

19. Confusion Matrix

Neural Networks - From Brain to Business

Evaluate classifier performance beyond simple accuracy using precision, recall, and F1 score.

A **confusion matrix** breaks down predictions into four categories: - **True Positive (TP)**: Predicted UP, actually UP - **False Positive (FP)**: Predicted UP, actually DOWN (Type I error) - **True Negative (TN)**: Predicted DOWN, actually DOWN - **False Negative (FN)**: Predicted DOWN, actually UP (Type II error)

This granular view reveals important patterns that overall accuracy hides: - **Precision**: When we predict UP, how often are we right? - **Recall**: Of all actual UP days, how many did we catch? - **F1 Score**: Harmonic mean of precision and recall

In trading, these distinctions matter: a false BUY signal (FP) costs money on a losing trade, while a missed opportunity (FN) is a foregone profit. Different strategies prioritize different metrics.

19_confusion_matrix/confusion_matrix.pdf

Precision (positive predictive value):

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

Recall (sensitivity, true positive rate):

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

F1 Score (harmonic mean):

$$F_1 = 2 \times \frac{\text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$

Accuracy:

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

Think of a spam filter:

- **High precision:** When it says "spam," it's almost always right. Few legitimate emails are blocked. - **High recall:** It catches almost all spam. Few spam emails get through.

You can't maximize both simultaneously: - More aggressive filtering = higher recall, lower precision - More conservative filtering = higher precision, lower recall

In trading: - High precision strategy: Trade rarely, but when you trade, usually win - High recall strategy: Capture most winning opportunities, but also some losers

Practice Problem 1

Problem 1

Given: $TP = 40$, $FP = 15$, $TN = 35$, $FN = 10$. Calculate precision, recall, and F1 score.

Solution

Precision:

$$\text{Precision} = \frac{TP}{TP + FP} = \frac{40}{40 + 15} = \frac{40}{55} = 0.727 = 72.7\%$$

Recall:

$$\text{Recall} = \frac{TP}{TP + FN} = \frac{40}{40 + 10} = \frac{40}{50} = 0.800 = 80.0\%$$

F1 Score:

$$F_1 = 2 \times \frac{0.727 \times 0.800}{0.727 + 0.800} = 2 \times \frac{0.582}{1.527} = 0.762 = 76.2\%$$

Accuracy (for reference):

$$\text{Accuracy} = \frac{40 + 35}{40 + 15 + 35 + 10} = \frac{75}{100} = 75\%$$

Problem 2

Model A: Precision = 90%, Recall = 50%. Model B: Precision = 70%, Recall = 70%. Which is better for a conservative trading strategy?

Solution

Model A is better for conservative trading.

Why:

Model A (High precision): - When it says BUY, it's right 90% of the time - Only catches 50% of opportunities - Trades less frequently but with higher confidence - Lower risk of losing trades

Model B (Balanced): - When it says BUY, it's right 70% of the time - Catches 70% of opportunities - More trades, more wrong signals - Higher risk but also catches more winners

F1 comparison: - Model A: $F_1 = 2 \times \frac{0.90 \times 0.50}{0.90 + 0.50} = 0.643$ - Model B: $F_1 = 2 \times \frac{0.70 \times 0.70}{0.70 + 0.70} = 0.700$

By F1, Model B is "better" overall. But for conservative trading that prioritizes avoiding losses, Model A's 90% precision is preferable.

Business context determines the right choice.

- Confusion matrix reveals four types of prediction outcomes
- Accuracy alone can be misleading (especially with imbalanced classes)
- Precision: "When I predict positive, am I right?"
- Recall: "Do I catch all the actual positives?"
- F1 balances precision and recall
- Choose metrics based on business costs of different errors