

# Link Stealing Attacks on Inductive Trained Graph Neural Networks

Bachelor Thesis Introduction - Philipp Zimmermann

#### Outline



- Graphs
- Graph Neural Networks
- Our Approach: Link Stealing Attacks on Inductive Trained Graph Neural Networks
- Experimental Setup
- Goal





- Data Structure
  - Model large data and relationships between entities
  - Nodes with features
  - Edges



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- Chemical Networks
  - Protein-protein interactions



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- **Data Structure** 
  - Model large data and relationships between entities
  - Nodes with features
  - Edges







- **Chemical Networks** 
  - Protein-protein interactions









- Social Networks
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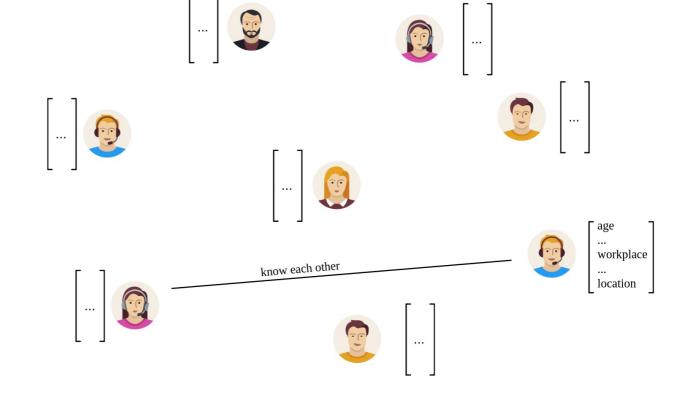






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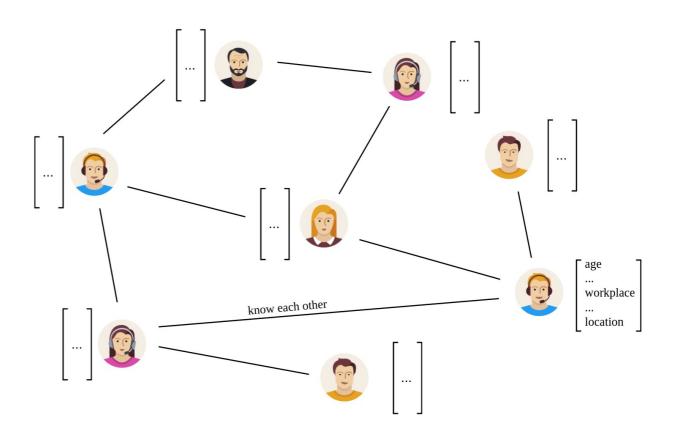
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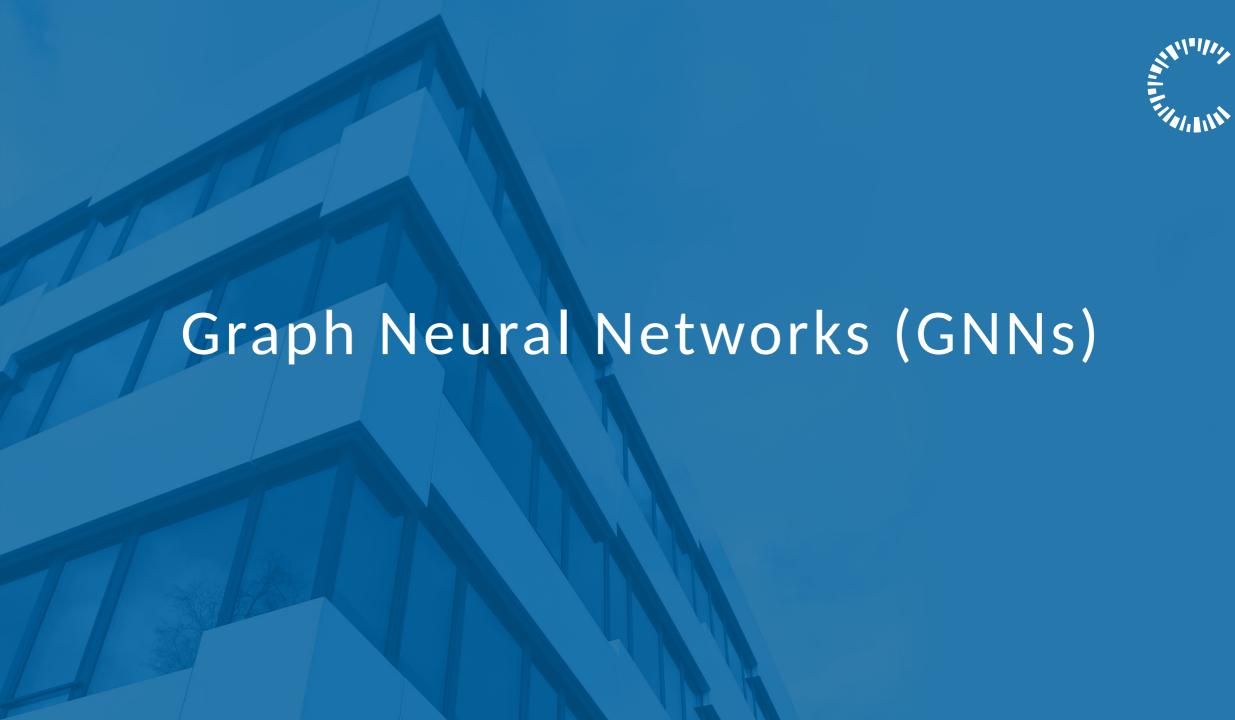




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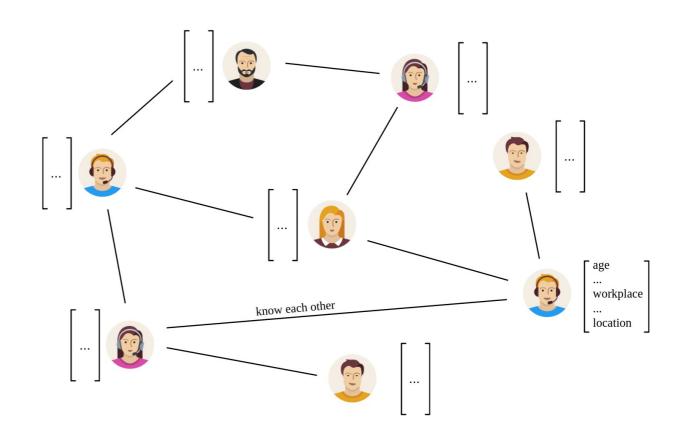




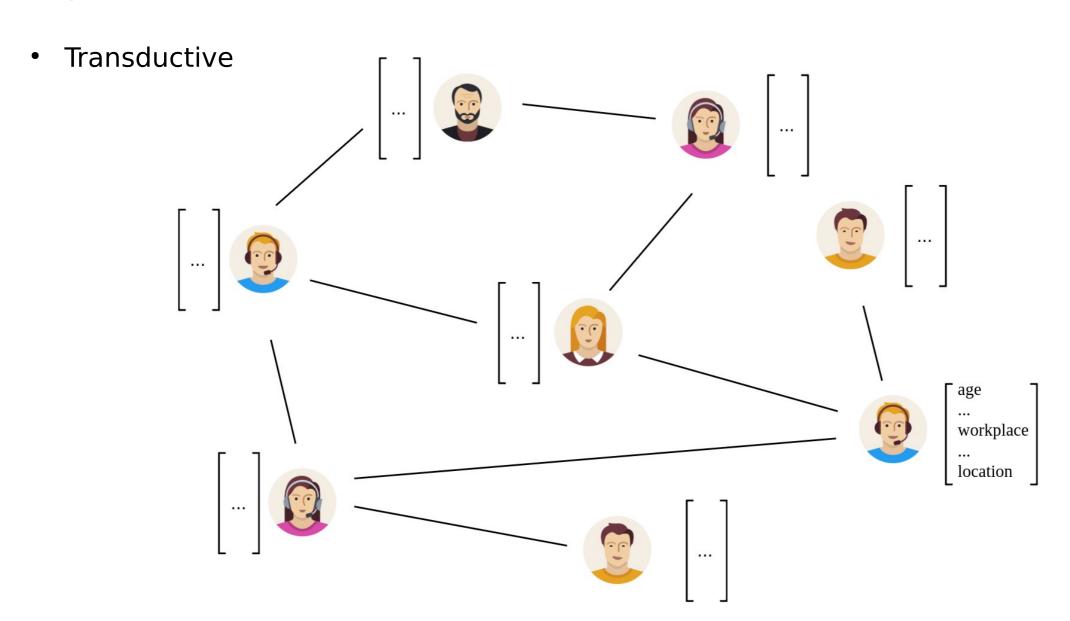


Machine Learning Model over Graphs

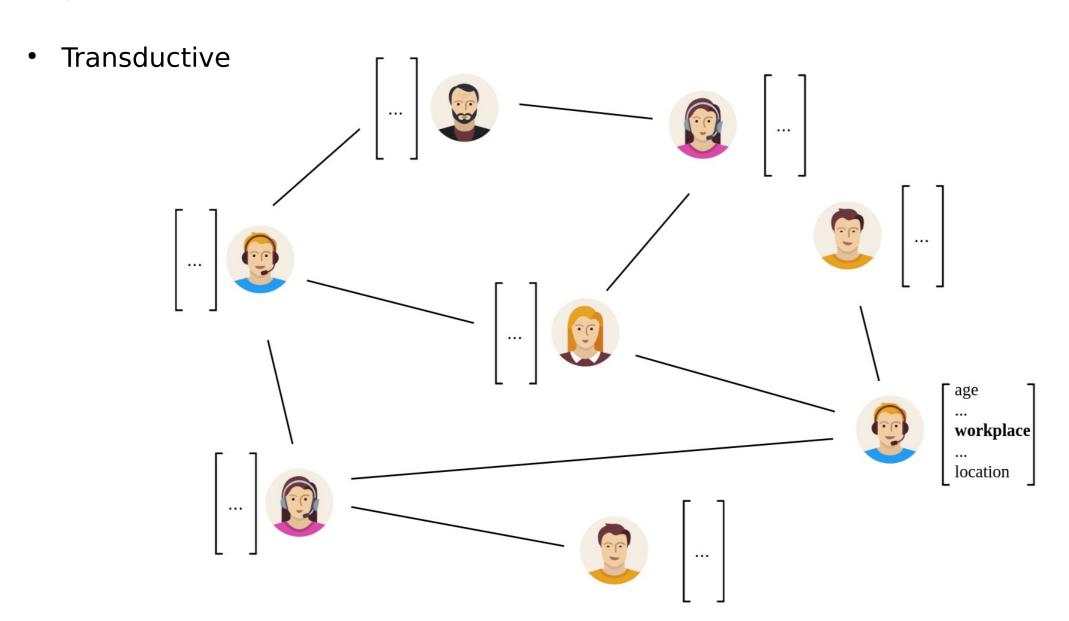
- Different Tasks
  - Node classification
  - Graph classification
  - Link prediction
- Different Learning Methods
  - Transductive
  - Inductive



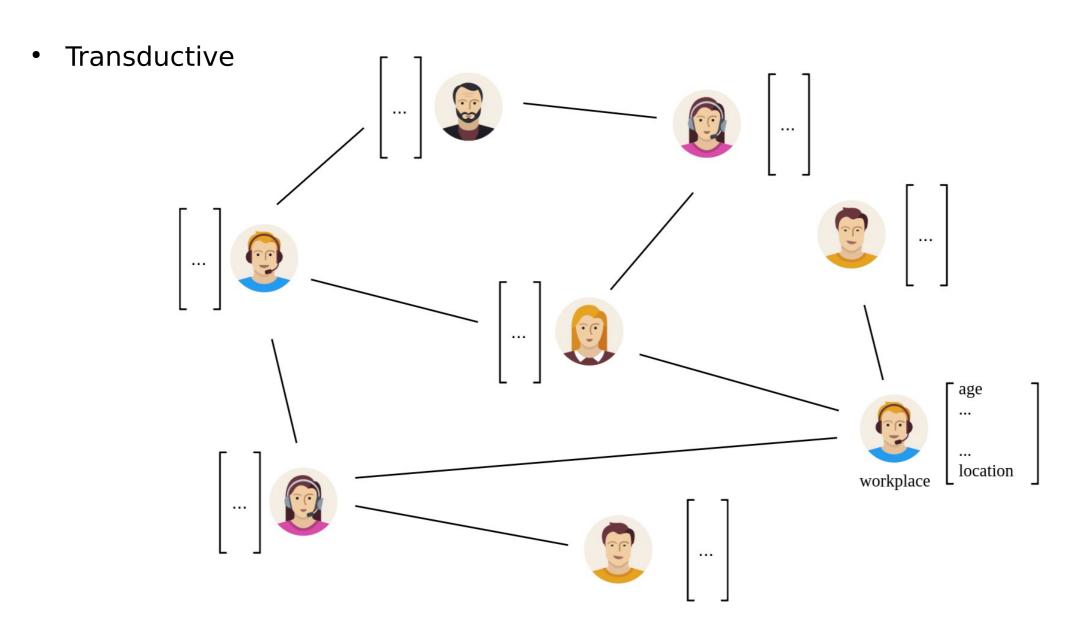




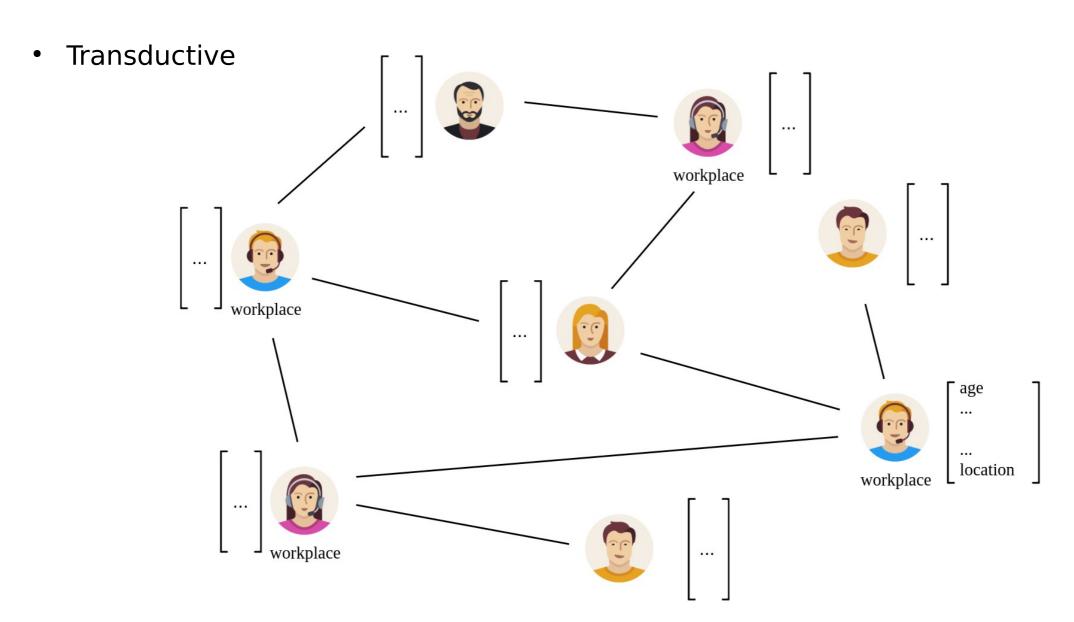




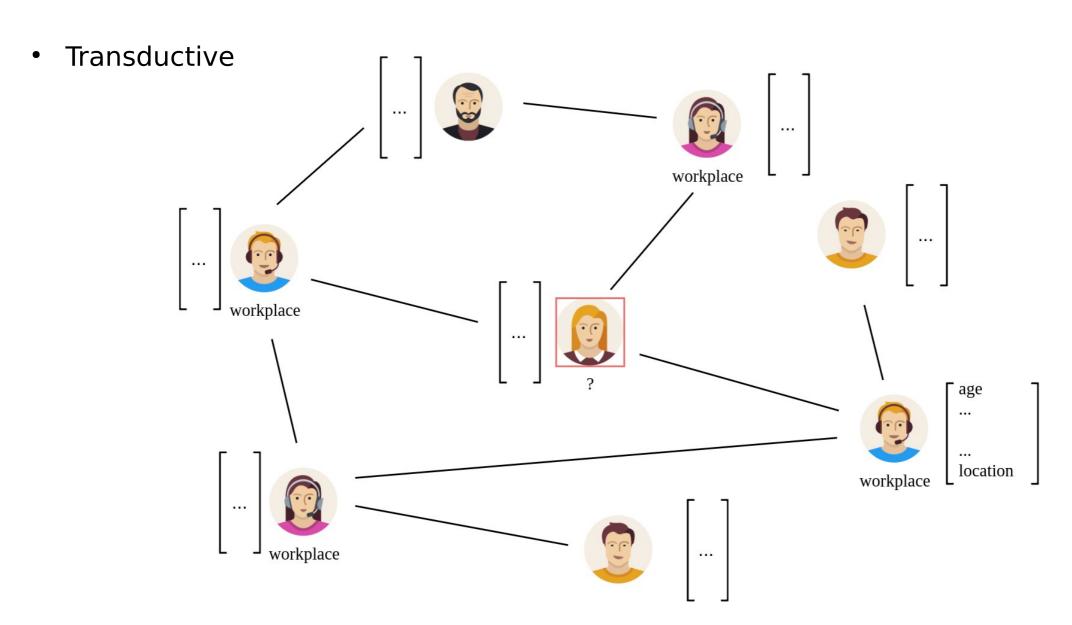






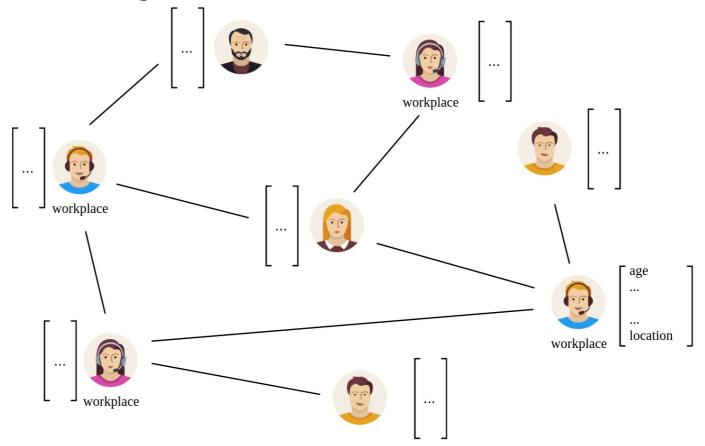




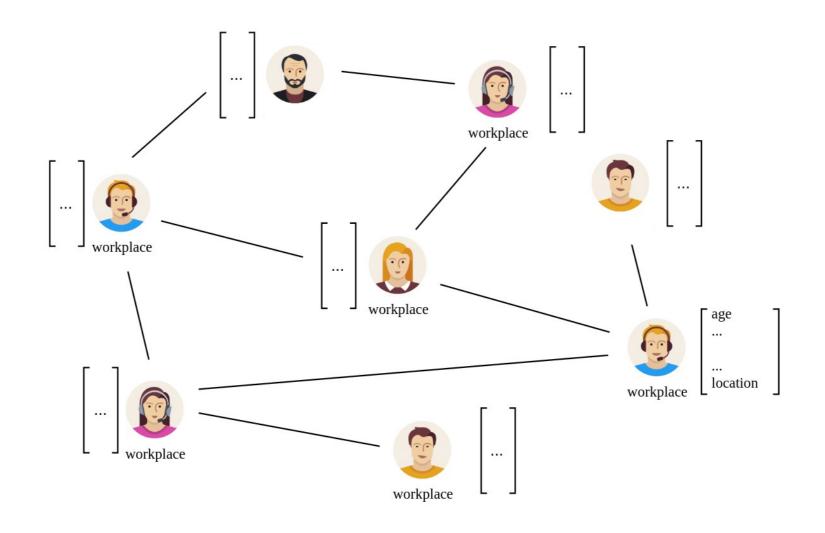




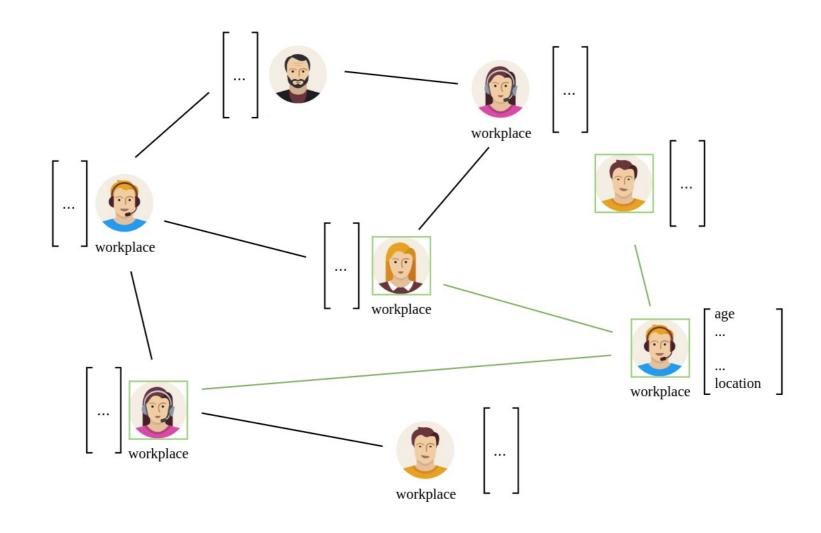
- Transductive
  - Fix graph (features and link)
  - Some nodes' labels are missing
  - Limited scenario



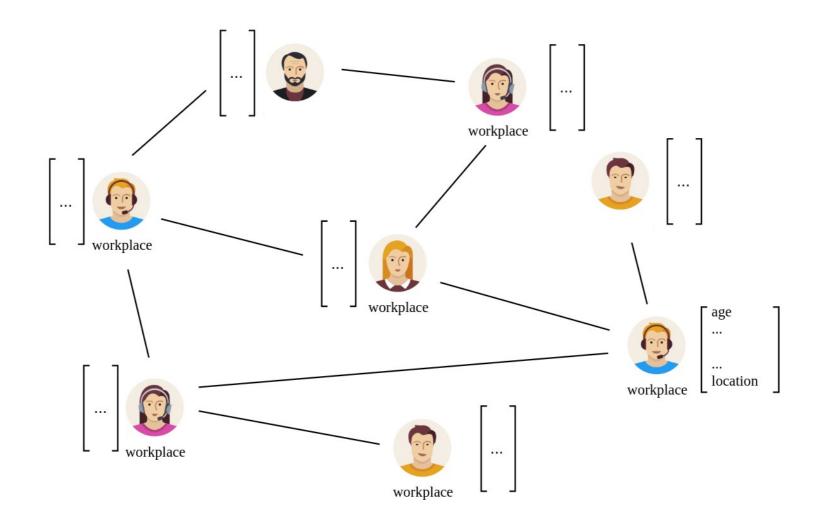




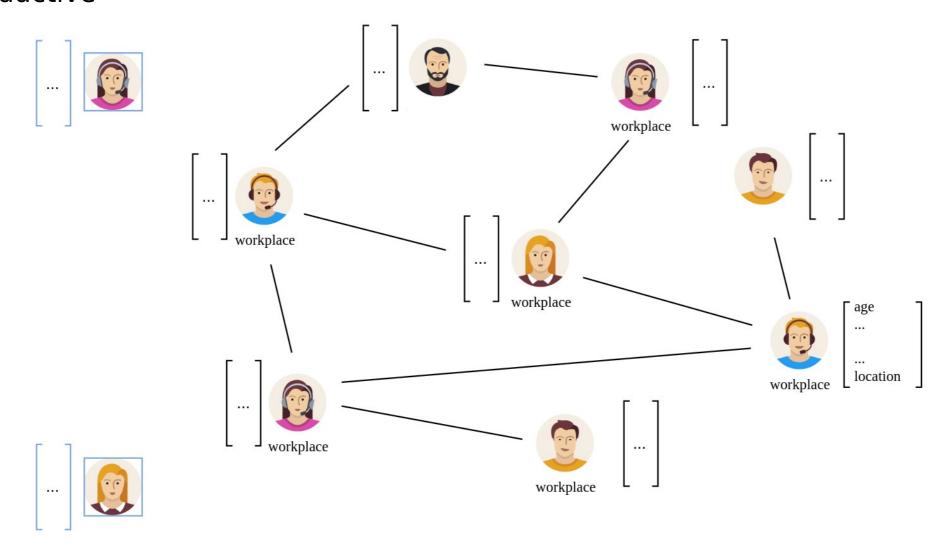




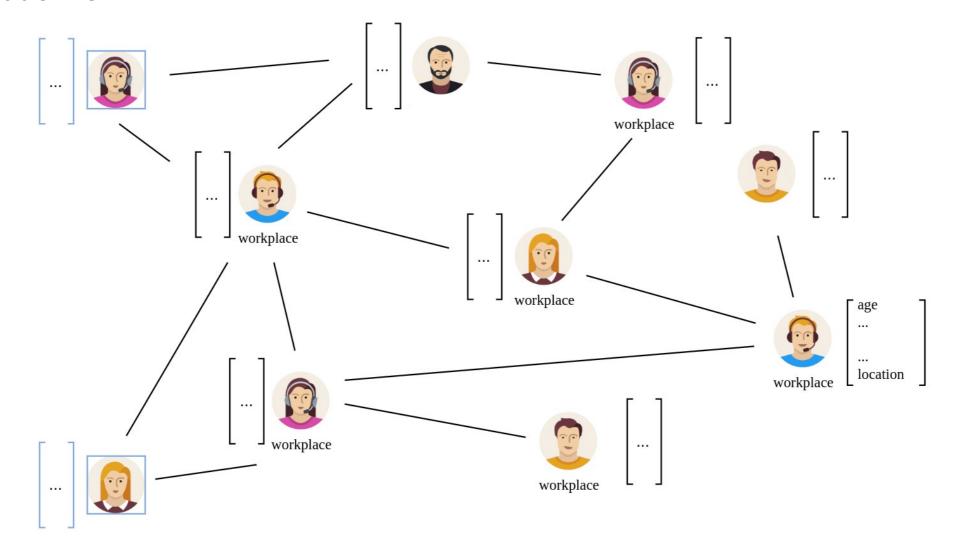




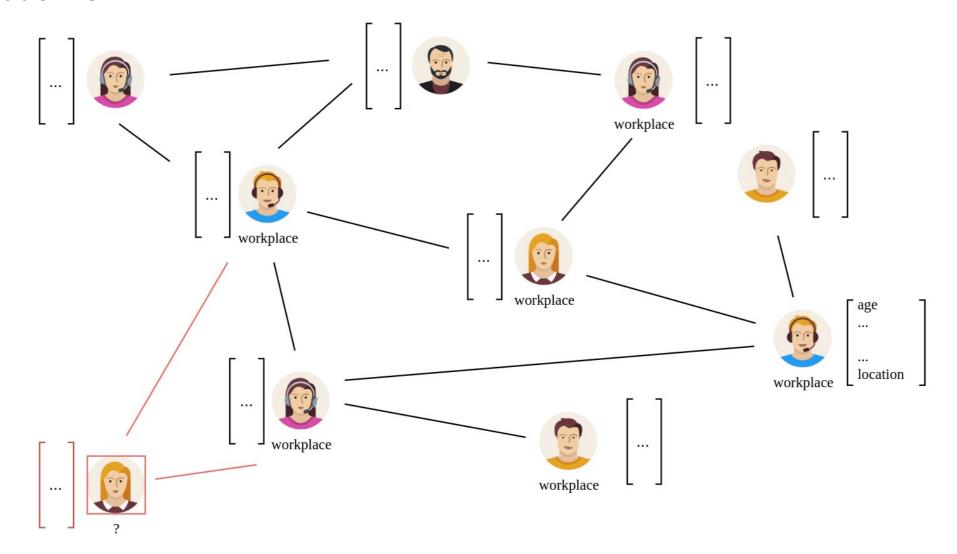




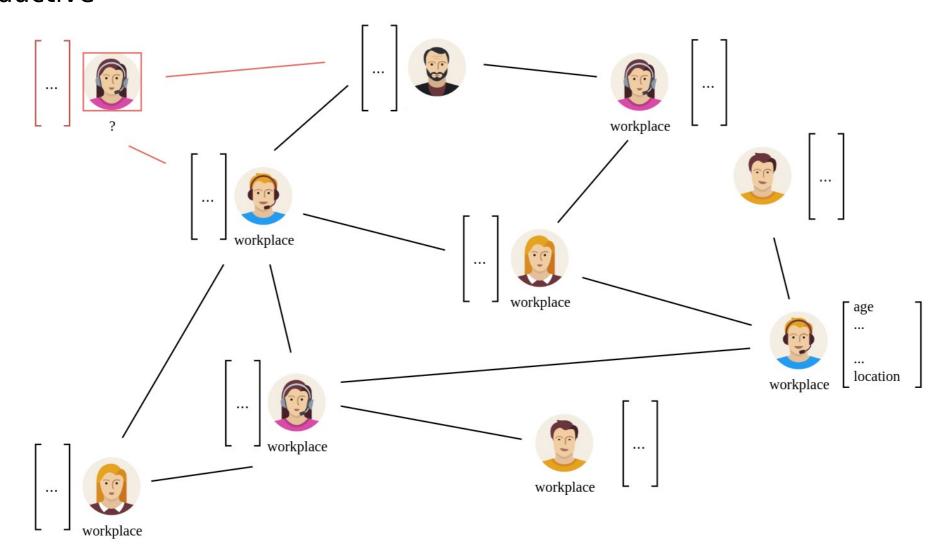




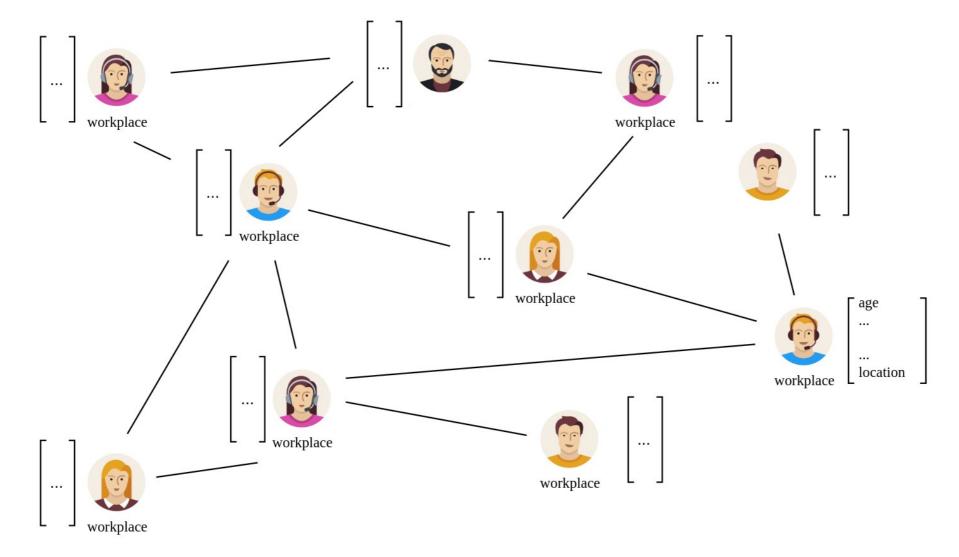














- Inductive
  - Extends transductive setting
  - Able to generalize to unseen nodes
  - Unnecessary to retrain the model
  - Broader scenario

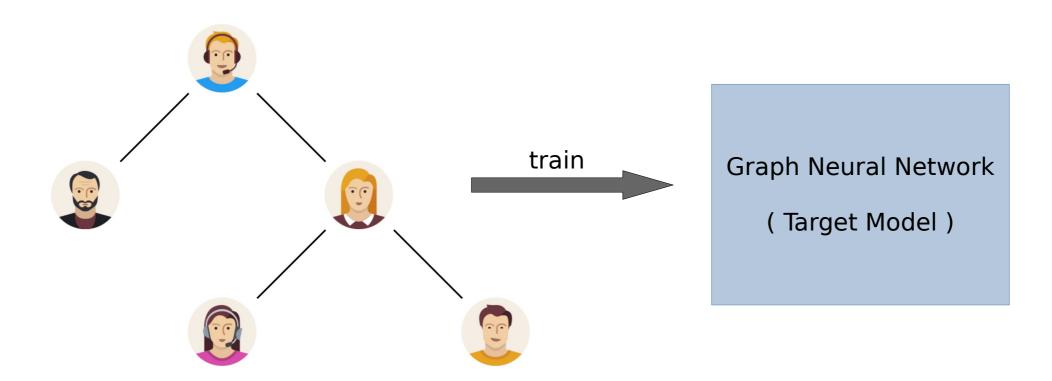




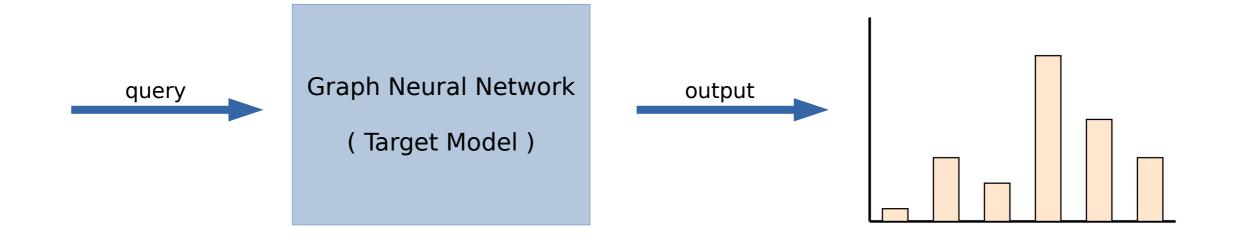


- Scenario:
  - GNN trained on graph G to perform downstream task
  - Attacker
    - Black box access to target model
    - Partial graph with incomplete set of edges
- Goal:
  - Recover missing links from partial graph



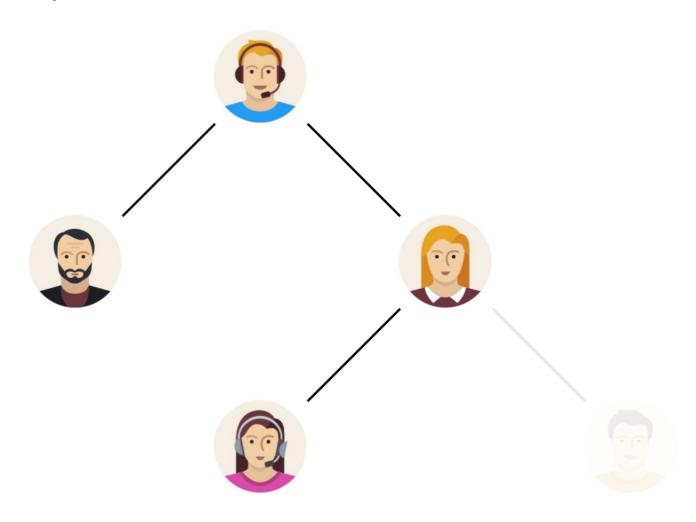






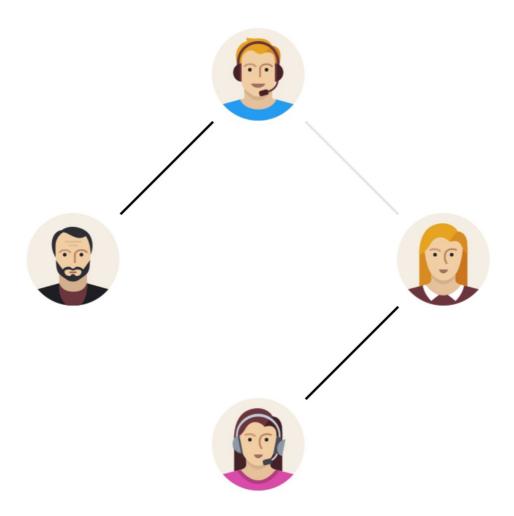


Attacker Graph



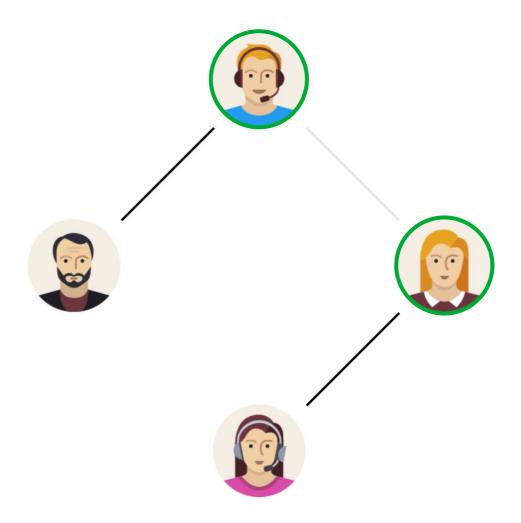


Attacker Graph with One Missing Link

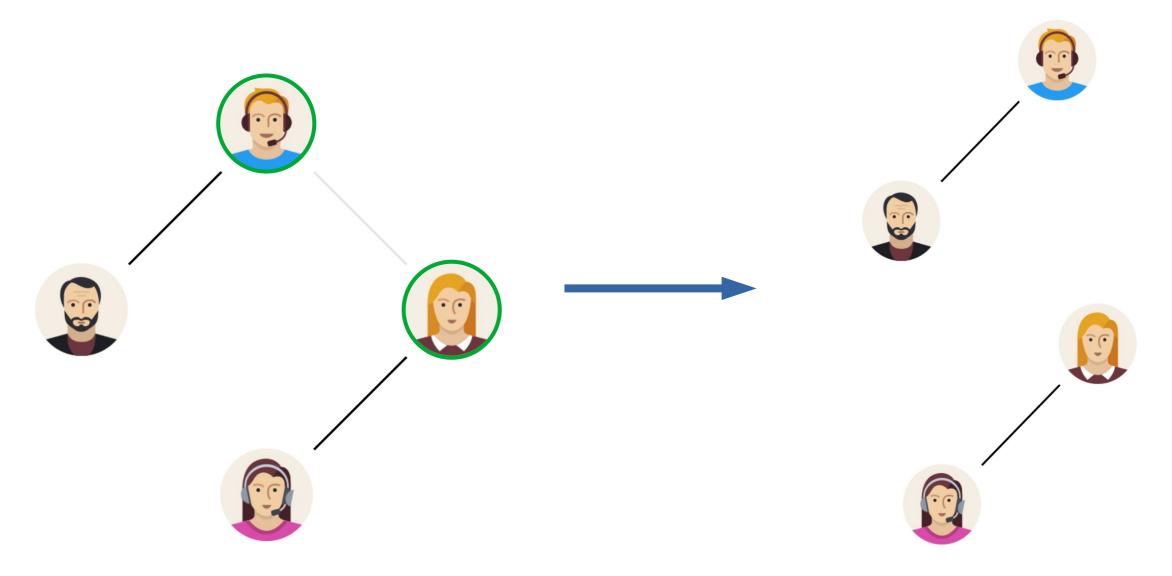




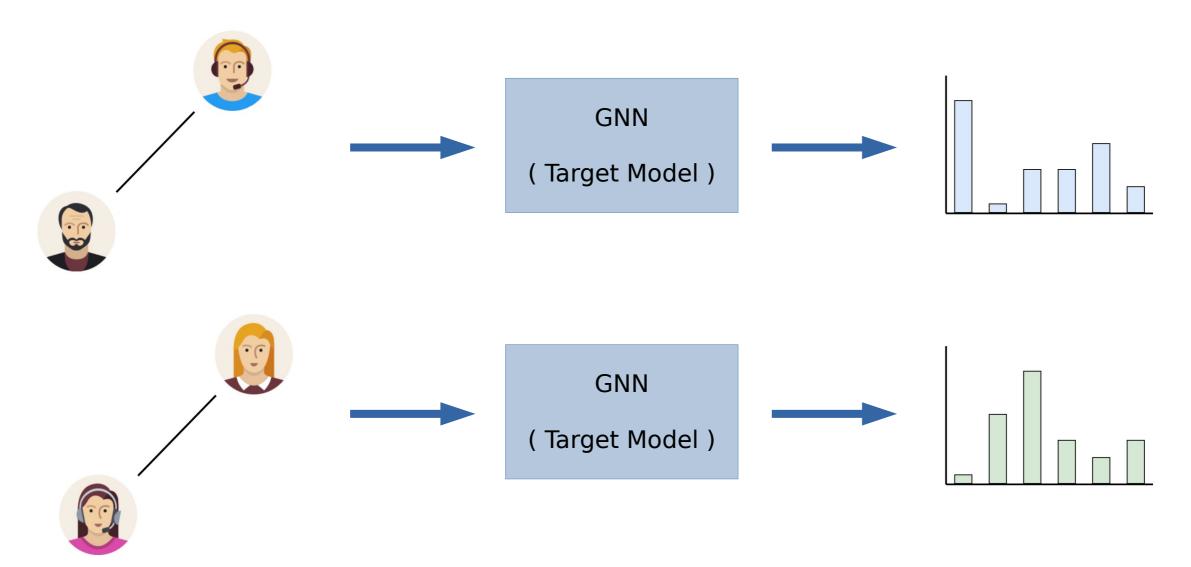
Attacker Graph



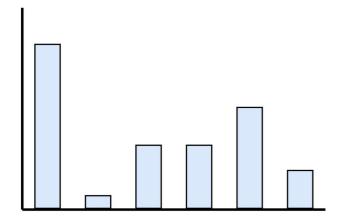


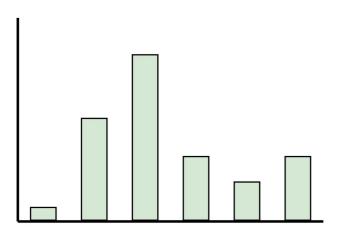




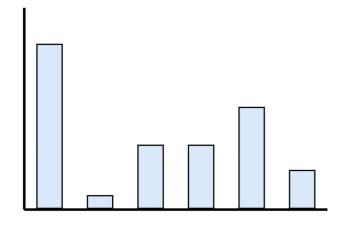




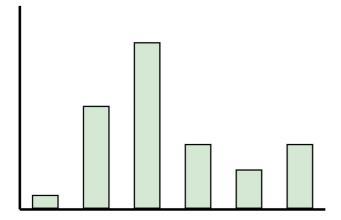






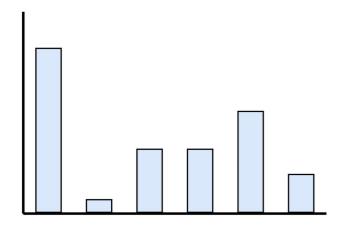


[ 0.3, 0.2, 0.3, ..., 0.1 ]



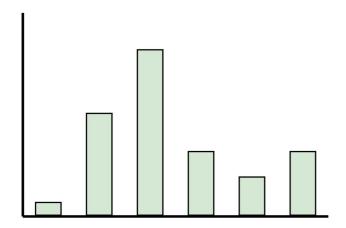
[ 0.2, 0.1, 0.4, ..., 0.1 ]





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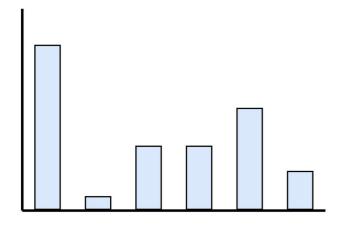




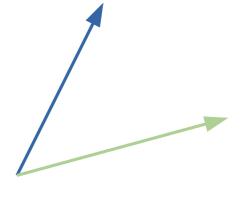
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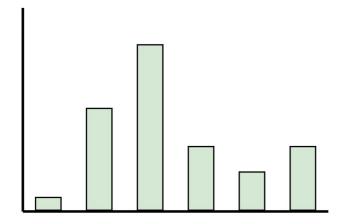






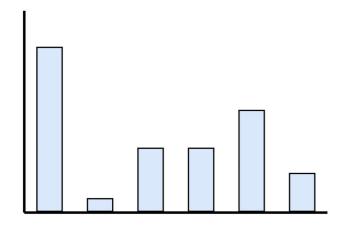
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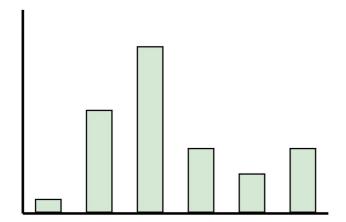


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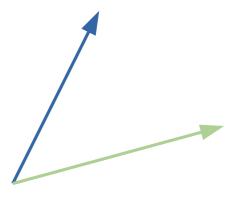




[ 0.3, 0.2, 0.3, ..., 0.1 ]



[0.2, 0.1, 0.4, ..., 0.1]



[ Cosine, Manhattan, ..., Euclidean ]



[Cosine, Manhattan, ..., Euclidean]



[Cosine, Manhattan, ..., Euclidean]





[ Cosine, Manhattan, ..., Euclidean ]



MLP ( Attack Model )



Prediction whether two nodes are connected or not





- Three Datasets
  - Cora
  - CiteSeer
  - Pubmed

- Three Graph Neural Network Types
  - GraphSAGE
  - GAT
  - GCN (inductive)



- Attack 1
  - Same distribution



[0.3, 0.2, 0.3, ..., 0.1, 0.2, 0.1, 0.4, ..., 0.1]





- Attack 2
  - Same Distribution

[Cosine, Manhattan, ..., Euclidean]





- Attack 3
  - Different Distribution

[Cosine, Manhattan, ..., Euclidean]





#### Goal



- Observation
  - Inductive trained GNNs are likely to reveal sensitive information about their training graph
- Serious Concerns
  - Intellectual property
  - Confidentiality
  - Privacy

