

# Structure Recognition with Graph Neural Networks

A project for the lab-course  
Advanced Projects in Computational Physics 2



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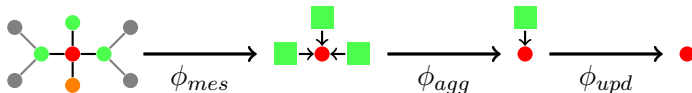
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## Theoretical Background - Message Passing

### Overview:

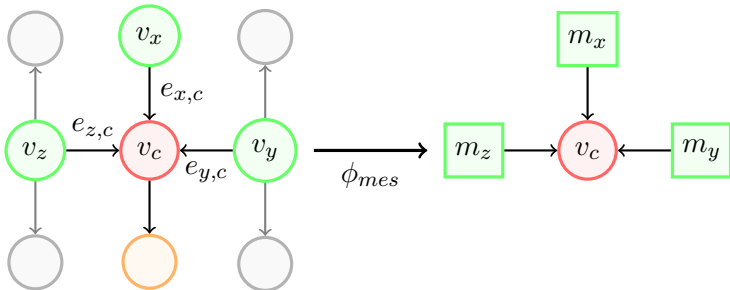


### Message passing consists of three steps:

1. Computing messages ( $\phi_{mes}$ )
2. Aggregating messages ( $\phi_{agg}$ )
3. Updating node values ( $\phi_{upd}$ )

## Theoretical Background - Message Passing

### Step 1: Compute Messages

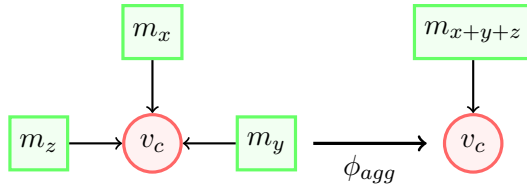


For  $i \in \{x, y, z\}$  calculate:

$$m_i := \phi_{mes}(v_c, v_i, e_{i,c})$$

## Theoretical Background - Message Passing

### Step 2: Aggregate Messages

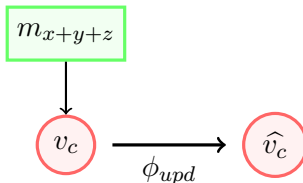


Calculate total message:

$$m_{x+y+z} := \phi_{agg}(m_x, m_y, m_z)$$

## Theoretical Background - Message Passing

### Step 3: Update node value



Calculate new node value:

$$\hat{v}_c := \phi_{upd}(v_c, m_{x+y+z})$$

**Question:** What exactly are  $\phi_{mes}$ ,  $\phi_{agg}$ ,  $\phi_{upd}$ ?

## Theoretical Background - Message Passing

**Common examples for  $\phi_{mes}, \phi_{agg}, \phi_{upd}$ :**

	$\phi_{mes}$	$\phi_{agg}$	$\phi_{upd}$
GCNConv ([2])	?	?	
GINEConv ([1])	?	?	

In general,  $\phi_{mes}, \phi_{agg}, \phi_{upd}$  can be anything

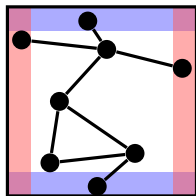
## Theoretical Background - Percolation

Given are

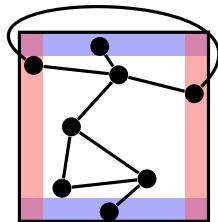
- ▶ graph  $G = (V, E)$
- ▶ one position  $(n_x, n_y) \in [0, 1] \times [0, 1]$  for each node  $n \in V$
- ▶  $0 < r < \frac{1}{2}$

$G$  is called percolating, if there are nodes  $n, m \in V$  such that

1. there is a cycle containing  $n$  and  $m$ ,
2.  $(n, m) \in E$ ,
3. either  $n_x < r$  and  $m_x > 1 - r$  or  $n_y < r$  and  $m_y > 1 - r$ .



(a) non-percolating



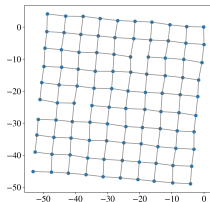
(b) percolating



## Goals

### Experiment 1

**Input:** Bravais lattice (in 2d, 3d), e.g.:



**Expected output:** Bravais class, e.g. square

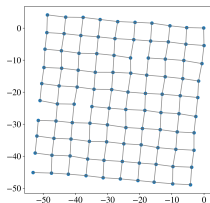
**Question:** What are suitable choices for  $\phi$ 's?

### Experiment 2

## Goals

### Experiment 1

**Input:** Bravais lattice (in 2D, 3D), e.g.:

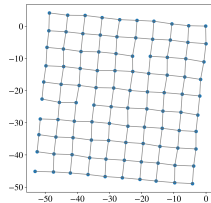


**Expected output:** Bravais class, e.g. square

**Question:** What are suitable choices for  $\phi$ 's?

### Experiment 2

**Input:** Graph inside unit-square, e.g.:



**Expected output:** percolating or not

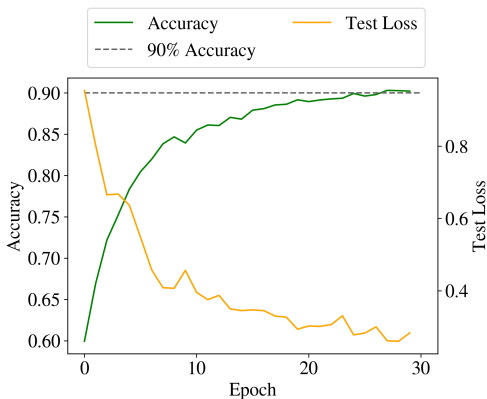
**Question:** Can this problem be solved by a GNN?

## Results - Bravais Lattices 2D (Experiment 1)

which parameters have been varied, what data looks like

## Results - Bravais Lattices 2D (Experiment 1)

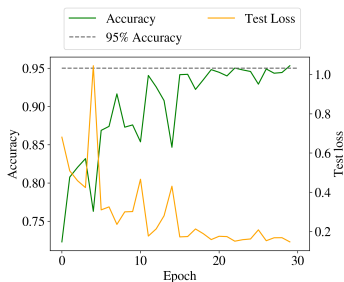
**Averaged performance (over all 36 models):**



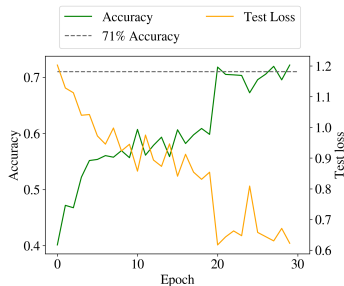
**Question:** Do different models lead to different accuracies?

## Results - Bravais Lattices 2D (Experiment 1)

### Best and worst performing models:



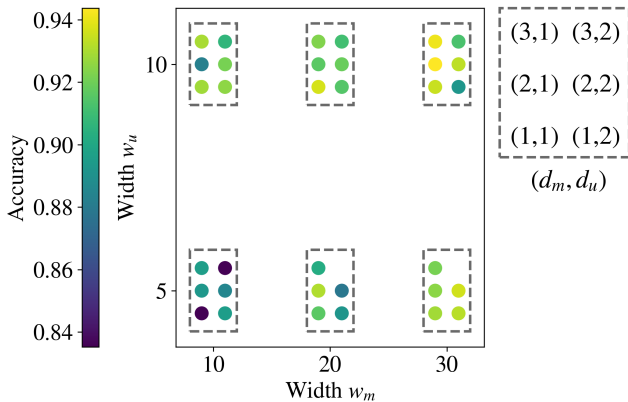
(a) Best performing model  
 $((d_m, w_m, d_u, w_u) =$   
 $(2, 30, 1, 10)).$



(b) Worst performing model  
 $((d_m, w_m, d_u, w_u) = (3, 30, 2, 5))$

## Results - Bravais Lattices 2D (Experiment 1)

Correlation between  $w_m$  and  $w_u$ :

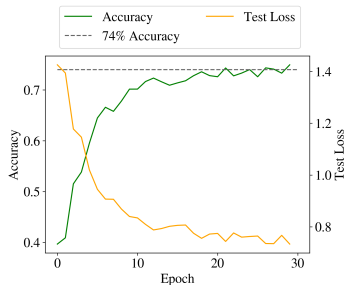


## Results - Bravais Lattices 3D (Experiment 1)

**Training of best and worst performing models on 3D dataset:**



**(a)** Best performing model  
 $((d_m, w_m, d_u, w_u) =$   
 $(2, 30, 1, 10)).$



**(b)** Worst performing model  
 $((d_m, w_m, d_u, w_u) = (3, 30, 2, 5))$

## Results - Percolation (Experiment 2)

why it is not possible to detect if two nodes are connected, what data looks like

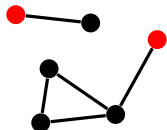
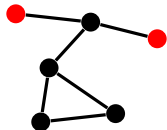


## Results - Percolation (Experiment 2)

**Claim:** GNN can solve percolation problem  $\implies$  GNN can solve connection problem

### Procedure:

1. Start with arbitrary Graph  $G$ , select two nodes  $n, m$

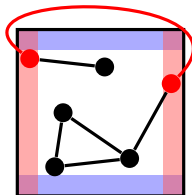
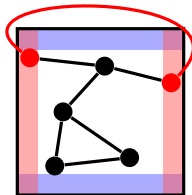


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1. Start with arbitrary Graph  $G$ , select two nodes  $n, m$
2. Place  $G$  inside unit square, move  $n, m$  to edges, add edge  $(n, m) \implies$  new graph  $\tilde{G}$

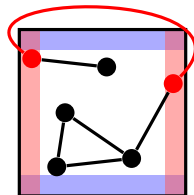
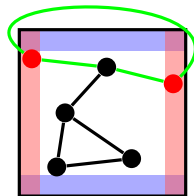


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3. Run GNN on  $\tilde{G}$   
 $\implies$  either  $\tilde{G}$  is percolating (i.e. there is a cycle containing  $n, m$  and the new edge) or not

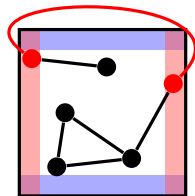
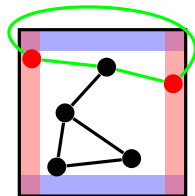


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3. Run GNN on  $\tilde{G}$
4.  $\tilde{G}$  percolating  $\iff n, m$  connected in  $G$



## Results - Percolation (Experiment 2)

ideas how to solve this problem top k pooling layer explained

## Results - Percolation (Experiment 2)

why we can not expect the layyer to work results (both with and without top k)



## Conclusion and Outlook



## Time for Discussion and Questions



## References

- [1] Weihua Hu et al. *Strategies for Pre-training Graph Neural Networks*. 2020. arXiv: 1905.12265 [cs.LG]. URL: <https://arxiv.org/abs/1905.12265>.
- [2] Thomas N. Kipf and Max Welling. *Semi-Supervised Classification with Graph Convolutional Networks*. 2017. arXiv: 1609.02907 [cs.LG]. URL: <https://arxiv.org/abs/1609.02907>.