Graph Convolutional Networks



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PhD student

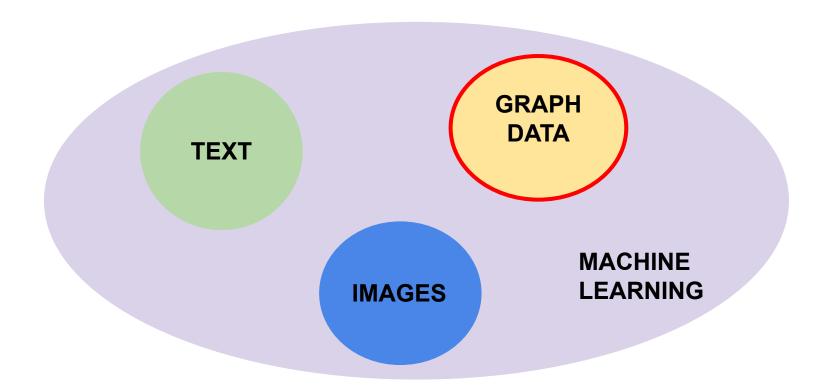


What is a GRAPH?

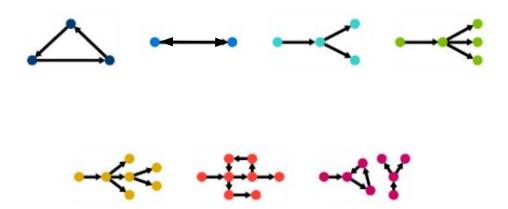
A Graph is anything with nodes connected by edges.



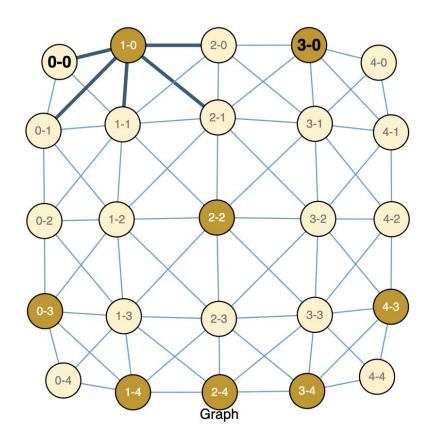
What are Graph Neural Networks?



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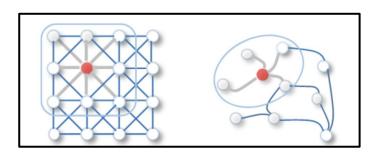


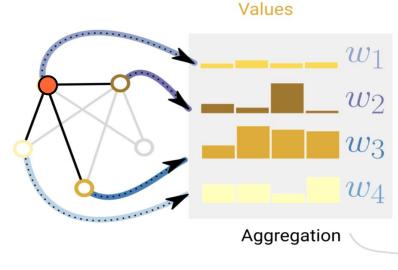
Graph Convolutional Networks (GCN)



- SAME GOAL of the convolution
- PIXELS = RGB values
- LINKS = spatial connections

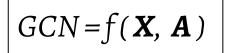
Graph Attention Networks (GAT)

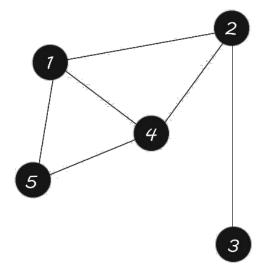




Updated node

How we define a Graph Convolutional Network?



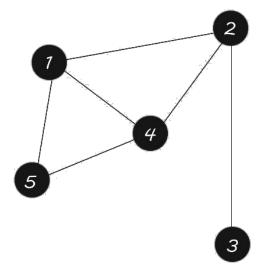


$$A = \begin{bmatrix} 7 & 2 & 3 & 4 & 5 \\ 7 & 0 & 1 & 0 & 1 & 1 \\ 2 & 1 & 0 & 1 & 1 & 0 \\ 3 & 0 & 1 & 0 & 0 & 0 \\ 4 & 1 & 1 & 0 & 0 & 1 \\ 5 & 1 & 0 & 0 & 1 & 0 \end{bmatrix}$$

X = input features

How we define a Graph Convolutional Network?

$$GCN = f(\mathbf{X}, \mathbf{A})$$



$$A = \begin{bmatrix} 7 & 0 & 1 & 0 & 1 & 1 \\ 2 & 1 & 0 & 1 & 1 & 0 \\ 3 & 0 & 1 & 0 & 0 & 0 \\ 4 & 1 & 1 & 0 & 0 & 1 \\ 5 & 1 & 0 & 0 & 1 & 0 \end{bmatrix}$$

$$X = \begin{bmatrix} 7 & 0 & \dots & 1 \\ 2 & 1 & \dots & 0 \\ 3 & 0 & \dots & 0 \\ 4 & 1 & \dots & 1 \\ 5 & 1 & \dots & 0 \end{bmatrix}$$
K dimensions

8

Defined in: https://arxiv.org/pdf/1609.02907.pdf

New layer
$$ightarrow$$
 $H(1)$ Previous layer $ightarrow$ $H(0)$
$$H^{(l+1)} = \sigma(\hat{D}^{-\frac{1}{2}}\hat{A}\hat{D}^{-\frac{1}{2}}H^{(l)}W(l))$$
 Weights we want to learn

For the first layer, we do not have previous layer... so, what do you think it will be H(0)?

$$H(0) == X$$

D = degree matrix

Left normalization

 \Rightarrow

 $\hat{D}^{-1}\hat{A}$

Symmetric normalization

$$H^{(l+1)} = \sigma(\hat{D}^{-\frac{1}{2}}\hat{A}\hat{D}^{-\frac{1}{2}}H^{(l)}W(l))$$

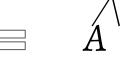
Symmetric normalization

D = degree matrix

$$H^{(l+1)} = \sigma(\hat{D}^{-\frac{1}{2}}\hat{A}\hat{D}^{-\frac{1}{2}}H^{(l)}W(l))$$

$$A = \begin{bmatrix} 7 & 2 & 3 & 4 & 5 \\ 7 & 0 & 1 & 0 & 1 & 1 \\ 2 & 1 & 0 & 1 & 1 & 0 \\ 3 & 0 & 1 & 0 & 0 & 0 \\ 4 & 1 & 1 & 0 & 0 & 1 \\ 5 & 1 & 0 & 0 & 1 & 0 \end{bmatrix}$$





D = degree matrix



Symmetric normalization

$$H^{(l+1)} = \sigma(\hat{D}^{-\frac{1}{2}}\hat{A}\hat{D}^{-\frac{1}{2}}H^{(l)}W(l))$$

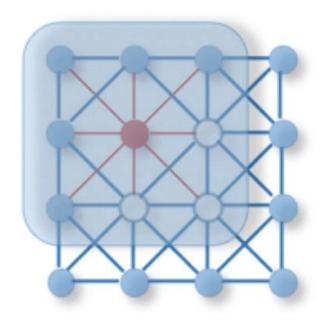
X = input features

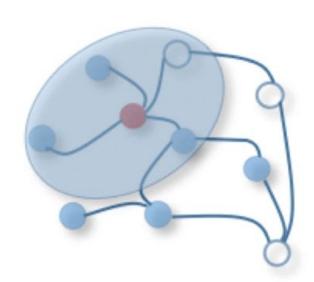
$$\hat{A} = \begin{bmatrix}
 1 & 2 & 3 & 4 & 5 \\
 1 & 1 & 0 & 1 & 1 \\
 2 & 1 & 1 & 1 & 1 & 0 \\
 3 & 0 & 1 & 1 & 0 & 0 \\
 4 & 1 & 1 & 0 & 1 & 1 \\
 5 & 1 & 0 & 0 & 1 & 1
\end{bmatrix}$$

$$D^{\wedge} = \begin{bmatrix} & 1 & 2 & 3 & 4 & 5 \\ & 4 & 0 & 0 & 0 & 0 \\ & 2 & 0 & 4 & 0 & 0 & 0 \\ & 3 & 0 & 0 & 2 & 0 & 0 \\ & 4 & 0 & 0 & 0 & 4 & 0 \\ & 5 & 0 & 0 & 0 & 0 & 3 \end{bmatrix}$$

Analogy with Image Convolution

$$H^{(l+1)} = \sigma(\hat{D}^{-\frac{1}{2}}\hat{A}\hat{D}^{-\frac{1}{2}}H^{(l)}W(l))$$





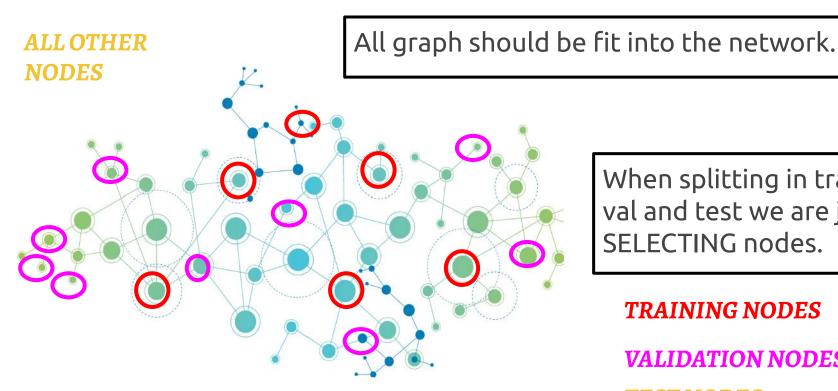
GCN: generating predictions

All graph should be fit into the network.

When splitting in train, val and test we are just SELECTING nodes.

TRAINING NODES

GCN: generating predictions



When splitting in train, val and test we are just SELECTING nodes.

TRAINING NODES

VALIDATION NODES **TEST NODES**

IMPORTANT

We just have ONE GRAPH, even we have 3 different sets.

Kick off the lab

- 1. Launch a web browser (Chrome recommended).
- 2. Login to a Google account. Create a new one if preferred.
- 3. Two options:
 - a. Create a new EMPTY notebook in <u>Google Colab</u> pressing <u>this link</u>.
 - b. You can copy the live code from this Jupyter notebook.
- 4. Change runtime type to work with **GPU**.

