```
# import library
import pandas as Nafasa
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.datasets import load wine
# Load dataset and convert to DataFrame
data = load wine()
# mengubah data load_wine menjadi dataframe
df = Nafasa.DataFrame(data.data, columns=data.feature names)
df['target'] = data.target
df.head(10)
\rightarrow
         alcohol malic_acid ash alcalinity_of_ash magnesium total_phenols flavanoids nonflavanoid_phenols proant
      0
            14.23
                          1.71 2.43
                                                   15.60
                                                              127.00
                                                                                 2.80
                                                                                             3.06
                                                                                                                     0.28
      1
            13.20
                          1.78 2.14
                                                    11.20
                                                              100.00
                                                                                 2.65
                                                                                             2.76
                                                                                                                     0.26
                          2.36 2.67
      2
            13.16
                                                   18.60
                                                              101.00
                                                                                 2.80
                                                                                             3.24
                                                                                                                     0.30
      3
            14.37
                          1.95 2.50
                                                   16.80
                                                              113.00
                                                                                 3.85
                                                                                             3.49
                                                                                                                     0.24
            13.24
                          2.59 2.87
                                                   21.00
                                                                                 2.80
                                                                                             2.69
                                                                                                                     0.39
      4
                                                              118.00
                          1.76 2.45
      5
            14.20
                                                   15.20
                                                              112.00
                                                                                 3.27
                                                                                                                     0.34
                                                                                             3.39
      6
            14.39
                          1.87 2.45
                                                   14.60
                                                               96.00
                                                                                 2.50
                                                                                             2.52
                                                                                                                     0.30
            14.06
      7
                                                   17.60
                                                                                 2.60
                          2.15 2.61
                                                              121.00
                                                                                             2.51
                                                                                                                     0.31
      8
                          1.64 2.17
                                                   14.00
                                                               97.00
                                                                                 2.80
                                                                                             2.98
                                                                                                                     0.29
            14.83
      9
            13.86
                          1.35 2.27
                                                    16.00
                                                               98.00
                                                                                 2.98
                                                                                             3.15
                                                                                                                     0.22
 Langkah berikutnya:
                       Buat kode dengan df

    Lihat plot yang direkomendasikan

                                                                                        New interactive sheet
#mengetahui berapa data/class pada target
df['target'].value_counts()
\overline{\Rightarrow}
               count
      target
         1
                  71
         0
                  59
         2
                  48
     dtvne: int64
df.shape
    (178, 14)
# Mengubah format tampilan angka desimal
Nafasa.options.display.float_format = '{:,.2f}'.format
```

```
# Melihat deskripsi statistik
df['alcohol'].describe()
```



	alcohol
count	178.00
mean	13.00
std	0.81
min	11.03
25%	12.36
50%	13.05
75%	13.68
max	14.83

dtvne: float64

df.info()

```
<<class 'pandas.core.frame.DataFrame'>
RangeIndex: 178 entries, 0 to 177
Data columns (total 14 columns):
# Column No
```

#	Column	Non-Null Count	Dtype
0	alcohol	178 non-null	float64
1	malic_acid	178 non-null	float64
2	ash	178 non-null	float64
3	alcalinity_of_ash	178 non-null	float64
4	magnesium	178 non-null	float64
5	total_phenols	178 non-null	float64
6	flavanoids	178 non-null	float64
7	nonflavanoid_phenols	178 non-null	float64
8	proanthocyanins	178 non-null	float64
9	color_intensity	178 non-null	float64
10	hue	178 non-null	float64
11	od280/od315_of_diluted_wines	178 non-null	float64
12	proline	178 non-null	float64
13	target	178 non-null	int64

dtypes: float64(13), int64(1)
memory usage: 19.6 KB

```
df['target'].unique()
```

 $\rightarrow$  array([0, 1, 2])

# Menerapkan StandardScaler
scaler = StandardScaler()

# Memuat Dataset

 $from \ sklearn.model\_selection \ import \ train\_test\_split$ 

```
wine = load_wine()

# Split the data into features (X) and target (y)

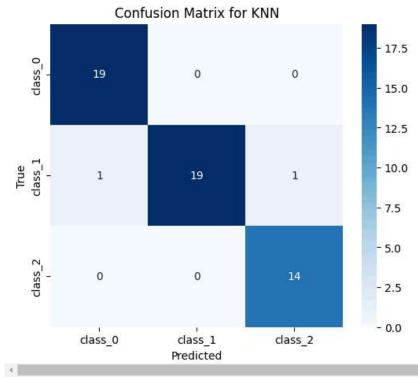
X = df.drop('target', axis=1) # Fitur (independent variables)
y = df['target'] # Target (dependent variable)

# Split the data into training and test sets (70% latih dan 30% uji)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)

from sklearn.preprocessing import StandardScaler
```

```
scaler. fit(X_train)
X_train = scaler.transform(X_train)
X_test = scaler.transform(X_test)
from sklearn.neighbors import KNeighborsClassifier
# Initialize K-Nearest Neighbor Classifier
knn = KNeighborsClassifier (n_neighbors=3)
# Train the model / Melatih model dengan data pelatihan
knn.fit(X_train, y_train)
₹
            KNeighborsClassifier
     KNeighborsClassifier(n_neighbors=3)
from sklearn.metrics import accuracy_score
# Predict on the test set
y_pred = knn.predict (X_test)
#menentukan probabilitas prediksi
knn.predict_proba(X_test)
→
      Tampilkan output tersembunyi
# Calculate accuracy
accuracy = accuracy_score(y_test, y_pred)
print(accuracy * 100)
→ 96.29629629629
from sklearn.metrics import classification_report, confusion_matrix
#Membuat confusion matrix
cm = print(confusion_matrix(y_test, y_pred))
#Membuat laporan klasifikasi
cf = print(classification_report(y_test, y_pred))
→ [[19 0 0]
      [ 1 19 1]
      [ 0 0 14]]
                   precision
                                recall f1-score
                                                   support
                0
                        0.95
                                  1.00
                                            0.97
                                                        19
                                  0.90
                        1.00
                                            0.95
                                                        21
                1
                        0.93
                                  1.00
                                            0.97
                                                        14
                                            0.96
                                                         54
         accuracy
                        0.96
                                  0.97
                                            0.96
                                                         54
        macro avg
     weighted avg
                        0.97
                                  0.96
                                            0.96
                                                         54
#Menampilkan visualisasi confusion matrix
plt.figure(figsize=(6, 5))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=data.target_names, yticklabels=data.target_names)
plt.xlabel('Predicted')
plt.ylabel('True')
plt.title('Confusion Matrix for KNN')
plt.show()
```



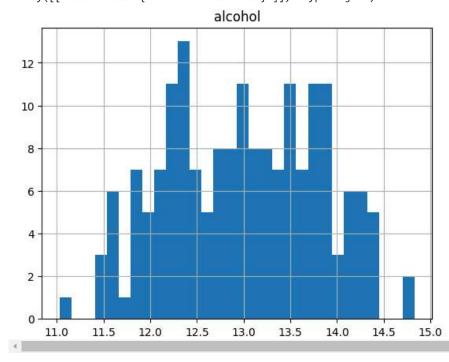


### ✓ alcohol

# @title alcohol

df.hist(column='alcohol',bins=30)

array([[<Axes: title={'center': 'alcohol'}>]], dtype=object)

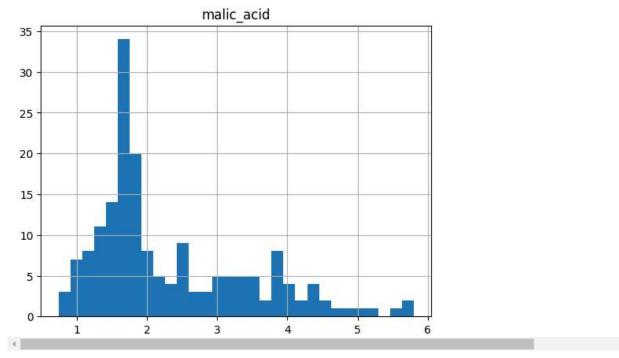


# ✓ malic\_acid

# @title malic\_acid

df.hist(column='malic\_acid',bins=30)

array([[<Axes: title={'center': 'malic\_acid'}>]], dtype=object)

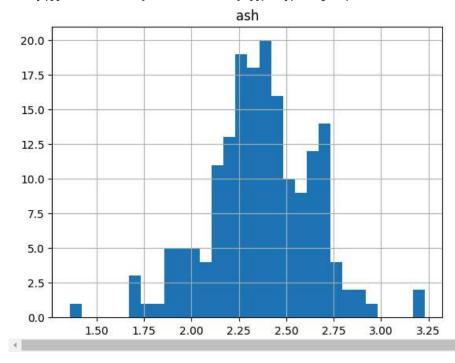


#### ✓ ash

# @title ash

df.hist(column='ash',bins=30)

array([[<Axes: title={'center': 'ash'}>]], dtype=object)

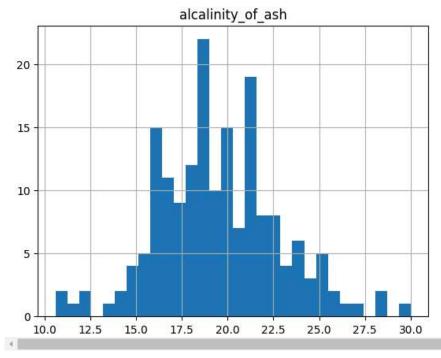


# ✓ alcalinity\_of\_ash

# @title alcalinity\_of\_ash

df.hist(column='alcalinity\_of\_ash',bins=30)

array([[<Axes: title={'center': 'alcalinity\_of\_ash'}>]], dtype=object)

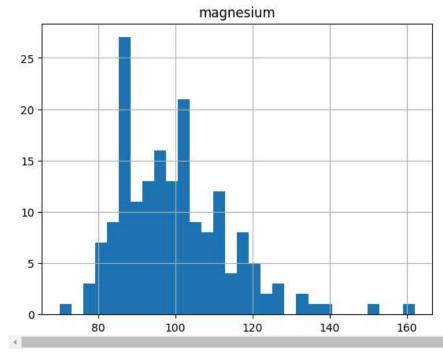


### → magnesium

# @title magnesium

df.hist(column='magnesium',bins=30)

array([[<Axes: title={'center': 'magnesium'}>]], dtype=object)

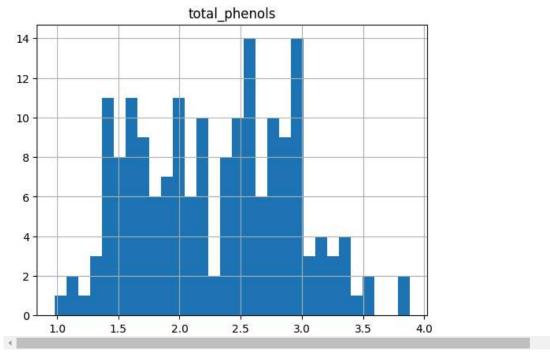


## 

# @title total\_phenols

df.hist(column='total\_phenols',bins=30)

array([[<Axes: title={'center': 'total\_phenols'}>]], dtype=object)

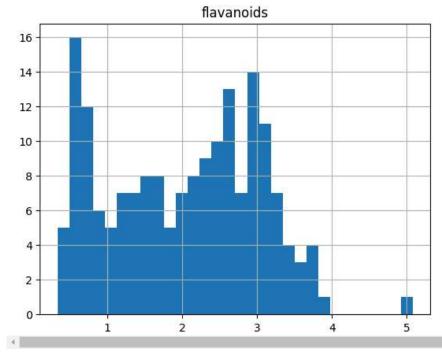


### ✓ flavanoids

# @title flavanoids

df.hist(column='flavanoids',bins=30)

array([[<Axes: title={'center': 'flavanoids'}>]], dtype=object)

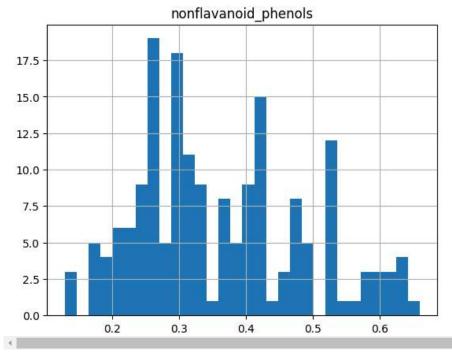


# → nonflavanoid\_phenols

# @title nonflavanoid\_phenols

df.hist(column='nonflavanoid\_phenols',bins=30)

array([[<Axes: title={'center': 'nonflavanoid\_phenols'}>]], dtype=object)

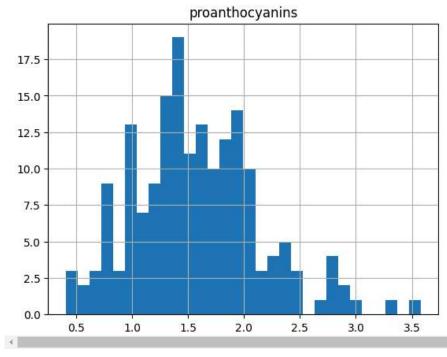


## → proanthocyanins

# @title proanthocyanins

df.hist(column='proanthocyanins',bins=30)

array([[<Axes: title={'center': 'proanthocyanins'}>]], dtype=object)

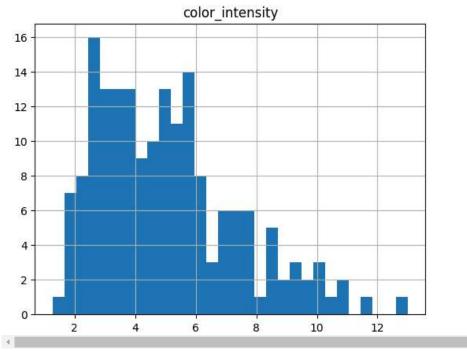


## ✓ color\_intensity

# @title color\_intensity

df.hist(column='color\_intensity',bins=30)

array([[<Axes: title={'center': 'color\_intensity'}>]], dtype=object)

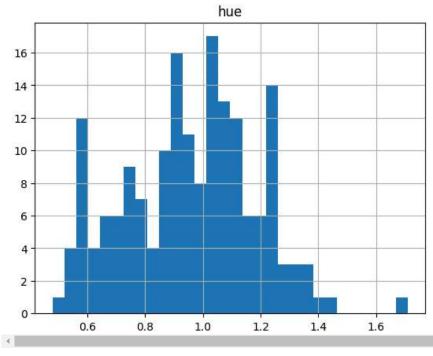


#### ✓ hue

# @title hue

df.hist(column='hue',bins=30)

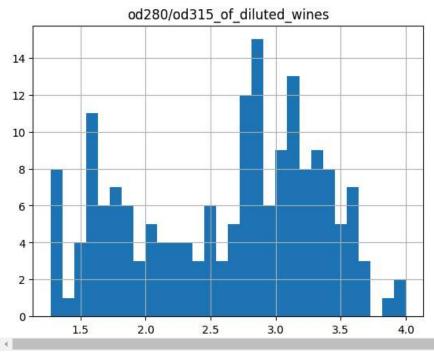
array([[<Axes: title={'center': 'hue'}>]], dtype=object)



# od280/od315\_of\_diluted\_wines

# @title od280/od315\_of\_diluted\_wines

df.hist(column='od280/od315\_of\_diluted\_wines',bins=30)



## ✓ proline

# @title proline

df.hist(column='proline',bins=30)

array([[<Axes: title={'center': 'proline'}>]], dtype=object)

