Canser Classifier

Imports:

```
In [1]: from src.homeworks.homework2.KNNClassifier import KNNClassifier
from src.homeworks.homework2.scalers import MinMaxScaler, MaxAbsScaler, S
from src.homeworks.homework2.score import MetricCalculator
from src.homeworks.homework2.train_test_split import train_test_split
import pandas as pd
import matplotlib.pyplot as plt
```

Read cancer.csv:

Out[2]

```
In [2]: data = pd.read_csv("src/homeworks/homework2/notebooks/cancer.csv")
    data.describe()
```

:		1	2	3	4	5	6	
	count	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569
	mean	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	0
	std	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	0
	min	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	0
	25%	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	0
	50%	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	0
	75%	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	0
	max	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	0

8 rows × 30 columns

Replace "M" to 1 and "B" to 0 in "label"

```
In [3]: data["label"] = data["label"].replace({"M": 1, "B": 0})
    data.describe()
```

/tmp/ipykernel_91621/2696528524.py:1: FutureWarning: Downcasting behavior in `replace` is deprecated and will be removed in a future version. To ret ain the old behavior, explicitly call `result.infer_objects(copy=False)`. To opt-in to the future behavior, set `pd.set_option('future.no_silent_downcasting', True)`

data["label"] = data["label"].replace({"M": 1, "B": 0})

Out[3]:		label	1	2	3	4	5	
	count	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569
	mean	0.372583	14.127292	19.289649	91.969033	654.889104	0.096360	0
	std	0.483918	3.524049	4.301036	24.298981	351.914129	0.014064	0
	min	0.000000	6.981000	9.710000	43.790000	143.500000	0.052630	0
	25%	0.000000	11.700000	16.170000	75.170000	420.300000	0.086370	0
	50%	0.000000	13.370000	18.840000	86.240000	551.100000	0.095870	0
	75%	1.000000	15.780000	21.800000	104.100000	782.700000	0.105300	0
	max	1.000000	28.110000	39.280000	188.500000	2501.000000	0.163400	0

8 rows × 31 columns

```
←
```

Divide the dataset into X and y:

Creat a function that performs normalization, splitting into train and test, and counts metrics.

```
f1_score.append(metric.f1_score())
return accuracy, f1_score
```

MinMaxScaler

```
In [6]: accuracy, f1_score = get_score(X, y, test_size=0.15, shuffle=True, random
In [7]: k_values = range(1, 21)
    plt.plot(k_values, accuracy, label='Accuracy', linestyle='-')
    plt.plot(k_values, f1_score, label='F1 Score', linestyle='-')
    plt.xlabel('k')
    plt.ylabel('metric')
    plt.xticks(k_values)

plt.grid(True)
    plt.legend()

plt.show()
```

MaxAbsScaler

0.90

```
In [8]: accuracy, f1_score = get_score(X, y, test_size=0.15, shuffle=True, random
In [9]: k_values = range(1, 21)
    plt.plot(k_values, accuracy, label='Accuracy', linestyle='-')
    plt.plot(k_values, f1_score, label='F1 Score', linestyle='-')
    plt.xlabel('k')
```

3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20

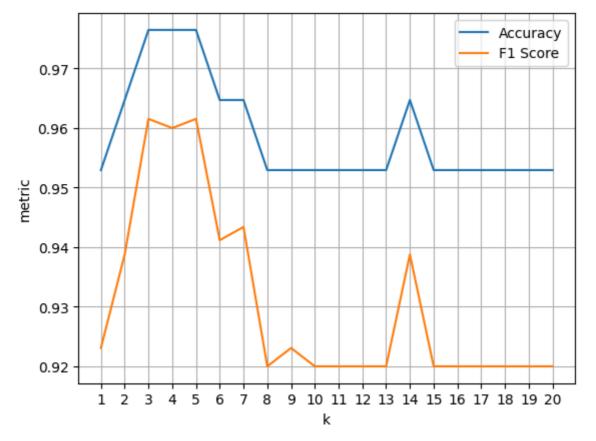
k

Accuracy F1 Score

```
plt.ylabel('metric')
plt.xticks(k_values)

plt.grid(True)
plt.legend()

plt.show()
```



StandardScaler

```
In [10]: accuracy, f1_score = get_score(X, y, test_size=0.15, shuffle=True, random
In [11]: k_values = range(1, 21)
    plt.plot(k_values, accuracy, label='Accuracy', linestyle='-')
    plt.plot(k_values, f1_score, label='F1 Score', linestyle='-')
    plt.xlabel('k')
    plt.ylabel('metric')
    plt.yticks(k_values)

plt.grid(True)
    plt.legend()

plt.show()
```

