

```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [2]: data = pd.read_csv('C:/Users/prajw/Downloads/SPPU-TE-COMP-SEM-2-2019-PATTERN-DSBDA-main/SPPU-TE-COMP-SEM-2-2019-PATTERN-DSBDA-m
```

```
In [3]: data
```

Out[3]:

	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 [4]: data.head(5)
```

Out[4]:

	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 [5]: data.tail()
```

Out[5]:

	User ID	Gender	Age	EstimatedSalary	Purchased
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

```
In [6]: data.shape
```

Out[6]: (400, 5)

```
In [7]: data.columns
```

Out[7]: Index(['User ID', 'Gender', 'Age', 'EstimatedSalary', 'Purchased'], dtype='object')

```
In [8]: data.describe()
```

Out[8]:

	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 [9]:

data.isnull().sum()

Out[9]:

User ID0
Gender0
Age0
EstimatedSalary0
Purchased0
dtype: int64

In [10]:

data.iloc[:,2:4]

Out[10]:

	Age	EstimatedSalary
0	19	19000
1	35	20000
2	26	43000
3	27	57000
4	19	76000
...
395	46	41000
396	51	23000
397	50	20000
398	36	33000
399	49	36000

400 rows × 2 columns

In [11]:

data.iloc[:,2:4].values

```
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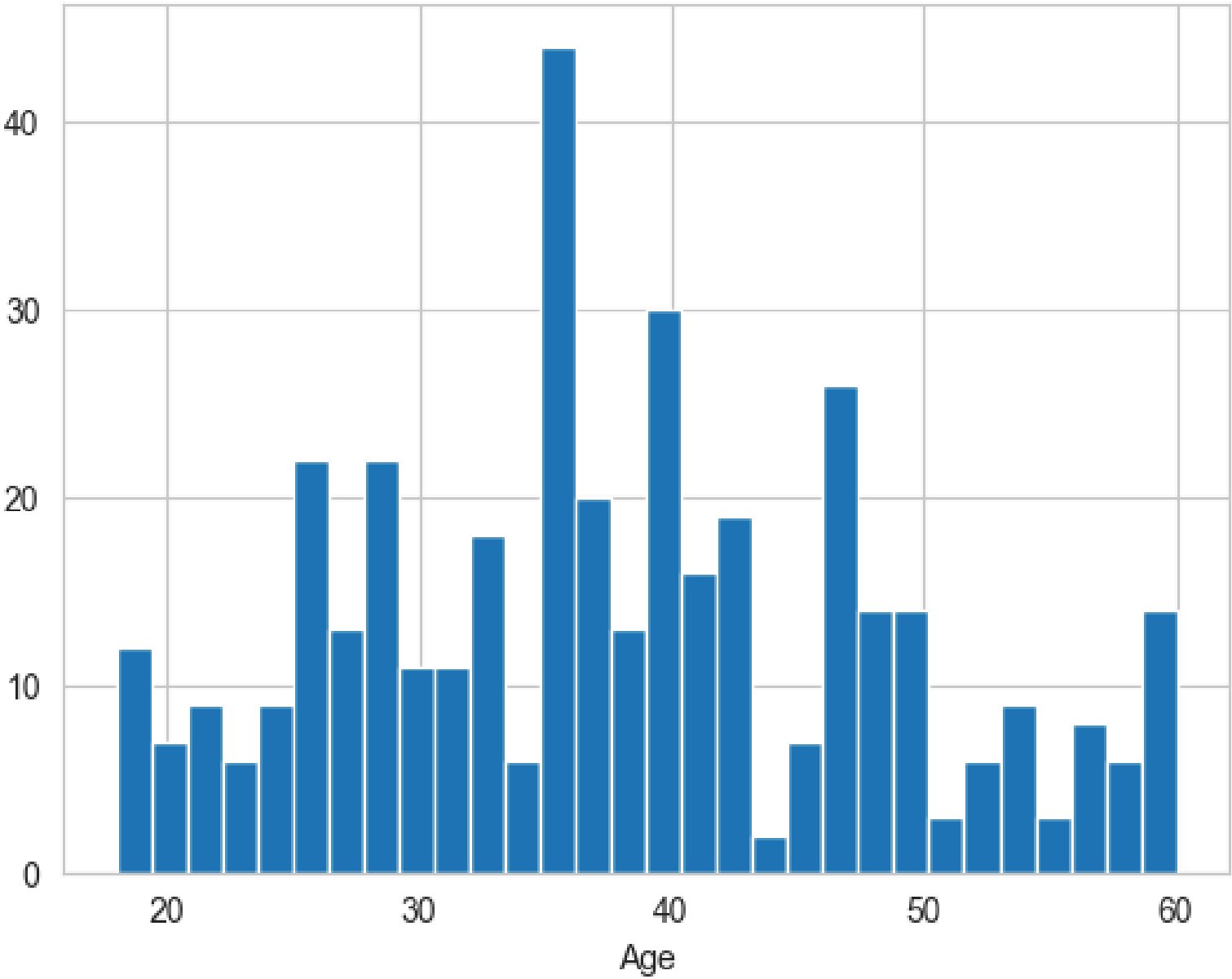
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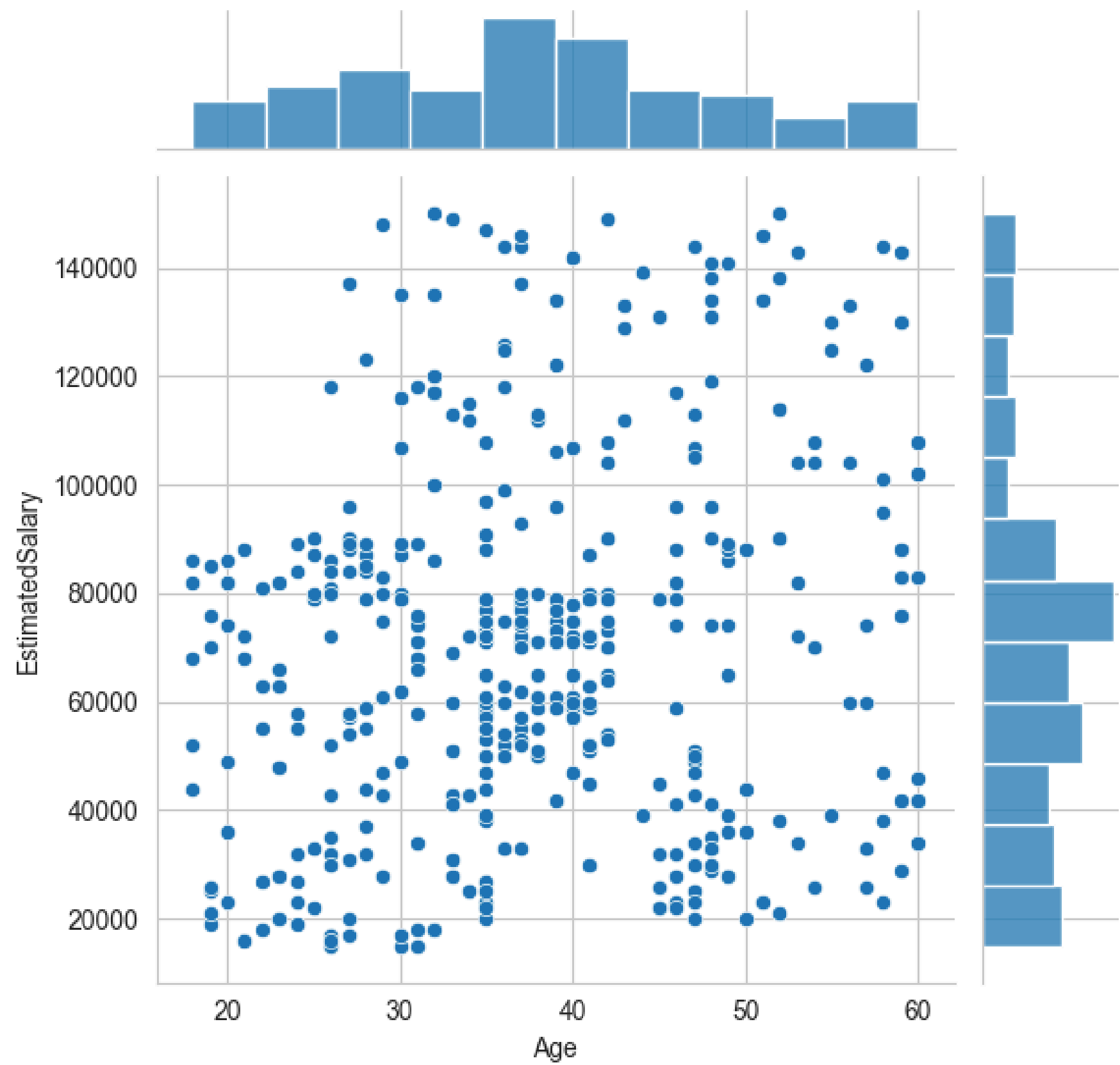
```
In [12]: sns.set_style('whitegrid')
data['Age'].hist(bins=30)
plt.xlabel('Age')
```

Out[12]: Text(0.5, 0, 'Age')



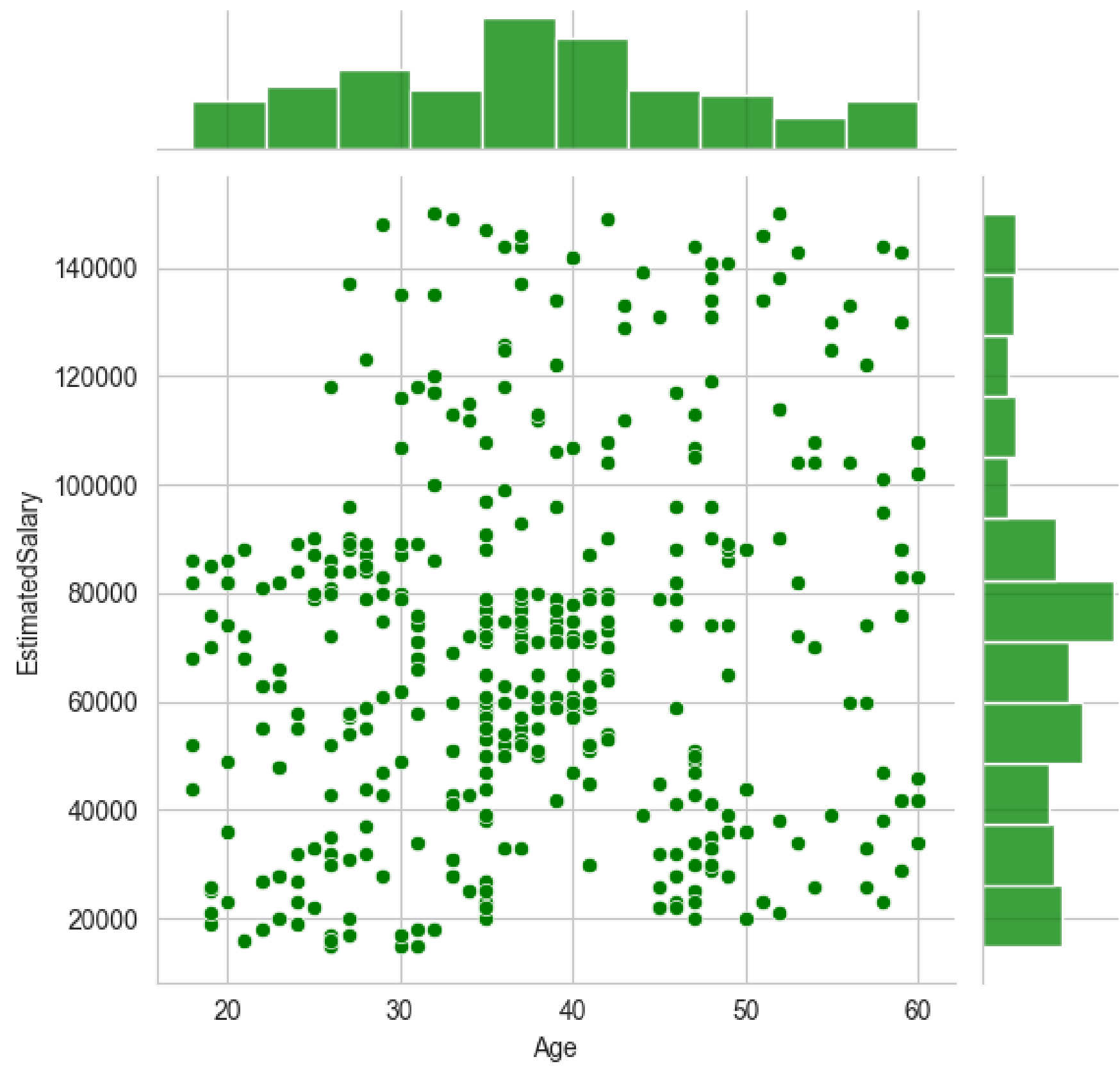
```
In [13]: sns.jointplot(x='Age', y='EstimatedSalary', data = data)
```

Out[13]: <seaborn.axisgrid.JointGrid at 0x262f75b5490>



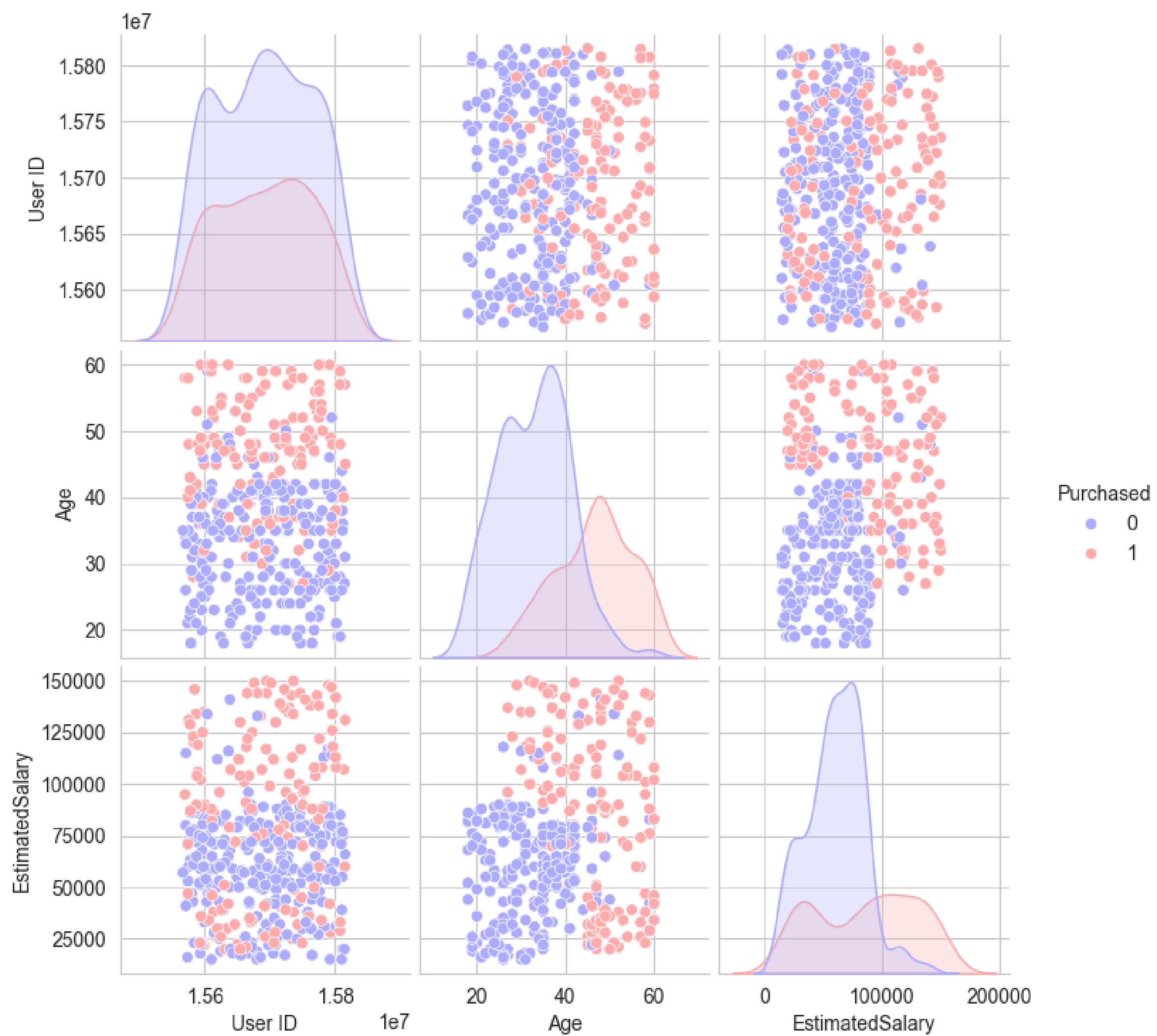
```
In [14]: sns.jointplot(x='Age',y='EstimatedSalary',data= data,color='green')
```

Out[14]: <seaborn.axisgrid.JointGrid at 0x262f748c770>



```
In [15]: sns.pairplot(data,hue='Purchased',palette='bwr')
```

Out[15]: <seaborn.axisgrid.PairGrid at 0x262f758a4e0>



```
In [16]: from sklearn.model_selection import train_test_split
```

```
In [17]: X = data[['Age', 'EstimatedSalary']]
y = data['Purchased']
```

```
In [19]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=0)
```

```
In [20]: X_train.shape
```

```
Out[20]: (300, 2)
```

```
In [21]: X_test.shape
```

```
Out[21]: (100, 2)
```

```
In [22]: y_train.shape
```

```
Out[22]: (300,)
```

```
In [23]: y_test.shape
```

```
Out[23]: (100,)
```

```
In [24]: from sklearn.linear_model import LogisticRegression
```

```
In [25]: logmodel = LogisticRegression()
logmodel.fit(X_train, y_train)
```

```
Out[25]: LogisticRegression
```

```
LogisticRegression()
```

```
In [26]: predictions = logmodel.predict(X_test)
```

```
In [27]: from sklearn.metrics import classification_report
```

```
In [28]: print(classification_report(y_test, predictions))
```

	precision	recall	f1-score	support
0	0.89	0.96	0.92	68
1	0.89	0.75	0.81	32
accuracy			0.89	100
macro avg	0.89	0.85	0.87	100
weighted avg	0.89	0.89	0.89	100

In [29]: `from sklearn.metrics import confusion_matrix`

```
cm = confusion_matrix(y_test, predictions)
print(cm)
```

```
[[65  3]
 [ 8 24]]
```

In []: