naive_bayes

October 18, 2024

0.1 1. Mengimpor Pustaka yang Diperlukan

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

0.2 2. Mengimpor Dataset

```
[2]: dataset = pd.read_csv('Social_Network_Ads.csv')
X = dataset.iloc[:, [2, 3]].values
y = dataset.iloc[:, -1].values
```

[3]: print(dataset.head())

	User ID	Gender	Age	${\tt EstimatedSalary}$	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

[4]: print(dataset.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	int.64

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

None

[5]: # periksa nilai yang hilang di setiap kolom print(dataset.isnull().sum())

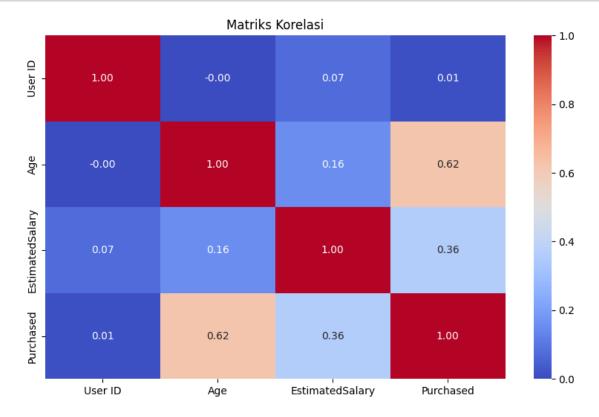
User ID 0
Gender 0
Age 0
EstimatedSalary 0
Purchased 0
dtype: int64

31

```
[6]: # Memilih hanya kolom numerik
numerical_data = dataset.select_dtypes(include=['int64', 'float64'])

# Menghitung korelasi
correlation_matrix = numerical_data.corr()

# Visualisasi korelasi
plt.figure(figsize=(10, 6))
sns.heatmap(correlation_matrix, annot=True, fmt=".2f", cmap='coolwarm')
plt.title('Matriks Korelasi')
plt.show()
```



0.3 3. Pembagian Dataset

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```
[7]: # Membagi dataset menjadi data pelatihan dan data pengujian
    from sklearn.model_selection import train_test_split
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.25,__
     →random_state = 0)
[8]: # Menampilkan nilai X_train dan y_train
    print("\nFitur Pelatihan (X_train):")
    print(X_train)
    print("\nTarget Pelatihan (y_train):")
    print(y_train)
    Fitur Pelatihan (X_train):
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[9]: # Menampilkan nilai X_test dan y_test
   print("\nFitur Pengujian (X test):")
   print(X test)
   print("\nTarget Pengujian (y_test):")
   print(y_test)
  Fitur Pengujian (X_test):
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0.4 4. Standardisasi Fitur

```
[10]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

0.5 5. Pelatihan model Naive Bayes pada data pelatihan

```
[11]: from sklearn.naive_bayes import GaussianNB
    classifier = GaussianNB()
    classifier.fit(X_train, y_train)
```

[11]: GaussianNB()

0.6 6. Prediksi hasil data pengujian

```
[12]: y_pred = classifier.predict(X_test)
```

0.7 7. Evaluasi Model

```
[13]: # Menghitung dan menampilkan confusion matrix dan akurasi
from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
print("\nConfusion Matrix:")
print(cm)
print("\nAkurasi:", accuracy_score(y_test, y_pred))
```

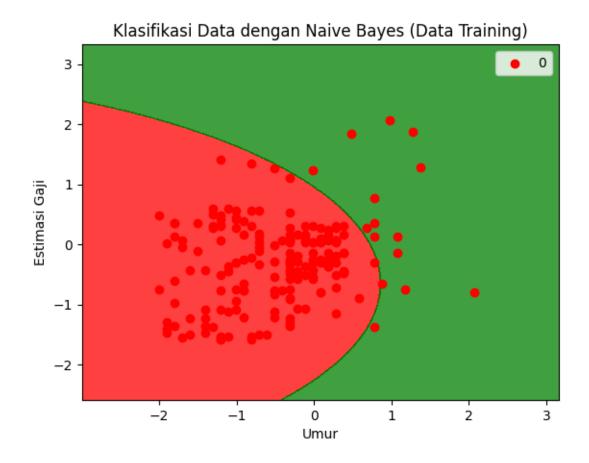
Confusion Matrix: [[65 3] [7 25]]

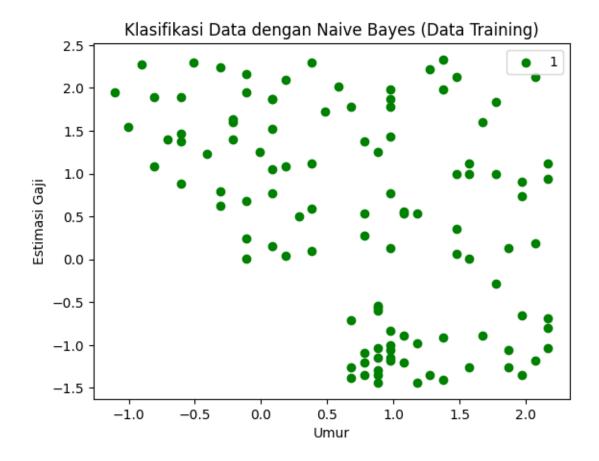
Akurasi: 0.9

0.8 8. Visualisasi hasil pelatihan

```
[16]: from matplotlib.colors import ListedColormap
      X_set, y_set = X_train, y_train
      X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, __
       0].max() + 1, step = 0.01),
                           np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:,__
       41].max() + 1, step = 0.01))
      plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).
       ⇔reshape(X1.shape),
                   alpha = 0.75, cmap = ListedColormap(('red', 'green')))
      plt.xlim(X1.min(), X1.max())
      plt.ylim(X2.min(), X2.max())
      for i, j in enumerate(np.unique(y_set)):
          plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],
                      c = ListedColormap(('red', 'green'))(i), label = j)
          plt.title('Klasifikasi Data dengan Naive Bayes (Data Training)')
          plt.xlabel('Umur')
          plt.ylabel('Estimasi Gaji')
          plt.legend()
          plt.show()
```

C:\Users\RESTU\AppData\Local\Temp\ipykernel_26372\3165894093.py:10: UserWarning: *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points. plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],





0.9 9. Visualisasi hasil pengujian

```
[15]: from matplotlib.colors import ListedColormap
      X_set, y_set = X_test, y_test
      X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, __
       0].max() + 1, step = 0.01),
                           np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:, __
       41].max() + 1, step = 0.01))
      plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).
       ⇒reshape(X1.shape),
                   alpha = 0.75, cmap = ListedColormap(('red', 'green')))
      plt.xlim(X1.min(), X1.max())
      plt.ylim(X2.min(), X2.max())
      for i, j in enumerate(np.unique(y_set)):
          plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],
                      c = ListedColormap(('red', 'green'))(i), label = j)
          plt.title('Klasifikasi Data dengan Naive Bayes (Data Training)')
          plt.xlabel('Umur')
          plt.ylabel('Estimasi Gaji')
```

plt.legend()
plt.show()

C:\Users\RESTU\AppData\Local\Temp\ipykernel_26372\1992769346.py:10: UserWarning:
c argument looks like a single numeric RGB or RGBA sequence, which should be
avoided as value-mapping will have precedence in case its length matches with
x & *y*. Please use the *color* keyword-argument or provide a 2D array with a
single row if you intend to specify the same RGB or RGBA value for all points.
plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],

Klasifikasi Data dengan Naive Bayes (Data Training)

