

assignment5

March 4, 2024

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

C:\Users\aaakas\AppData\Local\Temp\ipykernel_13864\555797462.py:1:

DeprecationWarning:

Pyarrow will become a required dependency of pandas in the next major release of pandas (pandas 3.0),

(to allow more performant data types, such as the Arrow string type, and better interoperability with other libraries)

but was not found to be installed on your system.

If this would cause problems for you,

please provide us feedback at <https://github.com/pandas-dev/pandas/issues/54466>

```
import pandas as pd
```

```
[2]: network = pd.read_csv("assignment5-dataset.csv")
```

```
[5]: network.head()
```

```
[5]:
```

	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

```
[3]: network.replace("?", np.nan, inplace=True)
```

```
[4]: network.isnull().sum()
```

```
[4]: User ID          0
Gender            0
Age              0
EstimatedSalary  0
Purchased        0
dtype: int64
```

```
[10]: network["Gender"].replace("Male", 0, inplace=True)
network["Gender"].replace("Female", 1, inplace=True)
```

C:\Users\aaakas\AppData\Local\Temp\ipykernel_13864\632014200.py:2: FutureWarning: Downcasting behavior in `replace` is deprecated and will be removed in a future version. To retain the old behavior, explicitly call `result.infer_objects(copy=False)`. To opt-in to the future behavior, set `pd.set_option('future.no_silent_downcasting', True)`
network["Gender"].replace("Female", 1, inplace=True)

```
[11]: network.head()
```

```
[11]:
```

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

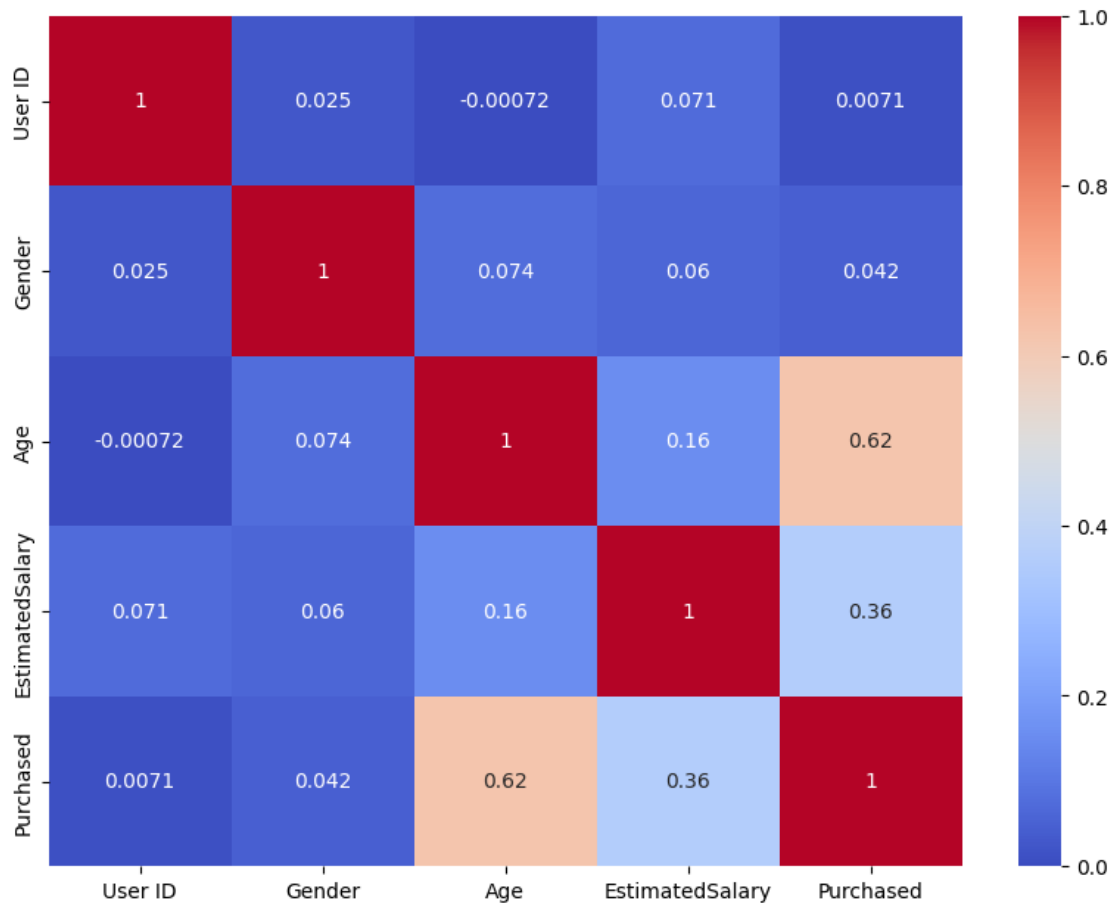
```
[13]: correlation = network.corr()
correlation
```

```
[13]:
```

	User ID	Gender	Age	EstimatedSalary	Purchased
User ID	1.000000	0.025249	-0.000721	0.071097	0.007120
Gender	0.025249	1.000000	0.073741	0.060435	0.042469
Age	-0.000721	0.073741	1.000000	0.155238	0.622454
EstimatedSalary	0.071097	0.060435	0.155238	1.000000	0.362083
Purchased	0.007120	0.042469	0.622454	0.362083	1.000000

```
[17]: plt.figure(figsize=(10, 7.5))
sns.heatmap(correlation, annot=True, cmap="coolwarm")
```

```
[17]: <Axes: >
```



```
[21]: x = network.iloc[:,1:-1]
      y = network.iloc[:, -1]
```

```
[25]: from sklearn.model_selection import train_test_split
      from sklearn.linear_model import LogisticRegression
```

```
[41]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.20,
      ↪ random_state=0)
```

```
[42]: from sklearn.preprocessing import StandardScaler
```

```
[43]: sc = StandardScaler()
      x_train = sc.fit_transform(x_train)
      x_test = sc.transform(x_test)
```

```
[44]: model = LogisticRegression(random_state=0)
```

```
[45]: model.fit(x_train, y_train)
```

```
[45]: LogisticRegression(random_state=0)
```

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

```
[50]: from sklearn.metrics import confusion_matrix, accuracy_score
```

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

```
[48]: array([[56,  2],
           [ 5, 17]], dtype=int64)
```

```
[49]: log_accuracy = accuracy_score(y_test, y_pred)
      log_accuracy
```

```
[49]: 0.9125
```