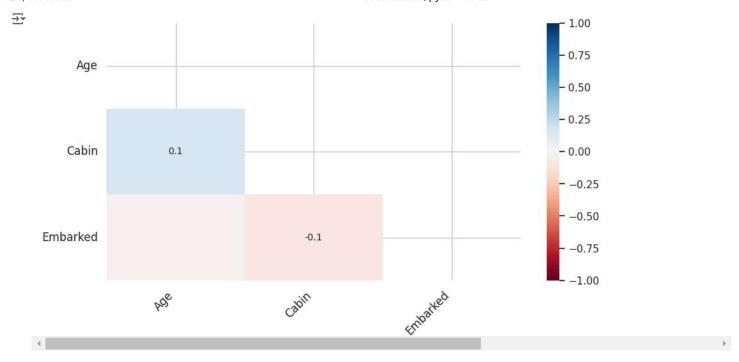
missingno.matrix(data, figsize=(10,5), fontsize=12, color=(1, 0.38, 0.27));



data.isnull().sum()



missingno.heatmap(data, figsize=(10,5), fontsize=12);

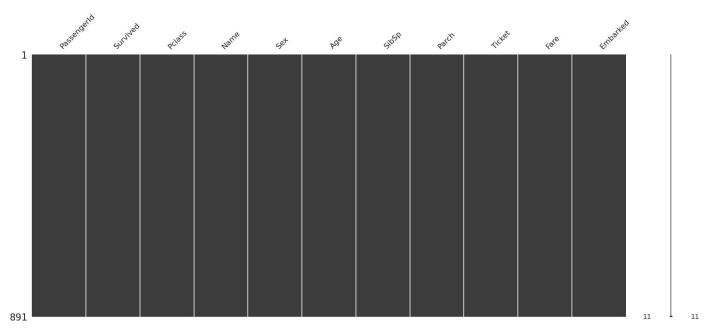


data['Age'] = data['Age'].fillna(data['Age'].mean())
data[data['Embarked'].isnull()]

→		PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
	61	62	1	1	Icard, Miss. Amelie	female	38.0	0	0	113572	80.0	B28	NaN
	829	830	1	1	Stone. Mrs. George Nelson (Martha Evelvn)	female	62.0	0	0	113572	80.0	B28	NaN
	4												

data['Embarked'] = data['Embarked'].fillna(method='bfill')
data = data.drop(['Cabin'],axis=1)
import missingno as msno
msno.matrix(data)
plt.show()





data = data.drop(['Name','Ticket'],axis=1)
data.head()

		PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
	0	1	0	3	male	22.0	1	0	7.2500	S
	1	2	1	1	female	38.0	1	0	71.2833	С
	2	3	1	3	female	26.0	0	0	7.9250	S
	3	4	1	1	female	35.0	1	0	53.1000	S
	4	5	0	3	male	35.0	0	0	8.0500	S

data = pd.get_dummies(data,columns=['Sex','Embarked'],drop_first=True)
data.head()

_ →		PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare	Sex_male	Embarked_Q	Embarked_S
	0	1	0	3	22.0	1	0	7.2500	True	False	True
	1	2	1	1	38.0	1	0	71.2833	False	False	False
	2	3	1	3	26.0	0	0	7.9250	False	False	True
	3	4	1	1	35.0	1	0	53.1000	False	False	True
	4	5	0	3	35.0	0	0	8.0500	True	False	True
	4										

```
X = data.drop(['Survived'],axis=1)
```

y = data['Survived']

 $from \ sklearn.model_selection \ import \ train_test_split$

 $\label{lem:continuous} X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=21)$

from sklearn.preprocessing import StandardScaler

scaler = StandardScaler()

X_train = scaler.fit_transform(X_train)

X_test = scaler.transform(X_test)

X_train = pd.DataFrame(X_train, columns=X.columns

X_test = pd.DataFrame(X_test, columns=X.columns) display(X_train.head())



	PassengerId	Pclass	Age	SibS	Parch	Fare	Sex_male	Embarked_Q	Embarked_S
0	1.360492	-1.584396	0.010681	-0.47969	-0.460682	-0.018600	0.728823	-0.311564	-1.611198
1	-1.632266	-1.584396	-0.119643	-0.47969	-0.460682	0.079245	0.728823	-0.311564	0.620656
າ	_1 3//650	_1 59/306	_Ი ᲜᲘ२1//요	_n 170692	0.810657	0.646624	0.728823	-0.311564	-1.611198
3	-1.686680	-0.381742	-1.193456	0.493365	-0.460682	-0.031329	-1.372075	-0.311564	-1.611198
4	-1.111449	0.820913	0.033758	-0.479698	-0.460682	-0.479818	0.728823	-0.311564	0.620656
	PassengerId	Pclass	Age	SibSp	Parch	Fare	Sex_male	Embarked_Q	Embarked_S
0	PassengerId 0.676433	Pclass 0.820913	Age -0.273045	SibSp 0.493365	Parch -0.460682	-0.315867	Sex_male -1.372075	Embarked_Q -0.311564	Embarked_S 0.620656
0				•			_		
-	0.676433	0.820913	-0.273045	0.493365	-0.460682	-0.315867	-1.372075	-0.311564	0.620656
1	0.676433	0.820913 0.820913	-0.273045 -0.809952	0.493365 -0.479698	-0.460682 -0.460682	-0.315867 -0.485419	-1.372075 0.728823	-0.311564 -0.311564	0.620656 0.620656

print(X_train.info())



<class 'pandas.core.frame.DataFrame'> RangeIndex: 712 entries, 0 to 711 Data columns (total 9 columns):

		ar > cora	
#	Column	Non-Null Count	Dtype
0	PassengerId	712 non-null	float64
1	Pclass	712 non-null	float64
2	Age	712 non-null	float64
3	SibSp	712 non-null	float64
4	Parch	712 non-null	float64
5	Fare	712 non-null	float64
6	Sex_male	712 non-null	float64
7	Embarked_Q	712 non-null	float64
8	Embarked_S	712 non-null	float64
dtype	es: float64(9)	

memory usage: 50.2 KB

None

print(X_test.info())



<class 'pandas.core.frame.DataFrame'> RangeIndex: 179 entries, 0 to 178 Data columns (total 9 columns):

Data	COTUMIS (COC	ar o corumns).	
#	Column	Non-Null Count	Dtype
0	PassengerId	179 non-null	float64
1	Pclass	179 non-null	float64
2	Age	179 non-null	float64
3	SibSp	179 non-null	float64
4	Parch	179 non-null	float64
5	Fare	179 non-null	float64
6	Sex_male	179 non-null	float64
7	Embarked_Q	179 non-null	float64
8	Embarked_S	179 non-null	float64
44	C1+C4/0	`	

dtypes: float64(9) memory usage: 12.7 KB

None