

kNN (k-Nearest-Neighbors)

Cara Kerja:

1. Tentukan jumlah tetangga k
2. Hitung jarak antara data test dengan training dataset menggunakan metrik jarak, misalnya, *Euclidean distance*, *Manhattan distance*, atau *Minkowski distance*.
3. Urutkan *training dataset* berdasarkan jarak.
4. Pilih k data terdekat.
5. Klasifikasikan data baru dengan mayoritas kelas dari k tetangga terdekat.

Perbandingan:

- *Scratch*
 - *Euclidean distance*

```
[19]: from knn import KNNScratch

knn_euclidean_scratch = KNNScratch(neighbors=3, metric='euclidean')
knn_euclidean_scratch.fit(X_train, y_train)
y_pred_knn_euclidean_scratch = knn_euclidean_scratch.predict(X_test)

validate_model(knn_euclidean_scratch, method_name="kNN with Euclidean distance from Scratch")
```

Hold-Out Validation (kNN with Euclidean distance from Scratch):
F1 Score: 0.88

	precision	recall	f1-score	support
0	0.89	1.00	0.94	72
1	1.00	0.79	0.88	42
accuracy			0.92	114
macro avg	0.94	0.89	0.91	114
weighted avg	0.93	0.92	0.92	114

K-Fold Cross-Validation (kNN with Euclidean distance from Scratch):
F1 Scores for each fold: [0.9230769230769231, 0.9393939393939394, 0.9295774647887324, 0.8235294117647058, 0.8135593220338984]
Mean F1 Score: 0.8858274122116399
Standard Deviation of F1 Score: 0.05527148218059224

- *Manhattan distance*

```
[21]: knn_manhattan_scratch = KNNScratch(neighbors = 3, metric = 'manhattan')
knn_manhattan_scratch.fit(X_train, y_train)
y_pred_knn_manhattan_scratch = knn_manhattan_scratch.predict(X_test)

validate_model(knn_manhattan_scratch, method_name="kNN with Manhattan distance from Scratch")
```

Hold-Out Validation (kNN with Manhattan distance from Scratch):
F1 Score: 0.8648648648648649

	precision	recall	f1-score	support
0	0.88	1.00	0.94	72
1	1.00	0.76	0.86	42
accuracy			0.91	114
macro avg	0.94	0.88	0.90	114
weighted avg	0.92	0.91	0.91	114

K-Fold Cross-Validation (kNN with Manhattan distance from Scratch):
F1 Scores for each fold: [0.9375, 0.9538461538461539, 0.9142857142857143, 0.8615384615384616, 0.8]
Mean F1 Score: 0.8934340659340659
Standard Deviation of F1 Score: 0.056166666995138954

- *Minkowski distance*

```
[23]: knn_minkowski_scratch = KNNScratch(neighbors = 5, metric = 'minkowski')
knn_minkowski_scratch.fit(X_train, y_train)
y_pred_knn_minkowski_scratch = knn_minkowski_scratch.predict(X_test)

validate_model(knn_minkowski_scratch, method_name="kNN with Minkowski distance from Scratch")
```

Hold-Out Validation (kNN with Minkowski distance from Scratch):
F1 Score: 0.868421052631579

	precision	recall	f1-score	support
0	0.89	0.99	0.93	72
1	0.97	0.79	0.87	42
accuracy			0.91	114
macro avg	0.93	0.89	0.90	114
weighted avg	0.92	0.91	0.91	114

K-Fold Cross-Validation (kNN with Minkowski distance from Scratch):
F1 Scores for each fold: [0.9230769230769231, 0.9090909090909091, 0.9459459459459459, 0.8695652173913043, 0.7636363636363637]
Mean F1 Score: 0.8822630718282891
Standard Deviation of F1 Score: 0.0643032552295382

- *Library*
 - *Euclidean distance*

```
[20]: from sklearn.neighbors import KNeighborsClassifier

knn_euclidean = KNeighborsClassifier(n_neighbors = 3, metric = 'euclidean')
knn_euclidean.fit(X_train, y_train)
y_pred_knn_euclidean = knn_euclidean.predict(X_test)

validate_model(knn_euclidean, method_name="kNN with Euclidean distance from Library")
```

Hold-Out Validation (kNN with Euclidean distance from Library):
F1 Score: 0.88

	precision	recall	f1-score	support
0	0.89	1.00	0.94	72
1	1.00	0.79	0.88	42
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Mean F1 Score: 0.8858274122116399
Standard Deviation of F1 Score: 0.05527148218059224

- *Manhattan distance*

```
[22]: knn_manhattan = KNeighborsClassifier(n_neighbors = 3, metric = 'manhattan')
knn_manhattan.fit(X_train, y_train)
y_pred_knn_manhattan = knn_manhattan.predict(X_test)

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```

Hold-Out Validation (kNN with Manhattan distance from Library):
F1 Score: 0.8648648648648649

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0	0.88	1.00	0.94	72
1	1.00	0.76	0.86	42
accuracy			0.91	114
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K-Fold Cross-Validation (kNN with Manhattan distance from Library):
F1 Scores for each fold: [0.9375, 0.9538461538461539, 0.9142857142857143, 0.8615384615384616, 0.8]
Mean F1 Score: 0.8934340659340659
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- *Minkowski distance*

Library

```
[24]: knn_minkowski = KNeighborsClassifier(n_neighbors = 5, metric = 'minkowski')
knn_minkowski.fit(X_train, y_train)
y_pred_knn_minkowski = knn_minkowski.predict(X_test)

validate_model(knn_minkowski, method_name="kNN with Minkowski distance from Library")
```

Hold-Out Validation (kNN with Minkowski distance from Library):
F1 Score: 0.868421052631579

	precision	recall	f1-score	support
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accuracy			0.91	114
macro avg	0.93	0.89	0.90	114
weighted avg	0.92	0.91	0.91	114

K-Fold Cross-Validation (kNN with Minkowski distance from Library):
F1 Scores for each fold: [0.9230769230769231, 0.9253731343283582, 0.9166666666666666, 0.8529411764705882, 0.7636363636363637]
Mean F1 Score: 0.8763388528357801
Standard Deviation of F1 Score: 0.06239301703039033

Dari hasil *F1 Score*, baik menggunakan metrik *Euclidean*, *Manhattan*, atau *Minkowski*, penggunaan kNN dengan *scratch* menghasilkan skor yang sama dengan penggunaan kNN secara *library*.

Improvement:

Improvement yang dapat dilakukan pada algoritma kNN secara *scratch* dapat dilakukan dengan memperbaiki cara untuk mencari jarak terdekat pada titik test.