



December 6, 2025

## Learning Goal

See how a single neuron makes buy/sell decisions using a threshold.

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**This slide establishes the learning objective for this topic**

## Key Concept (1/2)

A single neuron acts as a **decision maker** by comparing its output to a **threshold**. For binary classification with sigmoid activation:

- Output  $< 0.5$ : Predict Class 1 (BUY)
- Output  $\geq 0.5$ : Predict Class 0 (SELL)

The neuron computes a weighted combination of inputs, transforms it through the sigmoid function, and produces a probability. The threshold converts this probability into a discrete decision.

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Understanding this concept is crucial for neural network fundamentals

## Key Concept (2/2)

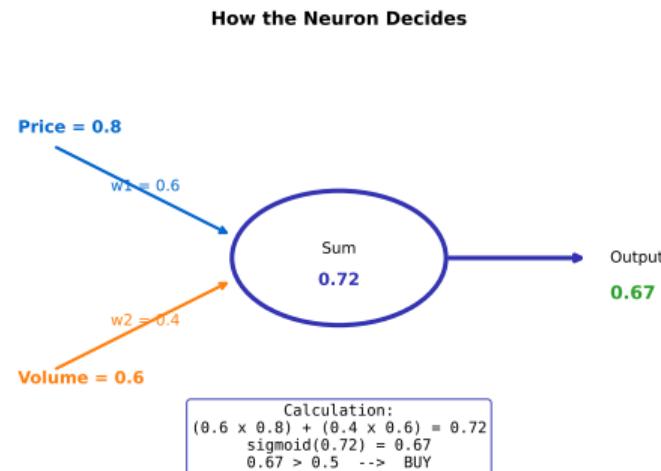
In trading terms: if the neuron outputs 0.67 (67% confidence in price increase), the decision is BUY. If it outputs 0.33 (33% confidence), the decision is SELL.

The position of the decision boundary in feature space corresponds to where the neuron output equals exactly 0.5 - the point of maximum uncertainty.

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Understanding this concept is crucial for neural network fundamentals

# Visualization



Visual representations help solidify abstract concepts

**Neuron computation:**

$$z = w_1x_1 + w_2x_2 + b$$

$$\hat{y} = \sigma(z) = \frac{1}{1 + e^{-z}}$$

**Decision rule:**

$$\text{Decision} = \begin{cases} \text{BUY} & \text{if } \hat{y} > 0.5 \\ \text{SELL} & \text{if } \hat{y} \leq 0.5 \end{cases}$$

**Boundary condition ( $\hat{y}$ -hat = 0.5):**

$$z = 0 \implies w_1x_1 + w_2x_2 + b = 0$$

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Mathematical formalization provides precision

Think of the neuron as a judge weighing evidence:

1. **Gather evidence:** Each input (price, volume) provides information
2. **Weight importance:** Some evidence matters more (larger weights)
3. **Combine and evaluate:** Sum weighted evidence, adjust by bias
4. **Confidence level:** Sigmoid converts to 0-100% confidence
5. **Make decision:** If confidence  $\geq 50\%$ , rule in favor (BUY)  
The weights encode what the neuron has learned about which inputs matter and how much.

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Intuitive explanations bridge theory and practice

## Practice Problem 1

### Problem 1

A neuron has weights  $w_1 = 0.6$  (for price),  $w_2 = 0.4$  (for volume), and bias  $b = -0.3$ . Given price = 0.8 and volume = 0.5, what is the decision?

### Solution

#### Step 1: Weighted sum

$$\begin{aligned}z &= 0.6(0.8) + 0.4(0.5) + (-0.3) \\z &= 0.48 + 0.20 - 0.30 = 0.38\end{aligned}$$

#### Step 2: Sigmoid activation

$$\hat{y} = \frac{1}{1 + e^{-0.38}} = \frac{1}{1 + 0.684} = 0.594$$

Step 3: Decision Since  $0.594 > 0.5$ : **Decision: BUY** (59.4% confidence in price increase)

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Practice problems reinforce understanding

## Practice Problem 2

### Problem 2

Using the same neuron, what values of (price, volume) lie exactly on the decision boundary?

### Solution

On the boundary,  $z = 0$ :

$$0.6 \cdot \text{price} + 0.4 \cdot \text{volume} - 0.3 = 0$$

Solving for volume:

$$\text{volume} = \frac{0.3 - 0.6 \cdot \text{price}}{0.4}$$

$$\text{volume} = 0.75 - 1.5 \cdot \text{price}$$

Example points on the boundary: - price = 0.0: volume = 0.75 - price = 0.5: volume = 0.0 - price = 0.3: volume = 0.3

These points all yield 50% confidence (maximum uncertainty).

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Practice problems reinforce understanding

## Key Takeaways

- A neuron outputs a probability via sigmoid activation
- Default threshold of 0.5 converts probability to binary decision
- The decision boundary is where output = 0.5 ( $z = 0$ )
- Threshold can be adjusted based on risk tolerance
- Single neuron = single linear decision boundary

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These key points summarize the essential learnings