# **CMLG Course Lab II**

Siqi Miao, Haoyu Wang CS@Purdue

#### **Outline**

- How to store graph data for ML?
- How to implement GNNs for different tasks?
- Q&A

#### How to store image data?

A sample:

$$x = ?, y = ?$$

$$x \in \mathbb{R}^?, y \in \mathbb{Z}^{+?}$$

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 $y = Dog (or, the k^{th} class)$ 

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$$\circ \ x \in \mathbb{R}^{W \times H \times 3}, y \in \mathbb{Z}^+$$

#### How to store image data?

• The entire dataset with **N** samples:

$$x \in \mathbb{R}^2, y \in \mathbb{Z}^{+?}$$

### How to train an image classifier?

- The entire dataset with **N** samples:
  - $\circ$  CNN: f , images: x , labels: y
  - o 1. Forward pass:
    - $\hat{y} = f(x)$
  - 2. Compute loss
    - lacksquare loss =  $\mathcal{L}(y, \hat{y})$
  - 3. Backward pass
    - Grad =  $\frac{d\mathcal{L}}{df}$
  - o **4. Update** fusing Grad

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  - $\circ$  **4. Update** f using Grad

```
def train_one_epoch(data):
    model.train()
    optimizer.zero_grad()
    y_hat = model(data.x)
    loss = criterion(y_hat, data.y)
    loss.backward()
    optimizer.step()
    return loss
```

It's impractical to do full-batch training when <u>N</u> is large!

#### How to store image data?

• A batch of **B** samples:

$$\circ x =$$

 $y = \text{all } \underline{\textbf{\textit{B}}} \text{ labels}$ 

$$x \in \mathbb{R}^?, y \in \mathbb{Z}^{+?}$$

#### How to train an image classifier?

• A batch has **B** samples; the dataset has **N** samples:

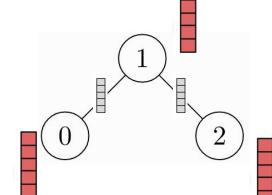
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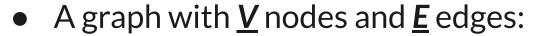
Full-batch training

```
def train_one_epoch():
    model.train()
    # what's the number of iterations per epoch?
    for batch in train_loader:
        ptimizer.zero_grad()
        y_hat = model(data.x)
        loss = criterion(y_hat, data.y)
        loss.backward()
        optimizer.step()
```

Mini-batch training

- A graph with <u>V</u> nodes and <u>E</u> edges:
  - x = ?, y = ?
  - What should be included as features?





$$x = ?, y = ?$$



$$x \in \mathbb{R}^?$$

■ Edge features 
$$edge attr \in \mathbb{R}^?$$

• Edge list! 
$$edge index \in \mathbb{Z}^{+?}$$

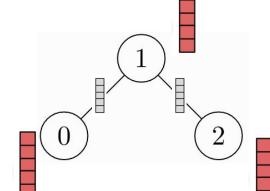
### How to store graph data?

- A graph with <u>V</u> nodes and <u>E</u> edges:
  - x = ?, y = ?

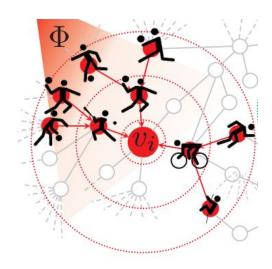


Node features

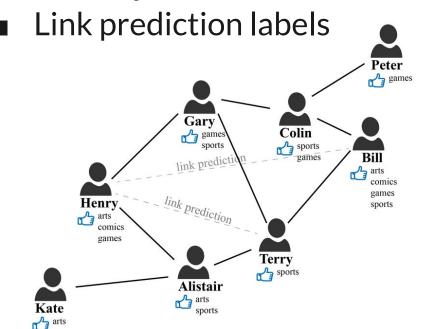
- $x \in \mathbb{R}^{?}$
- Edge features  $edge_attr \in \mathbb{R}^?$
- Adjacency matrix...?
  - Edge list! edge index  $\in \mathbb{Z}^{+?}$
- $\circ$  What can y be?



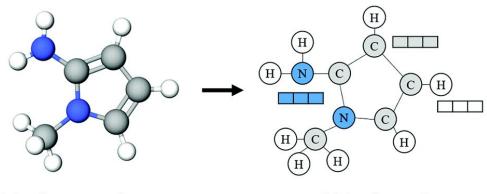
- A graph with <u>V</u> nodes and <u>E</u> edges:
  - $\circ$  What can y be?
    - Node classification labels



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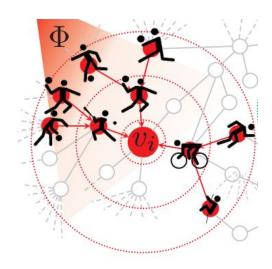


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Molecular graph

- A graph with <u>V</u> nodes and <u>E</u> edges:
  - $\circ$  What can y be?
    - Node classification labels



#### How to train a node classifier?

- A graph with <u>V</u> nodes and <u>E</u> edges:
  - Full-batch training
    - What if <u>V</u> is 1 million?
  - Mini-batch training
    - How?

#### How to train a node classifier?

- A graph with <u>V</u> nodes and <u>E</u> edges:
  - Mini-batch training

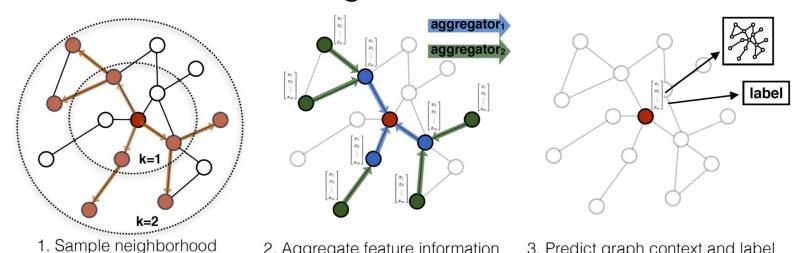


Figure 1: Visual illustration of the GraphSAGE sample and aggregate approach.

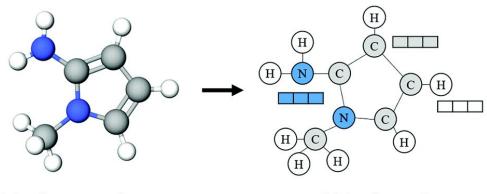
2. Aggregate feature information

from neighbors

3. Predict graph context and label

using aggregated information

- A graph with <u>V</u> nodes and <u>E</u> edges:
  - $\circ$  What can y be?
    - Graph classification labels

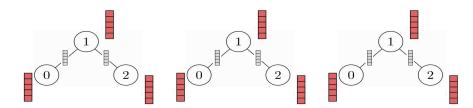


Molecular graph

- How about a batch of **B** graphs:
  - What should be included as features?
    - Node features

$$x \in \mathbb{R}^?$$

- Edge features  $edge_attr \in \mathbb{R}^?$
- Adjacency matrix...?
  - Edge list!  $edge_index \in \mathbb{Z}^{+?}$



### How to store graph data?

- How about a batch of <u>B</u> graphs:
  - What should be included as features?
    - Node features

$$x \in \mathbb{R}^{?}$$

Edge features

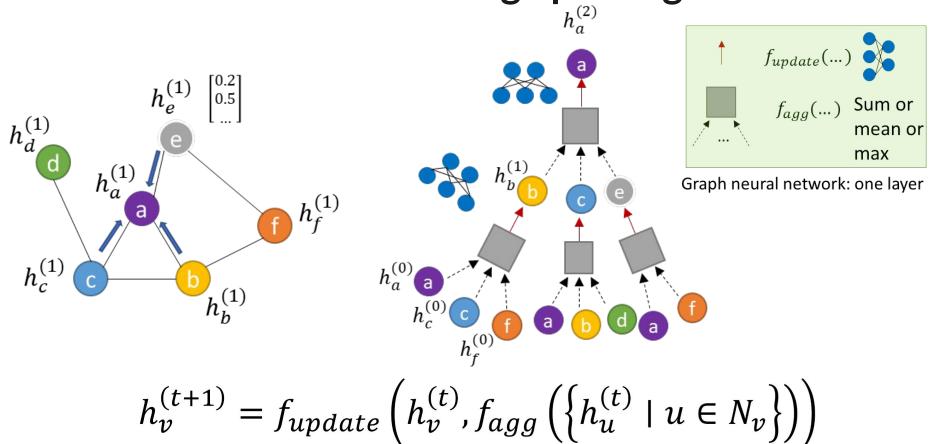
edge attr 
$$\in \mathbb{R}^?$$

- Adjacency matrix...?

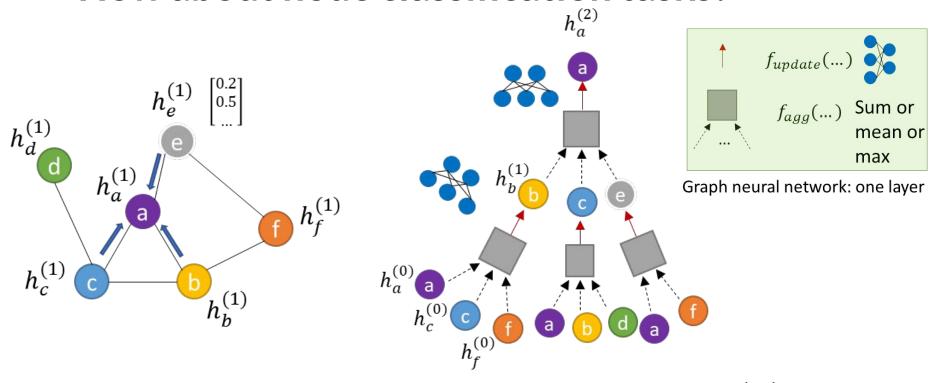
Edge list! 
$$edge_index \in \mathbb{Z}^{+?}$$

$$\mathcal{G}_1 = (\mathbf{X}_1, \mathbf{A}_1)$$
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Most GNNs use a message passing scheme

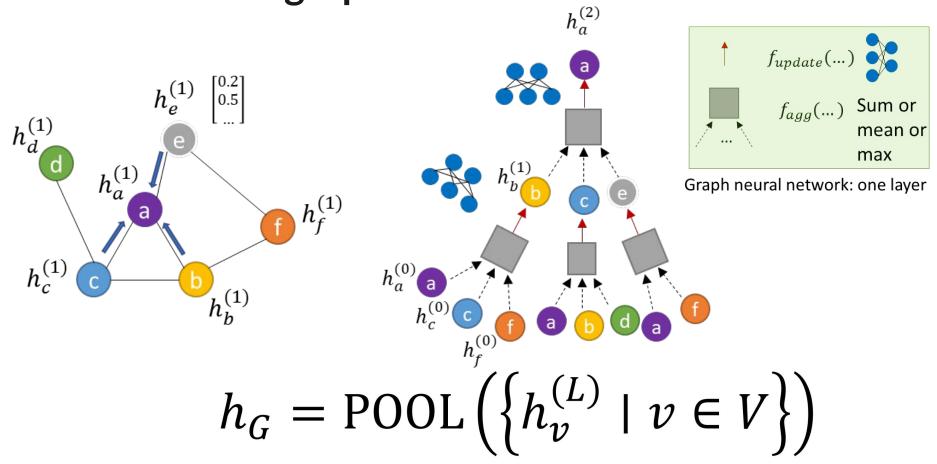


How about node classification tasks?



$$y_v = Softmax(MLP(h_v^{(L)}))$$

How about graph classification tasks?



Create your own GNNs!

$$h_v^{(t+1)} = f_{update}\left(h_v^{(t)}, f_{agg}\left(\left\{h_u^{(t)} \mid u \in N_v\right\}\right)\right)$$

# **Q&A**