

COMP396

Meeting 5

Summary

1. Recall
2. Relevant Papers
3. Supporting Material
4. JSD
5. Code
6. What I don't understand yet?

Recall

- We had an overview of DIM objective functions
- Discussed points to focus on (i.e infograph and JSD)

Relevant Papers

-LEARNING DEEP REPRESENTATIONS BY MUTUAL INFORMATION ESTIMATION AND MAXIMIZATION-><https://arxiv.org/pdf/1808.06670.pdf>

-DEEP GRAPH INFOMAX: <https://arxiv.org/pdf/1809.10341.pdf>

-INFOGRAPH: UNSUPERVISED AND SEMI-SUPERVISED GRAPH-LEVEL REPRESENTATION LEARNING VIA MUTUAL INFORMATION MAXIMIZATION-><https://arxiv.org/pdf/1908.01000.pdf>

Supporting Material

-Mean teachers are better role models: Weight-averaged consistency targets improve semi-supervised deep learning results-> <https://arxiv.org/pdf/1703.01780.pdf>

FASTGCN: FAST LEARNING WITH GRAPH CONVOLUTIONAL NETWORKS VIA IMPORTANCE SAMPLING-> <https://arxiv.org/pdf/1801.10247.pdf>

Mutual Information Neural Estimation-><https://arxiv.org/pdf/1801.04062.pdf>

f-GAN: Training Generative Neural Samplers using Variational Divergence Minimization
<https://arxiv.org/pdf/1606.00709.pdf>

HOW POWERFUL ARE GRAPH NEURAL NETWORKS?

JSD

The mutual information estimator modeled by discriminator T_ψ

$$I_{\phi,\psi}(\vec{h}_\phi^i(G); H_\phi(G)) := \mathbb{E}_{\mathbb{P}}[-\text{sp}(-T_{\phi,\psi}(\vec{h}_\phi^i(x), H_\phi(x)))] - \mathbb{E}_{\mathbb{P} \times \tilde{\mathbb{P}}}[\text{sp}(T_{\phi,\psi}(\vec{h}_\phi^i(x'), H_\phi(x)))] \quad (5)$$

Unsupervised learning

h is the summarized patch representation centered at node u and $H_\phi(G)$ is the global representation after applying READOUT.

$$\hat{\phi}, \hat{\psi} = \arg \max_{\phi, \psi} \sum_{G \in \mathbf{G}} \frac{1}{|G|} \sum_{u \in G} I_{\phi,\psi}(\vec{h}_\phi^u; H_\phi(G)). \quad (4)$$

Here we use GIN:

Semi-supervised learning

$$L_{\text{total}} = \sum_{i=1}^{|\mathbb{G}^L|} L_{\text{supervised}}(y_{\phi}(G_i), o_i) + \sum_{j=1}^{|\mathbb{G}^L|+|\mathbb{G}^U|} L_{\text{unsupervised}}(h_{\phi}(G_j); H_{\phi}(G_j)) \quad (7)$$

$$- \lambda \sum_{j=1}^{|\mathbb{G}^L|+|\mathbb{G}^U|} \frac{1}{|G_j|} \sum_{k=1}^K I(H_{\phi}^k(G_j); H_{\phi}^k(G_j)). \quad (8)$$

Code (setup)

<https://colab.research.google.com/drive/1XC2H9SzhT7myepPNJO8WX0hoDBRuhFcV>

NB: parameters are not correctly set as in the actual experiment

Materials & Resources

InfoGraph Code: <https://github.com/fanyun-sun/InfoGraph>

Deep Graph Infomax (Original): <https://github.com/PetarV-/DGI>

Deep Graph Infomax:

https://github.com/rusty1s/pytorch_geometric/blob/master/torch_geometric/nn/models/deep_graph_infomax.py

What I don't understand yet

- relationship between mean teacher and how mean teacher is used here (function) is ambiguous

To do:

Focus on global global (smiles + graph) + local global (node, readout) unsupervised DIM

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