In [1]:
```
import pandas as pd
```

In [7]:
```
df = pd.read_csv('Downloads//Social_Network_Ads.csv')
```

In [9]:
```
df
```

Out[9]:

	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
...	...	...	...	...	...
395	15691863	Female	46	41000	1
396	15706071	Male	51	23000	1
397	15654296	Female	50	20000	1
398	15755018	Male	36	33000	0
399	15594041	Female	49	36000	1

400 rows × 5 columns

In [11]:
```
 #input data
x=df[['Age','EstimatedSalary']]

#output data
y=df['Purchased']
```

In [13]:
```
 from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
x_scaled = scaler.fit_transform(x)
```

In [15]:
```
 #cross. validation

from sklearn.model_selection import train_test_split
```

In [19]:
```
 x_train, x_test, y_train, y_test = train_test_split(x_scaled,y,random_state=0,test
```

In [21]:
```
x_train
```

Out[21]:
```
array([[0.61904762, 0.17777778],
       [0.33333333, 0.77777778],
       [0.47619048, 0.25925926],
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       [0.47619048, 0.34074074],
       [1.        , 0.68888889],
       [0.04761905, 0.4962963 ],
```

```
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[0.92857143, 0.13333333],
[0.38095238, 0.42222222],
```

```
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       [0.71428571, 0.11111111],
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       [0.42857143, 0.27407407],
       [0.21428571, 0.28888889],
       [0.19047619, 0.76296296]])
```

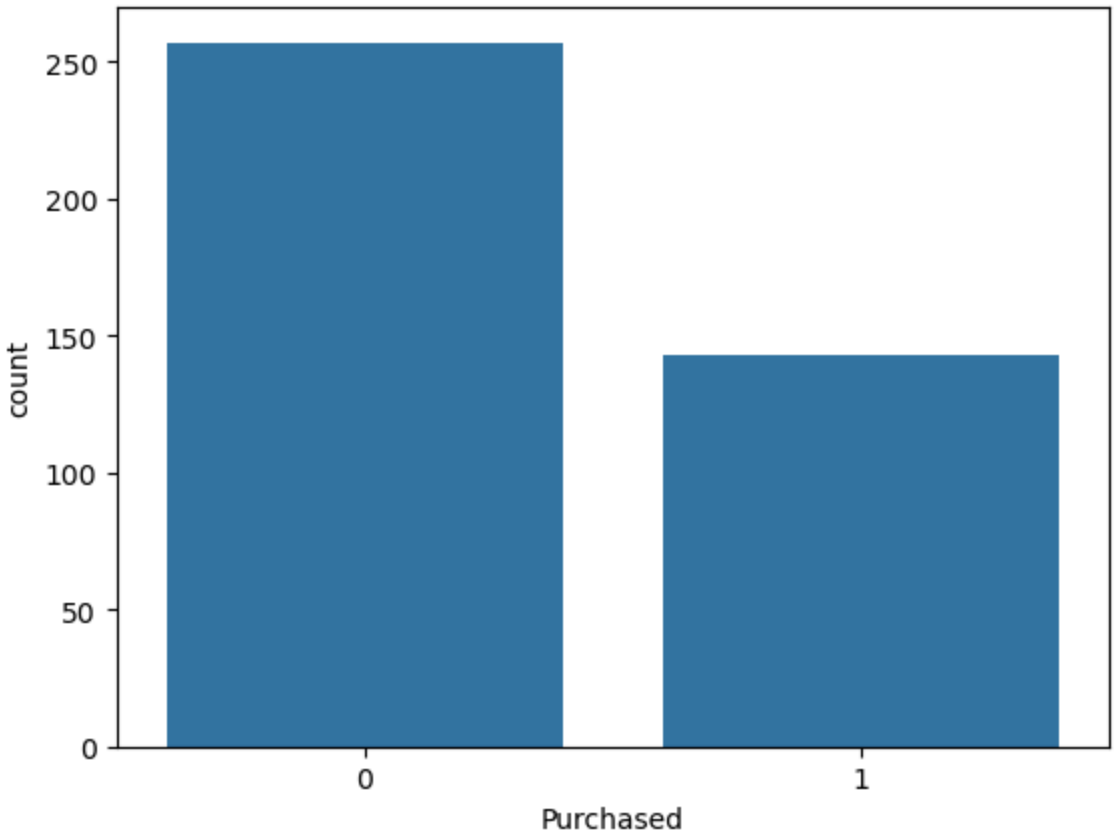
In [23]:
```
y_train
```

Out[23]:
```
250    0
63     1
312    0
159    1
283    1
      ..
323    1
192    0
117    0
47     0
172    0
Name: Purchased, Length: 300, dtype: int64
```

In [25]:
```
from sklearn.linear_model import LogisticRegression
```

In [27]:
```
import seaborn as sns
sns.countplot(x=y)
```

Out[27]:
```
<Axes: xlabel='Purchased', ylabel='count'>
```



In [29]:
```
y.value_counts()
```

Out[29]:
```
Purchased
0    257
1    143
Name: count, dtype: int64
```

In [31]:
```
#creat the object
classifier = LogisticRegression()
```

In [33]:
```
classifier.fit(x_train,y_train)
```

Out[33]:
```
LogisticRegression()
```

In [35]:
```
#predication
y_pred = classifier.predict(x_test)
```

In [37]:
```
y_train.shape
```

Out[37]:
```
(300,)
```

In [39]:
```
x_train.shape
```

Out[39]:
```
(300, 2)
```

In [41]:
```
y_pred
```

Out[41]:
```
array([0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
       0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1,
       0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1,
       0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1], dtype=int64)
```

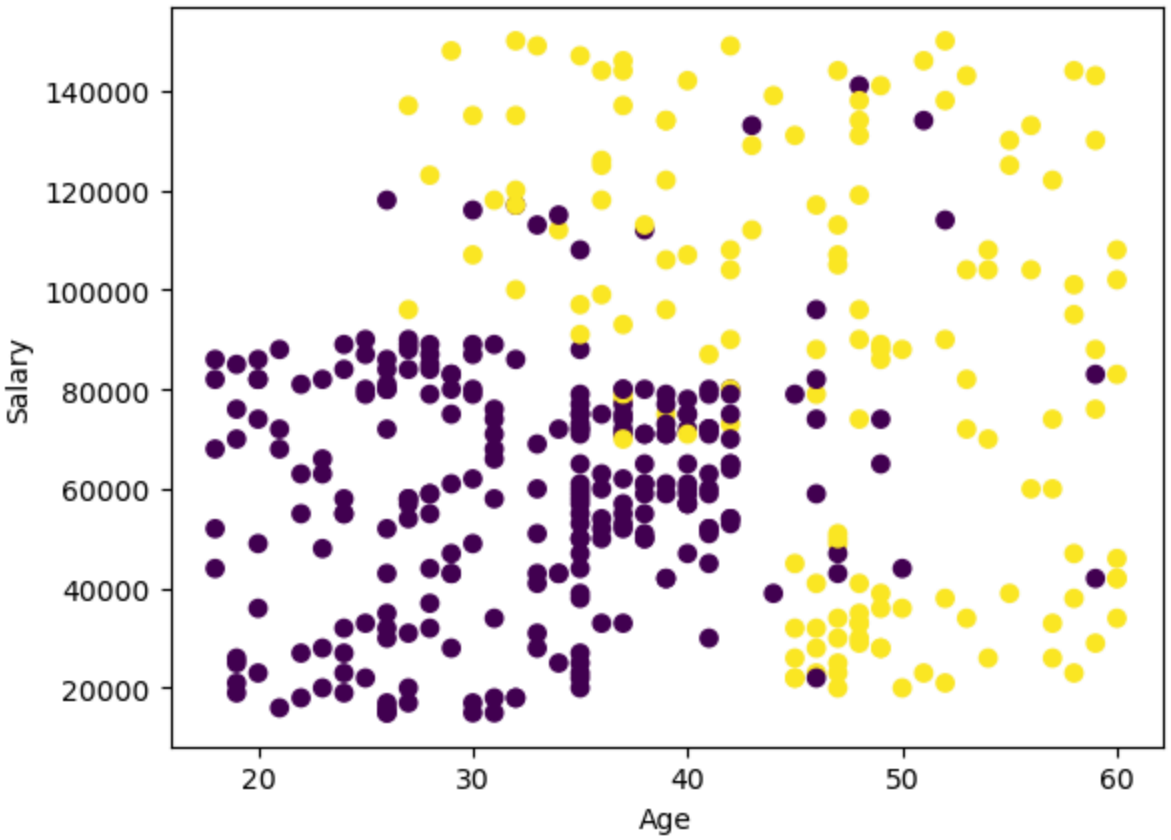
In [43]:
```
y_test
```

Out[43]:
```
132    0
309    0
341    0
196    0
246    0
      ..
146    1
135    0
390    1
264    1
364    1
Name: Purchased, Length: 100, dtype: int64
```

In [45]:
```
import matplotlib.pyplot as plt
```

In [47]:
```
plt.xlabel('Age')
plt.ylabel('Salary')
plt.scatter(x['Age'],x['EstimatedSalary'],c=y)
```

Out[47]:
```
<matplotlib.collections.PathCollection at 0x26c46625f10>
```



In [49]:
```
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
x_scaled = scaler.fit_transform(x)
```

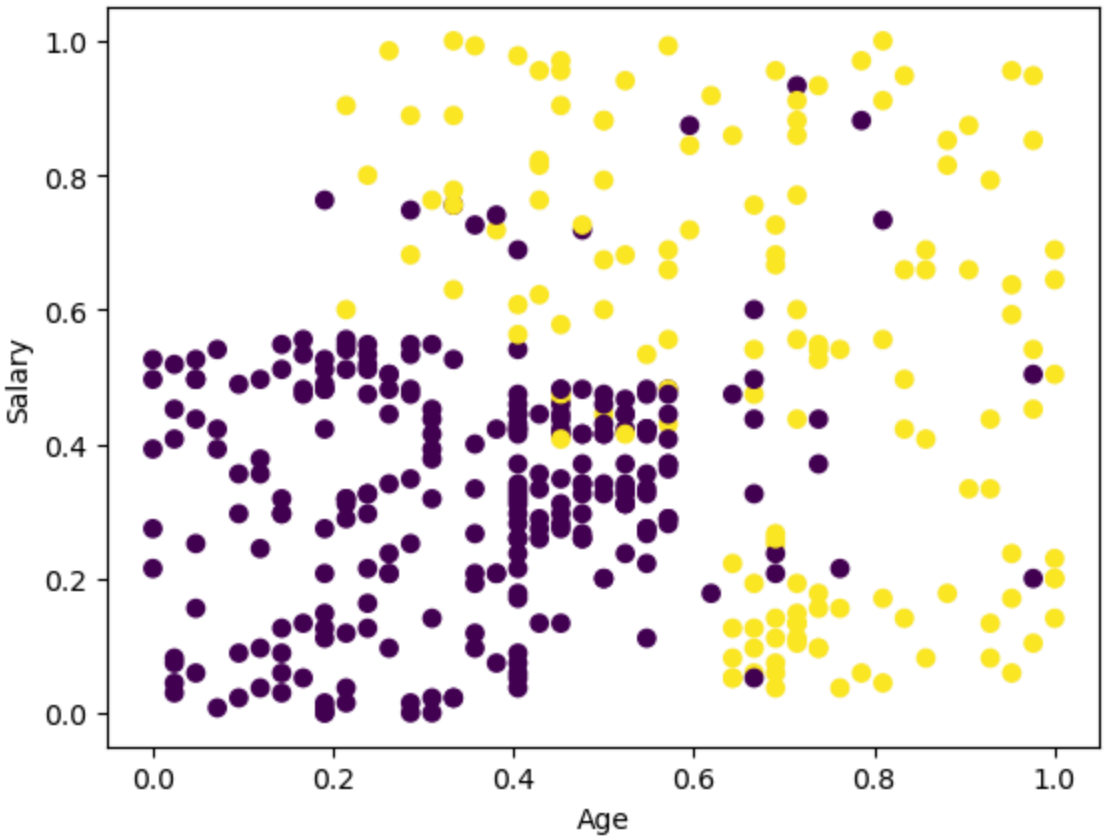
In [51]:
```
pd.DataFrame(x_scaled).describe()
```

Out[51]:

	0	1
count	400.000000	400.000000
mean	0.467976	0.405500
std	0.249592	0.252570
min	0.000000	0.000000
25%	0.279762	0.207407
50%	0.452381	0.407407
75%	0.666667	0.540741
max	1.000000	1.000000

In [53]:
```
plt.xlabel('Age')
plt.ylabel('Salary')
plt.scatter(x_scaled[:,0],x_scaled[:,1],c=y)
```

Out[53]:
```
<matplotlib.collections.PathCollection at 0x26c46f50830>
```



In [55]:
```
from sklearn.metrics import confusion_matrix
```

In [57]:
```
confusion_matrix(y_test,y_pred)
```

Out[57]:
```
array([[67,  1],
       [10, 22]], dtype=int64)
```

In [59]:
```
y_test.value_counts()
```

Out[59]:
```
Purchased
0    68
1    32
Name: count, dtype: int64
```

In [77]:
```
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import ConfusionMatrixDisplay, confusion_matrix
```

In [79]:
```
model = LogisticRegression()
model.fit(x_train, y_train)
```

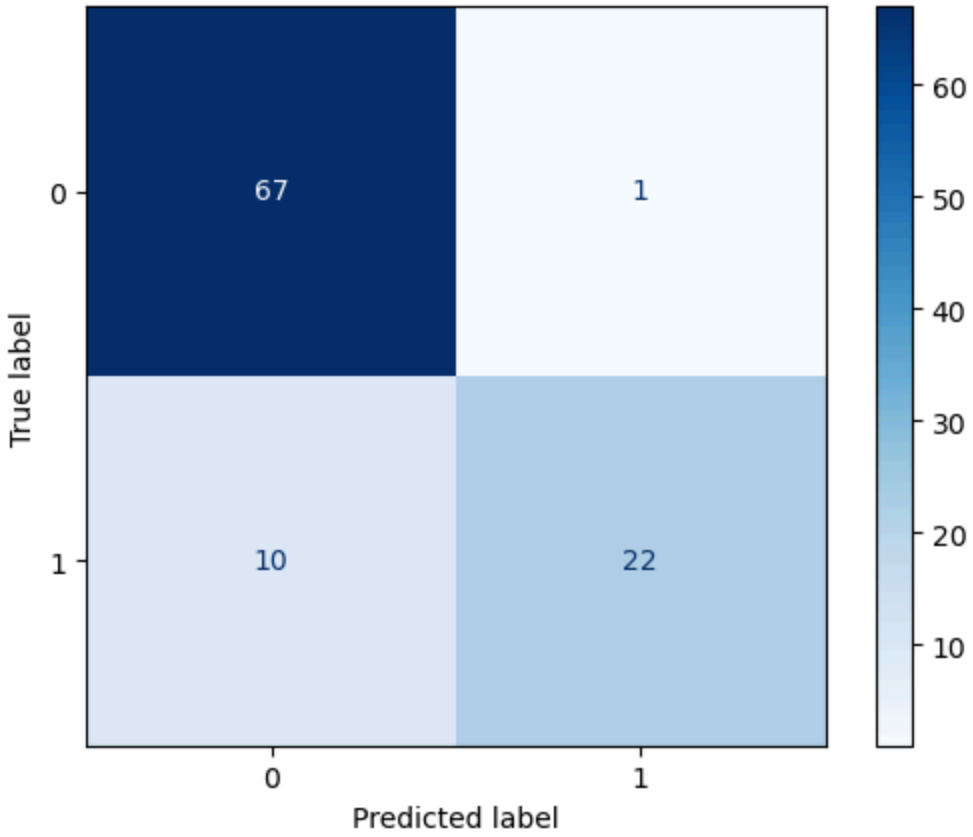
Out[79]:
```
LogisticRegression()
```

In [81]:
```
y_pred = model.predict(x_test)
```

In [83]:
```
cm = confusion_matrix(y_test, y_pred)
```

In [85]:
```
disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=model.classes_)
disp.plot(cmap='Blues')
```

Out[85]:
```
<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x26c46885460>
```



In [87]:
```
from sklearn.metrics import accuracy_score
accuracy_score(y_test,y_pred)
```

Out[87]:
```
0.89
```

In [89]:
```
from sklearn.metrics import classification_report
```

In [91]:
```
print(classification_report(y_test,y_pred))
```

```
              precision    recall  f1-score   support

           0       0.87      0.99      0.92        68
           1       0.96      0.69      0.80        32

    accuracy                           0.89       100
   macro avg       0.91      0.84      0.86       100
weighted avg       0.90      0.89      0.88       100
```

In [93]:
```
new1=[[26,34000]]
new2=[[57,138000]]
```

In [95]:
```
classifier.predict(scaler.transform(new1))
```

```
C:\ProgramData\anaconda3\Lib\site-packages\sklearn\base.py:493: UserWarning: X does
not have valid feature names, but MinMaxScaler was fitted with feature names
  warnings.warn(
```

Out[95]:
```
array([0], dtype=int64)
```

In [97]:
```
classifier.predict(scaler.transform(new2))
```

```
C:\ProgramData\anaconda3\Lib\site-packages\sklearn\base.py:493: UserWarning: X does
not have valid feature names, but MinMaxScaler was fitted with feature names
  warnings.warn(
```

Out[97]:
```
array([1], dtype=int64)
```

In [ ]: