

GNNs based on **local features** capture a logic heavily used in **Knowledge Representation**.

GNNs mixing **local + global** features can express First Order logic **with 2 variables and counting**.

## Logical Expressiveness of Graph Neural Networks

### Aggregate-Combine GNNs (AC-GNNs)

$$\mathbf{x}_v^{(i)} = \text{COM}^{(i)} \left( \mathbf{x}_v^{(i-1)}, \text{AGG}^{(i)} \left( \{\{\mathbf{x}_u^{(i-1)} \mid u \in \mathcal{N}_G(v)\}\} \right) \right)$$

features of node  $v$  at layer  $i$       features of node  $v$  at layer  $i-1$       multiset of features of  $v$  neighbors

combine node and neighbor features      aggregate  $v$ -neighbors (local) features

### ALCQ: a Description Logic to define concepts

**Example:** The following is an ALCQ formula

$$\text{Man} \sqcap \text{married.Doctor} \sqcap \exists^{\geq 3} \text{child.Happy}$$

defining “men married with a doctor that have at least three happy children”.

#### Theorem 1

A formula is captured by an **AC-GNN** if and only if it is expressible in **ALCQ**

### First Order logic with counting (FOC)

**Example:** A node  $v$  satisfies the FOC formula

$$\text{Red}(x) \wedge \exists y \left( \neg E(x, y) \wedge \exists^{\geq 2} x [E(y, x) \wedge \text{Blue}(x)] \right)$$

If and only if:  $v$  is red, and there is a node not connected with  $v$  that has at least two blue neighbors

**FOC- $k$ :** FOC with only  $k$  variables in every formula.

### ACR-GNNs: adding global features (Readouts)

$$\mathbf{x}_v^{(i)} = \text{COM}^{(i)} \left( \mathbf{x}_v^{(i-1)}, \text{AGG}^{(i)} \left( \{\{\mathbf{x}_u^{(i-1)} \mid u \in \mathcal{N}_G(v)\}\} \right), \text{READ}^{(i)} \left( \{\{\mathbf{x}_u^{(i-1)} \mid u \in G\}\} \right) \right)$$

considers the features of all nodes in the graph

#### Theorem 2

Every formula expressible in **FOC2** can be captured by an **ACR-GNN**

AC-GNNs cannot always implement FOC-2 formulas

#### Proposition

There are (infinite) **FOC-2** formulas that cannot be captured by **AC-GNNs**

### Results on synthetic data and FOC-2 formulas

	$\alpha_1$ Train	$\alpha_1$ Test		$\alpha_2$ Train	$\alpha_2$ Test	
		same-size	bigger		same-size	bigger
AC 10-layers	0.839	0.826	0.671	0.694	0.695	0.667
GIN 10-layers	0.567	0.566	0.536	0.689	0.693	0.672
ACR 1-layer	1.000	1.000	1.000	0.827	0.834	0.726
ACR 2-layers	1.000	1.000	1.000	0.895	0.897	0.770
ACR 3-layers	1.000	1.000	1.000	0.903	0.902	0.836