



# **Structural Analysis with Graph Neural Network**

## 圖神經網路於結構分析之應用

周遠同

2022/01/05

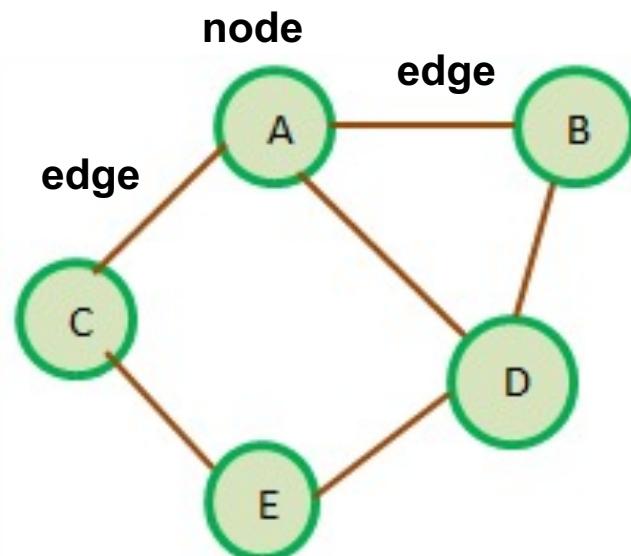
[r10521608@ntu.edu.tw](mailto:r10521608@ntu.edu.tw)

# Outline

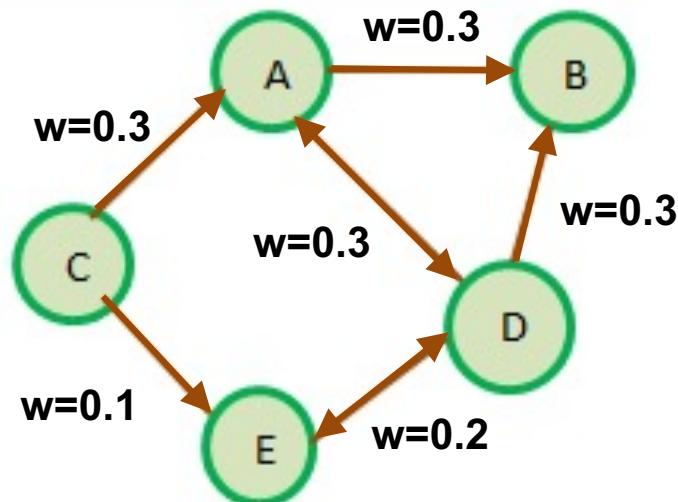
- **Graph Neural Network**
  - ✓ **What**
  - ✓ **Why**
  - ✓ **How**
- **Structural Analysis with GNN**
  - ✓ **What**
  - ✓ **Why**
  - ✓ **How**

# What is graph?

- data structure

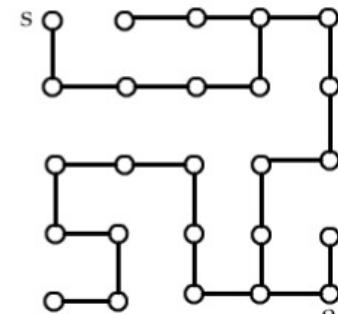
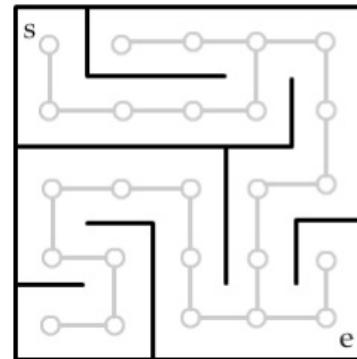
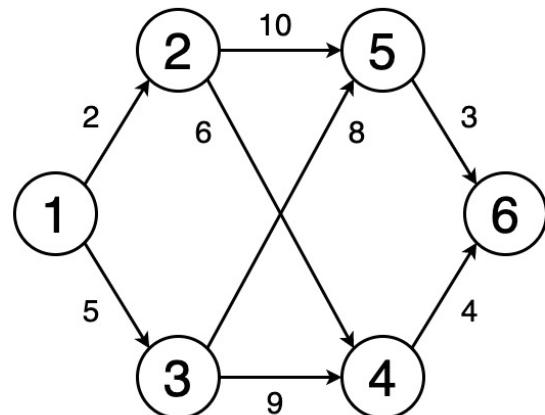


- **directed**
- **weighted**



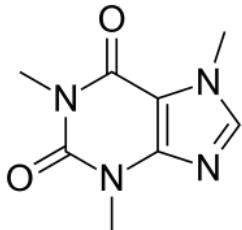
# What is graph?

- Solve optimized problems
  - Travelling salesman problem (Dijkstra's shortest path algorithm)
  - Mazes (depth-first search)

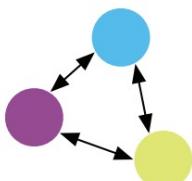


# What is graph?

- These can also be graph!!!

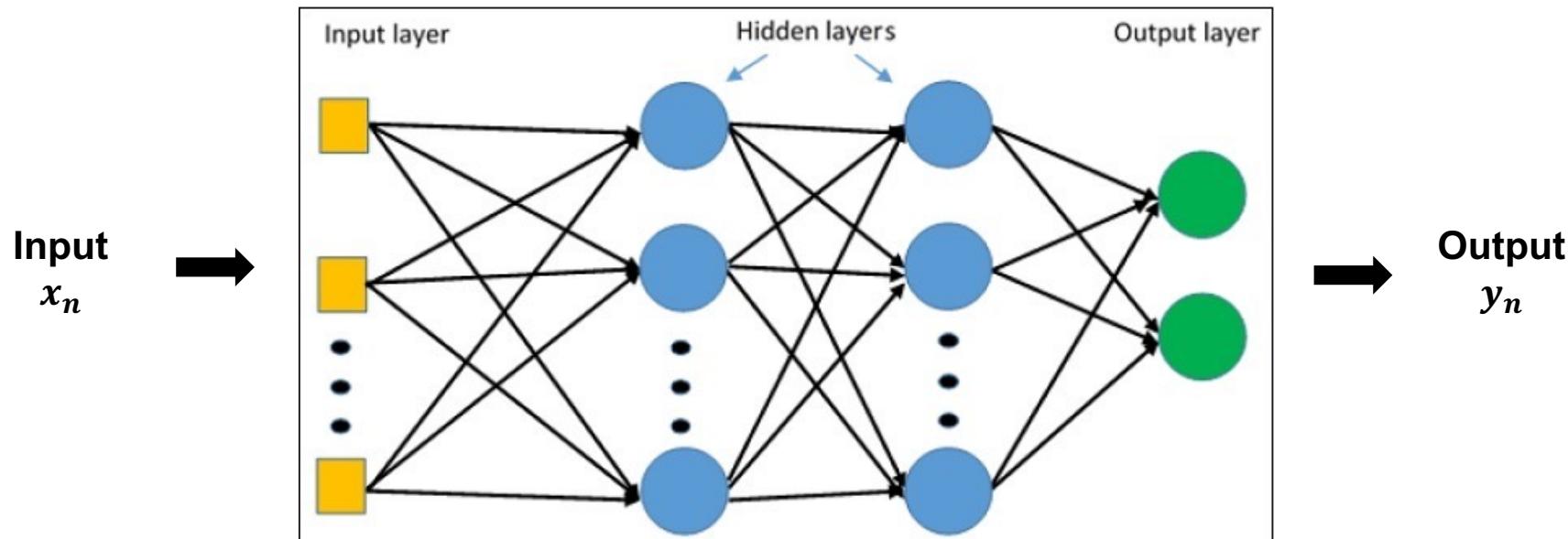


*n*-body System



# Why need Graph Neural Network (GNN)?

- We have just learned Neural Networks (MLP, CNN, .....)



# Why GNN?

- Consider the case, who stole Nobita's iPhone?



10歲  
胖  
成績普通  
貪吃



10歲  
瘦  
成績很好  
喜歡洗澡



10歲  
瘦  
成績超好  
???



10歲  
瘦  
成績不錯  
有錢



10歲  
瘦  
成績差  
常被欺負



???歲  
胖  
成績???  
機器貓

# Why GNN?

- We can train a binary classifier

$$f( \text{ (image of Nobita) } ) = False$$

$$f( \text{ (image of Doraemon) } ) = False$$

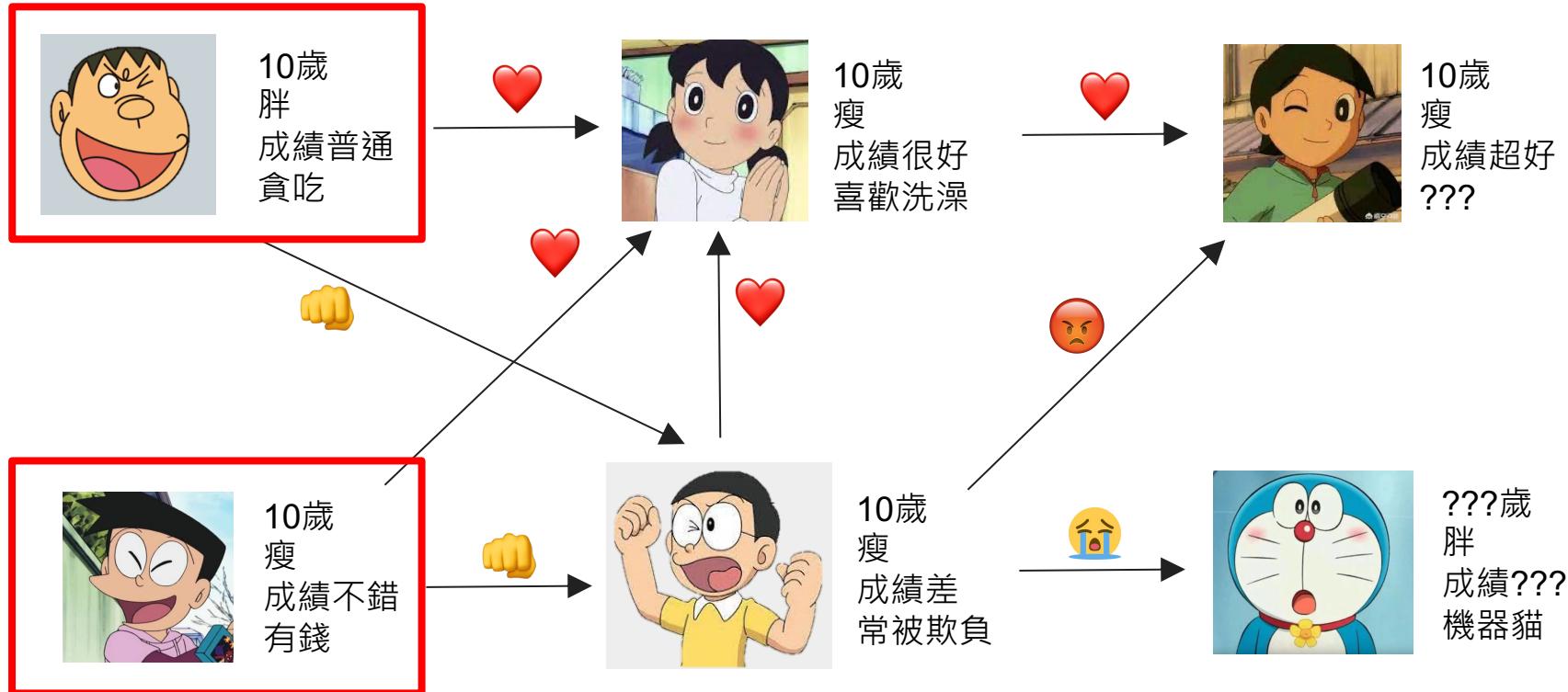
$$f( \text{ (image of Shizuka) } ) = False$$

$$f( \text{ (image of Chota) } ) = ???$$

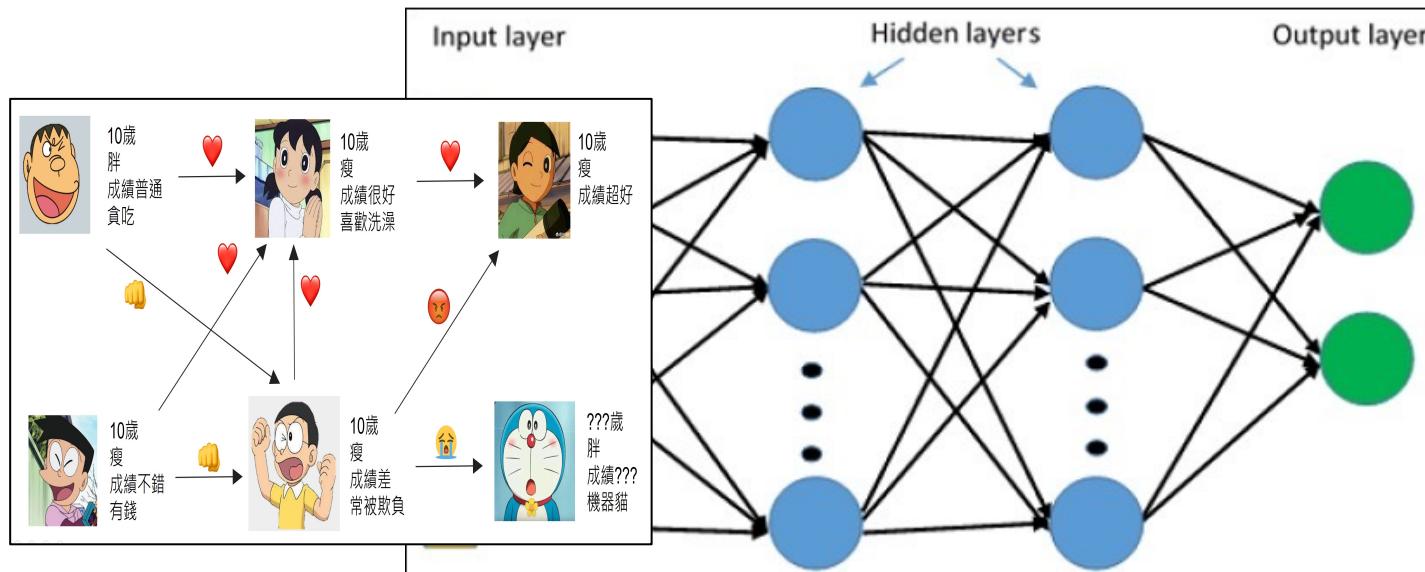
- It's not easy to predict by only using their personal features

# Why GNN?

- Missing something? Their relationship!



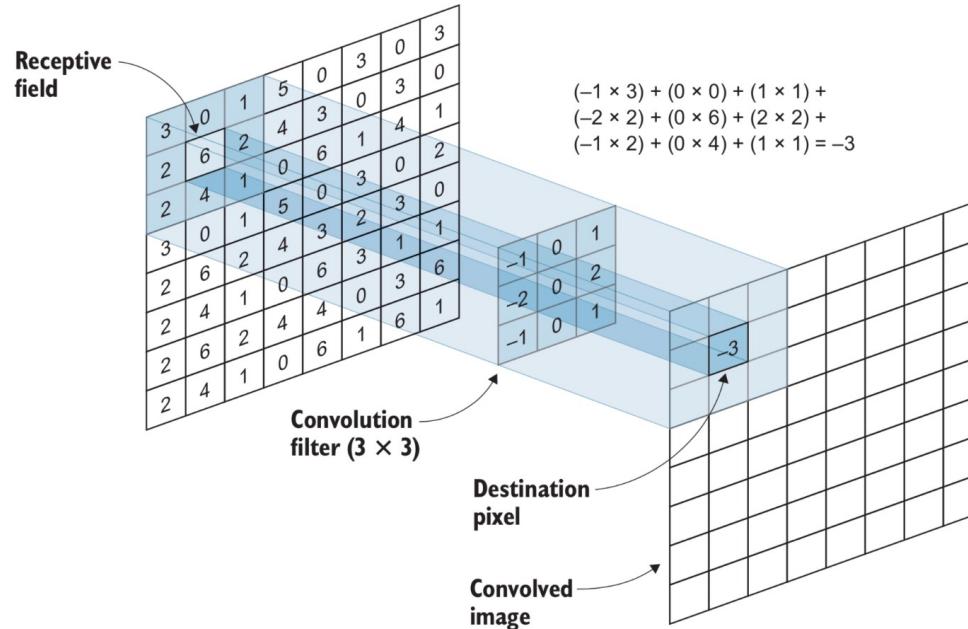
# Why GNN?



- Node clf
- Node reg
- Edge clf
- Edge reg
- Graph clf
- Graph reg
- Graph gen

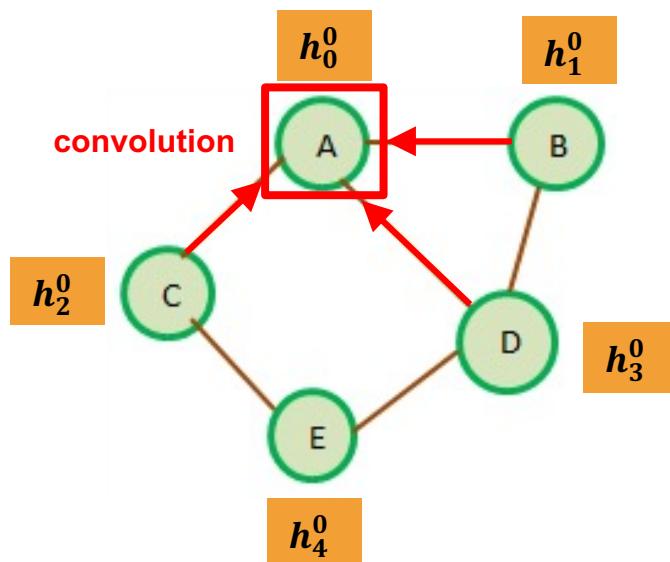
# How to consider relationship between nodes??

- In CNN, we use convolution to get the local feature

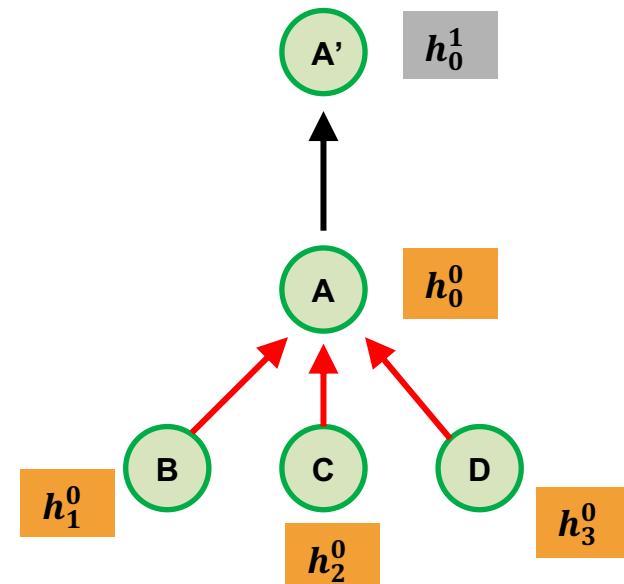


# How to consider relationship between nodes??

- In graph, we can also use convolution!



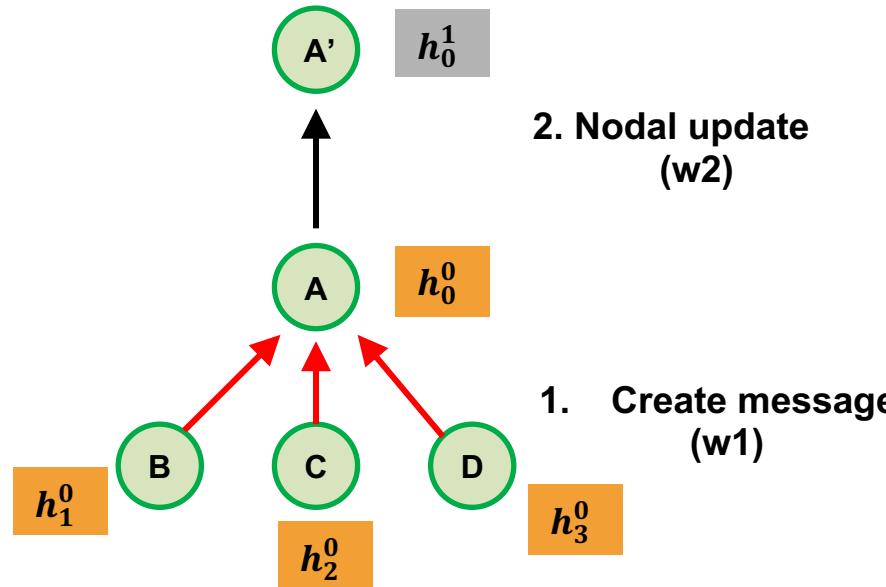
For node A,



Convolution → Message Passing

# How to consider relationship between nodes??

- Message passing is done in two stages:



1.

$$m_{0,1}^0 = w_1 (h_0^0 + h_1^0)$$

$$m_{0,2}^0 = w_1 (h_0^0 + h_2^0)$$

$$m_{0,3}^0 = w_1 (h_0^0 + h_3^0)$$

$$m_0^0 = \text{aggr} (m_{0,1}^0, m_{0,2}^0, m_{0,3}^0)$$

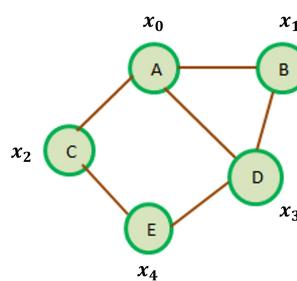
$\text{aggr} = \{\text{max}, \text{add}, \text{mean}\}$

2.

$$h_0^1 = w_2 (h_0^0 + m_0^0)$$

# Graph Neural Networks

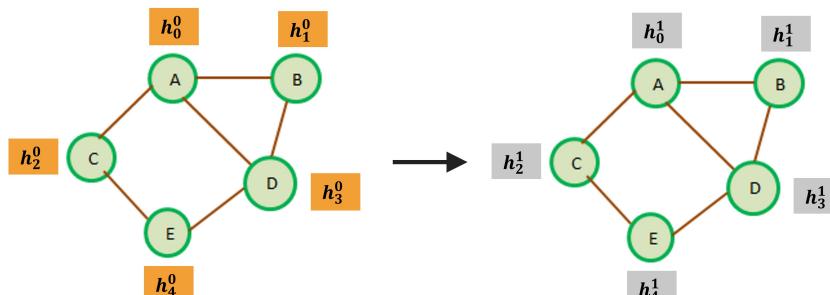
$$\begin{array}{ccc} \text{Node feat.} & \xrightarrow{\text{transform}} & \{h_i^0\} \\ \text{Edge feat.} & \xrightarrow{\text{transform}} & \{e_{ij}^0\} \end{array}$$



$$m_{i,j}^t = \varphi_{message}(h_i^t, h_j^t)$$

$$m_i^t = aggr(m_{i,0}^t, m_{i,1}^t, \dots, m_{i,N}^t)$$

$$h_i^{t+1} = \varphi_{update}(h_i^t, m_i^t)$$



Input

Message Passing Layers

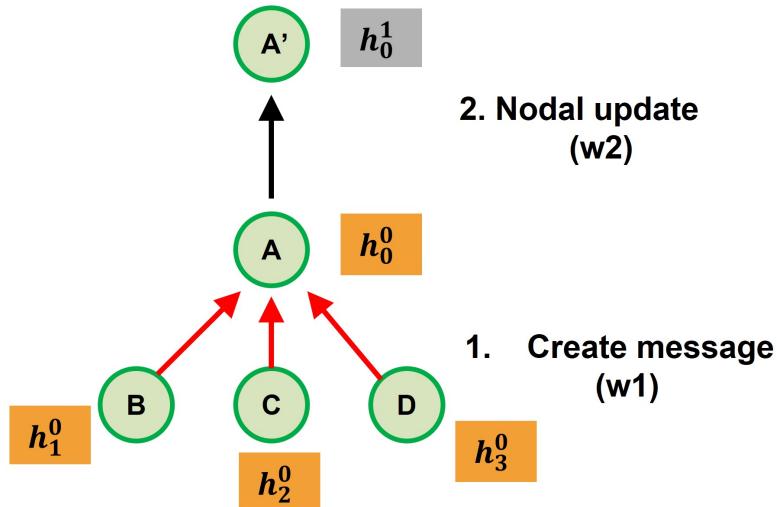
Output

$$h_i^L \xrightarrow{\text{MLP}} \text{Node Predictions}$$

$$\frac{1}{\mathcal{V}} \sum_{i=0}^{\mathcal{V}} h_i^L \xrightarrow{\text{MLP}} \text{Graph Prediction}$$

$$\text{Concat}(h_i^L, h_j^L) \xrightarrow{\text{MLP}} \text{Edge Predictions}$$

# What is the message, update function?



Different message passing mechanism

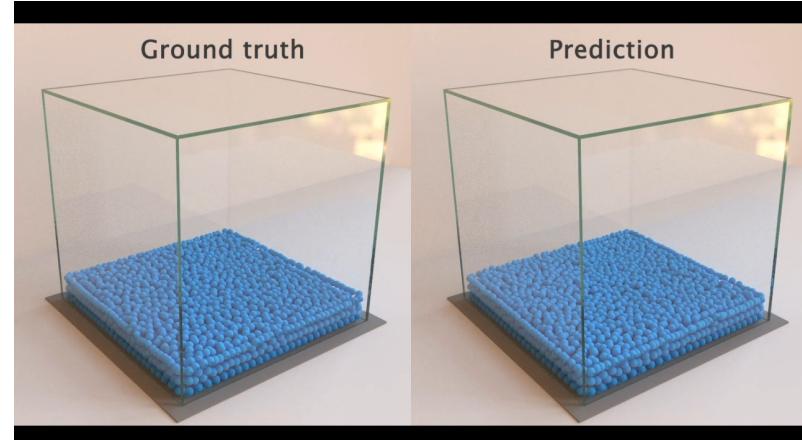
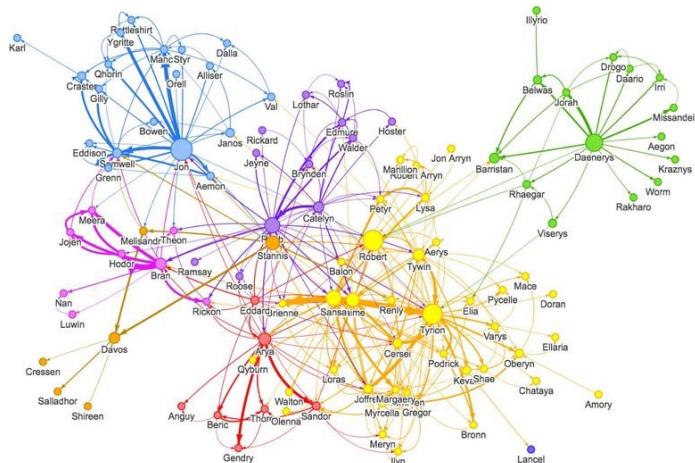
- GCN (Graph Convolution Network)
- GAT (Graph Attention Network)
- GIN (Graph Isomorphism Network)
- GN (Graph Network)

⋮

<https://arxiv.org/pdf/1609.02907.pdf>  
<https://arxiv.org/pdf/1710.10903.pdf>  
<https://arxiv.org/pdf/1810.00826.pdf>  
<https://arxiv.org/pdf/1806.01261.pdf>

# What can message passing do?

- Pass k-hop by using k layers
- Interaction between nodes
  - Simulate physics

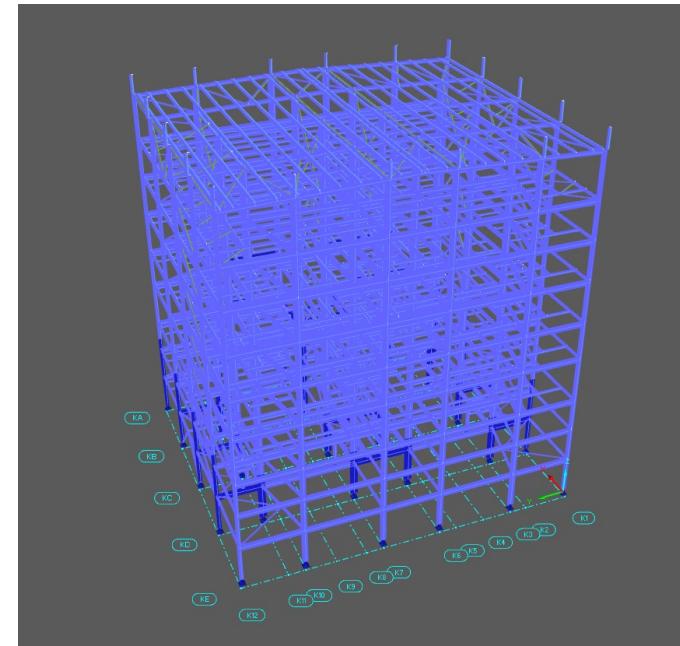


# Structural Analysis with GNN?



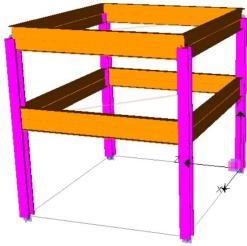
# Why ML in structural analysis?

- It's time-consuming
- Have to wait several hours to know the output response

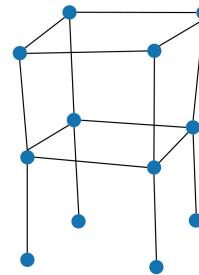


# Represent structure as graph

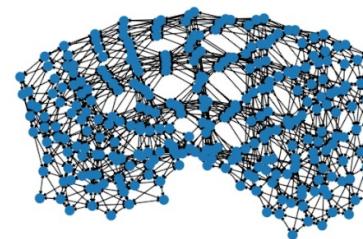
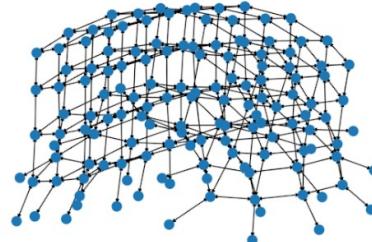
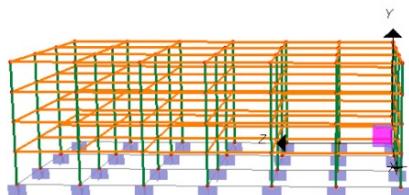
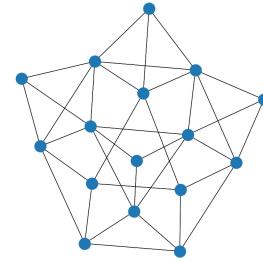
- Structure



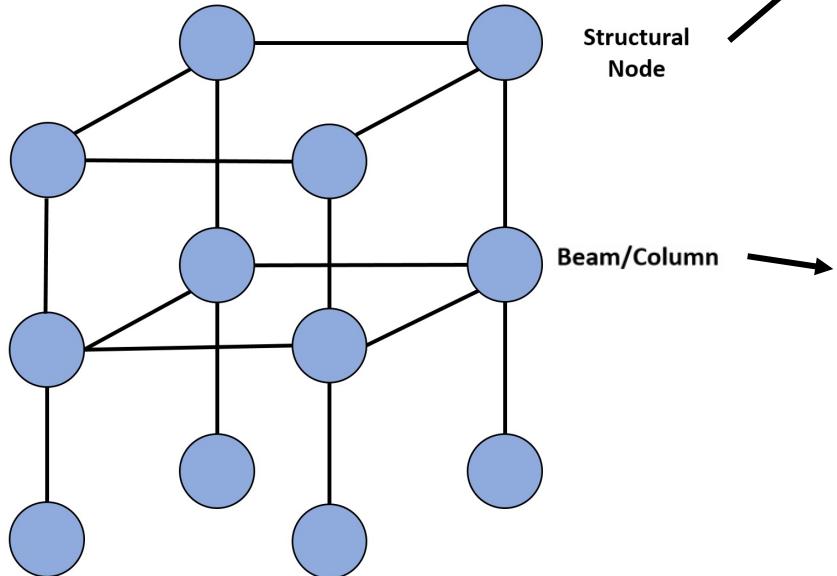
- NodeAsNode



- ElemAsNode



# Node feature and output labels



## Node Features:

- span/story number of structure
- nodal grid coordinates (x, y, z)
- degree of freedom
- nodal mass
- load

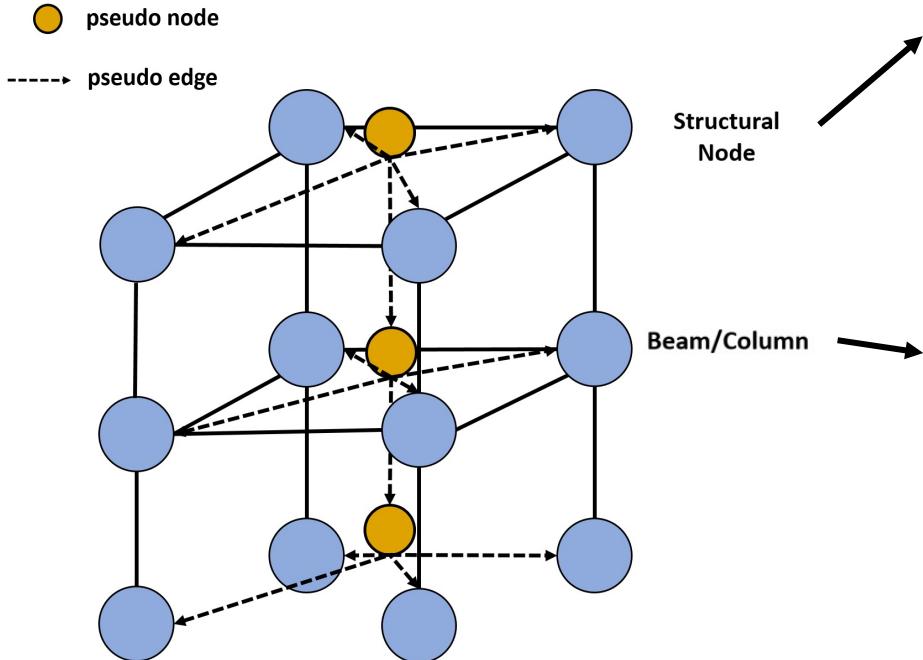
## Edge Attributes:

- beam or column
- beam/column length

## Node Outputs:

- displacement
- bending moment
- shear force

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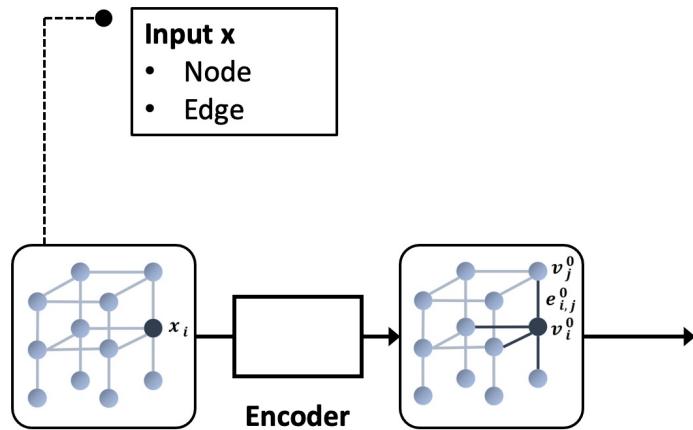
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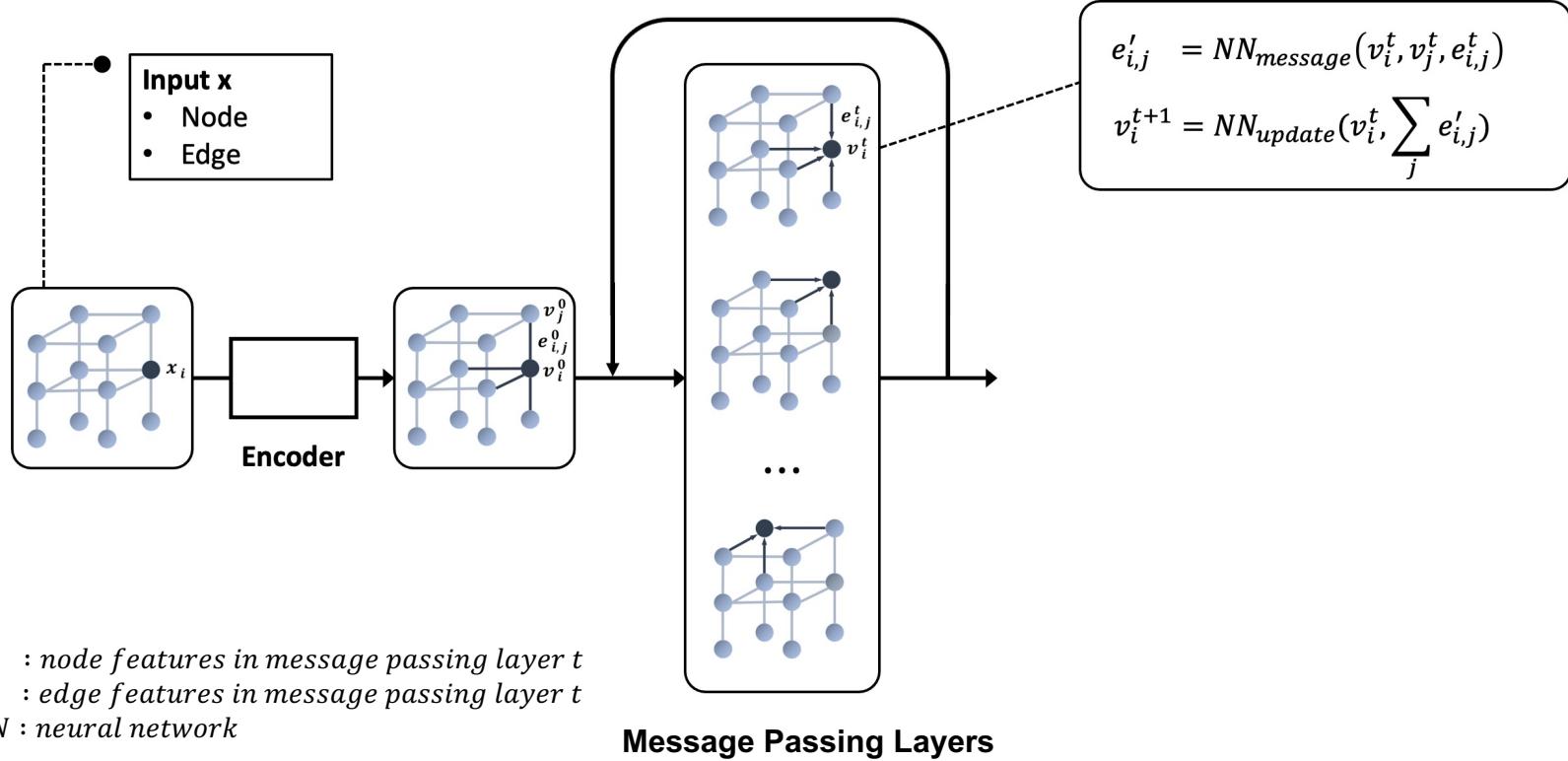
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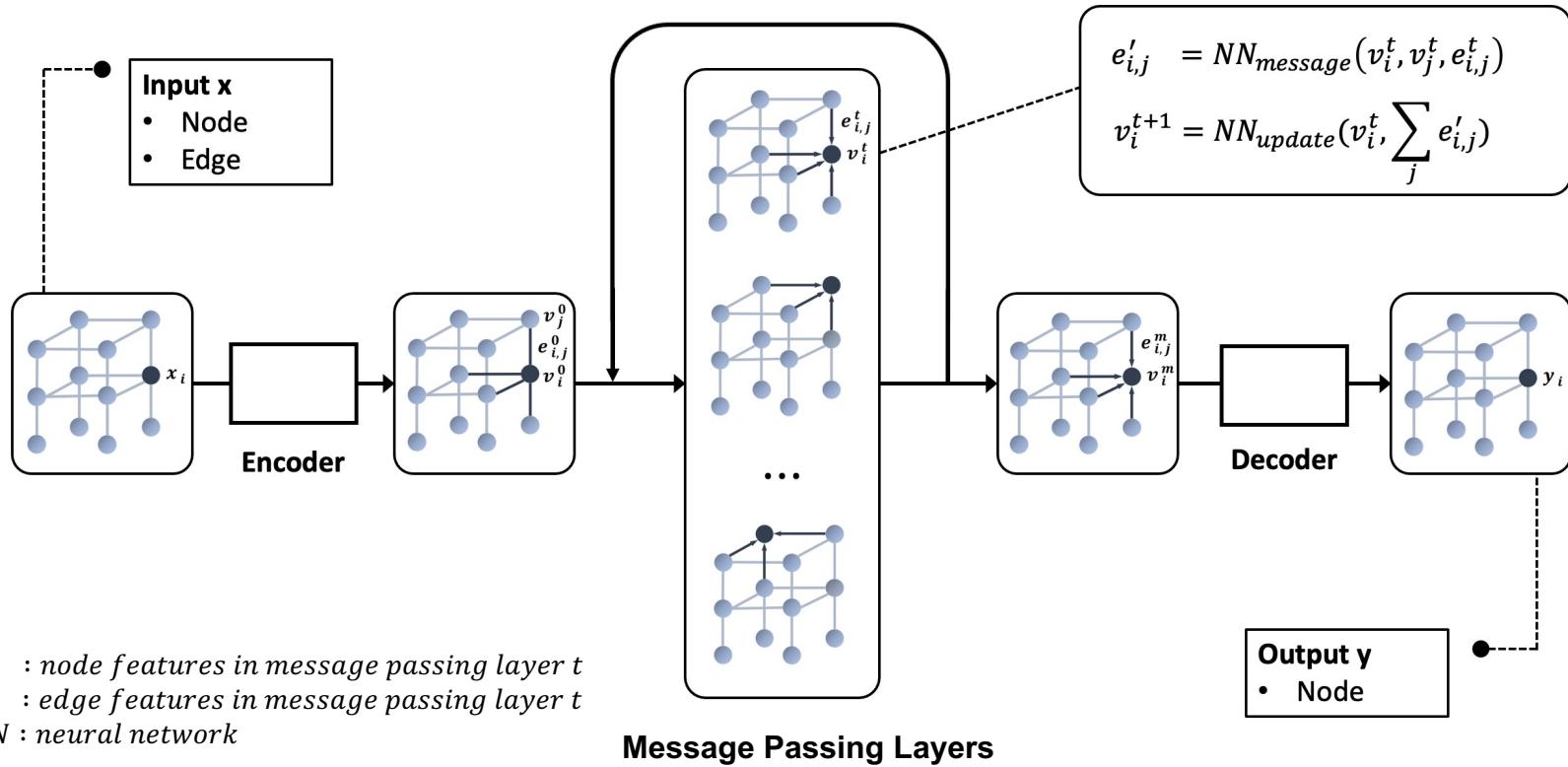
# GNN Architecture



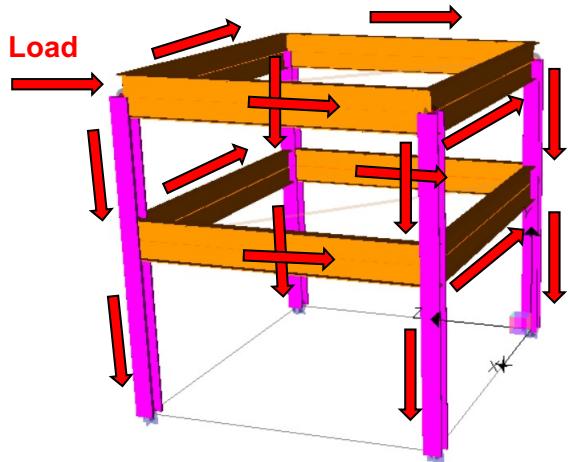
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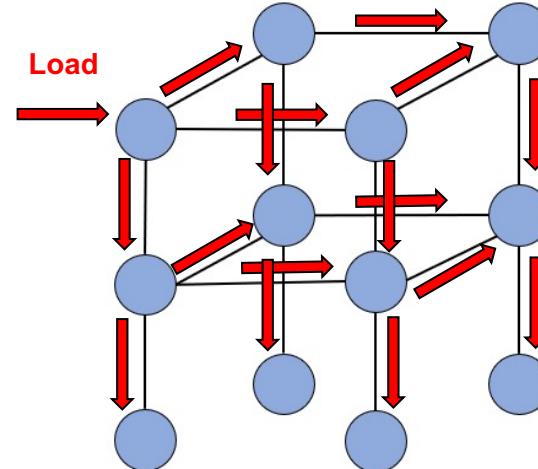
# GNN Architecture



# Force Transmission vs. Message Passing



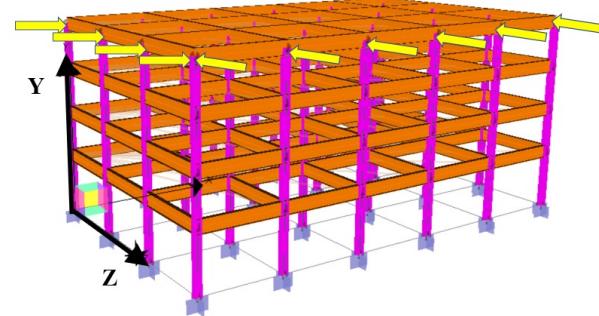
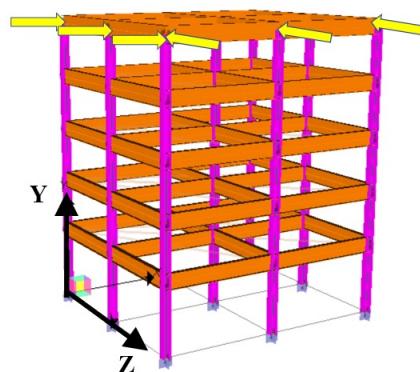
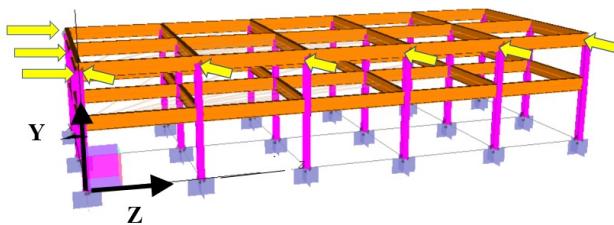
Structure



Graph

# Data Generation

- random structure



# Linear Static Analysis

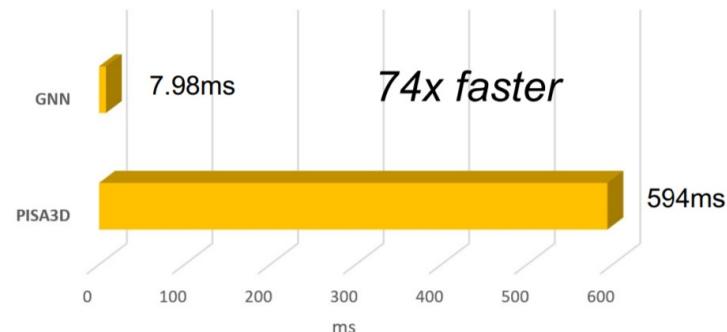
*Structural Load → Displacement, Bending Moment, Shear Force*

Model	Pseudo Nodes	$\delta_X$	$\delta_Y$	$M_Y$	$M_Z$	$V_Y$	$V_Z$
GCN	True	97.04	96.82	90.80	90.18	87.42	90.83
GAT	True	97.94	98.26	92.12	91.18	90.02	92.22
GIN	True	98.73	98.44	98.93	98.66	96.92	98.90
StrucGNN	True	<b>99.31</b>	<b>99.42</b>	<b>99.40</b>	<b>99.29</b>	<b>99.19</b>	<b>99.43</b>

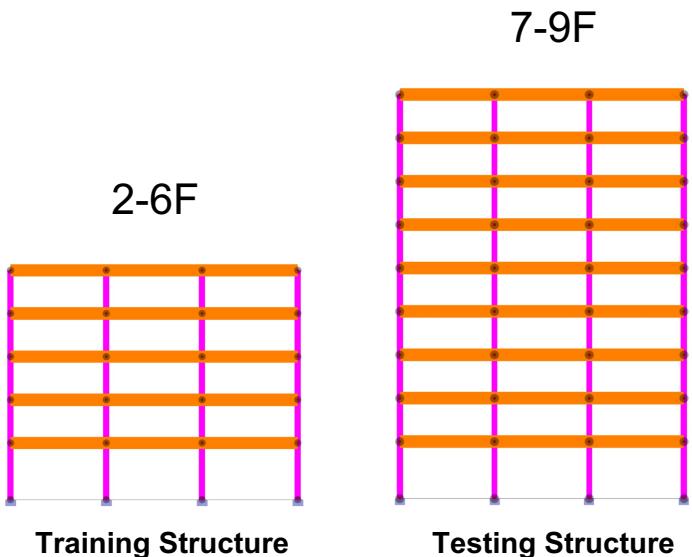
Table 3: Training performance on 2-6 story (relative accuracy, %)

Method	Total Time	Total Iterations	Time / Iteration
PISA3D	9m54s	1000	0.594s
GNN	7.98s	1000	0.00798s

Table 2: Time comparison using PISA3D and GNN



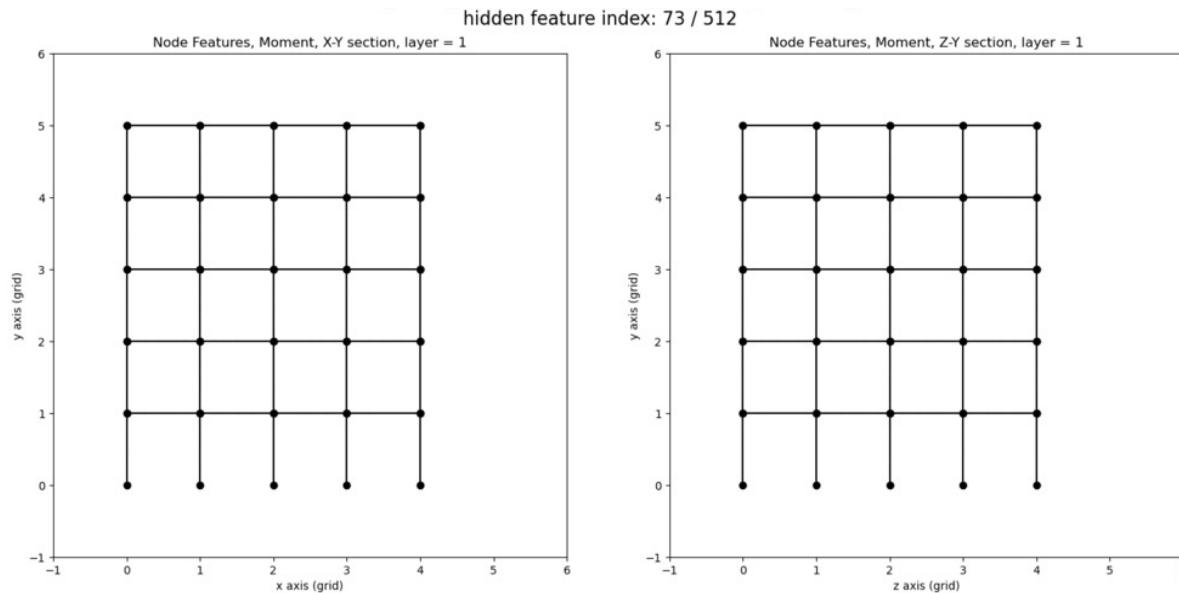
# Generalizability



Model	Pseudo Nodes	$\delta_X$	$\delta_Y$	$M_Y$	$M_Z$	$V_Y$	$V_Z$
GCN	True	85.92	85.01	77.42	76.98	75.03	77.37
GAT	True	89.81	91.24	79.28	81.88	77.03	79.38
GIN	True	93.58	93.12	88.81	90.42	86.01	89.53
<b>StrucGNN</b>	<b>True</b>	<b>97.41</b>	<b>97.62</b>	<b>96.19</b>	<b>95.14</b>	<b>95.13</b>	<b>96.50</b>

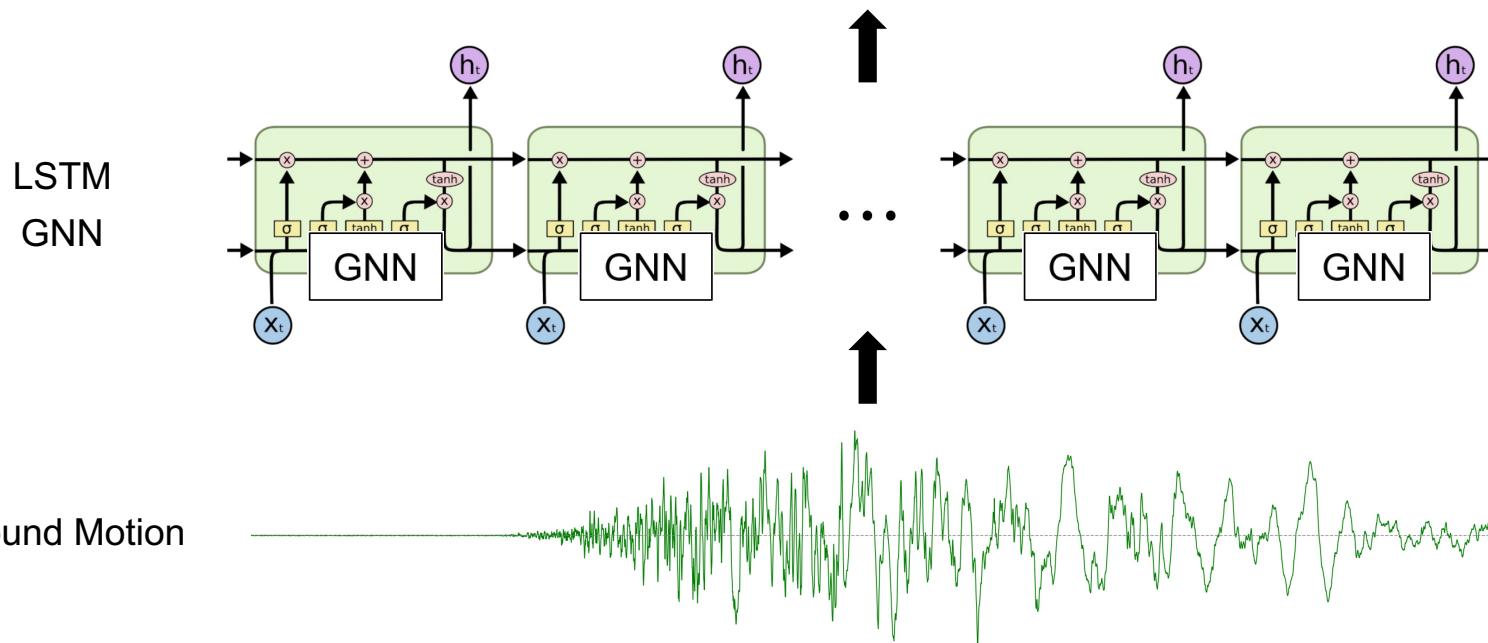
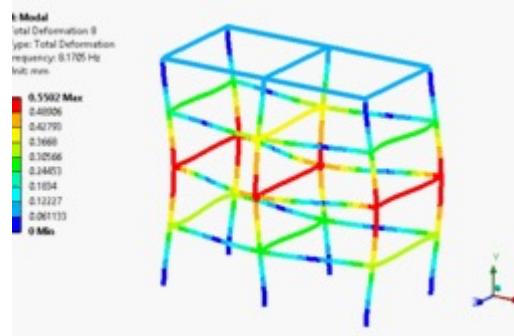
Table 5: Generalizability performance on 7-9 story (relative accuracy, %)

# GNN: An Iterative Method



# What else?

Response



# Summary

GNN is

- mature
- widely used in many field
- straightforward and explainable
- Fun!

# Thanks for your attention!

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