#### In [1]:

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

#### In [2]:

```
df = pd.read_csv("C:/Users/ameya/OneDrive/Desktop/DSBDAL/Social_Network_Ads.csv")
```

### In [3]:

df.head()

#### Out[3]:

	User ID	Gender	Age	EstimatedSalary	Purchased	Unnamed: 5
0	15624510	Male	19	19000	0	NaN
1	15810944	Male	35	20000	0	NaN
2	15668575	Female	26	43000	0	NaN
3	15603246	Female	27	57000	0	NaN
4	15804002	Male	19	76000	0	NaN

# In [4]:

```
df2 = df.iloc[:,:-1]
```

#### In [6]:

df2.head()

#### Out[6]:

	User ID	Gender	Age	EstimatedSalary	Purchased
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15668575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0

## In [7]:

```
df2.describe()
```

# Out[7]:

	User ID	Age	EstimatedSalary	Purchased
count	4.000000e+02	400.000000	400.000000	400.000000
mean	1.569154e+07	37.655000	69742.500000	0.357500
std	7.165832e+04	10.482877	34096.960282	0.479864
min	1.556669e+07	18.000000	15000.000000	0.000000
25%	1.562676e+07	29.750000	43000.000000	0.000000
50%	1.569434e+07	37.000000	70000.000000	0.000000
75%	1.575036e+07	46.000000	88000.000000	1.000000
max	1.581524e+07	60.000000	150000.000000	1.000000

#### In [8]:

```
df2.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 5 columns):

#	Column	Non-Null Count	Dtype
0	User ID	400 non-null	int64
1	Gender	400 non-null	object
2	Age	400 non-null	int64
3	EstimatedSalary	400 non-null	int64
4	Purchased	400 non-null	int64

dtypes: int64(4), object(1)
memory usage: 15.8+ KB

#### In [9]:

```
df2['Purchased'].value_counts()
```

#### Out[9]:

0 2571 143

Name: Purchased, dtype: int64

```
In [10]:
df2.isnull().sum()
Out[10]:
User ID
                    0
Gender
                    0
Age
                    0
EstimatedSalary
                    0
Purchased
                    0
dtype: int64
In [11]:
df2['EstimatedSalary'].value_counts()
Out[11]:
72000
          12
80000
          11
79000
          10
75000
           9
71000
           9
123000
           1
           1
37000
115000
           1
148000
           1
           1
139000
Name: EstimatedSalary, Length: 117, dtype: int64
In [12]:
df2['Gender'].value_counts()
```

### Out[12]:

Female 204 Male 196

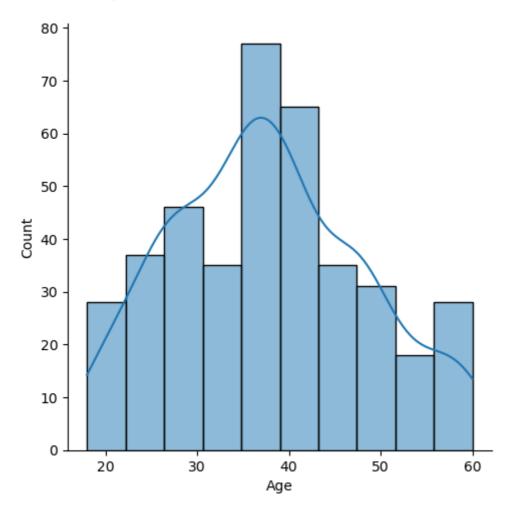
Name: Gender, dtype: int64

# In [16]:

```
sns.displot(data = df2,x='Age',kde=True)
```

# Out[16]:

<seaborn.axisgrid.FacetGrid at 0x18f29bb4550>

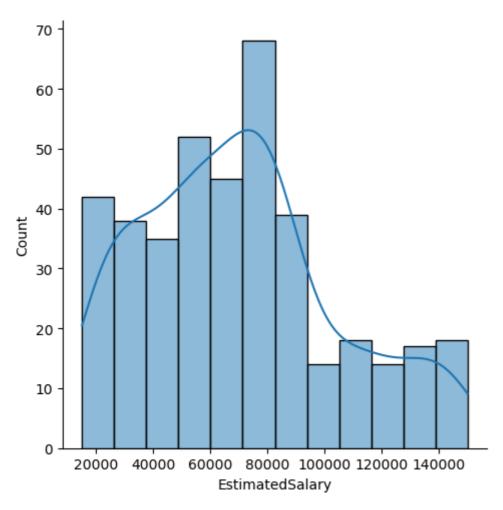


#### In [17]:

```
sns.displot(df['EstimatedSalary'], kde=True)
```

#### Out[17]:

<seaborn.axisgrid.FacetGrid at 0x18f284d1b20>



# In [18]:

```
x = df2.iloc[:,2:4].values
y = df2.iloc[:,-1].values
```

#### In [19]:

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=0)
```

# Normalizing the dataset

```
In [20]:
```

```
from sklearn.preprocessing import StandardScaler
scalar = StandardScaler()
x_train = scalar.fit_transform(x_train)
x_test = scalar.fit_transform(x_test)
```

# Model

```
In [21]:
```

```
from sklearn.linear_model import LogisticRegression
classifier = LogisticRegression()
classifier.fit(x_train,y_train)
```

#### Out[21]:

LogisticRegression()

#### In [22]:

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

#### In [23]:

```
y_pred
```

#### Out[23]:

#### In [24]:

```
print("Training Accuracy: ",classifier.score(x_train,y_train)*100)
print("Testing Accuracy: ",classifier.score(x_test,y_test)*100)
```

Training Accuracy: 82.1875
Testing Accuracy: 88.75

# **Computing Accuracy**

```
In [25]:
```

```
from sklearn.metrics import accuracy_score
Acc = accuracy_score(y_true= y_test, y_pred = y_pred)
```

```
In [26]:
Acc
Out[26]:
0.8875
In [27]:
from sklearn.metrics import confusion matrix
cm = confusion_matrix(y_true=y_test,y_pred = y_pred)
print(cm)
[[52 6]
[ 3 19]]
In [28]:
from sklearn.metrics import precision_score
ps = precision_score(y_test,y_pred)
In [29]:
ps
Out[29]:
0.76
In [30]:
from sklearn.metrics import precision_recall_fscore_support
prfc = precision_recall_fscore_support(y_test,y_pred)
print('precision:',prfc[0])
print('Recall:',prfc[1])
print('fscore:',prfc[2])
print('support:',prfc[3])
precision: [0.94545455 0.76
                                  ]
Recall: [0.89655172 0.86363636]
fscore: [0.92035398 0.80851064]
support: [58 22]
```

# In [31]:

```
from sklearn.metrics import classification_report
cr = classification_report(y_test,y_pred)
print(cr)
```

support	f1-score	recall	precision	
58	0.92	0.90	0.95	0
22	0.81	0.86	0.76	1
80	0.89			accuracy
80	0.86	0.88	0.85	macro avg
80	0.89	0.89	0.89	weighted avg

# In [ ]: