In [1]:	import pandas as pd import numpy as np import seaborn as sns
In [4]:	<pre>import matplotlib.pyplot as plt df = pd.read_csv("used.csv") df.head(10)</pre>
Out[4]:	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 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 5 6 0 3 Moran, Mr. James male NaN 0 0 330877 8.4583 NaN Q 6 7 0 1 McCarthy, Mr. Timothy J male 54.0 0 0 17463 51.8625 E46 S
In [6]:	7 8 0 3 Palsson, Master. Gosta Leonard male 2.0 3 1 349909 21.0750 NaN S 8 9 1 3 Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg) female 27.0 0 2 347742 11.1333 NaN S 9 10 1 2 Nasser, Mrs. Nicholas (Adele Achem) female 14.0 1 0 237736 30.0708 NaN C
Out[6]:	PassengerId Survived Pclass Name Sex Age SibSp Parch Ticket Fare Cabin Embarked 881 882 0 3 Markun, Mr. Johann male 33.0 0 0 349257 7.8958 NaN S 882 883 0 3 Dahlberg, Miss. Gerda Ulrika female 22.0 0 0 7552 10.5167 NaN S
	883 884 0 2 Banfield, Mr. Frederick James male 28.0 0 0 C.A./SOTON 34068 10.5000 NaN S 884 885 0 3 Sutehall, Mr. Henry Jr male 25.0 0 0 SOTON/OQ 392076 7.0500 NaN S 885 886 0 3 Rice, Mrs. William (Margaret Norton) female 39.0 0 5 382652 29.1250 NaN Q 886 887 0 2 Montvila, Rev. Juozas male 27.0 0 0 211536 13.0000 NaN S
	887 888 1 1 Graham, Miss. Margaret Edith female 19.0 0 0 112053 30.0000 B42 S 888 889 0 3 Johnston, Miss. Catherine Helen "Carrie" female NaN 1 2 W./C. 6607 23.4500 NaN S 889 890 1 1 Behr, Mr. Karl Howell male 26.0 0 0 111369 30.0000 C148 C 890 891 0 3 Dooley, Mr. Patrick male 32.0 0 0 370376 7.7500 NaN Q
In [8]:	<pre>print(df.info()) <class 'pandas.core.frame.dataframe'=""> RangeIndex: 891 entries, 0 to 890 Data columns (total 12 columns):</class></pre>
	# Column Non-Null Count Dtype O PassengerId 891 non-null int64 Survived 891 non-null int64 Polass 891 non-null int64 Non-Null object Survived 891 non-null object Survived 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
In [9]:	<pre>dtypes: float64(2), int64(5), object(5) memory usage: 83.7+ KB None df.count()</pre>
Out[9]:	PassengerId 891 Survived 891 Pclass 891 Name 891 Sex 891 Age 714 SibSp 891
	Parch 891 Ticket 891 Fare 891 Cabin 204 Embarked 889 dtype: int64
In [11]: Out[11]: In [12]:	<pre>df.shape (891, 12) df.size</pre>
Out[12]: In [15]:	<pre>10692 missing_values=df.isnull().sum()</pre>
In [20]:	<pre>print(missing_values) PassengerId 0 Survived 0 Pclass 0 Name 0</pre>
	Sex 0 Age 177 SibSp 0 Parch 0 Ticket 0 Fare 0 Cabin 687
In [21]:	<pre>Embarked 2 dtype: int64 missing_percentage = (df.isnull().sum() / len(df)) * 100 print(missing_percentage) PassengerId 0.000000</pre>
	Survived 0.000000 Pclass 0.000000 Name 0.000000 Sex 0.000000 Age 19.865320 SibSp 0.000000 Parch 0.000000
In [24]:	Ticket 0.000000 Fare 0.000000 Cabin 77.104377 Embarked 0.224467 dtype: float64 sns.heatmap(df.isnull(), cmap='viridis', cbar=False)
	plt.show()
	0 - 43 - 43 - 42 - 42 - 42 - 42 - 42 - 42
	Survived - Prilate - Prila
In [25]:	<pre>numerical_columns = df.select_dtypes(include=['int64', 'float64']).columns df[numerical_columns] = df[numerical_columns].fillna(df[numerical_columns].mean())</pre>
	<pre>categorical_columns = df.select_dtypes(include='object').columns df[categorical_columns] = df[categorical_columns].fillna('Unknown') # Chcek if any missing value is left after handling print(df.isnull().sum())</pre>
	PassengerId 0 Survived 0 Pclass 0 Name 0 Sex 0 Age 0
	SibSp 0 Parch 0 Ticket 0 Fare 0 Cabin 0 Embarked 0 dtype: int64
In [26]: Out[26]:	PassengerId 891 Survived 891 Pclass 891 Name 891
	Sex 891 Age 891 SibSp 891 Parch 891 Ticket 891 Fare 891 Cabin 891
In [27]: Out[27]:	Embarked dtype: int64
	PassengerId12345Survived01110Pclass331313NameBraund, Mr. Owen HarrisCumings, Mrs. John Bradley (Florence Briggs ThHeikkinen, Miss. LainaFutrelle, Mrs. Jacques Heath (Lily May Peel)Allen, Mr. William Henry
	Sex male female female female male Age 22.0 38.0 26.0 35.0 35.0 SibSp 1 0 1 0 Parch 0 0 0 0 Tight 0/5 21171 DC 17500 STON/02 2101292 112902 272450
	Ticket A/5 21171 PC 17599 STON/O2. 3101282 113803 373450 Fare 7.25 71.2833 7.925 53.1 8.05 Cabin Unknown C123 Unknown Embarked S C S S S
In [28]: In [29]:	<pre>df.to_csv('usedcleaned_dataset.csv', index=False) sns.boxplot(x=df.Fare, showfliers=False)</pre>
Out[29]:	<pre><axessubplot:xlabel='fare'></axessubplot:xlabel='fare'></pre>
In [31]:	0 10 20 30 40 50 60 Fare sns.boxplot(x=df.Age, showfliers=False)
Out[31]:	<pre><axessubplot:xlabel='age'></axessubplot:xlabel='age'></pre>
To [00].	10 20 30 40 50 Age
In [32]:	<pre>def remove_outliers_IQR(df, Fare, threshold=1.5): q1 = df["Fare"].quantile(0.25) q3 = df["Fare"].quantile(0.75) iqr = q3 - q1 lower_bound = q1 - threshold * iqr upper_bound = q3 + threshold * iqr return df[(df["Fare"] >= lower_bound) & (df['Fare'] <= upper_bound)]</pre>
	<pre>columns_to_check = ['Fare'] for col in columns_to_check: df1 = remove_outliers_IQR(df, col) #here df1 is new dataframe with no missing values and without outliers in fare column #comparsion between with and without outliers in fare column for better understanding plt.figure(figsize=(10, 6)) plt.boxplot([df['Fare'], df1['Fare']], labels=['With Outliers', 'Without Outliers'])</pre>
	plt.stitle('Comparison of Fare with and without outliers') plt.xlabel('Data with and without outliers') plt.ylabel('Fare Values') plt.show() Comparison of Fare with and without outliers
	500 - C
	300 - CO
In [36]:	With Outliers Data with and without outliers Without Outliers Data with and without outliers def remove_outliers_IQR(df, Fare, threshold=1.5): q1 = df1["Age"].quantile(0.25) q3 = df1["Age"].quantile(0.75) igr = q2 = q3
In [37]:	<pre>iqr = q3 - q1 lower_bound = q1 - threshold * iqr upper_bound = q3 + threshold * iqr return df1[(df1["Age"] >= lower_bound) & (df1['Age'] <= upper_bound)]</pre> columns_to_check = ['Age']
In [38]:	<pre>for col in columns_to_check: df2 = remove_outliers_IQR(df1, col) plt.figure(figsize=(10, 6)) plt.boxplot([df1['Age'], df2['Age']], labels=['With Outliers', 'Without Outliers'])</pre>
	plt.title('Comparison of Age with and without outliers') plt.xlabel('Data with and without outliers') plt.ylabel('Age Values') plt.show() Comparison of Age with and without outliers
	80 - O 70 - O 60 - O
	So - Age Values 30 - Age Value
	20 - 10 - 10 - With Outliers Without Outliers
	Data with and without outliers Data with and without outliers