

Active GNN

Reading list

- [Transfer Active Learning For Graph Neural Networks](#)
- [S2: An Efficient Graph Based Active Learning Algorithm with Application to Nonparametric Classification](#)
- [GRAPH-BASED ACTIVE LEARNING: A NEW LOOK AT EXPECTED ERROR MINIMIZATION](#)
- [Active Learning for Networked Data](#)
- [Towards Active Learning on Graphs: An Error Bound Minimization Approach](#)
- [Active Learning on Graphs via Spanning Trees](#)
- [Bayesian Semi-supervised Learning with Graph Gaussian Processes](#)
- [Batch Mode Active Learning for Networked Data](#)
- [FEW-SHOT LEARNING ON GRAPHS VIA SUPER-CLASSES BASED ON GRAPH SPECTRAL MEASURES](#)
- [Active Learning with Partial Feedback](#)
- [Active Learning for Graph Neural Networks via Node Feature Propagation](#)
- [Active Discriminative Network Representation Learning](#)

Reading notes

- [Transfer Active Learning For Graph Neural Networks](#)
 - Problem setup
 - Learn a policy that can sequentially select nodes to query the label
 - Transfer the policy to unseen graphs
 - Datasets
 - Cora, Citeseer, Pubmed
 - 1 citation network.
 - Active learning on single graphs.
 - Coauthor-CS, Cora-full
 - 1 co-author network, 1 citation network.
 - Transfer active learning on heterogeneous graphs.
 - PPI
 - 20 protein-protein interaction networks.

- Transfer active learning on homologous graphs.
- Method
 - Use reinforcement learning to learn the policy
 - State: graph, node features, and predictions of the GNN
 - Action: next node to query
 - Reward: performance gain on validation set
 - Policy network: $\pi(a = v|s) = \frac{1}{Z} \exp(w^T z_v^s)$
 - Transfer to unseen graphs
 - Train a universal policy by joint training on multiple graphs
 - Then train a graph-specific policy using distillation
- Baselines
 - Random, Entropy, Centrality
 - AGE: [Active learning for graph embedding](#)
- Open reviews
 - Missing baseline: [Active Discriminative Network Representation Learning](#)