Evaluation

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Introduction



Introduction

- The assignment is to calculate various evaluation methods using classification results and labels. (See page 19 for specifics)
- In this PDF, we will explain various evaluation methods and how to use the functions of scikit-learn library for evaluation.



Evaluation Methods

- There are several evaluation methods as below.
 - Confusion Matrix
 - Accuracy
 - Precision
 - Recall
 - F1-score
 - Macro Averaging
 - Micro Averaging



Confusion Matrix

Confusion matrix is a matrix of prediction and actual labels.

• Example of confusion matrix (ex. Classification for cat and non-cat)

- True Positive (TP) : predict cat as cat

- True Negative (TN) : predict non-cat as non-cat

- False Positive (FP) : predict cat as non-cat

- False Negative (FN): predict non-cat as cat

Prediction Label	Predict as cat	Predict as non-cat
Actually cat	True Positive (TP)	False Negative (FN)
Actually non-cat	False Positive (FP)	True Negative (TN)



Accuracy

• Accuracy is ratio of correctly predicted samples to the whole samples.

$$Accuracy = \frac{TP + TN}{TP + FN + FP + TN}$$

Prediction Label	Î	Ê
Т	True Positive (TP)	False Negative (FN)
F	False Positive (FP)	True Negative (TN)



Precision

• Precision is ratio of actual true samples to the samples predicted as true.

$$Precision = \frac{TP}{TP + FP}$$

Prediction	Î	\widehat{F}
T	True Positive (TP)	False Negative (FN)
F	False Positive (FP)	True Negative (TN)



❖ Recall

• Recall is ratio of samples predicted as true to the actually true samples.

$$Recall = \frac{TP}{TP + FN}$$

Prediction Label	\widehat{T}	\widehat{F}
Т	True Positive (TP)	False Negative (FN)
F	False Positive (FP)	True Negative (TN)



- ❖ F1-score
 - F1-score is harmonic mean between precision and recall.

$$F1-score = \frac{2 \times precision \times recall}{precision + recall}$$



Macro Averaging

• To calculate Macro averaging scores, we first calculate precision, recall and F1-score for each label respectively and then divide each score with the number of labels. (see next page)

Prediction Label	Â	Â	Ĉ
A	9	3	0
В	3	5	1
С	1	1	4

 \hat{T}, \hat{F} : Prediction

T, F: Labels

A	\widehat{T}	\widehat{F}
T	9	3
F	4	11

В	\widehat{T}	Ê
T	5	4
F	4	14

С	\widehat{T}	\widehat{F}
T	4	2
F	1	20



Macro Averaging

A	\widehat{T}	\widehat{F}
T	9	3
F	4	11

В	\widehat{T}	\widehat{F}
T	5	4
F	4	14

С	\widehat{T}	Ê
T	4	2
F	1	20

$$Precision_A = 0.69$$

 $Recall_A = 0.75$
 $F1 - score_A = 0.72$

$$Precision_B = 0.55$$

 $Recall_B = 0.55$
 $F1 - score_B = 0.55$

$$Precision_C = 0.66$$

 $Recall_C = 0.80$
 $F1 - score_C = 0.72$

Macro averaging Precision =
$$\frac{0.69 + 0.55 + 0.66}{3} = 0.63$$
Macro averaging Recall =
$$\frac{0.75 + 0.55 + 0.80}{3} = 0.70$$
Macro averaging F1 - score =
$$\frac{0.72 + 0.55 + 0.72}{3} = 0.66$$



Micro Averaging

• To calculate Micro-averaging scores, we first add up all confusion matrices into one confusion matrix, and then calculate precision, recall and F1-score using the confusion matrix. (see next page)

Prediction Label	Â	Â	Ĉ
A	9	3	0
В	3	5	1
С	1	1	4

 \hat{T}, \hat{F} : Prediction

T, F: Labels

A	\widehat{T}	\widehat{F}
T	9	3
F	4	11

В	\widehat{T}	Ê
T	5	4
F	4	14

С	\widehat{T}	\widehat{F}
T	4	2
F	1	20



Micro Averaging

A	\widehat{T}	\widehat{F}
T	9	3
F	4	11

В	\widehat{T}	Ê
T	5	4
F	4	14

С	\widehat{T}	\widehat{F}
T	4	2
F	1	20





SUM

Micro	\widehat{T}	\widehat{F}
T	18	9
F	9	45

 $Micro\ averaging\ Precision=0.67$

 $Micro\ averaging\ Recall=0.67$

 $Micro\ averaging\ F1-score=0.67$



Evaluation with Scikit-Learn

- 'sklearn.metrics' provides following evaluation python classes.
 - confusion_matrix
 - accuracy_score
 - precision_score
 - recall_score
 - f1_score



Examples of confusion_matrix, accuracy_score

```
1 from sklearn.metrics import confusion_matrix, accuracy_score
2
3 # Label
4 y_true = [1, 0, 0, 1, 1, 0, 0, 1, 1, 0]
5 # Prediction
6 y_pred = [1, 1, 1, 1, 1, 0, 0, 1, 0, 1]
7
8 # confusion matrix
9 print(confusion_matrix(y_true, y_pred))
10 # Accuracy
11 print('Accuracy : ', accuracy_score(y_true, y_pred))
```

[[2 3] [1 4]]

[1 3 1]

Accuracy : 0.6

Prediction Label	0	1
0	2	3
1	1	4

1 from sklearn.metrics import confusion_matrix, accuracy_score			
3 # Label			
4 y_true = [2, 1, 0, 2, 0, 1, 1, 2, 2, 0, 0, 1, 2, 1, 0]			
5 # Prediction			
6 y_pred = [2, 1, 2, 2, 2, 1, 1, 2, 0, 0, 0, 2, 1, 0, 1]			
7			
8 # confusion matrix			
9			
10 # Accuracy			
11 print('Accuracy : ', accuracy_score(y_true, y_pred))			
[[2 + 2]			

T0.533333333333333333

Prediction Label	0	1	2
0	2	1	2
1	1	3	1
2	1	1	3



- precision_score, recall_score, f1_score
 - Parameters
 - average : string (Default : 'binary')
 - Parameters used for multi-label classification
 - macro: use 'macro averaging'
 - · micro: use 'micro averaging'



Examples of precision_score, recall_score, f1_score

```
1 from sklearn, metrics import confusion matrix, precision score, recall score, f1 score
 3 # Label
 4 y_true = [1, 0, 0, 1, 1, 0, 0, 1, 1, 0]
 5 # Prediction
 6 y_pred = [1, 1, 1, 1, 1, 0, 0, 1, 0, 1]
 8 # confusion matrix
 9 print(confusion_matrix(y_true, y_pred))
10 # Precision
11 print('Precision : ', precision_score(y_true, y_pred))
12 # Recall
13 print('Recall : ', recall_score(y_true, y_pred))
14 # F1-score
15 print('F1-score : ', f1_score(y_true, y_pred))
[[2 3]
[1 4]]
Precision: 0.5714285714285714
Recall: 0.8
```



Examples of precision_score, recall_score, f1_score

```
1 from sklearn.metrics import confusion_matrix, precision_score, recall_score, f1_score
 2
 3 # Label
 4 y_true = [2, 1, 0, 2, 0, 1, 1, 2, 2, 0, 0, 1, 2, 1, 0]
 5 # Prediction
 6 y_pred = [2, 1, 2, 2, 2, 1, 1, 2, 0, 0, 0, 2, 1, 0, 1]
 8 # confusion matrix
 9 print(confusion_matrix(y_true, y_pred))
10 # Precision
11 print('Macro averaging Precision : ', precision_score(y_true, y_pred, average='macro'))
12 print('Micro averaging Precision : ', precision_score(y_true, y_pred, average='micro'))
13 # Recall
14 print('Macro averaging Recall : ', recall score(y true, y pred, average='macro'))
15 print('Micro averaging Recall : ', recall score(y true, y pred, average='micro'))
16 # F1-score
17 print('Macro averaging F1-score : ', f1_score(y_true, y_pred, average='macro'))
18 print('Micro averaging F1-score : ', f1_score(y_true, y_pred, average='micro'))
[[2 1 2]
[1 3 1]
[1 1 3]]
Macro averaging Recall: 0.533333333333333333
Macro averaging F1-score : 0.52996632996633
```

Assignment



Assignment

- 1) Use the 'Evaluation.json' file as an input.
 - See page 22 for input file format.
- 2) Construct 'confusion matrix', and calculate 'accuracy', 'precision', 'recall', and 'F1-score' with respect to 'macro averaging' and 'micro averaging'.
- 3) You have to create .txt file containing results of above metrics in form of percentage (file I/O) by using your code.
 - See page 23 for output file format.
- 4) You are allowed to use only 'Scikit-Learn' for calculating above metrics.
- 5) You cannot use any library except scikit-learn and json.

Assignment



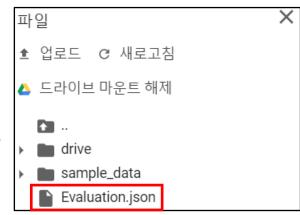
Assignment

- 6) Teaching assistants have been checking out "Copy" to protect plagiarism by a copy detecting system. 0 scores will be assigned to all "Copy" results.
- 7) Round down to the fourth digit after the decimal point.

```
Ex) 78.42159 \rightarrow 78.4215 / 51.172426 \rightarrow 51.1724
```

- 8) For this assignment, If there is any evidence that you upload your source code or download it from any public internet site (Stack Overflow, Github, etc.), we will regard it as copy and you will get 0 score.
- 9) Upload the input file to the same folder where the source code exists as shown on the right.
- 10) All paths in the code use a relative path as below.

```
Ex) with open('./Evaluation.json') as f:
```



Assignment

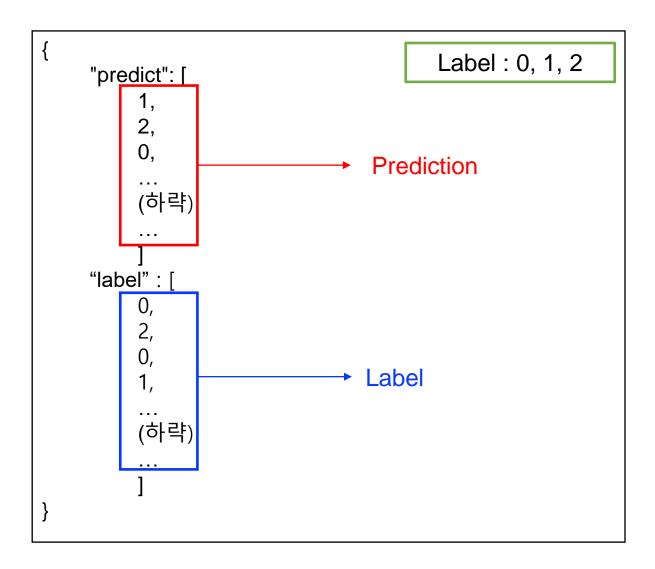


- Submission File (submit following files as a compressed .zip file)
 - 1) Python code file (.py) (python version 3.x, not .ipynb file)
 - Format: "Student Number_Name_EVAL.py".
 - Ex) "2020000000_홍길동_EVAL.py"
 - You will get deduction score if you submit the code with any other format such as .ipynb.
 - Should develop your code on Colab
 - You will get **deduction score** if you submit different outputs from different environments (not Colab).
 - 2) TEXT file (.txt)
 - Format: "Student Number_Name_EVAL.txt".
 - You will get deduction score if you copy and paste the results from print()
 function. This means that your code has to create .txt file by file I/O
 mechanism as below.

```
Ex) fw = open('./Student-Number_NAME.txt', 'w', encodig='UTF-8')
fw.write(result)
fw.close()
```

JSON Input File

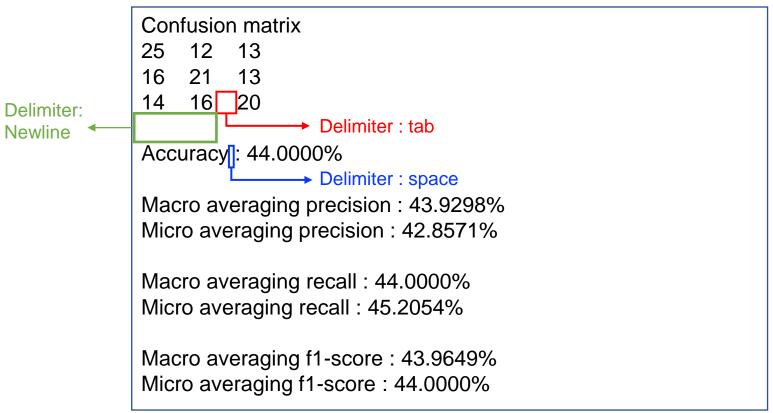




An Example of Output File



Student Number_Name_EVAL.txt



- Above examples can be different with actual answers.
- Round down to the fourth digit after the decimal point.