

naive__bayes__pet

October 18, 2024

0.1 Mengimpor pustaka yang diperlukan

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

0.2 Mengimpor dataset

```
[2]: dataset = pd.read_csv('pet_adoption_data.csv')
```

```
[3]: print(dataset.head(10))
```

	PetID	PetType	Breed	AgeMonths	Color	Size	WeightKg	\
0	500	Bird	Parakeet	131	Orange	Large	5.039768	
1	501	Rabbit	Rabbit	73	White	Large	16.086727	
2	502	Dog	Golden Retriever	136	Orange	Medium	2.076286	
3	503	Bird	Parakeet	97	White	Small	3.339423	
4	504	Rabbit	Rabbit	123	Gray	Large	20.498100	
5	505	Dog	Labrador	70	Brown	Large	20.986261	
6	506	Bird	Parakeet	169	Brown	Small	10.902613	
7	507	Cat	Siamese	13	Orange	Large	7.252683	
8	508	Bird	Parakeet	49	Brown	Medium	24.597598	
9	509	Bird	Parakeet	60	Gray	Large	7.295994	

	Vaccinated	HealthCondition	TimeInShelterDays	AdoptionFee	PreviousOwner	\
0	1	0	27	140	0	
1	0	0	8	235	0	
2	0	0	85	385	0	
3	0	0	61	217	1	
4	0	0	28	14	1	
5	0	0	87	301	1	
6	1	0	70	440	1	
7	1	0	3	137	0	
8	1	1	69	405	0	
9	0	0	73	231	1	

	AdoptionLikelihood
0	0

1	0
2	0
3	0
4	0
5	0
6	0
7	1
8	0
9	0

```
[4]: print(dataset.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2007 entries, 0 to 2006
Data columns (total 13 columns):
#   Column                Non-Null Count  Dtype
---  -
0   PetID                 2007 non-null   int64
1   PetType               2007 non-null   object
2   Breed                 2007 non-null   object
3   AgeMonths             2007 non-null   int64
4   Color                 2007 non-null   object
5   Size                  2007 non-null   object
6   WeightKg              2007 non-null   float64
7   Vaccinated            2007 non-null   int64
8   HealthCondition        2007 non-null   int64
9   TimeInShelterDays     2007 non-null   int64
10  AdoptionFee           2007 non-null   int64
11  PreviousOwner          2007 non-null   int64
12  AdoptionLikelihood     2007 non-null   int64
dtypes: float64(1), int64(8), object(4)
memory usage: 204.0+ KB
None
```

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

PetID	0
PetType	0
Breed	0
AgeMonths	0
Color	0
Size	0
WeightKg	0
Vaccinated	0
HealthCondition	0
TimeInShelterDays	0
AdoptionFee	0
PreviousOwner	0

```
AdoptionLikelihood    0
dtype: int64
```

0.3 Encoding data kategori (Atribut)

```
[6]: from sklearn.preprocessing import LabelEncoder

categorical_cols = dataset.select_dtypes(include=['object']).columns
label_encoders = {}

for col in categorical_cols:
    label_encoders[col] = LabelEncoder()
    dataset[col] = label_encoders[col].fit_transform(dataset[col])
```

```
[7]: # definisikan kolom kategorikal dan numerik
# kolom_kategorikal = ['PetType', 'Breed', 'Color', 'Size']
# kolom_numerik = ['AgeMonths', 'WeightKg', 'TimeInShelterDays', 'AdoptionFee',
#                 ↪ 'PreviousOwner']
```

```
[8]: # from sklearn.preprocessing import LabelEncoder

# encoding kolom kategorikal
# label_encoder = LabelEncoder()
# for kolom in kolom_kategorikal:
#     dataset[kolom] = label_encoder.fit_transform(dataset[kolom])
```

```
[9]: print("Tipe data setelah encoding:")
print(dataset.dtypes)
```

```
Type data setelah encoding:
PetID                int64
PetType              int32
Breed                int32
AgeMonths            int64
Color                int32
Size                 int32
WeightKg             float64
Vaccinated            int64
HealthCondition       int64
TimeInShelterDays     int64
AdoptionFee           int64
PreviousOwner         int64
AdoptionLikelihood    int64
dtype: object
```

```
[10]: # Menampilkan dataset yang sudah diolah
dataset.head()
```

```
[10]:
```

	PetID	PetType	Breed	AgeMonths	Color	Size	WeightKg	Vaccinated	\
0	500	0	2	131	3	0	5.039768	1	
1	501	3	5	73	4	0	16.086727	0	
2	502	2	0	136	3	1	2.076286	0	
3	503	0	2	97	4	2	3.339423	0	
4	504	3	5	123	2	0	20.498100	0	

	HealthCondition	TimeInShelterDays	AdoptionFee	PreviousOwner	\
0	0	27	140	0	
1	0	8	235	0	
2	0	85	385	0	
3	0	61	217	1	
4	0	28	14	1	

	AdoptionLikelihood
0	0
1	0
2	0
3	0
4	0

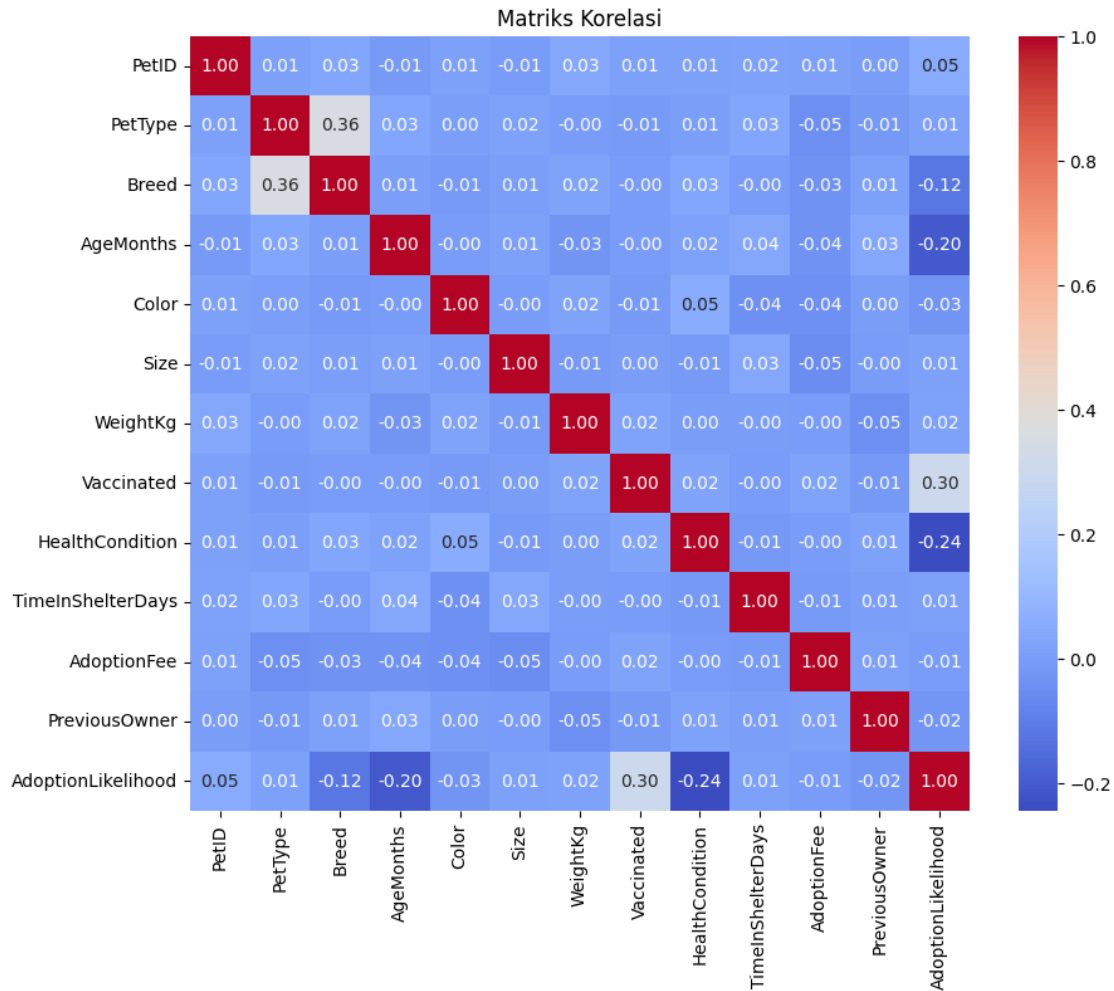
```
[11]: # Memilih fitur dan label
X = dataset.iloc[:, [3, 9]].values
y = dataset.iloc[:, -1].values
```

```
[12]: import seaborn as sns

numerical_data = dataset.select_dtypes(include=['number'])

# Menghitung matriks korelasi
correlation_matrix = dataset.corr()

# Visualisasi matriks korelasi
plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f')
plt.title('Matriks Korelasi')
plt.show()
```



0.4 Pembagian dataset

```
[13]: # 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 = 0.2,
random_state = 42)
```

```
[14]: # 7. Memeriksa bentuk dan tipe data setelah pembagian
print("\nBentuk X_train:", X_train.shape)
print("Tipe data X_train:", X_train.dtype)
```

Bentuk X_train: (1605, 2)

Tipe data X_train: int64

```
[15]: # Menampilkan hasil pembagian
print("Ukuran X_train:", X_train.shape)
print("Ukuran X_test:", X_test.shape)
print("Ukuran y_train:", y_train.shape)
print("Ukuran y_test:", y_test.shape)
```

```
Ukuran X_train: (1605, 2)
Ukuran X_test: (402, 2)
Ukuran y_train: (1605,)
Ukuran y_test: (402,)
```

```
[16]: # 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):

```
[[ 68  74]
 [172  44]
 [169  43]
 ...
 [ 19  31]
 [155  87]
 [ 34  83]]
```

Target Pelatihan (y_train):

```
[1 0 0 ... 1 0 0]
```

```
[17]: # 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):

```
[[116  71]
 [165  52]
 [106  22]
 [ 31  52]
 [125  39]
 [161  64]
 [  3  42]
 [ 56   2]
 [117  18]
 [  5  32]
 [151  19]]
```

[172 31]
[59 20]
[120 1]
[3 86]
[98 2]
[152 14]
[65 6]
[84 9]
[82 8]
[68 57]
[141 71]
[139 30]
[94 81]
[139 55]
[170 49]
[114 13]
[129 15]
[41 35]
[74 77]
[44 24]
[55 19]
[35 67]
[92 37]
[159 13]
[137 60]
[69 37]
[115 16]
[170 11]
[65 28]
[166 4]
[34 66]
[35 10]
[35 46]
[33 89]
[153 23]
[46 60]
[121 33]
[155 52]
[100 29]
[95 85]
[40 53]
[21 48]
[148 47]
[4 78]
[165 34]
[16 15]
[148 48]
[65 29]

[142 13]
[149 31]
[149 81]
[44 59]
[128 53]
[8 63]
[144 54]
[110 2]
[175 46]
[66 26]
[149 81]
[56 6]
[137 19]
[51 6]
[107 49]
[86 34]
[173 15]
[134 58]
[29 7]
[46 75]
[169 66]
[22 8]
[126 20]
[89 51]
[113 46]
[77 44]
[161 44]
[25 83]
[45 12]
[96 27]
[51 9]
[64 65]
[141 21]
[28 35]
[162 65]
[120 59]
[118 40]
[28 7]
[74 56]
[34 55]
[55 36]
[55 20]
[54 6]
[163 47]
[112 51]
[6 23]
[41 35]
[116 72]

[125 39]
[64 66]
[172 14]
[82 8]
[137 37]
[170 84]
[86 60]
[47 78]
[92 43]
[87 10]
[9 34]
[16 13]
[174 85]
[113 70]
[3 27]
[107 76]
[177 50]
[122 35]
[109 65]
[169 4]
[129 57]
[122 45]
[125 40]
[147 76]
[42 3]
[42 21]
[179 3]
[106 15]
[49 58]
[70 74]
[162 78]
[19 52]
[156 11]
[41 19]
[149 62]
[129 25]
[65 88]
[24 56]
[78 77]
[146 66]
[78 26]
[73 57]
[109 83]
[162 18]
[70 8]
[171 23]
[40 5]
[30 20]

[33 12]
[154 68]
[87 31]
[98 33]
[34 89]
[21 79]
[136 73]
[101 57]
[88 14]
[2 5]
[101 70]
[79 82]
[159 7]
[73 29]
[149 38]
[163 45]
[102 28]
[143 54]
[137 17]
[129 71]
[136 86]
[163 64]
[167 68]
[30 38]
[87 73]
[134 10]
[86 26]
[18 40]
[22 67]
[161 5]
[149 65]
[29 42]
[143 1]
[4 74]
[74 38]
[116 30]
[173 10]
[153 66]
[119 51]
[120 22]
[175 60]
[124 31]
[112 18]
[108 34]
[148 85]
[172 70]
[161 84]
[140 45]

[58 34]
[131 84]
[128 77]
[137 13]
[138 66]
[41 47]
[82 21]
[149 51]
[8 49]
[75 37]
[10 8]
[17 49]
[8 89]
[81 14]
[17 42]
[86 30]
[140 72]
[121 86]
[43 6]
[124 7]
[82 1]
[1 24]
[56 63]
[46 38]
[49 64]
[82 6]
[118 32]
[116 46]
[152 67]
[132 33]
[168 7]
[179 38]
[154 57]
[158 52]
[178 35]
[149 7]
[76 34]
[139 10]
[29 61]
[72 66]
[20 73]
[41 69]
[8 41]
[2 66]
[127 61]
[46 17]
[115 37]
[13 11]

[105 76]
[113 22]
[115 36]
[173 52]
[20 66]
[109 74]
[30 51]
[94 2]
[108 82]
[27 53]
[158 63]
[37 72]
[52 81]
[85 87]
[124 68]
[116 15]
[165 83]
[26 24]
[19 39]
[98 20]
[89 17]
[133 82]
[3 56]
[98 61]
[175 42]
[82 74]
[51 42]
[121 10]
[84 57]
[145 57]
[47 7]
[66 14]
[160 62]
[70 17]
[28 63]
[17 54]
[141 89]
[130 2]
[5 3]
[127 52]
[104 41]
[36 48]
[10 74]
[69 37]
[118 54]
[158 80]
[128 47]
[58 69]

[57 62]
[144 9]
[100 24]
[172 55]
[111 36]
[115 36]
[66 67]
[96 84]
[79 80]
[27 48]
[100 82]
[14 47]
[23 1]
[127 30]
[98 5]
[38 52]
[54 8]
[107 73]
[54 15]
[47 27]
[79 85]
[179 72]
[81 40]
[90 51]
[126 30]
[175 3]
[55 48]
[4 52]
[159 18]
[126 49]
[57 2]
[12 6]
[57 2]
[163 68]
[118 43]
[77 59]
[54 54]
[176 61]
[12 81]
[16 54]
[24 61]
[4 5]
[137 22]
[17 47]
[108 47]
[69 31]
[141 83]
[178 71]

[176 21]
[142 15]
[30 19]
[59 23]
[28 10]
[84 66]
[66 2]
[59 9]
[156 60]
[30 41]
[11 72]
[4 87]
[14 7]
[13 17]
[41 59]
[102 53]
[109 51]
[178 79]
[127 81]
[153 73]
[169 44]
[178 10]
[49 2]
[88 61]
[8 41]
[32 63]
[141 49]
[87 16]
[163 9]
[79 67]
[150 88]
[133 6]
[165 3]
[153 87]
[63 87]
[176 8]
[120 64]
[143 30]
[143 48]
[136 72]
[144 87]
[104 68]
[3 70]
[19 25]
[119 48]
[170 42]
[87 87]
[126 66]

```
[ 59 17]
[ 61 69]
[ 20 32]
[  8 17]
[163 43]
[175 17]
[ 49 38]]
```

Target Pengujian (y_test):

```
[0 0 0 0 0 1 1 0 0 0 0 0 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 1 1 0 0 0 0 1 1
 0 0 1 0 0 1 0 0 1 0 0 0 0 0 1 1 1 1 0 1 1 1 0 0 0 0 0 1 0 0 0 0 0 1 0 1 0
 0 0 1 0 1 0 0 0 1 0 0 1 0 0 0 1 0 0 0 0 0 1 0 1 0 1 0 0 0 1 1 1 1 1 0 1 1
 0 1 0 0 1 0 1 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0
 1 1 0 1 0 0 0 0 0 0 0 1 1 0 1 0 1 0 0 0 0 0 0 0 0 1 0 1 0 1 0 0 0 0 0 0 1 0
 1 1 0 1 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 0 0 0 0 1 0 1 0 0 0 1
 0 0 1 0 1 0 0 0 0 0 1 1 0 0 0 0 0 0 0 0 0 1 1 1 1 0 0 0 1 1 0 0 0 1 1 0 0 1
 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 0 1 0 1 0 0 0 0 0 1 1 1 1 1 0 0 0 0 1 1 0
 1 1 1 0 1 0 0 0 0 0 1 0 0 1 1 1 1 1 0 0 1 1 1 0 0 0 0 0 0 0 1 1 0 0 1 0 0
 0 0 0 0 0 1 0 1 0 1 0 0 0 1 0 1 1 0 1 0 0 0 0 0 1 0 1 1 0 0 0 0 1 0 0 1 0
 0 1 0 0 0 0 1 0 0 0 0 1 0 0 1 0 0 1 0 0 1 0 0 0 0 1 0 1 1 1 0 0]
```

0.5 Standardisasi fitur

```
[18]: # Standarisasi kolom numerik
from sklearn.preprocessing import StandardScaler

scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

0.6 Pelatihan model Naive Bayes pada data pelatihan

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

```
[19]: GaussianNB()
```

0.7 Prediksi hasil data pengujian

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

0.8 Evaluasi model

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

```

print(cm)
# Menghitung akurasi
accuracy = accuracy_score(y_test, y_pred)
print(f'Akurasi Model Naive Bayes: {accuracy * 100:.2f}%)

```

Confusion Matrix:

```

[[260  10]
 [106  26]]

```

Akurasi Model Naive Bayes: 71.14%

0.9 Visualisasi hasil pelatihan

```

[22]: 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[:, 1].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('AgeMonths (Scaled)')
plt.ylabel('TimeInShelterDays (Scaled)')
plt.legend(title='Adoption Likelihood', loc='upper right')
plt.show()

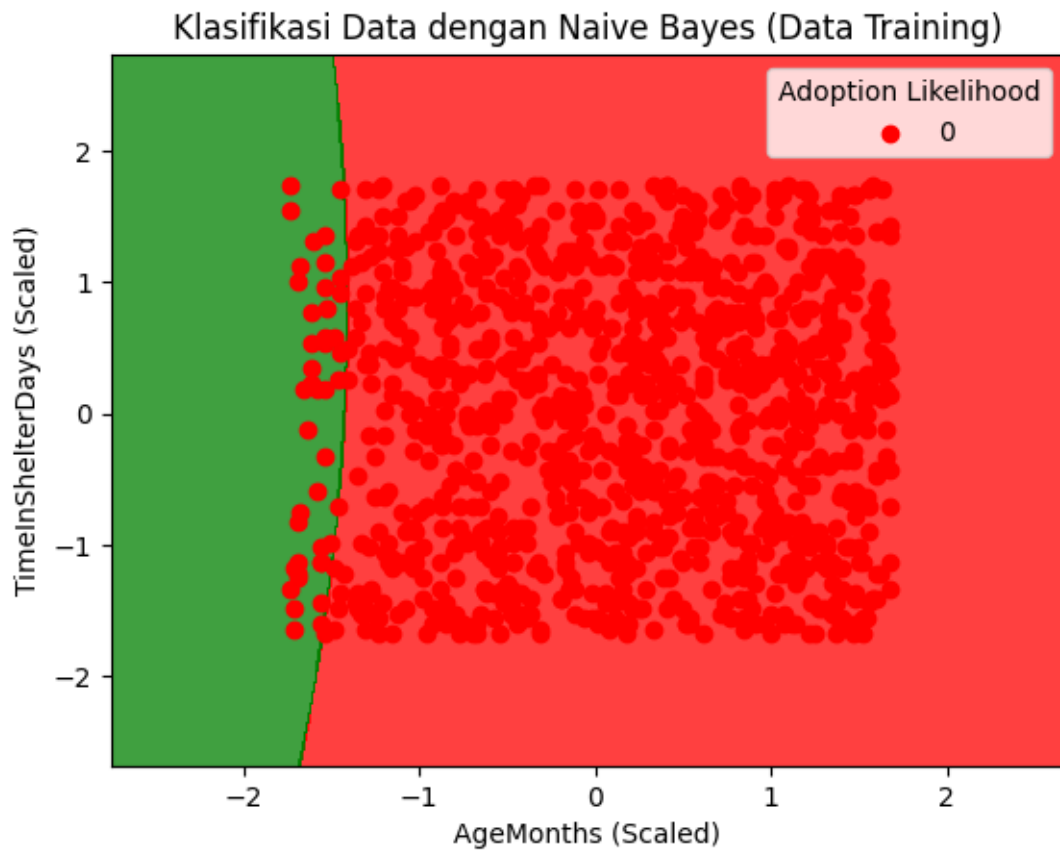
```

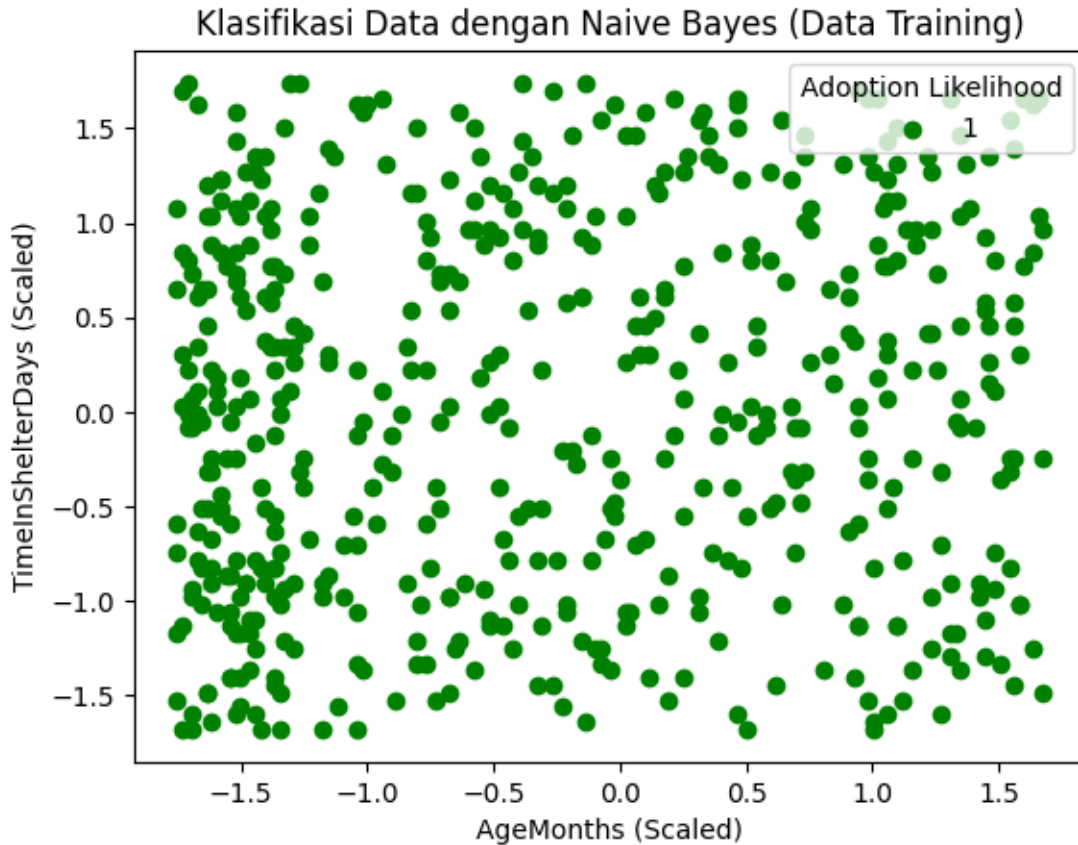
C:\Users\RESTU\AppData\Local\Temp\ipykernel_6536\1858157641.py:10: UserWarning:
 ** argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 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],

```



0.10 Visualisasi hasil pengujian

```
[23]: 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[:, 1].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('AgeMonths (Scaled)')
plt.ylabel('TimeInShelterDays (Scaled)')
```

```
plt.legend(title='Adoption Likelihood', loc='upper right')
plt.show()
```

C:\Users\RESTU\AppData\Local\Temp\ipykernel_6536\2972547322.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],

