

TTIC 31230, Fundamentals of Deep Learning

David McAllester, Autumn 2020

Deep Graphical Models Summary and Review

Deep Graphical Models

We are given a set of pairs $\langle x_1, \mathcal{Y}_1 \rangle, \dots, \langle x_N, \mathcal{Y}_T \rangle$ where each \mathcal{Y}_t is a “colored graph” — a graph with a label (color) assigned to each node. Semantic segmentation is an example where pixels of an image x are labeled with semantic categories.

We have a scoring function on colorings of a fixed graph.

$$s_{\Phi, x}(\hat{\mathcal{Y}}) = \sum_{n \in \text{Nodes}} s_{\Phi, x}^N[n, \hat{\mathcal{Y}}[n]] + \sum_{\langle n, m \rangle \in \text{Edges}} s_{\Phi, x}^E[\langle n, m \rangle, \hat{\mathcal{Y}}[n], \hat{\mathcal{Y}}[m]]$$

Deep Graphical Models

$$s_{\Phi,x}(\hat{\mathcal{Y}}) = \sum_{n \in \text{Nodes}} s_{\Phi,x}^N[n, \hat{\mathcal{Y}}[n]] + \sum_{\langle n, m \rangle \in \text{Edges}} s_{\Phi,x}^E[\langle n, m \rangle, \hat{\mathcal{Y}}[n], \hat{\mathcal{Y}}[m]]$$

$$P_{\Phi,x}(\hat{\mathcal{Y}}) = \underset{\hat{\mathcal{Y}}}{\text{softmax}} \ s_{\Phi,x}(\hat{\mathcal{Y}})$$

$$\mathcal{L}_{\Phi,x}(\mathcal{Y}) = -\ln P_{\Phi,x}(\mathcal{Y})$$

Deep Graphical Models

$$P_{\Phi,x}(\hat{\mathcal{Y}}) = \underset{\hat{\mathcal{Y}}}{\text{softmax}} \ s_{\Phi,x}(\hat{\mathcal{Y}})$$

To back-propagate from the loss $-\ln P_{\Phi,x}(\mathcal{Y})$ through the exponential softmax to the tensors $S_{\Phi,x}^N$ and $s_{\Phi,x}^E$ we need to compute or estimate

$$P_{\Phi,x}(\hat{\mathcal{Y}}[n] = y) \quad \text{and} \quad P_{\Phi,x}(\hat{\mathcal{Y}}[\langle n, m \rangle] = \langle y, y' \rangle)$$

for each node n and value y and each edge $\langle n, m \rangle$ and value $\langle y, y' \rangle$.

END