

# Conformal Inductive Graph Neural Networks

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### **TLDR**

- Conformal Prediction (CP) produces prediction sets guaranteed to include the true label with high probability.
- In inductive GNNs the graph evolution breaks the CP guarantee.
- We restore the guarantee under node and edge exchangeablility.

### **Motivation**

- Quantification of uncertainty is crucial yet hard for graphs.
- Model softmax is uncalibrated, therefore unreliable.
- Conformal sets are interpretable and reliable.

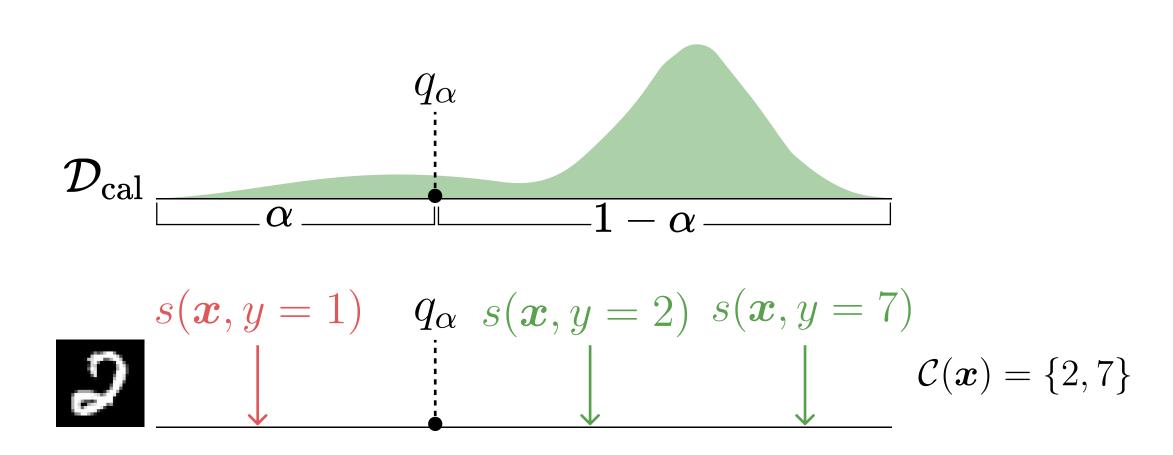
### **Standard Conformal Prediction**

Given Black-box model  $f(\cdot)$ , holdout and exchangeable calibration set  $\mathcal{D}_{\operatorname{cal}} = \{(\boldsymbol{x}_i, y_i)\}_{i=1}^n$ , user-specified  $1-\alpha$ , and new input  $\boldsymbol{x}_{n+1}$ . Algorithm With  $q_{\alpha} := \operatorname{Quant}(\alpha; \{s(\boldsymbol{x}_i, y_i)\}_{i=1}^n; 1)$ , define prediction sets

$$C_{\alpha}(\boldsymbol{x}_{\text{test}}) = \{ y : s(\boldsymbol{x}_{\text{test}}, y) \geq q_{\alpha} \}$$

where Quant  $(\cdot; \cdot; \cdot)$  is the quantile function, and  $s(\cdot, \cdot)$  is the conformity score function capturing a heuristic notion of uncertainty (e.g. softmax outputs).

Guarantee Prob  $[y_{\text{true}} \in C(x_{\text{test}})] \geq 1 - \alpha$ .



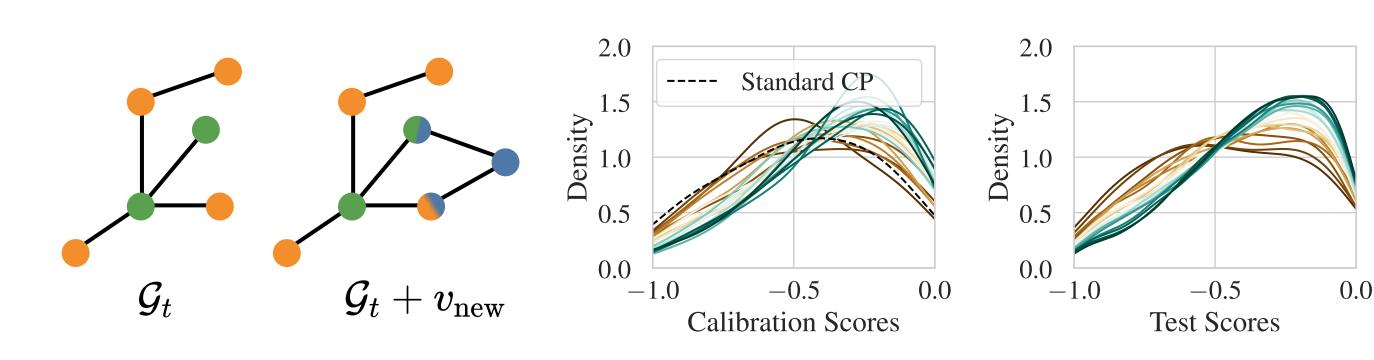
#### **Conformal Prediction for GNNs**

Transductive setting Permutation-equivariant GNN + Exchangeable calibration subset  $\rightarrow$  Guarantee is valid.

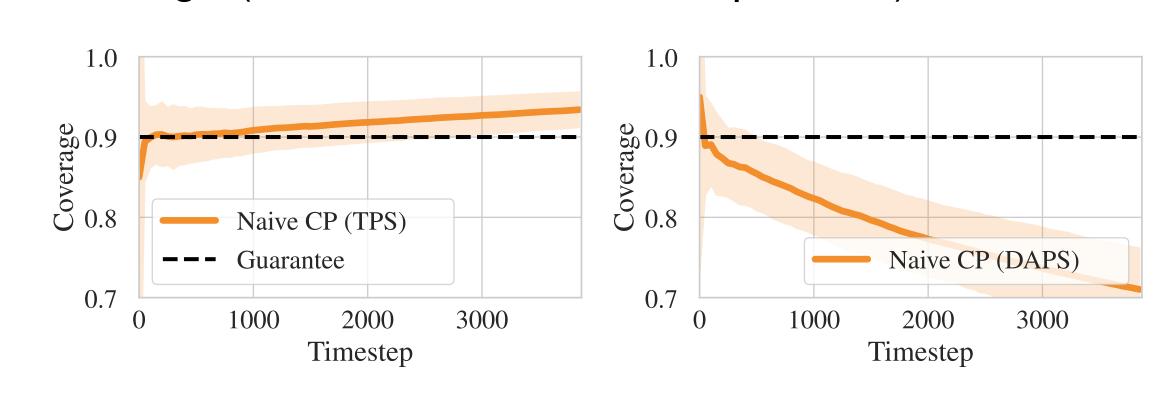
Inductive setting Message passing can break the exchangeability of conformity scores even with an exchangeable calibration set.

#### **Conformal Prediction for Inductive GNNs**

By introducing new nodes or edges, the embedding and scores will shift. This breaks the exchangeability and CP guarantee.

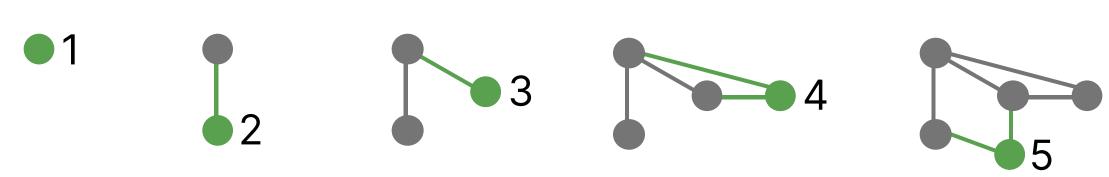


Coverage deviates from the desired level, resulting in over- or under-coverage (direction is unknown in practice).



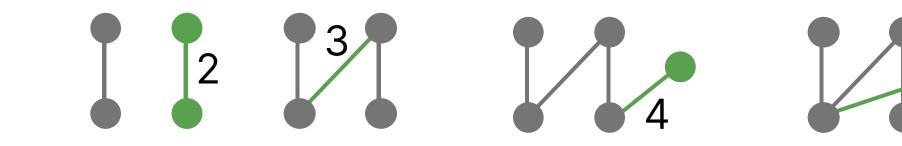
# **Exchangeable Graph Sequences**

Node-exchangeable Any node can appear at any time. Nodes come with all their edges to the existing nodes.



Node exchangeable generators result in dense or empty graphs.

Edge-exchangeable Any edge can appear at any time. Edges might connect new or existing nodes.



# **Our Results in Summary**

Observation The shift in conformity scores is **symmetric**.

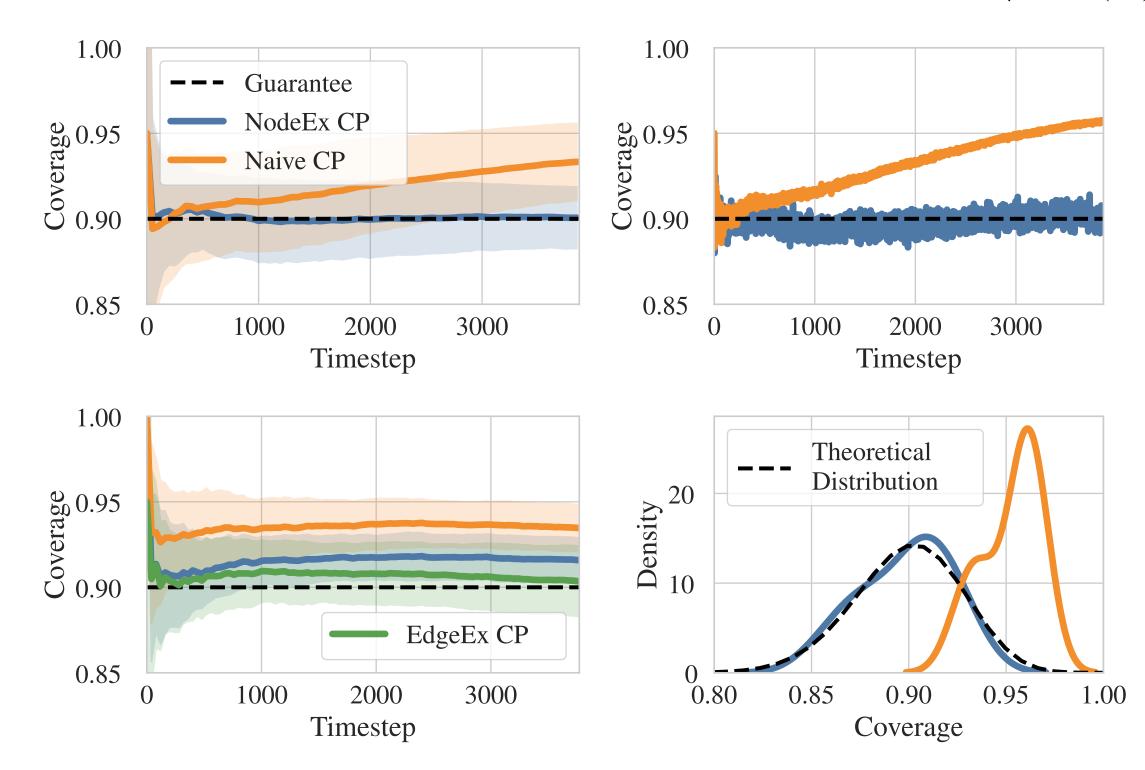
Solution Exact correction for the shift to recover the guarantee.

### Results

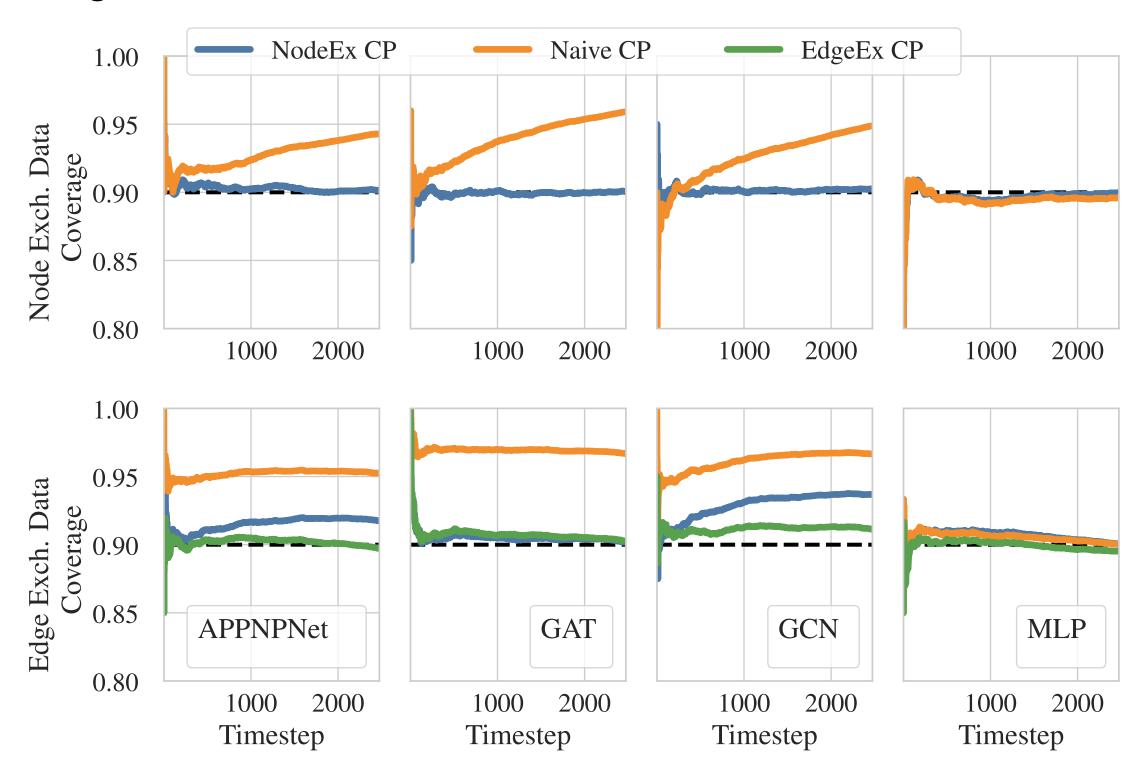
NodeEx CP **Recompute** all calibration scores given the current subgraph  $\mathcal{G}_t$ , **recalibrate** CP and predict.

$$C^{(t)}(v_j) = \left\{ y : s(v_j, y \mid \mathcal{G}_t) \ge \text{Quant} \left( \alpha; \left\{ s(v_i, y_i \mid \mathcal{G}_t) \right\}_{i \in \mathcal{V}_{cal}}; w_i = 1 \right) \right\}$$

EdgeEx CP Same as NodeEx CP but with weights  $w_i = 1/\deg(v_i)$ .



## Recovers guarantee for all models.



### Recovers guarantee for all scores.

