TTIC 31230, Fundamentals of Deep Learning

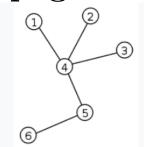
David McAllester, Autumn 2020

Loopy Belief Propagation (Loopy BP)

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We design an algorithm that is correct for tree graphs and use it on non-tree (loopy) graphs.

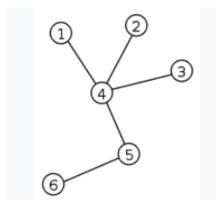
Belief Propagation on Trees



Belief Propagation is a message passing procedure (actually dynamic programming).

For each edge $\{n, m\}$ and possible value y for node n we define the message $\mathbb{Z}_{m \to n}[y]$ from m to n to be the partition function for the subtree attached to n through m and with $\hat{\mathcal{Y}}[n]$ restricted to y.

Dynamic Programming Computes the Messages



$$Z_{m \to n}[y] = \sum_{y'} e^{s^N[m,y'] + s^E[\langle m, n \rangle, y', y]} \left(\prod_{k \in N(m), k \neq n} Z_{k \to m}[y'] \right)$$

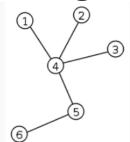
Loopy BP

In a Loopy Graph we can initializing all message $Z_{n\to m}[y] = 1$ and then repeating (until convergence) the updates

$$\tilde{Z}_{m \to n}[y] = \frac{1}{Z_{m \to n}} Z_{m \to n}[y] \qquad Z_{m \to n} = \sum_{y} Z_{m \to n}[y]$$

$$Z_{m\to n}[y] = \sum_{y'} e^{s^N[m,y']+s^E[m,n,y',y]} \left(\prod_{k\in N(m),\ k\neq n} \tilde{Z}_{k\to m}[y'] \right)$$

Computing Node Marginals from Messages



$$Z^{N}(y) = \sum_{\hat{\mathcal{Y}}: \hat{\mathcal{Y}}[n]=y} e^{s(\hat{\mathcal{Y}})}$$

$$= e^{s^{N}[y]} \left(\prod_{m \in N(n)} Z_{m \to n}[y] \right)$$

$$P^{N}(y) = Z^{N}(y)/Z, \quad Z = \sum_{y} Z^{N}(y)$$

Computing Edge Marginals from Messages

$$Z_{n,m}(y,y') \doteq \sum_{\hat{\mathcal{Y}}: \hat{\mathcal{Y}}[n]=c, \hat{\mathcal{Y}}[m]=y'} e^{s(\mathcal{Y})}$$

$$= e^{s^{N}[n,y]+s^{N}[m,y']+s^{E}[n,m,y,y']}$$

$$\prod_{m \in N(n), k \neq m} Z_{m \to n}[y]$$

$$m \in N(m), k \neq m$$

$$\sum_{m \in N(m), k \neq n} Z_{m \to n}[y']$$

$$p_{n,m}(y,y') = Z_{n,m}(y,y')/Z \quad Z = \sum_{y,y'} Z_{n,m}(y,y')$$

\mathbf{END}