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Understand why a single neuron cannot solve non-linearly separable problems like XOR.

This slide establishes the learning objective for this topic

Key Concept (1/2)

A single neuron with linear activation can only create a **straight line** (or hyperplane) to separate data into two categories. This works perfectly when data can be divided by a straight line - called "linearly separable" problems. However, many real-world problems are **non-linearly separable**. The classic example is XOR (exclusive or): output is 1 when exactly one input is 1, but 0 when both inputs are the same. No single straight line can separate XOR's four points correctly.

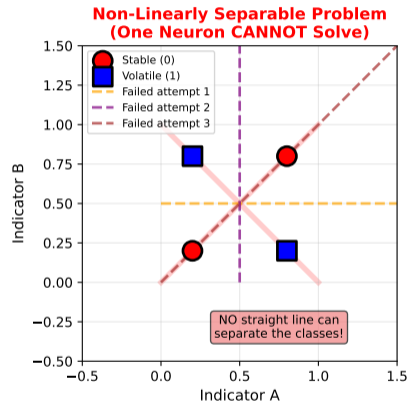
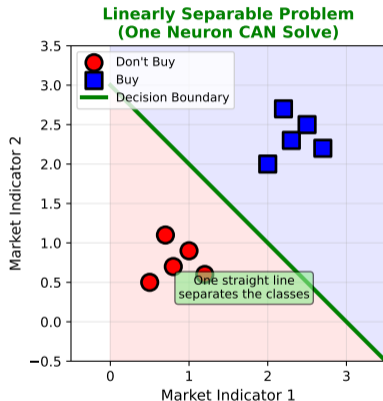
Understanding this concept is crucial for neural network fundamentals

Key Concept (2/2)

This fundamental limitation motivated the development of **multi-layer networks**. By stacking neurons in layers, each with non-linear activation functions, networks can learn curved decision boundaries that separate complex patterns. The XOR problem was historically important - in 1969, Minsky and Papert showed that single-layer perceptrons couldn't solve XOR, which temporarily dampened enthusiasm for neural networks. The solution came with backpropagation and hidden layers in the 1980s.

Understanding this concept is crucial for neural network fundamentals

Why One Neuron Is Not Enough



Solution: Use Multiple Layers (Hidden Layers) to Create Non-Linear Decision Boundaries

Visual representations help solidify abstract concepts

A single neuron defines a linear decision boundary:

$$w_1x_1 + w_2x_2 + b = 0$$

Points are classified as: - **Class 1** if $w_1x_1 + w_2x_2 + b > 0$ - **Class 0** if $w_1x_1 + w_2x_2 + b < 0$

This equation describes a straight line in 2D (or hyperplane in higher dimensions).

Mathematical formalization provides precision

Imagine trying to separate red and blue marbles on a table using only a straight ruler. If you can place the ruler so all red marbles are on one side and all blue on the other, the problem is linearly separable.

Now imagine four marbles arranged in a square: two diagonal marbles are red, two are blue. No matter how you place a single ruler, you'll always have at least one marble on the wrong side. This is the XOR pattern.

The solution? Use **two rulers** (two neurons in a hidden layer) to create a more complex boundary. Together, they can isolate the diagonal patterns that one ruler cannot.

Intuitive explanations bridge theory and practice

Practice Problem 1

Problem 1

Consider the AND function: output is 1 only if both inputs are 1. The four input combinations are (0,0)-0, (0,1)-0, (1,0)-0, (1,1)-1. Is this linearly separable? Can you find weights and bias that work?

Solution

Yes, AND is linearly separable.

The point (1,1) must be on one side, and (0,0), (0,1), (1,0) on the other.

One solution: $w_1 = 1$, $w_2 = 1$, $b = -1.5$

Check: - (0,0): $1(0) + 1(0) - 1.5 = -1.5 < 0$ - Class 0 (correct) - (0,1): $1(0) + 1(1) - 1.5 = -0.5 < 0$ - Class 0 (correct) - (1,0): $1(1) + 1(0) - 1.5 = -0.5 < 0$ - Class 0 (correct) - (1,1): $1(1) + 1(1) - 1.5 = 0.5 > 0$ - Class 1 (correct)

A single neuron can solve AND.

Practice problems reinforce understanding

Practice Problem 2

Problem 2

Now consider XOR: $(0,0) \rightarrow 0$, $(0,1) \rightarrow 1$, $(1,0) \rightarrow 1$, $(1,1) \rightarrow 0$. Try to find weights and bias that separate the classes. Why is this impossible?

Solution

XOR is NOT linearly separable.

We need: - $(0,0)$ and $(1,1)$ on one side (Class 0) - $(0,1)$ and $(1,0)$ on the other side (Class 1)

The Class 1 points are diagonally opposite! Any line separating $(0,1)$ from $(0,0)$ would also separate $(1,0)$ from $(1,1)$, but $(0,0)$ and $(1,1)$ should be together.

No single straight line can achieve this - the classes are "interleaved" diagonally.

This is why XOR requires at least 2 neurons in a hidden layer.

Practice problems reinforce understanding

- A single neuron can only create linear (straight) decision boundaries
- XOR is the classic example of a non-linearly separable problem
- Most real-world problems (including financial prediction) are non-linear
- Solution: Use multiple neurons in hidden layers with non-linear activation

These key points summarize the essential learnings