# Graph Neural Network Lecture 2



- 1 Introduction to PyG
- 2 Basic Code Examples
- 3 Advanced Features
- 4 Node Classification using Cora dataset
- **5** Conclusion

# What is PyTorch Geometric?

- **PyG** (PyTorch Geometric) is a library for deep learning on graphs built on PyTorch.
- Provides tools for graph creation, transformations, and GNN layers.
- Efficient handling of large-scale graphs.
- Installation:
  - pip install torch\_geometric
  - conda install pyg c pyg

# **Key Features**

- **Data Handling**: Easy-to-use data structures for graphs.
- Predefined Layers: GCN, GAT, GraphSAGE, etc.
- Datasets: Built-in datasets like Cora, Citeseer, and more.
- Extensibility: Custom layers and datasets.

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## Creating a simple graph

```
import torch
   from torch geometric.data import Data
3
  # Define node features and edges
  x = torch.tensor([[1], [2], [3], [4]],
6
                    dtype=torch.float)
   edge index = torch.tensor([[0, 1, 2, 3],
8
                               [1, 0, 3, 2]],
9
                              dtype=torch.long)
10
   # Create a graph data object
11
  data = Data(x=x, edge_index=edge_index)
  print (data)
```

## Accessing Graph Properties

```
print (f'Number of nodes: {data.num_nodes}')
print (f'Number of edges: {data.num_edges}')
print (f'Has isolated nodes: {data.has_isolated_nodes()}
print (f'Is undirected: {data.is_undirected()}')
```

# Adding Edge Weights

# Convert to Undirected Graph

```
from torch_geometric.utils import to_undirected
description
d
```

## Using Built-in Datasets

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## **Batching Graphs**

- PyG supports batching multiple graphs into a single graph object.
- Useful for training on datasets with multiple small graphs.

# **Graph Transformations**

• PyG provides utilities for graph transformations (e.g., normalization, augmentation).

## Using a GCN Layer

```
from torch_geometric.nn import GCNConv
   import torch.nn.functional as F
4
5
   class GCN(torch.nn.Module):
6
       def init (self):
8
           super(). init ()
9
           self.conv1 = GCNConv(1, 16)
10
           self.conv2 = GCNConv(16, 2)
11
12
       def forward(self, x, edge_index):
13
           x = self.conv1(x, edge index)
14
           x = F.relu(x)
```

# Defining a Custom GNN Layer

```
1
2
3
# -----
from torch_geometric.nn import MessagePassing
5
6
7 class CustomGNN (MessagePassing):
    def __init__(self):
```

## Training a GNN Model

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#### The Cora Dataset

- The **Cora dataset** is a well-known citation network dataset used for graph-based machine learning tasks.
- It consists of a collection of scientific publications in computer science, with citation links between papers.
- Key features of the dataset:
  - 2,708 scientific publications (nodes).
  - 5,429 citation links (edges).
  - Each paper is represented by a feature vector with 1,433 binary attributes.
  - 7 classes representing different research topics in computer science (e.g., neural networks, machine learning).

#### **Node Classification Problem**

- Each **publication** is a **node** in the graph, and the **citation between papers** is represented as an **edge**.
- The **node classification** task aims to predict the class/topic of each node/paper based on its feature vector and graph structure.

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#### Conclusion

- PyTorch Geometric simplifies working with GNNs.
- Supports various graph neural network architectures.
- Efficient for large-scale graph data processing.