# **LAPORAN**

# PENERAPAN NAIVE BAYES TERHADAP DATASET PARKINSONS

Mata Kuliah Kecerdasan Buatan



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# **PENDAHULUAN**

#### A. Tugas

- Penerapan Naive Bayes terhadap dataset Parkinson yang ada di "parkinsons.xlsx".
   Ketentuan sesuai PPT, 100 sampel pertama sebagai training set dan sisanya sebagai data uji/test.
- Manualisasi di excel seperti contoh di ppt yg 7 langkah itu. Terus buat model machine learningnya.
- Dibuat laporan (Penjelasan DataSet, Kode, Hasil)

## **B.** Pengertian

Naive Bayes adalah algoritma klasifikasi yang didasarkan pada teorema Bayes dengan asumsi bahwa setiap fitur dalam data bersifat independen satu sama lain (naive). Algoritma ini menghitung probabilitas terjadinya suatu kelas berdasarkan nilai fitur-fitur input, kemudian memilih kelas dengan probabilitas tertinggi sebagai hasil prediksi. Naive Bayes sering digunakan dalam aplikasi seperti klasifikasi teks (misalnya, spam email) dan analisis sentimen karena kesederhanaannya dan kemampuannya untuk bekerja dengan data besar secara efisien.

## **PEMBAHASAN**

#### A. Dataset

Dataset terdiri dari 195 data orang yand diklasifikasikan terkena penyakit Parkinson (PD) dan tidak terkena penyakit Parkinson (Healthy). Terdapat 23 kolom dengan kolom ke 23 berupa class identifikasi sehat atau tidak (PD atau Healthy). Jumlah data ialah 195, namun 100 pertama digunakan sebagai training dan 95 sisanya digunakan sebagai testing. Dataset disimpan dalam file excel dengan nama Parkinson.xlsx.

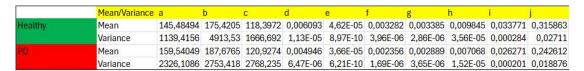
#### B. Manualisasi di Excel

#### 1. Menghitung Prior Probability

class	jumlah	prior	
Healthy		51	0,51
PD		49	0,49

## 2. Menghitung Mean dan Variance tiap Class

Semua fitur, a - v



## 3. Menghitung Likelihood

Semua fitur, a - v dengan dataset 101-195 (testing data)

Healthy:

```
0,009943125 0,0050473 0,00976 9,08E-13 1,15E-07 1,68E-16 5,26E-09 5,31E-17 0,063394 0,00543
Data-101
Data-102
                   0,010405714 0,0053414 0,005646 2,323353 641,3621 3,005202 0,79064 0,993374 5,997191 0,643357
Data-103
                   0,011617119 1,926E-10 0,004309 9,12E-10 0,000651 3,45E-11 5,33E-11 1,11E-11 0,038361 0,002308
Data-104
                   0,011701114 0,0054459 0,005605 66,59863 9076,114 106,8937 115,1152 35,63114 9,740282 1,083739
Data-105
                   0,011448304 0,0055599 0,00947 53,06732 6406,845 89,82046 87,17085 29,99957 8,976603 0,977403
Data-106
                   0,011727372 0,0055616 0,009117 68,52351 9076,114 113,5551 113,4857 37,92107 11,45148 1,246253
Data-107
                   0,011350911 0,0056128 0,008009 50,09014 6406,845 85,2911 81,44384 28,42522 9,230453 1,001531
                   0,011608209 0,0055136 0,009105 68,73774 9076,114 112,947 103,8753 37,71816 12,31194 1,343057
Data-108
                   0,011601309 0,0055051 0,009178 51,27271 6406,845 89,24927 84,28231 29,74545 8,932159 0,969419
Data-109
Data-110
                   0,004382903 0,0050778 0,006322 106,5688 13035,79 166,2092 225,8002 55,35793 23,21149 2,349486
Data-111
                    0,00309927 0,0044771 0,007631 118,7668 11500,88 195,869 236,0912 65,33898 19,19504 1,947247
Data-112
                   0,002067035 0,004636 0,00139 118,8392 11500,88 196,483
                                                                                236.1 65.49845 14.94698 1.503404
                   0,002541763 0,004594 0,002133 93,6468 13035,79 136,8895 162,1554 45,63061 19,8558 1,97052
Data-113
                   0,001887455 0,0040755 0,002556 115,8889 11500,88 200,326 222,3532 66,80865 14,2574 1,449858
Data-114
                    0.00232861 0.0043708 0.007931 112,1495 9076,114 198,7299 203,227 66,2579 17,80307 1,962804
Data-115
                                   2E-07 0,004712 90,71189 11986,06 186,4233 227,3387 62,20159
Data-116
                   0,011006915 3,995E-06 0,005115 109,8334 11500,88 170,0393 174,3902 56,69957 12,37879 1,440935
Data-117
Data-118
                   0,008930196 2,614E-06 0,006139 117,2962 11500,88 183,3044 212,5114 61,08254 14,64065 1,680397
                   0,007371181 3,937E-06 0,006576 107,936 11500,88 138,4767 163,7675 46,2346 10,90725 1,272503
Data-119
                   0,001243693 0,0040386 0,008171 98,57479 9076,114 120,2637 108,6427 40,09152 11,06877 1,228828
Data-120
                   0,010480963 4,609E-07 0,00744 118,4185 13216,59 182,1188 233,6262 60,7303 17,01815 2,052421
Data-121
Data-122
                   0,007680608 0,0048412 0,006847 107,6524 11500,88 167,5531 185,3708 55,92601 10,03481 1,168378
                   0,011545883 0,0052519 0,006759 114,2011 13216,59 185,7015 221,9212 61,92101 19,11488 1,81517
Data-123
Data-124
                   0,006579785 0,0054238 0,006161 93,45403 13216,59 134,4802 190,6697 44,89387 20,69115 2,121595
```

PD:

```
d
                                            е
                                                                       h
                                                              g
Data-101 0,006461 0,005125 0,007554 2,95E-25 5,55E-14 4,12E-45 1,87E-07 1,36E-45 0,000666 4,23E-05
         0,00672 0,005911 0,005234 0,028112 59,08716 0,000525 1,075588 0,000173 1,247386 0,132495
Data-102
Data-103 0,007569 2,15E-15 0,004411 8,63E-20 2,74E-08 5,99E-32 4,47E-09 1,97E-32
                                                                                  0,0003 1,06E-05
Data-104 0,00812 0,006234 0,00521 98,02688 12821,62 185,3117 151,6228 61,76415 16,12688 1,783984
Data-105 0,008217 0,006633 0,007502 74,0501 9051,113 144,4162 128,1069 48,28727 14,9238 1,617648
Data-106 0.008101 0.00664 0.007365 101,3857 12821,62 201,0549 150,3498 67,16886 18,70267 2,025453
Data-107 0,008236 0,006851 0,006879 68,76772 9051,113 133,6393 122,8042 44,54395 15,32709 1,655917
Data-108 0,008168 0,006463 0,007361 101,758 12821,62 199,6292 142,6046 66,69394 19,9277 2,160945
Data-109 0,008171 0,006433 0,00739 70,86346 9051,113 143,0526 125,455 47,68104 14,85286 1,604919
           0,0065 0,007006 0,005627 88,76834 15866,71 78,62755 182,0181 26,10825 26,91921 2,824186
Data-111 0.005746 0.005984 0.006357 138,6439 15459.82 178,3509 202,2983 59,61645 27,30555
                                                                                          2.80783
Data-112 0,004939 0,006266 0,00252 141,7755 15459,82 182,6489 201,7542 60,73051 23,3315 2,369495
         0,00534 0,006192 0,003233 62,04433 15866,71 37,47785 123,2774 12,5001 27,67601 2,823095
Data-114 0,004771 0,005261 0,003591 154,9964 15459,82 247,3962 208,6005 82,3262 22,49642 2,30222
Data-115 0,005167 0,005794 0,00652 156,8663 12821,62 269,4795 203,6357 89,94096 26,26705 2,818147
Data-116 0.008168 3.42E-10 0.004666 57.15557 10306.56 132.5036 183.9865 44.32452 24.12831 2.899307
Data-117 0,008269 5,72E-08 0,004916 156,447 15459,82 302,9198 190,5089 100,9598 20,02074 2,290769
Data-118 0,008051 2,78E-08 0,005522 152,5061 15459,82 306,4799 206,6165 102,1651 22,96583 2,572589
Data-119
         0,00767 5,58E-08 0,005772 155,5821 15459,82 255,7796 184,6065 85,38012 17,90268 2,062851
Data-120 0,004056 0,005194 0,00665 147,2193 12821,62 216,5768 146,4982 72,18335 18,14206 2,000368
Data-121 0,006764 1,43E-09 0,006252 148,0727 13860,85 306,8577 207,0241 102,2741 25,55528 2,868054
Data-122 0,007757 0,00662 0,005924 155,4205 15459,82 300,7667 196,0586 100,2942 16,58247 1,911801
           0,0075 0,007264 0,005875 111,7722 13860,85 129,8691 177,3813 43,29938 12,00004 1,023723
Data-123
Data-124
         0,00742 0,007479 0,005535 61,71246 13860,85 35,16722 146,9591 11,79242 27,9983 2,893407
Data-125 0,008252 0,007527 0,005588 115,3607 15866,71 133,8289 178,9577 44,6192 27,54812 2,877237
```

#### 4. Menghitung perkalian likelihood

	Healthy	PD
Data-101	1,52031E-95	0
Data-102	0,002406252	3,321E-47
Data-103	3,25785E-83	0
Data-104	6725808963	1,893E+13
Data-105	301281767,2	1,378E+12
Data-106	5,50499E+11	3,695E+14
Data-107	1582934740	3,028E+12
Data-108	9,77431E+11	1,666E+15
Data-109	1080281884	2,558E+12
Data-110	3,2146E+15	4,482E+15
Data-111	8,51888E+14	4,828E+16
Data-112	2,07954E+12	3,055E+15
Data-113	7,23545E+13	2,502E+14
Data-114	2,27125E+13	1,811E+16
Data-115	5,70921E+12	3,206E+15
Data-116	120500129,6	4,4869889
Data-117	1199818272	6,309E+10
Data-118	626485635,3	3,37E+09
Data-119	29868073,2	4,118E+09
Data-120	3554318,755	1,226E+13
Data-121	15682717340	5,157E+09
Data-122	2,07655E+11	1,31E+15
Data-123	3,09204E+15	9,204E+12

## 5. Menghitung Evidence

Evidence = (Prior Healthy \* Likelihood Healthy) + (Prior PD \* Likelihood PD)

	Evidence
Data-101	7,75357E-96
Data-102	0,001227189
Data-103	1,6615E-83
Data-104	9,28133E+12
Data-105	6,75237E+11
Data-106	1,81314E+14
Data-107	1,48438E+12
Data-108	8,17027E+14
Data-109	1,2539E+12
Data-110	3,83562E+15
Data-111	2,40907E+16
Data-112	1,498E+15
Data-113	1,59498E+14
Data-114	8,88744E+15

## 6. Menghitung Posterior Probability

Posterior = (Prior \* Likelihood) / Evidence

	Healthy	PD
Data-101	1	0
Data-102	1	1,326E-44
Data-103	1	0
Data-104	0,000369577	0,9996304
Data-105	0,000227555	0,9997724
Data-106	0,001548448	0,9984516
Data-107	0,00054386	0,9994561
Data-108	0,000610127	0,9993899
Data-109	0,000439383	0,9995606
Data-110	0,427426719	0,5725733
Data-111	0,018034485	0,9819655

## 7. Class Decision

pred Class = - jika posterior Healthy > posterior PD, maka classnya ialah Healthy,

- jika posterior PD > posterior Healthy, maka classnya ialah PD

Akurasi = perbandingan pred Class dengan class pada dataset

	Pred Class	True Class	Accuracy
Data-101	Healthy	Healthy	1
Data-102	Healthy	Healthy	1
Data-103	Healthy	Healthy	1
Data-104	PD	Healthy	0
Data-105	PD	Healthy	0
Data-106	PD	Healthy	0
Data-107	PD	Healthy	0
Data-108	PD	Healthy	0
Data-109	PD	Healthy	0
Data-110	PD	Healthy	0

## Hasil akhir:

Data-191	PD	PD	1
Data-192	PD	PD	1
Data-193	Healthy	PD	0
Data-194	Healthy	PD	0
Data-195	PD	PD	1
		Accuracy	0,494737

#### C. Model Machine Learning di Python

#### Code:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import accuracy_score
data = pd.read_excel('parkinsons.xlsx')
data['Class'] = data['Class'].fillna('Healthy')
data = data.replace({',': '.'}, regex=True)
data_train = data.iloc[:100]
data_test = data.iloc[100:]
x_train = data_train.drop(columns=['Class'])
y_train = data_train['Class']
x_test = data_test.drop(columns=['Class'])
y_test = data_test['Class']
y_train.hist()
plt.show()
model = GaussianNB()
model.fit(x_train, y_train)
y_pred = model.predict(x_test)
accuracy = accuracy_score(y_test, y_pred)
print(f'Accuracy: {accuracy} || Accuracy %: {accuracy * 100}%')
print(f'Error Rate: {1 - accuracy} || Error Rate %: {(1 - accuracy) * 100}%')
```

Hasil training:

plt.show()



**※ ← → | 中 Q 至 | 凹** (x, y) = (Healthy, 25.88)

#### Hasil Terminal:

Accuracy: 0.5368421052631579 || Accuracy %: 53.68421052631579%

Error Rate: 0.4631578947368421 || Error Rate %: 46.31578947368421%

PS D:\kuliah\sem3\AI\tugas1> python -u "d:\kuliah\sem3\AI\tugas1\tugas1.py"
Accuracy: 0.5368421052631579 || Accuracy %: 53.68421052631579%
Error Rate: 0.4631578947368421 || Error Rate %: 46.31578947368421%
PS D:\kuliah\sem3\AI\tugas1>