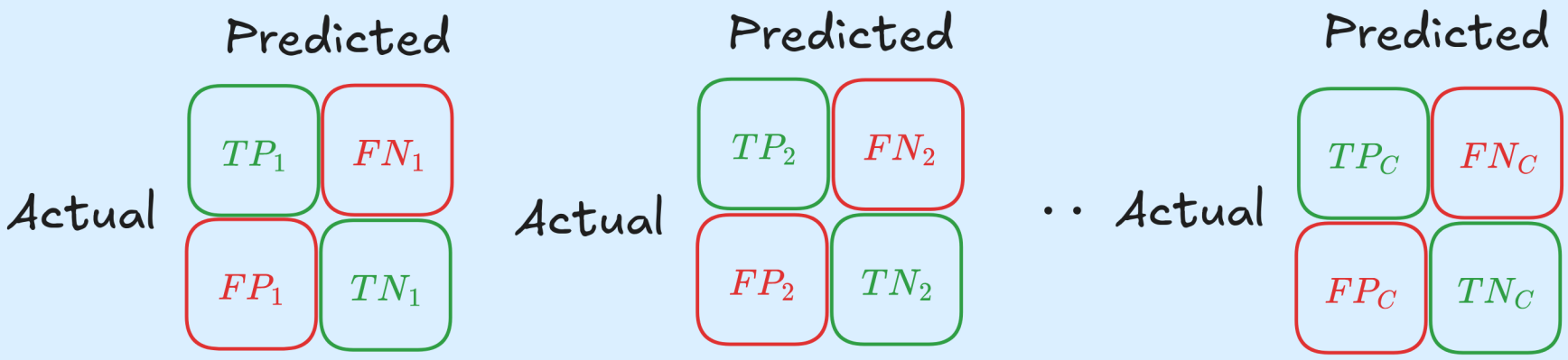
Precision, Recall & F1 for Multi-class Classification

Multi-class Classification Model (C classes)

Confusion Matrix for Each Class



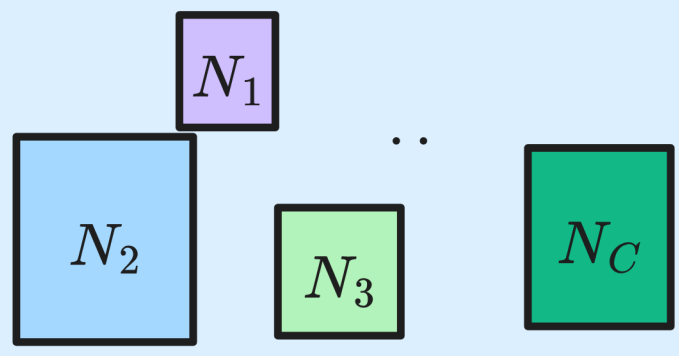
Let $P_i = \dfrac{TP_i}{TP_i + FP_i}$ be the precision for the i^{th} class

Let $R_i = \dfrac{TP_i}{TP_i + FN_i}$ be the recall for the i^{th} class

Let $F_i = \dfrac{2 \cdot P_i \cdot R_i}{P_i + R_i}$ be the F1-score for the i^{th} class

Let N_i = Number of true instances of class-i (support)

$$N = N_1 + N_2 + N_3 + .. + N_C$$



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Macro-averaged Metrics

Idea: The metric is computed per class, then all values are averaged (unweighted).

Feature: Treats all classes equally, so minority class performance matters just as much.

$$\text{Macro Precision} = \frac{1}{C}\sum_{i=1}^{C} P_i = \frac{1}{C}\sum_{i=1}^{C} \frac{TP_i}{TP_i + FP_i}$$

$$\text{Macro Recall} = \frac{1}{C}\sum_{i=1}^{C} R_i = \frac{1}{C}\sum_{i=1}^{C} \frac{TP_i}{TP_i + FN_i}$$

$$\text{Macro F1-score} = \frac{1}{C}\sum_{i=1}^{C} F_i$$

Micro-averaged Metrics

Idea: Combine all confusion matrices (sum), then compute the metric on the grand matrix.

Feature: Focuses on overall performance, not per-class. Can be high even if minority class performance is poor.

$$\text{Micro Precision} = \text{Recall} = \text{F1} = \frac{\sum_{i=1}^{C} TP_i}{\sum_{i=1}^{C} TP_i + \sum_{i=1}^{C} FP_i}$$

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Weighted Metrics

Idea: Compute the metric per class,
then all values are averaged (weighted by class support).

Feature: Reflects the representative importance
of each class (by its support).

$$\text{Weighted Precision} = \sum_{i=1}^{C} \frac{N_i}{N} \cdot P_i = \sum_{i=1}^{C} \frac{N_i}{N} \cdot \frac{TP_i}{TP_i + FP_i}$$

$$\text{Weighted Recall} = \sum_{i=1}^{C} \frac{N_i}{N} \cdot R_i = \sum_{i=1}^{C} \frac{N_i}{N} \cdot \frac{TP_i}{TP_i + FN_i}$$

$$\text{Weighted F1-score} = \sum_{i=1}^{C} \frac{N_i}{N} \cdot F_i$$

```python
from sklearn.metrics import confusion_matrix, f1_score

y_true = [2] * 1000 + [1] * 100 + [0] * 10
y_pred = [2] * 900 + [1] * 209 + [0] * 1

confusion_matrix(y_true, y_pred)
```

```
array([[  1,   9,   0],
       [  0, 100,   0],
       [  0, 100, 900]])
```

```python
# Macro
f1_score(y_true, y_pred, average='macro')
```

```
0.5921452646031082
```

```python
# Micro
f1_score(y_true, y_pred, average='micro')
```

```
0.9018018018018018
```

```python
# Weighted
f1_score(y_true, y_pred, average='weighted')
```

```
0.9134338035717698
```

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