

22/7/25

## Preprocessing using Pandas and Simple Imputer.

### AIM:

To load titanic dataset from csv, handle missing values using simple imputer, analyze key passenger features, filter passenger based on candidates, and prepare data for model training and testing.

### Procedure / Algorithm:

Step 1: load titanic.csv into a dataframe

Step 2: Explore dataset shape, info and summary statistics.

Step 3: use Simple Imputer to fill missing Age.

Step 4: Fill missing cabin with "unknown" and embarked with mode.

Step 5: visualize passenger class.

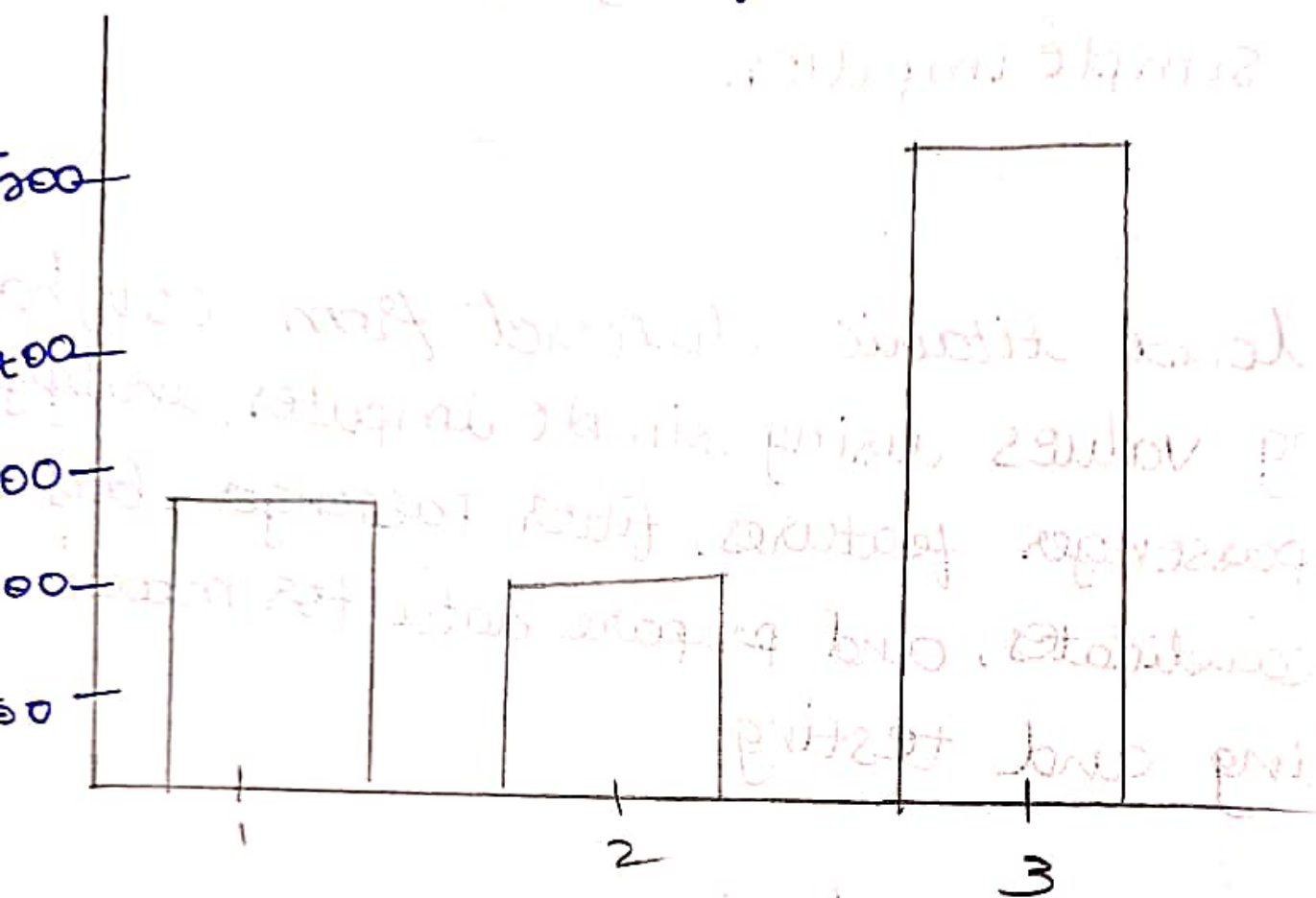
Step 6: Filter Passangers by genders, Survival, class, age fav embarking, family abroad, and survival status.

Step 7: Identity top oldest survivors and zeros - five Passangers

Step 8: Split training and testing sets.

output:

## passanger class distribution



Pclass



```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.impute import SimpleImputer
from sklearn.model_selection import train-  
test-split
```

```
df = sns.load_dataset('titanic')
df['age'] = SimpleImputer(strategy='mean')  
fit_transform(df[['age']])
```

```
df['deck'] = df['deck'].cat.add_categories('unknown')  
df['deck'] = df['deck'].fillna('unknown')
```

```
df['embarked'] = df['embarked'].fillna  
(df['embarked'].mode()[0])
```

```
Sns.countplot(x='Pclass', data=df) plt.title  
( 'Passenger class distribution')  
plt.show()
```

```
Print ("Females who survived:", df[(df,  
sex = 'female') & (df.survived == 1)].index.  
to_list())
```

```
Print(" 3rd class passengers under 18: ",  
df[(df.pclass == 3) & (df.age < 18)].  
index.to_list())
```

Passengers who paid zero tax: 15 passengers

Training set size: 712

Testing set size = 179

```
Print ("1st class passengers older than "  
dt[(dt.pclass == 1) & (dt.age > 40)].  
index.tolist())
```

```
Print ("1st class passengers older than 40  
who survived: ", dt[(dt.pclass == 1) & (dt.age >  
40) & (dt.survived == 1)].index.tolist())
```

### RESULT:

The Program successfully identifies passengers with zero fare and efficiently splits the datasets into 80% training and 20% testing sets, ensuring reproducibility and readiness for machine learning tasks.



## EX.NO:3 Model Planning and Building

### Aim:

To describe the model planning and Building of the whole data set.

### Code:

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
from sklearn.metrics import selection import
train-test split
from sklearn.metrics import mean_squared
from sklearn.metrics import mean_squared
df = pd.read_csv('advertising.csv')
Print (df.head())
Print (df.describe())
X = df[['TV', 'Radio', 'Newspaper']]
Y = df['Sales']
# Split data
```

	TV	Radio	Newspaper	Sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.7	69.3	12.0
3	151.5	41.3	58.5	16.5
4	180.8	10.8	58.4	17.9

	TV	Radio	Newspaper	Sale
Count	200.0000	2000.0000	2000.0000	200.0000
mean	147.0425	23.2640	30.55400	15.130500
Std	85.884	14.846	21.7786	5.2889
min	0.2000	0.000	0.3000	1.00000
25%	79.33130	9.975	12.78000	11.00000
50%	149.7500	22.900	25.78000	16.00000
max	296.4000	49.600	114.000	27.00000

Linear Regression MSE: 4.522582562041291



```
x_train, x_test, y_train, y_test = train_test_split  
(x, y, test_size = 0.2, random_state = 0)
```

```
# Train model
```

```
model = Linear Regression()  
model.fit(x_train, y_train)
```

```
# Predict & Evaluate
```

```
y_pred = model.predict(x_test)
```

```
mse = mean_squared_error(y_test, y_pred)
```

```
print("Linear Regression MSE, mse")
```

```
plt.figure(figsize=(8,5))
```

```
sns.scatterplot(x=y_test, y=y_pred)
```

```
plt.xlabel("Actual Sales")
```

```
plt.ylabel("Predicted Sales")
```

```
plt.title("Linear Regression: Actual vs  
Predicted Sales")
```

```
plt.show()
```

```
# Apply k-mean
```

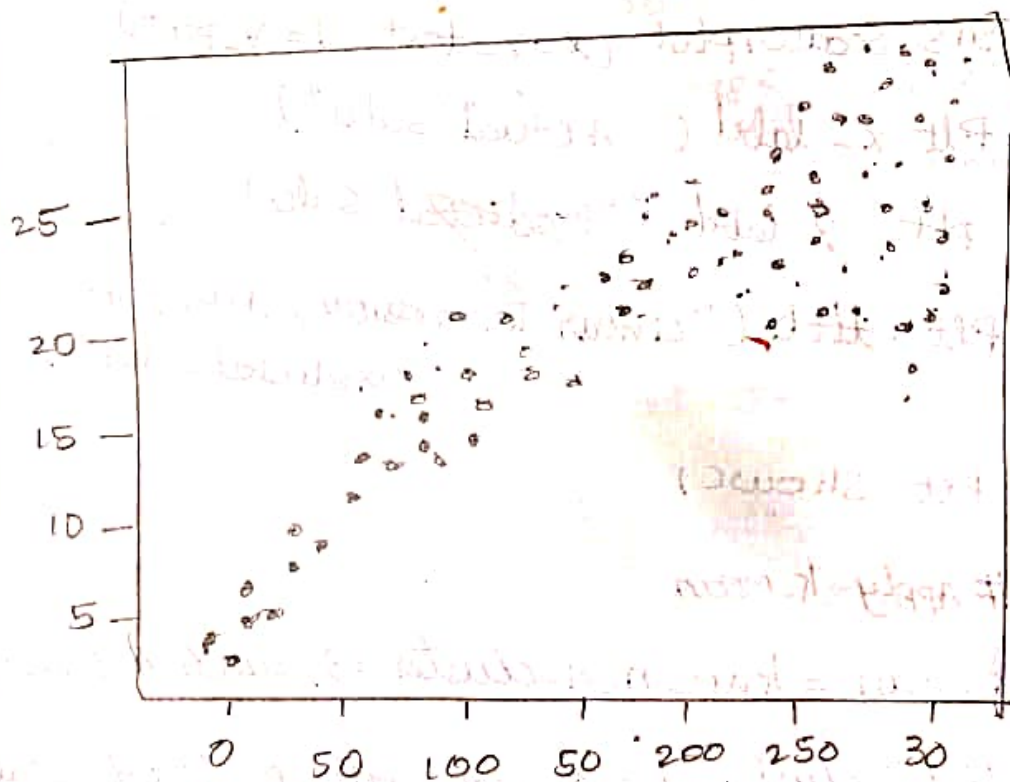
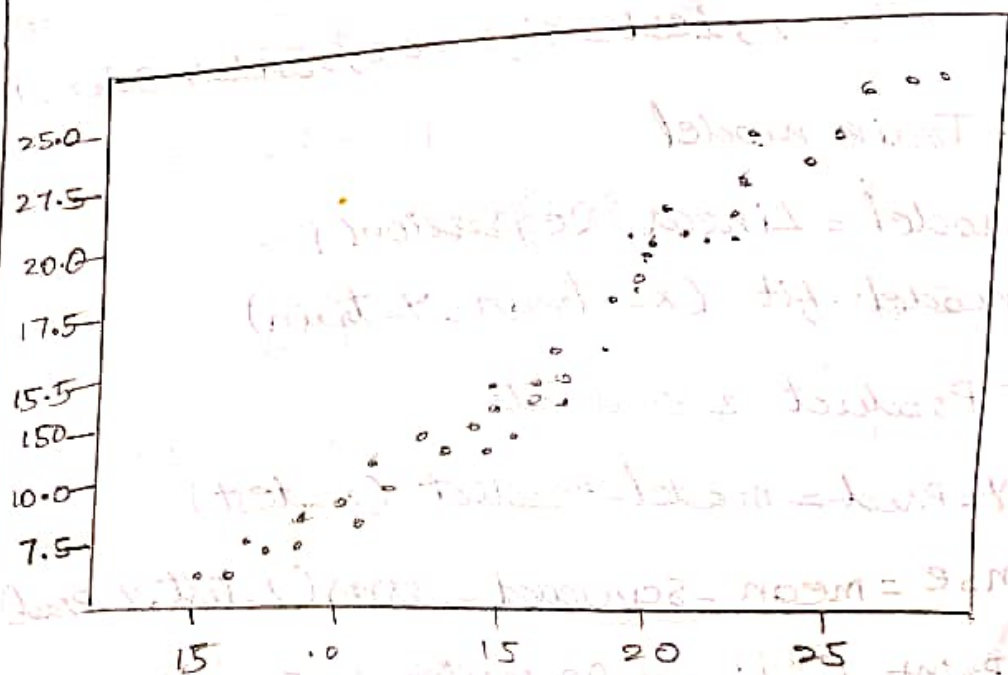
```
kmean = kmean(n_clusters=3, random_state=0)
```

```
dt["cluster"] = kmean.fit_predict(scaled)
```

```
plt.figure(figsize=(8,6))
```



# Actual vs Predicted Sale



```
Sns.scatter plot (data=df, x='TV',  
y='sales')
```

```
plt.title ('k-mean clustering: TV Budget  
vs Sales')
```

```
plt.show()
```

RESULT:

The program has been executed  
successful.