



For Those Who Don't Know the Jargon, Data Enthusiasts
Welcome!

What exactly is Graph Neural Networks (GNN) ?



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What Is a Graph?



A **graph** is a way to show **connections**.

-  **Nodes (Vertices)** = Items (e.g., people, cities, proteins)
-  **Edges (Links)** = Relationships between items (e.g., friendships, roads, chemical bonds)



Real-Life Examples:

-  **Facebook**: People are nodes, friendships are edges
-  **Maps**: Cities are nodes, roads are edges
-  **Molecules**: Atoms are nodes, bonds are edges

What Is a Graph Neural Network (GNN)?



GNN is a **deep learning model** that works on **graphs**.

Unlike regular neural networks (which work on tables or images), GNNs learn from **connections**.

📌 Key Idea:

A node's new information = its own info + neighbors' info

📊 It captures the **structure** of the data

🧠 GNN helps machines think more like humans – "*Who are your friends?*"



How GNN Works - In 4 Steps



Let's break it down 

- 1 Message Passing:** Each node sends messages to its neighbors
- 2 Aggregation:** Collect messages from all neighbors
- 3 Update:** Combine old info with new info
- 4 Repeat:** Do this for multiple layers/rounds

 It's like learning not just from your own homework, but from your friends' too!

Step 1 - Message Passing



What Happens?

Each node **sends a message** to its neighbors.

Imagine you're in a class 🧑. You tell your friends what you know about a subject. Each friend also shares what *they* know. 💬

In GNN:

Each node **shares its data (features)** with all connected nodes.

Step 1 - Message Passing



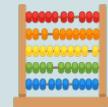
Example:

Node A = 2, Node B = 5 → A sends 2 to B, and B sends 5 to A.



This builds a network of shared knowledge!

Step 2 - Aggregation



What Happens?

Each node **collects** the messages it got from its neighbors and **merges** them.

 Like when your teacher collects everyone's ideas and combines them into one summary



 In GNN:

Use math like **sum**, **mean**, or **max** to combine the values.

Step 2 - Aggregation



Example:

If Node A receives 5 and 7 from its neighbors, initial value = 4

$$\text{Aggregation} = (5 + 7) / 2 = 6$$

 Node A now has a sense of what its neighborhood is like.

Step 3 - Update



What Happens?

Each node uses the aggregated message to **update** its own feature.

📘 Like you learning something new after hearing your friends' thoughts. 🧠

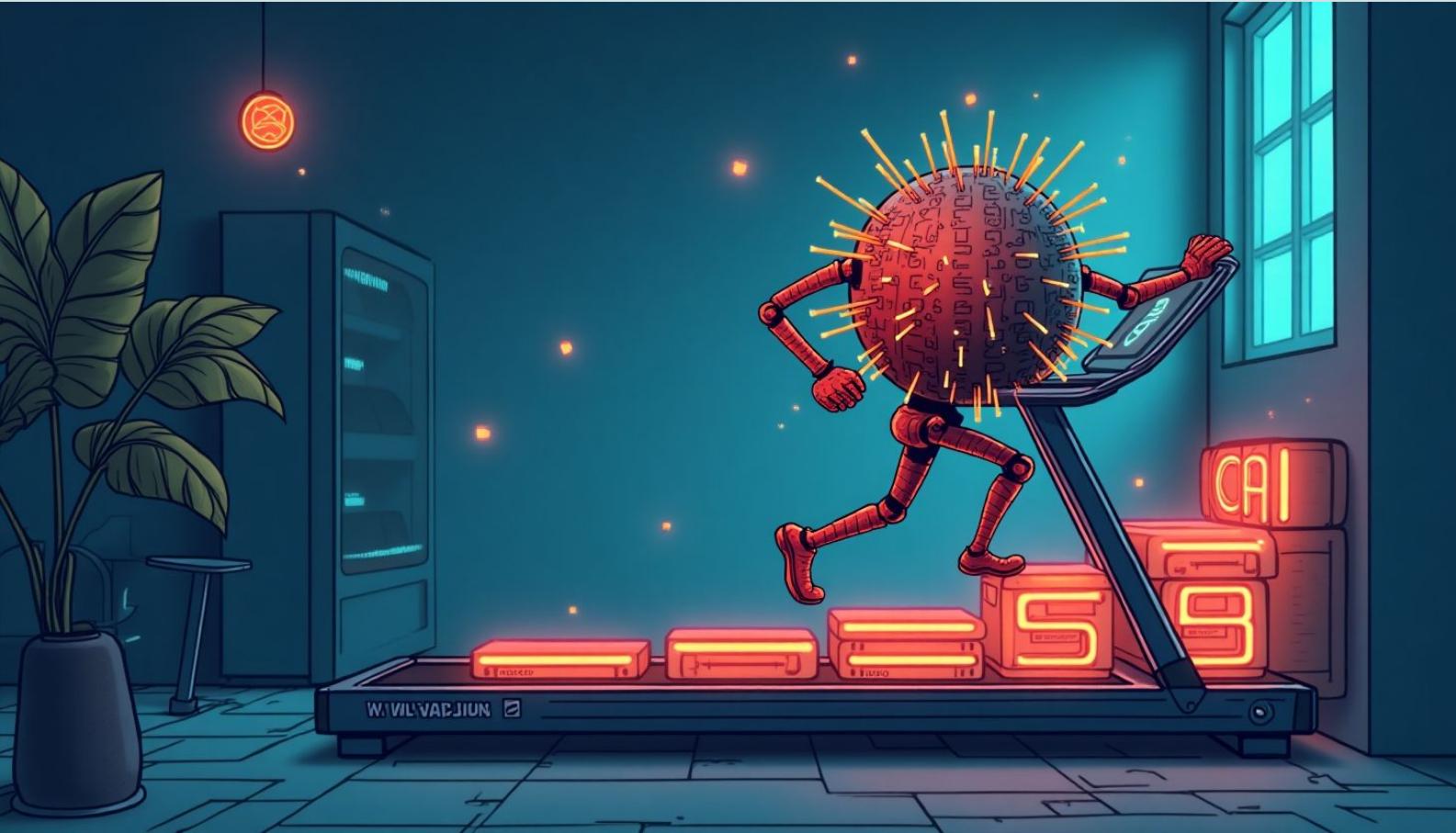
You *mix* your original idea with their ideas to get a better one.



In GNN:

Use a small neural network layer (like linear + ReLU) to update the node's features.

Step 3 - Update



Example:

Original value = 2

Aggregated value = 6

Updated value = **some function of (2, 6)**,
maybe:

$$\text{Updated} = \text{ReLU}(W_1 \times 2 + W_2 \times 6)$$

This helps each node learn a **better representation**.

Step 4 - Repeat



What Happens?

The GNN repeats the process for **multiple rounds** (called layers).

 It's like discussing with friends again tomorrow — now you all know more! 

Each round allows you to learn from friends of friends.

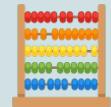


In GNN:

- **1 layer** = info from direct neighbors
- **2 layers** = info from neighbors of neighbors
- **3 layers** = info from even further out!

 More rounds = **wider understanding** of the graph.

Simple Math Behind GNN



Let's say node A has two neighbors: B = 5, C = 7
Node A's value = average of neighbors

$$\text{New_A} = (B + C)/2 = (5 + 7)/2 = 6$$

🧠 Now, A learns the *context* from its neighbors.

You can also use:

- **Sum**
- **Max**
- **Weighted Average (if some friends are more important)**

More Real-Life Examples



🛒 **E-commerce:** Products bought together → Graph

🚕 **Traffic:** Roads as nodes, connections as edges

🎬 **Netflix:** Users & movies connected by watch history

🧪 **Chemistry:** Atoms in molecules form a graph

🔒 **Banking:** Detecting fraud by checking transaction patterns

Difference from Other Neural Networks



◆ **CNN (Convolutional Neural Network)**

Used for images. It learns by looking at nearby pixels — like how your eyes notice patterns in a photo.

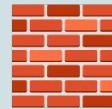
◆ **RNN (Recurrent Neural Network)**

Used for sequences like text or time series. It learns from the past — each step depends on the one before it.

◆ **GNN (Graph Neural Network)**

Used for graphs. It learns from connected neighbors — like how you learn from your friends and their friends.

GNN Layer Example



📌 Each layer updates a node like this:

$$h_i = f(h_i, \text{Aggregate}(h_j \text{ for all neighbors } j))$$

Let's say:

- Node has value **2**
- Two neighbors: **4** and **6**
- Aggregation = mean
- Then:

$$\text{New value} = f(2, \text{mean}(4, 6)) = f(2, 5)$$

Where **f** is a simple neural network layer (e.g., linear + ReLU)

GNN Use Case - Predict Disease

Spread



A person is infected.



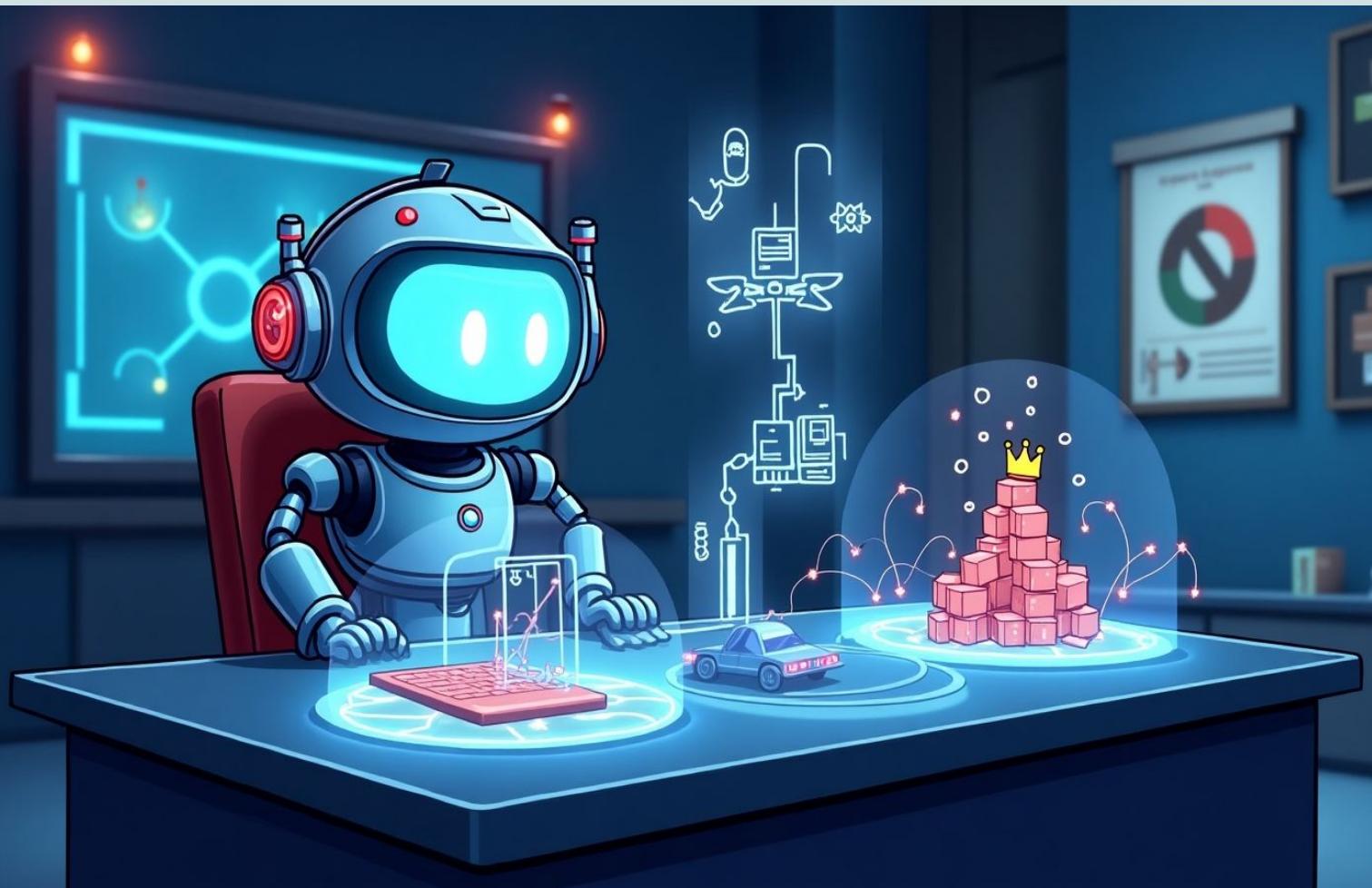
GNN sees who they met (nodes + edges).



It learns the pattern of spread to **predict the next possible infected person**.

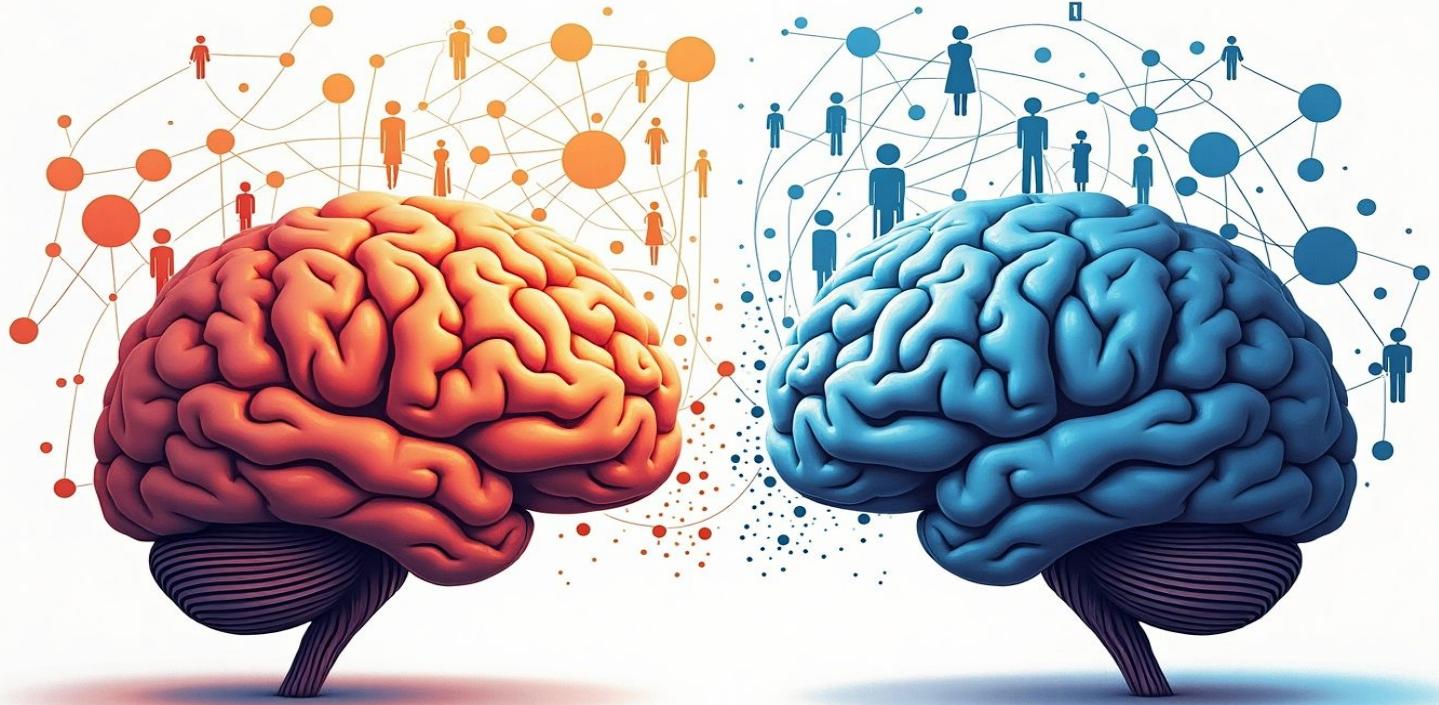
Used in **COVID models, network biology**, etc.

Real-World Use Cases



- Social Networks** – Who might you know?
- Search Engines** – Which page is important?
- Drug Discovery** – Which molecule works better?
- Traffic Flow Prediction** – Predicting congestion
- Fraud Detection** – Suspicious connections in financial transactions

GNN vs Traditional Neural Network



Traditional NN: Good for tables/images

GNN: Good for networks and connections

If your data involves *who is connected to whom*, GNN is the best choice.

Types of GNNs (For Curious Readers)



- ◆ **GCN (Graph Convolutional Network)** – Basic model
- ◆ **GAT (Graph Attention Network)** – Gives importance to certain neighbors
- ◆ **GraphSAGE** – Learns from a *sample* of neighbors
- ◆ **PinSage** – Used by Pinterest for recommendations!

Summary ❤



- **Graph** = Nodes + Edges
- **GNN** = Learns by looking at **neighbors**
- Used in social media, science, maps, fraud detection
- Math is simple (Average or Sum of neighbors)
- **GNN** = Powerful for **networked data**

Decode Connections with GNN



From friends to formulas, GNN sees what others miss.

Let's turn networks into knowledge — one node at a time. 🤝📊

✉️ Reach out — and let's build smarter systems, one connection at a time.



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