

Uine Recognition
Classification: A
Comparative Study Using
Logistic Regression, KNN,
and SIM



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Beaujolais Nouveau









# Hout Me

I am a 6th-semester Chemistry student passionate about learning new things, particularly in Data Science. I aim to combine my scientific background with modern technologies, aspiring to become a researcher and contribute to groundbreaking discoveries that solve real-world challenges.









# About The Project



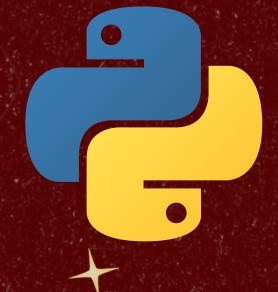
"Create a simple machine learning (classification)

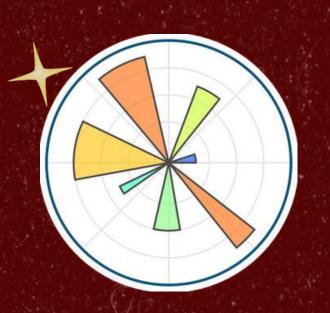
program using a datasets that has already been provided

by scikit-learn"



Yools











#### +Import Library & Loaf Dataset

```
# Load dataset
wine = load_wine()
df = pd.DataFrame(wine.data, columns=wine.feature_names)
df['target'] = wine.target

# Simpan ke CSV
df.to_csv('wine_dataset.csv', index=False)
print("Dataset berhasil disimpan sebagai wine_dataset.csv")
```



```
url = "https://github.com/ErikaYosepinS/Wine-Dataset/raw/refs/heads/main/wine_dataset.csv"
```

```
# Load dataset
data = pd.read_csv(url)

# Tampilkan 5 baris pertama
print(data.head())

# Cek informasi dataset
print("\nInfo Dataset:\n")
print(data.info())
```

### Data Processing



100		AND THE RESERVE OF		man a series of the series		Acres of the Oracles	was defeated below the		
	alcohol	malic_acid	ash al	calinity_	of_ash m	nagnesium	total_phen	ols	1
0	14.23	1.71	2.43		15.6	127.0	2	.80	
1	13.20	1.78	2.14	11.		100.0		.65	
2	13.16	2.36	2.67		18.6	101.0	2	.80	
3 4	14.37	1.95	2.50		16.8	113.0	3	.85	
4	13.24	2.59	2.87	87 21.0 1		118.0	2.80		
	flavanoi	ds nonflava	noid_pher	nols proa	nthocyani	ns color	intensity	hue	1
0	3.	96	(	3.28	2.	29	5.64	1.04	
1	2.76		0.26		1.	28	4.38	1.05	
3 4	3.	24	0		2.	81	5.68	1.03	
3	3.	49	6	3.24	2.	18	7.80	0.86	
4	2.	69	6	3.39	1.	82	4.32	1.04	
	od280/od	315_of_dilut	ed_wines	proline	target				
0			3.92	1065.0	0				
1			3.40	1050.0	0				
0 1 2 3 4		3.17		0 1050.0 0 7 1185.0 0 5 1480.0 0 3 735.0 0					
3			3.45	1480.0	0				
4			2.93	735.0	0				

Info Dataset:

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 178 entries, 0 to 177
Data columns (total 14 columns):

# Column

Non-Null Count Dtype

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 178 entries, 0 to 177
Data columns (total 14 columns):

#	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
dtom	oc. float64/43\ int64/4\		

dtypes: float64(13), int64(1)

memory usage: 19.6 KB

None

# Data Preparation & Training Data

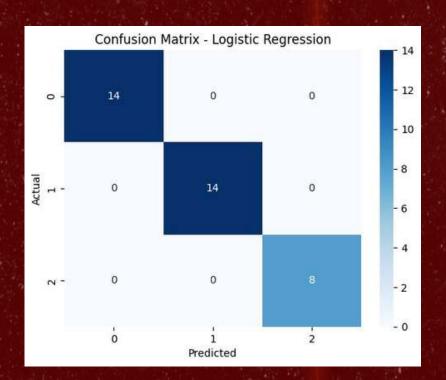
```
from sklearn.model selection import train test split
from sklearn.preprocessing import StandardScaler
# Pisahkan fitur (X) dan target (y)
X = data.drop(columns=['target']) # Semua kecuali target
y = data['target'] # Kolom target
# Split dataset menjadi training (80%) dan testing (20%)
X train, X test, y train, y test = train_test_split(X, y, test_size=0.2, random_state=42)
# Standardisasi fitur
                                                from sklearn.linear model import LogisticRegression
scaler = StandardScaler()
                                                 from sklearn.neighbors import KNeighborsClassifier
X_train = scaler.fit_transform(X_train)
                                                 from sklearn.svm import SVC
X test = scaler.transform(X test)
                                                 # Model Logistic Regression
                                                 log_reg = LogisticRegression(max_iter=1000)
                                                 log reg.fit(X train, y train)
                                                 # Model KNN (dengan k=5)
                                                 knn = KNeighborsClassifier(n neighbors=5)
                                                 knn.fit(X_train, y_train)
                                                 # Model SVM
                                                 svm = SVC(kernel='linear') # Kernel linear agar hasilnya bisa dibandingkan dengan Logistic Regression
```

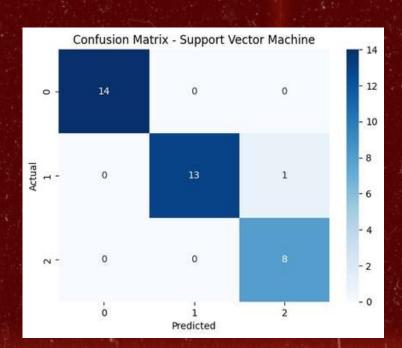
svm.fit(X\_train, y\_train)

print("Ketiga model berhasil dilatih!")

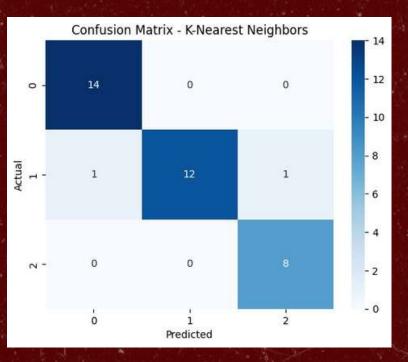
#### Evaluate Models

```
from sklearn.metrics import classification report, confusion matrix, accuracy score
import seaborn as sns
import matplotlib.pyplot as plt
# Prediksi
y pred log reg = log reg.predict(X test)
y pred knn = knn.predict(X test)
y_pred_svm = svm.predict(X_test)
# Evaluasi fungsi
def evaluate model(model name, y test, y pred):
    print(f"\n=== {model name} ===")
    print("\nClassification Report:\n", classification_report(y_test, y_pred))
    print("\nConfusion Matrix:")
    cm = confusion_matrix(y_test, y_pred)
    sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')
    plt.title(f'Confusion Matrix - {model_name}')
    plt.ylabel('Actual')
    plt.xlabel('Predicted')
    plt.show()
    accuracy = accuracy_score(y_test, y_pred)
    print(f"Model Accuracy: {accuracy:.2f}")
# Evaluasi setiap model
evaluate_model("Logistic Regression", y_test, y_pred_log_reg)
evaluate_model("K-Nearest Neighbors", y_test, y_pred_knn)
```

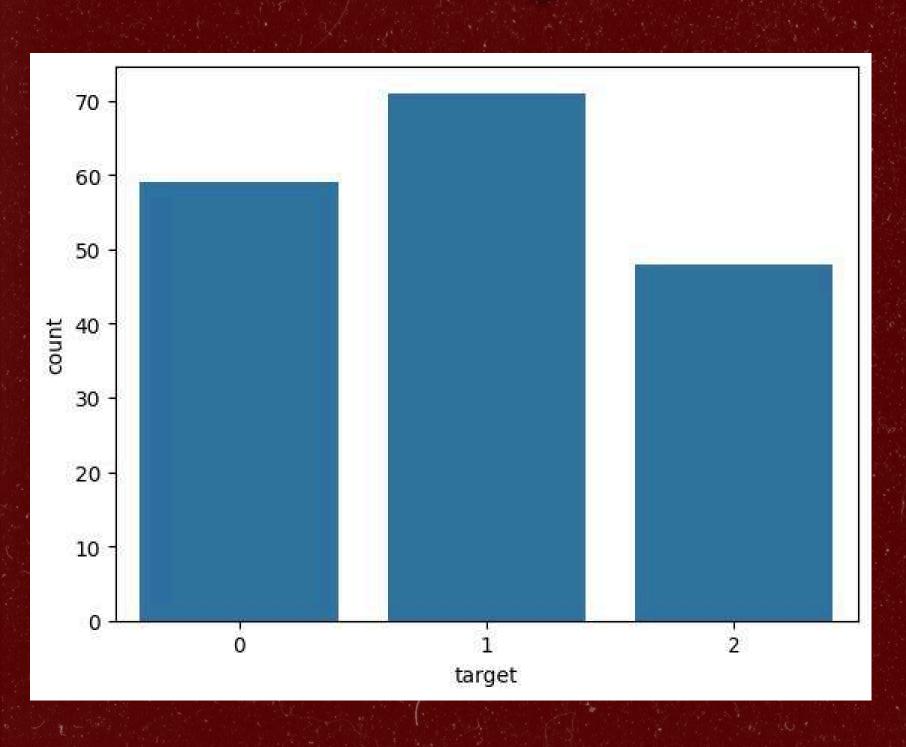


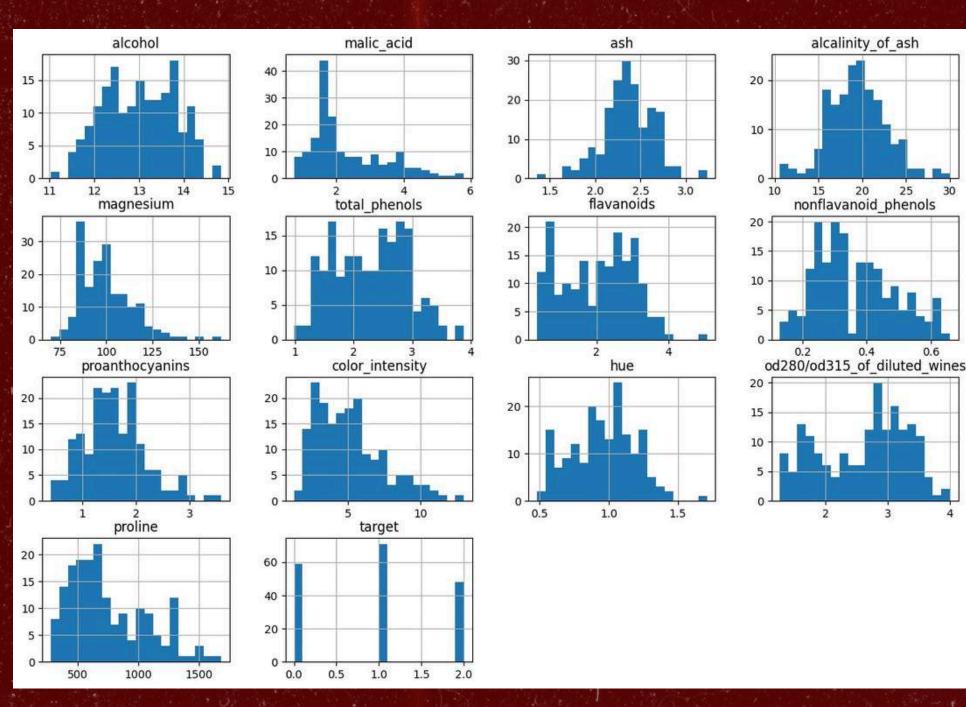






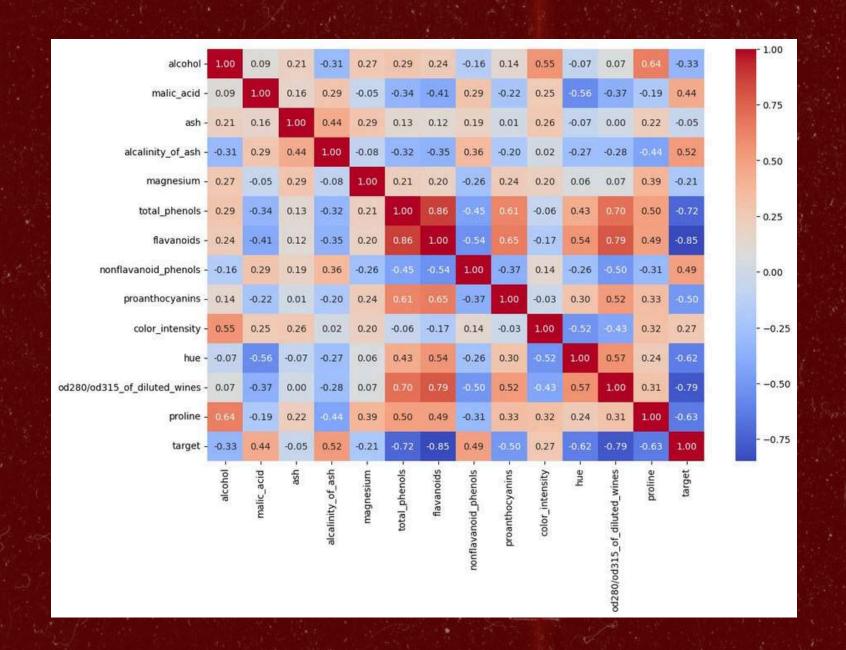
## Exploratory Data Analysis (EDA)

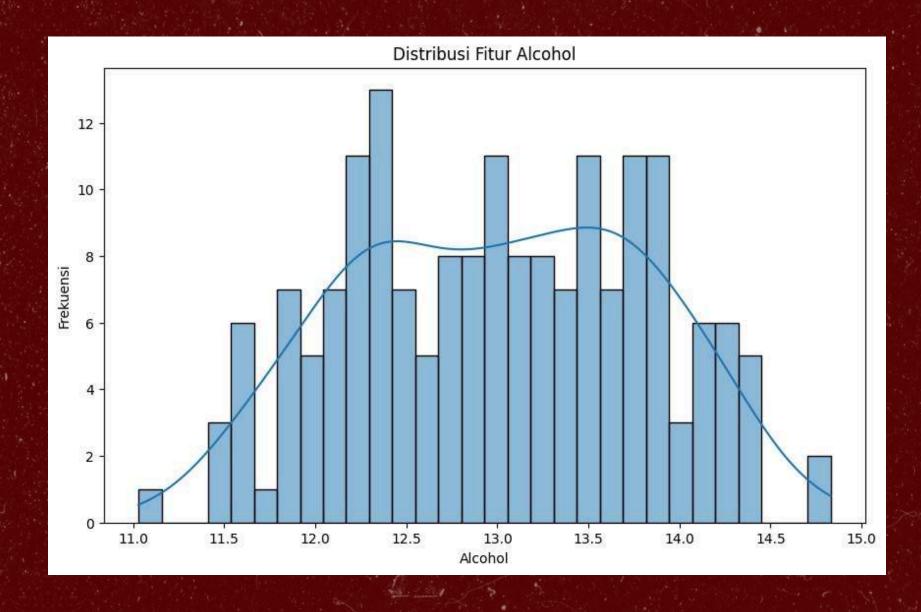




#### EDA & Data Visualization









## Conclusion

```
# Calculate accuracy for each model
log_reg_accuracy = accuracy_score(y_test, log_reg_pred)
knn_accuracy = accuracy_score(y_test, knn_pred)
svm_accuracy = accuracy_score(y_test, svm_pred)
# Print out the accuracies
print(f"Logistic Regression Accuracy: {log_reg_accuracy:.4f}")
print(f"K-Nearest Neighbors Accuracy: {knn_accuracy:.4f}")
print(f"Support Vector Machine Accuracy: {svm_accuracy:.4f}")
# Find the best model
accuracies = { 'Logistic Regression': log reg accuracy, 'K-Nearest Neighbors':
best_model = max(accuracies, key=accuracies.get)
print(f"The best model is: {best model} with an accuracy of {accuracies[best_model]
Logistic Regression Accuracy: 1.0000
K-Nearest Neighbors Accuracy: 0.7407
Support Vector Machine Accuracy: 0.7593
The best model is: Logistic Regression with an accuracy of 1.0000
```

The conclusion from the results above shows that Logistic Regression achieves a perfect accuracy of 1.0000 on the tested dataset, making it the best model in this experiment. Meanwhile, K-Nearest Neighbors (KNN) and Support Vector Machine (SVM) provide lower accuracies of 0.7407 and 0.7593, respectively. This indicates that Logistic Regression is more effective in modeling this data, while KNN and SVM are less optimal in terms of accuracy.

# Thank Mou FOR YOUR ATTENTION