

# ***GLEMOS: Benchmark for Instantaneous Graph Learning Model Selection***

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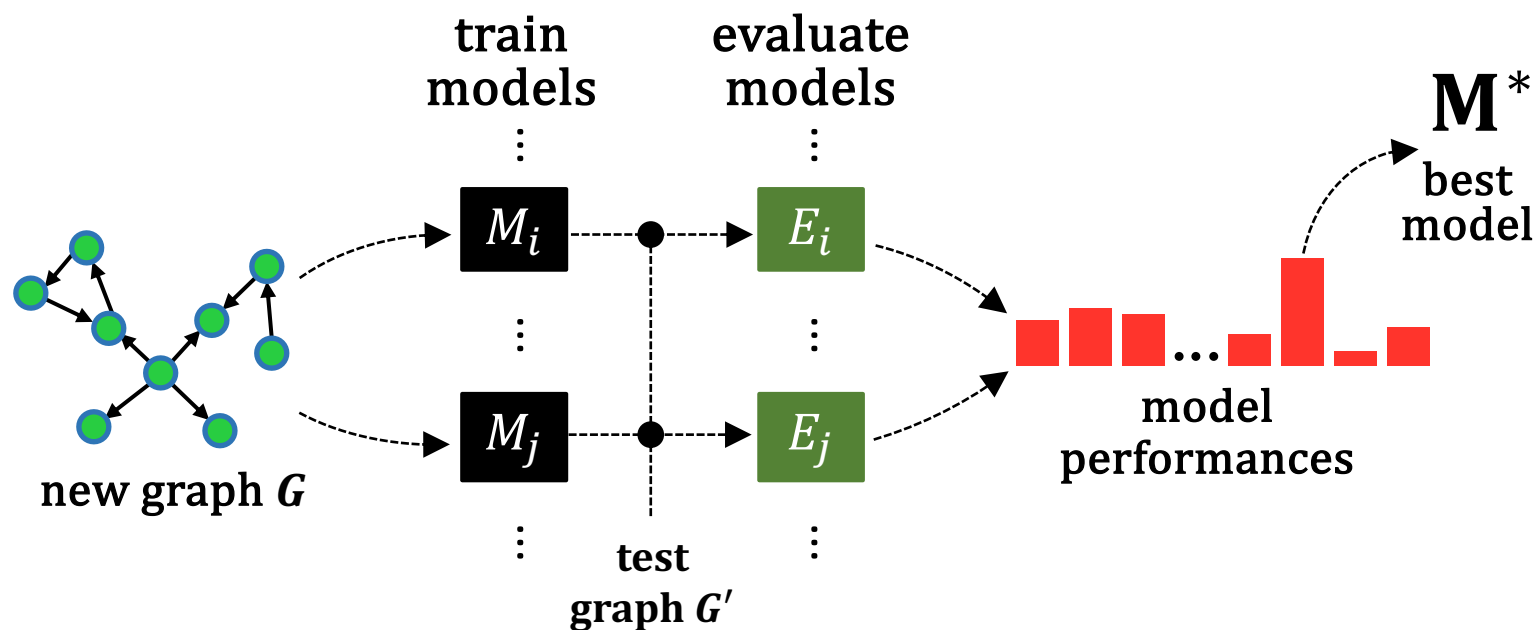
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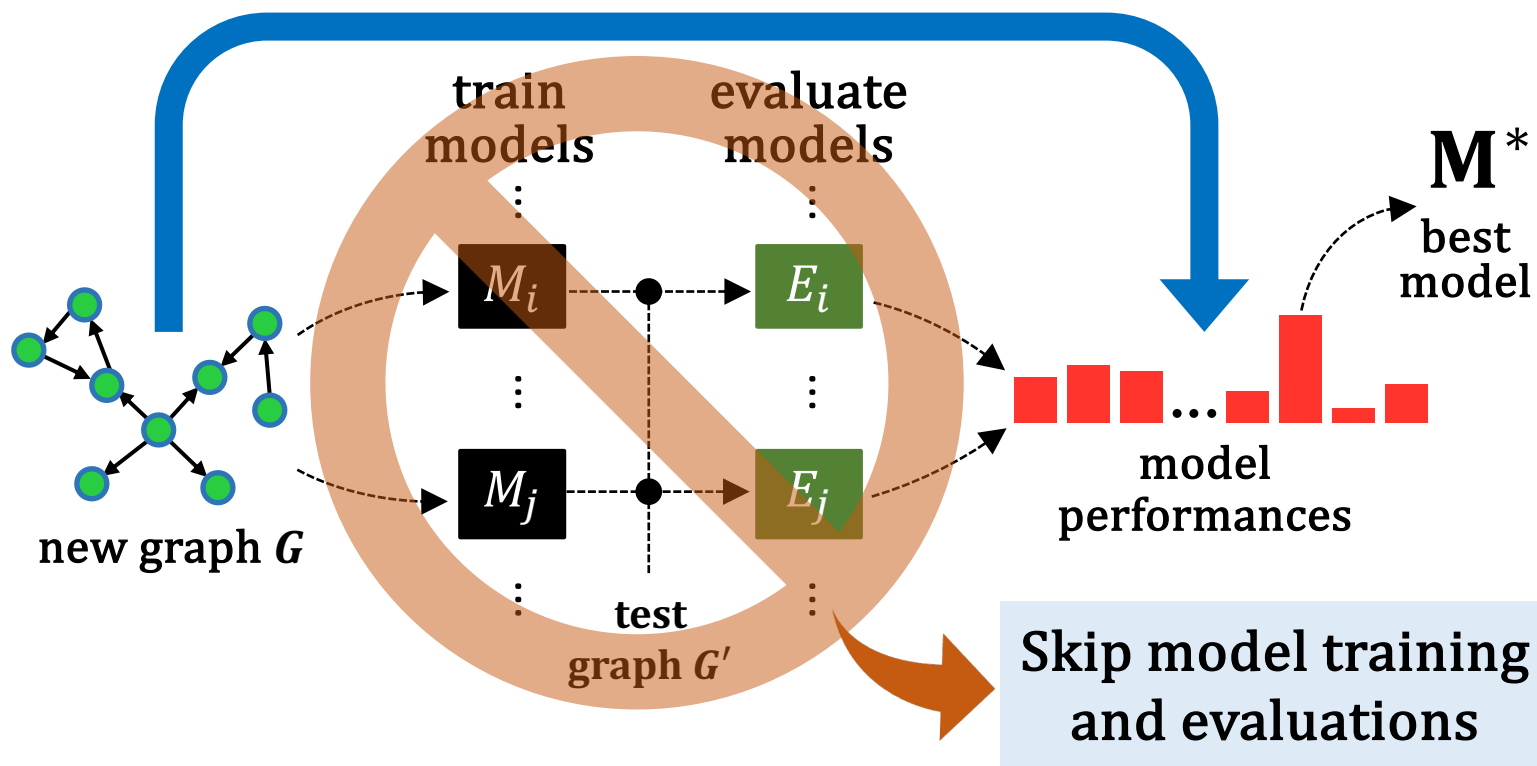
# Graph Learning (GL) Model Selection

## Existing Approach for Graph Learning Model Selection



# Graph Learning Model Selection

## Instantaneous Graph Learning Model Selection



# *Existing Data Banks of GNNs*

	<b>Benchmark Testbeds</b>	<b>Instantaneous Selection Algorithms</b>	<b>Meta-Graph Features</b>	<b>Graph Learning Models</b>	<b># Graph Datasets</b>	<b>Graph Size (max # nodes)</b>	<b># Data Domains</b>
GNN-Bank-101 (ICLR '23)	X	X	X	GNNs	12	34k	5
NAS-Bench-Graph (NeurIPS '22)	X	X	X	GNNs	9	170k	4
GraphGym (NeurIPS '20)	X	X	X	GNNs	32	34k	7

# *Contributions of The Proposed GLEMOS*

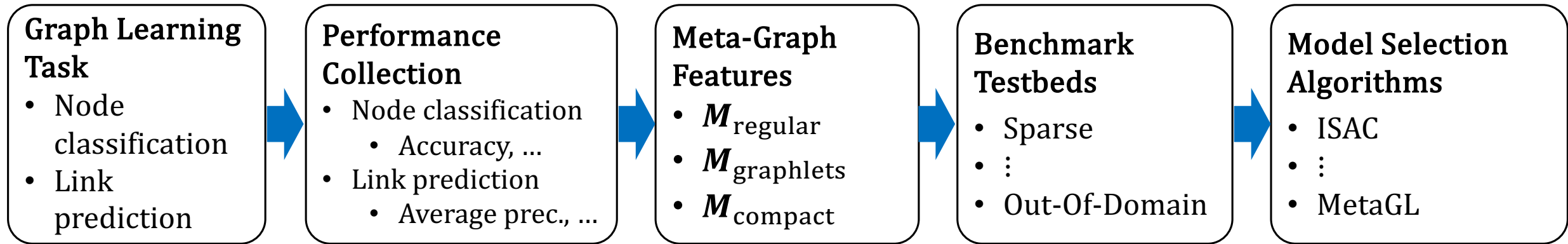
- Extensive Benchmark Data with Multiple GL Tasks

	Benchmark Testbeds	Instantaneous Selection Algorithms	Meta-Graph Features	Graph Learning Models	# Graph Datasets	Graph Size (max # nodes)	# Data Domains
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<b>GLEMOS (Ours)</b>	✓	✓	✓	<b>GNNs &amp; Non-GNNs</b>	<b>457</b>	<b>496k</b>	<b>37</b>

# ***Contributions of The Proposed GLEMOS***

- Extensive Benchmark Data with Multiple GL Tasks
- Comprehensive Evaluation Testbeds
- Extensible Benchmark Environment
- Future Research Directions

# Overview of GLEMOS



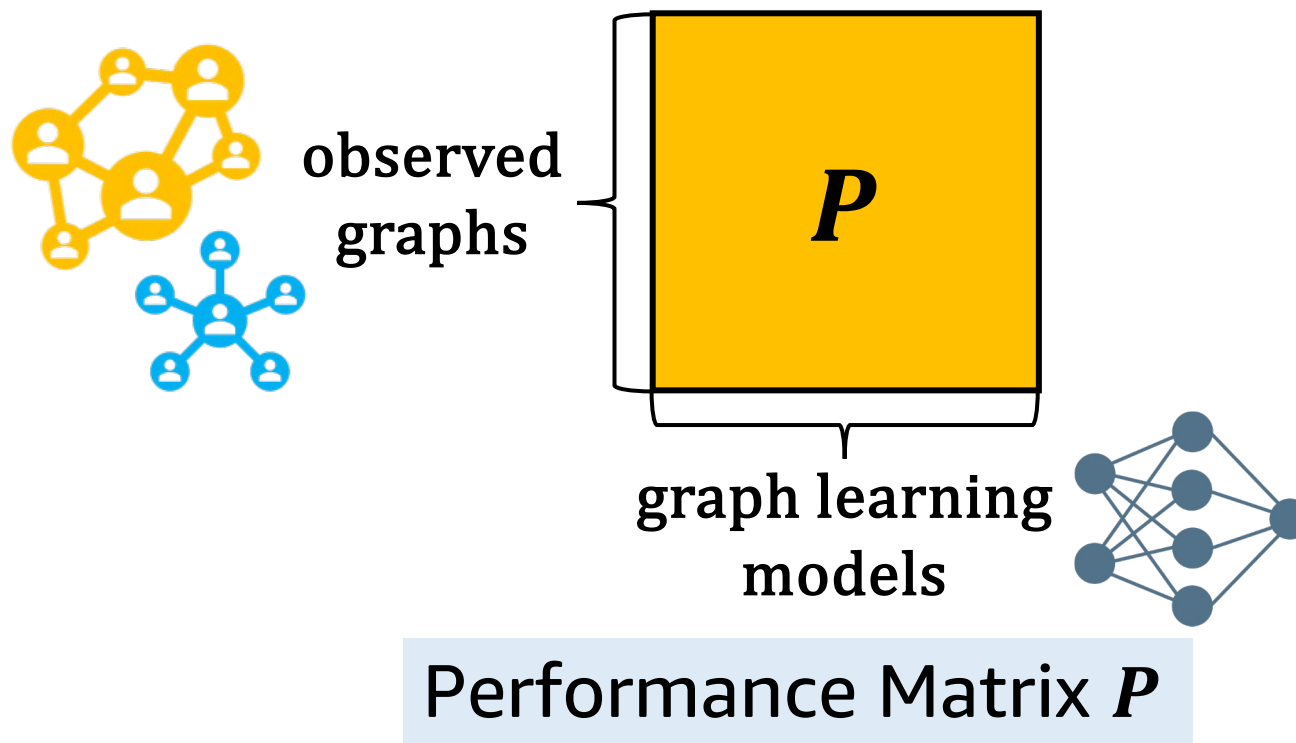
# *Roadmap*

- Introduction
- **Performance Collection**
- Meta-Graph Features
- Benchmark Testbeds & Algorithms
- Future Directions & Conclusion

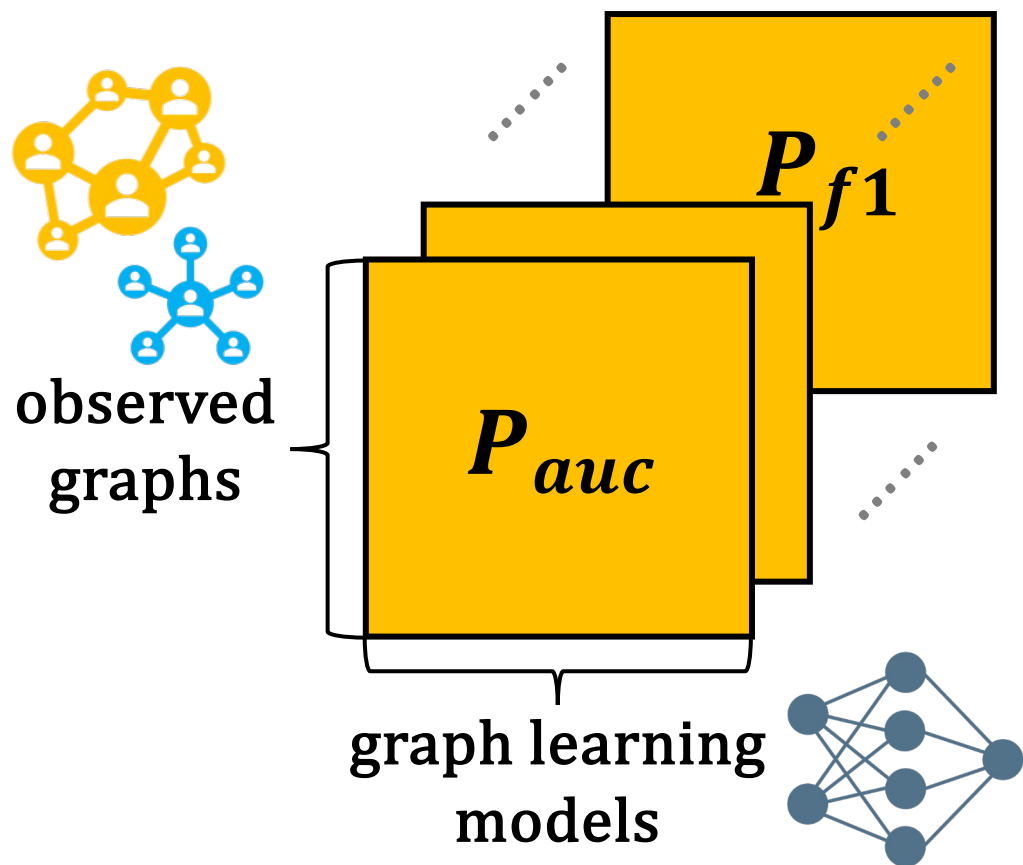




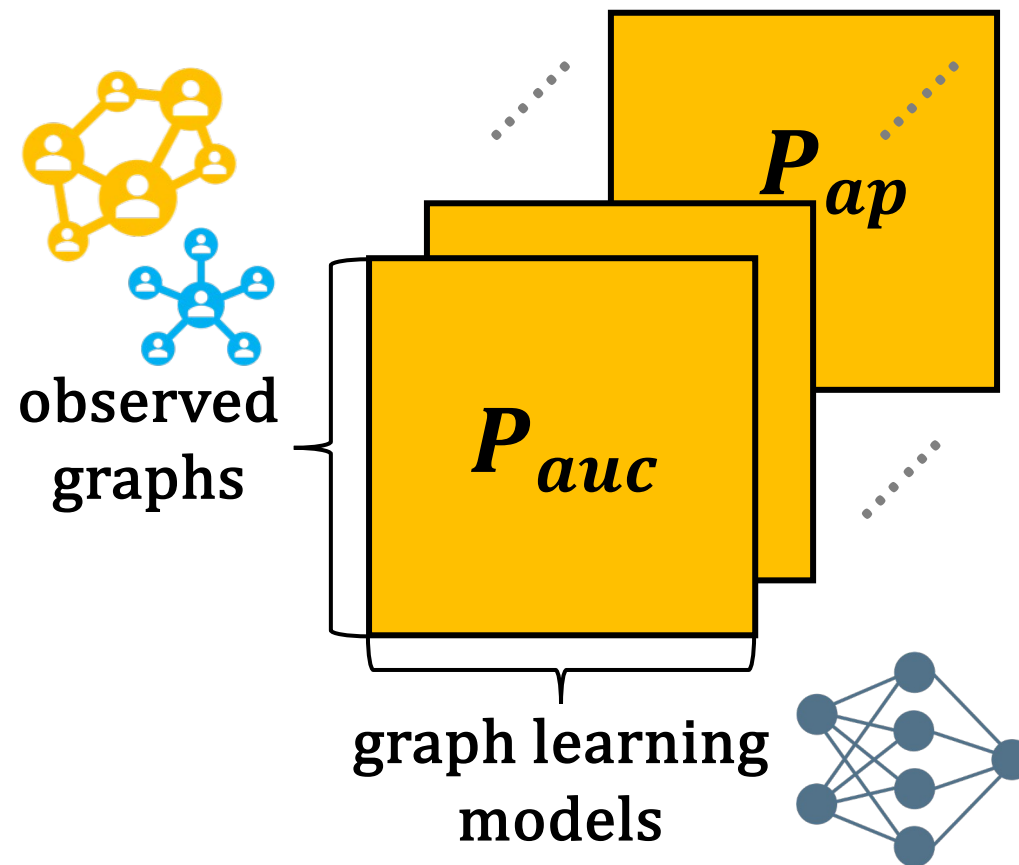
# *Performance Collection*



# Performance Collection

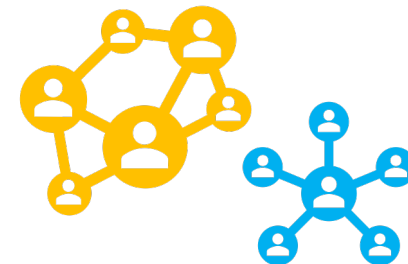


Node Classification Perfs.



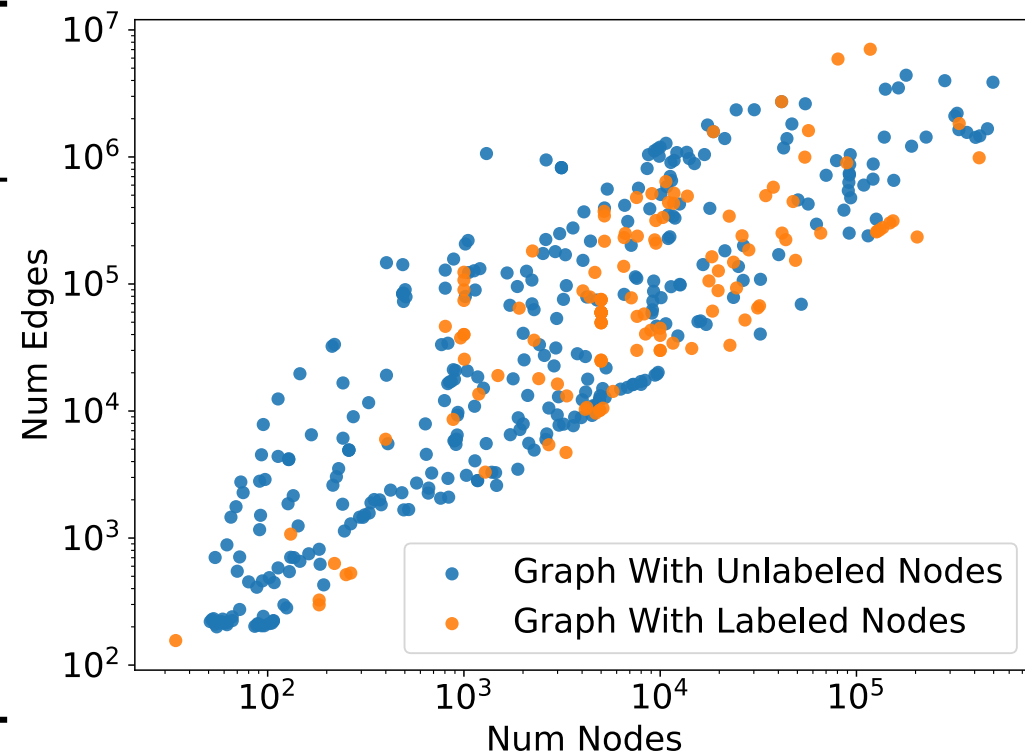
Link Prediction Perfs.

# Performance Collection: Graphs

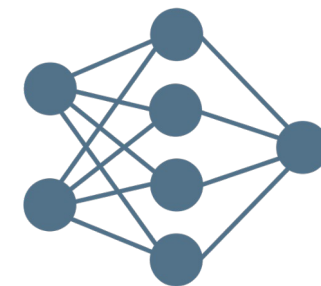


GLEMOS covers representative and diverse graph datasets

	Node Classification	Link Prediction
<b>Total graphs</b>	128	457
• # nodes	34–422k	34–496k
• # edges	156–7M	156–7M
• # node feats	2–61k	2–61k
• # node classes	2–195	N/A
• # data domains	25	37



# Performance Collection: Models



GLEMOS covers representative and diverse GL models

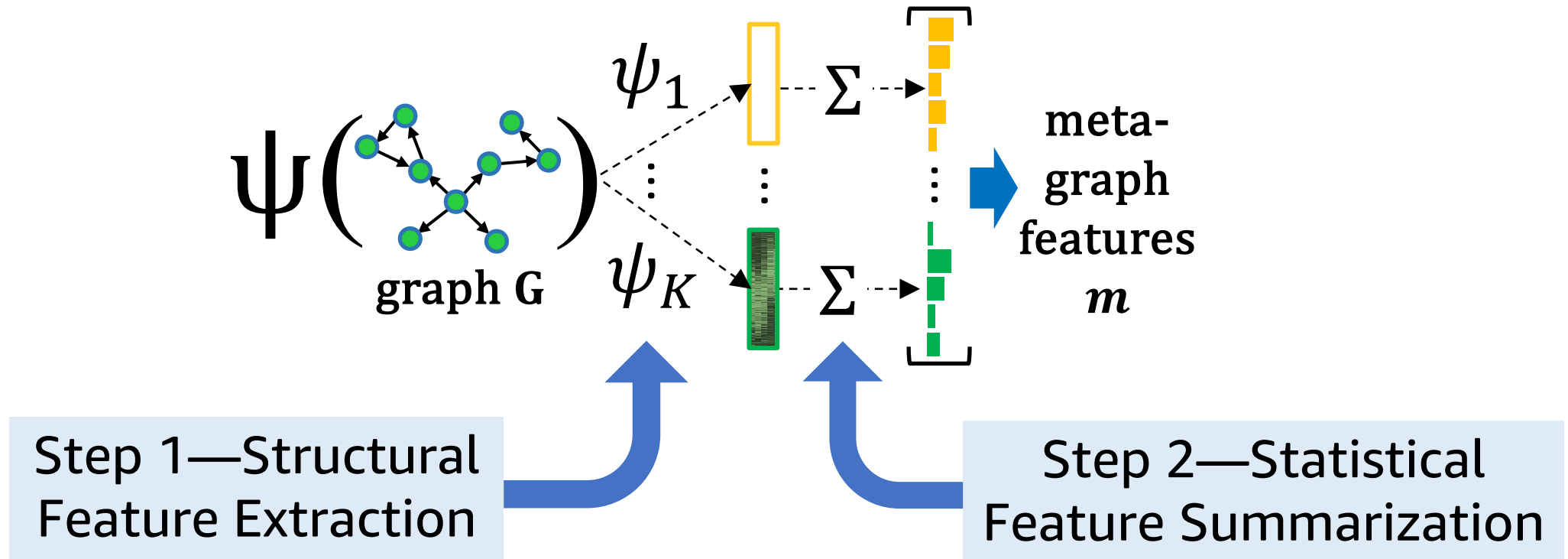
Method	NC	LP	Count	Method	NC	LP	Count
GCN	✓	✓	30	GraRep	✓	✓	6
GraphSAGE	✓	✓	24	DGI	✓	✓	24
GAT	✓	✓	40	node2vec	✓	✓	72
GIN	✓	✓	10	Label Prop.	✓		16
EGC	✓	✓	28	Jaccard's Coeff		✓	1
SGC	✓	✓	10	Resource Alloc.		✓	1
ChebNet	✓	✓	27	Adamic/Adar		✓	1
PNA	✓	✓	32	SEAL		✓	36
Spectral Emb.	✓	✓	8				
				<b>Total Count</b>			<b>366</b>

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- Benchmark Testbeds & Algorithms
- Future Directions & Conclusion



# Meta-Graph Features: Feature Generation



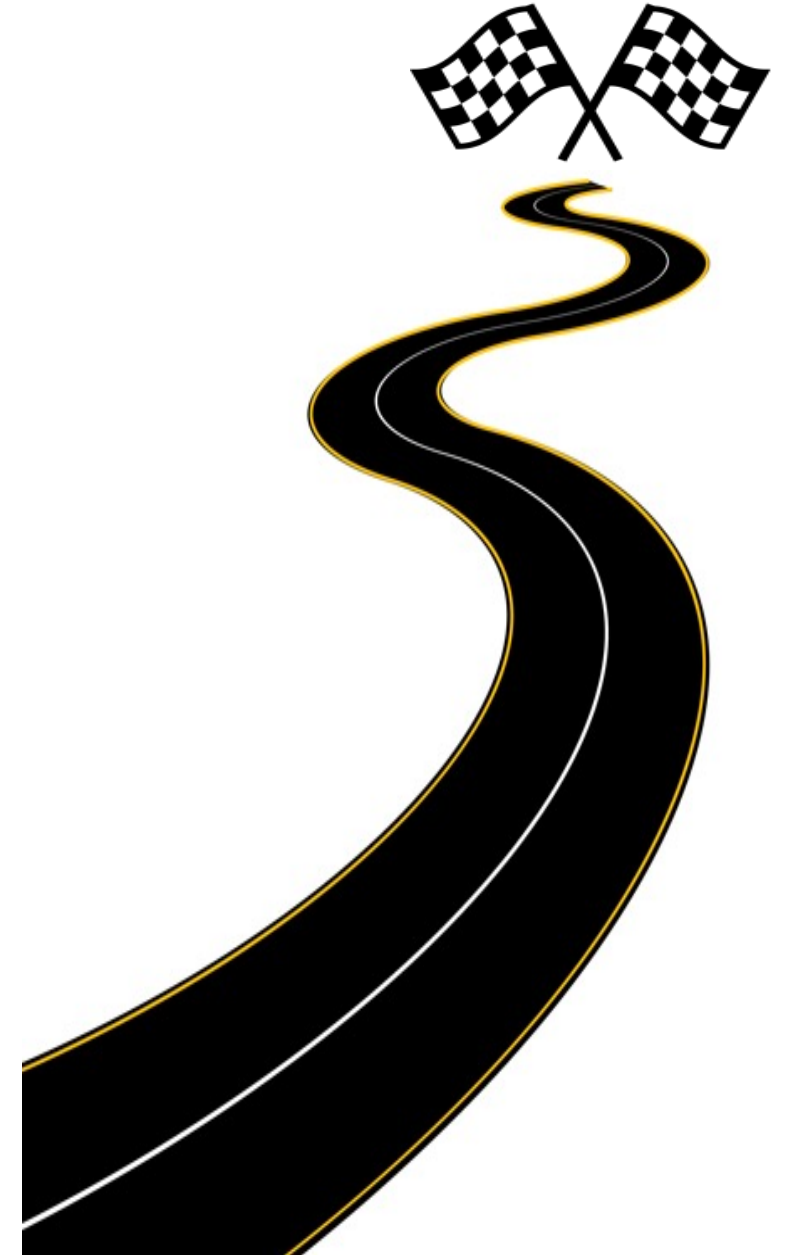
# ***Meta-Graph Features: Collections***

## **Predefined sets of meta-graph features**

- **$M_{regular}$** : 318 features
- **$M_{graphlets}$** : 756 features
- **$M_{compact}$** : 58 features
- **$M_{reg+raphlets}$** : 1,074 features

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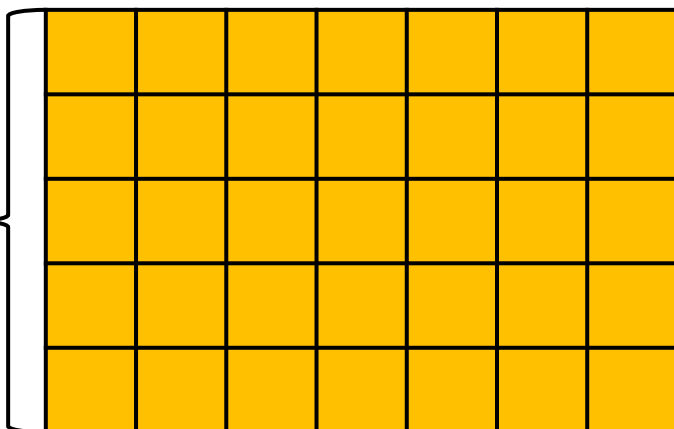




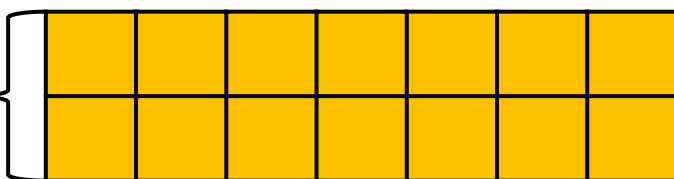
# Benchmark Testbeds

## Fully-Observed Testbed

Training  
perf.  
matrix

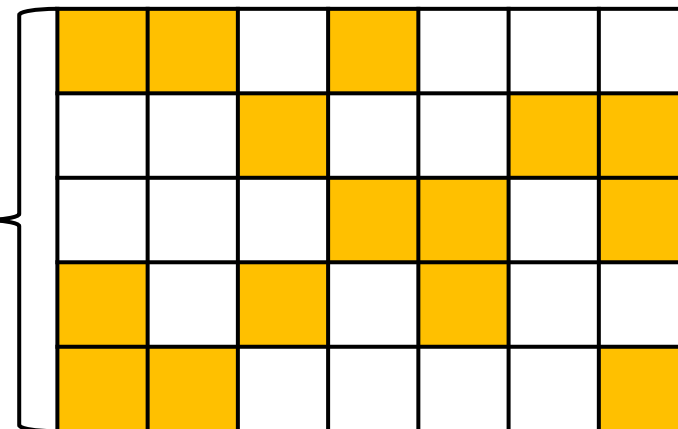


Testing  
perf.  
matrix

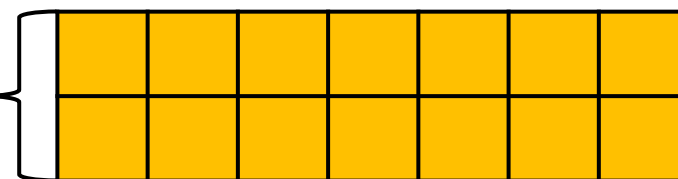


## Sparse Testbed

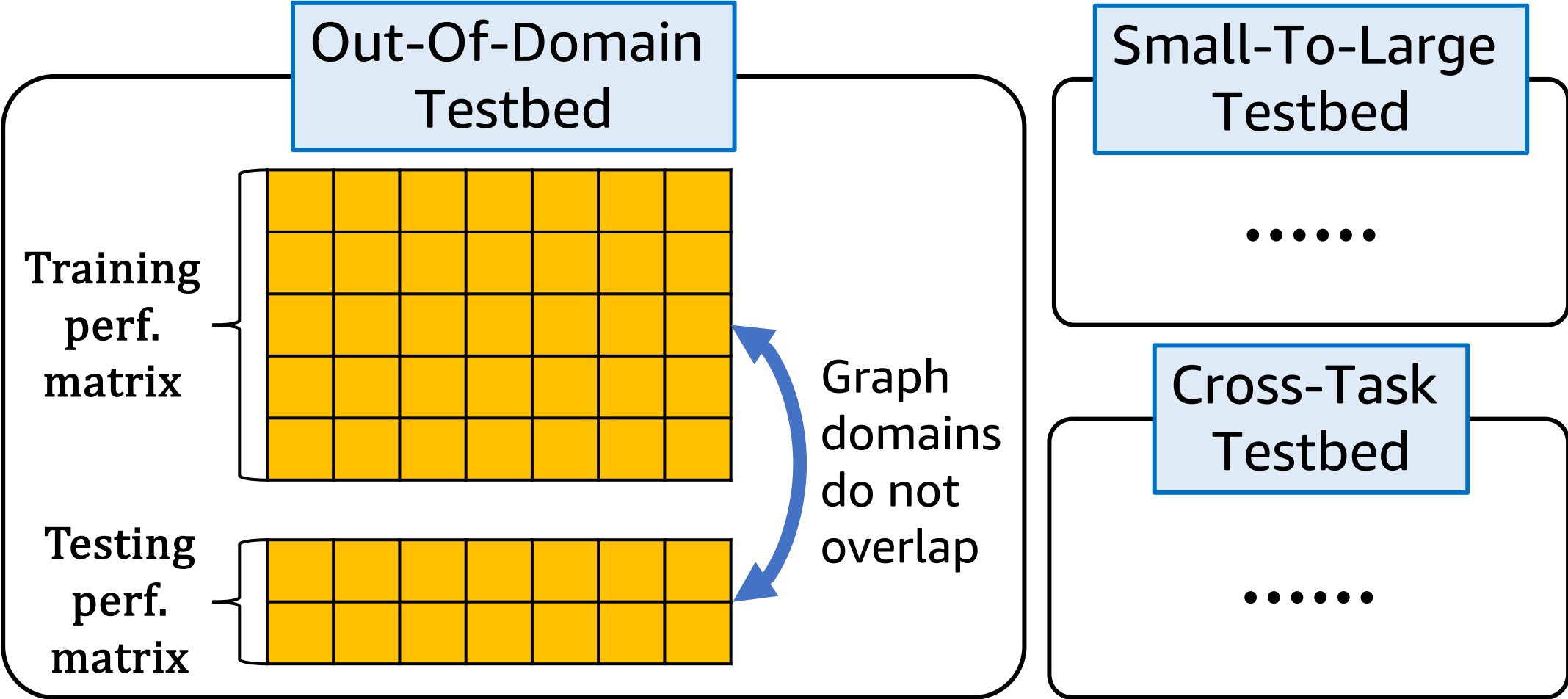
Training  
perf.  
matrix



Testing  
perf.  
matrix



# Benchmark Testbeds



# *Instantaneous Model Selection Algorithms*

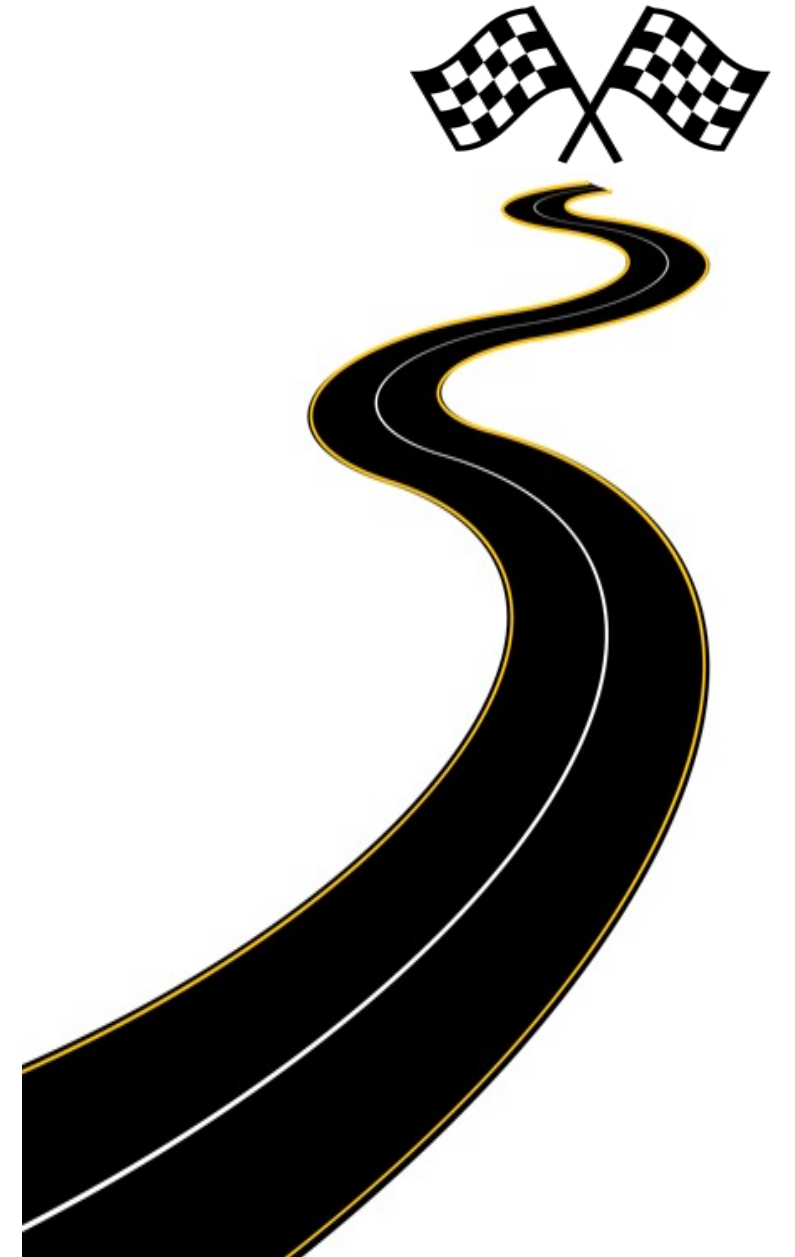
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<b>Algorithm</b>	<b>C1. Use meta-features</b>	<b>C2. Use prior performances</b>	<b>C3. Optimizable</b>
Random Selection			
GB-Avg. Perf		✓	
GB-Avg. Rank		✓	
ISAC	✓	✓	
AS	✓	✓	
Supervised Surrogates	✓	✓	✓
ALORS	✓	✓	✓
NCF	✓	✓	✓
MetaOD	✓	✓	✓
MetaGL	✓	✓	✓

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- **Future Directions & Conclusion**



# *Future Directions*

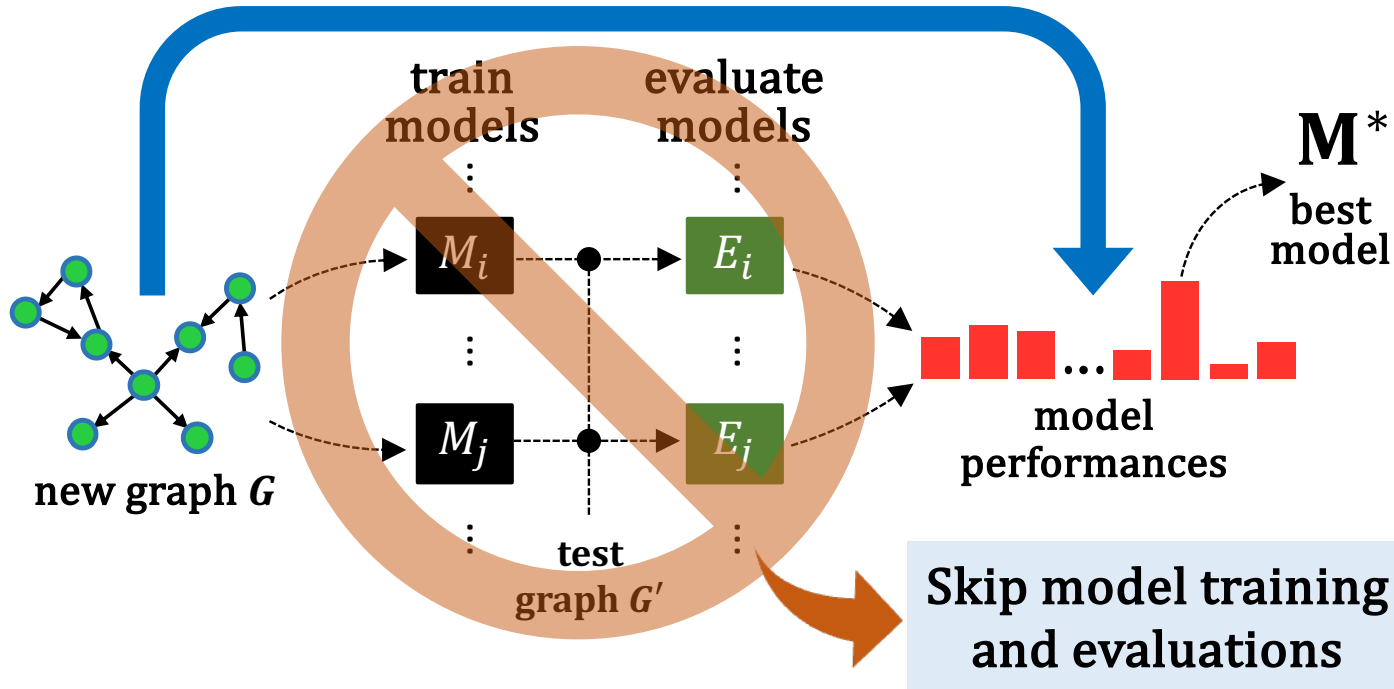


- D1: Enabling the use of additional graph data
- D2: Developing data augmentation techniques
- D3: Handling out-of-distribution settings
- D4: Collecting performance more efficiently

# Conclusion

Extensive  
Benchmark Data

Instantaneous Graph Learning Model Selection



Model  
Selection  
Algorithms

Evaluation  
Testbeds

Extensible  
Benchmark  
Environment