

# CRF w/ feature-based potentials

Data:

CRF:  $D = \{x^{(n)}, y^{(n)}\}_{n=1}^N$   
 MRF:  $D = \{y^{(n)}\}_{n=1}^N$

Model:

CRF:  $p(\vec{y}|\vec{x}, \vec{\theta}) = \frac{1}{Z(\vec{x}, \vec{\theta})} \prod_{c \in C} \psi_c(\vec{y}_c, \vec{x})$  where  $\psi_c(\vec{y}_c, \vec{x}) = \exp[\vec{\theta} \cdot \hat{f}_c(\vec{y}_c, \vec{x})]$

MRF:  $p(\vec{y}|\vec{\theta}) = \frac{1}{Z(\vec{\theta})} \prod_c \psi_c(\vec{y}_c)$  where  $\psi_c(\vec{y}_c) = \exp[\vec{\theta} \cdot \hat{f}_c(\vec{y}_c)]$

Avg. Log-likelihood:

CRF:  $\tilde{\ell}(\vec{\theta}; D) = \frac{1}{N} \sum_{n=1}^N \log p(\vec{y}^{(n)} | \vec{x}^{(n)}, \vec{\theta})$

MRF:  $\tilde{\ell}(\vec{\theta}; D) = \frac{1}{N} \sum_{n=1}^N \log p(\vec{y}^{(n)} | \vec{\theta})$

Derivative:

CRF:  $\frac{d\tilde{\ell}(\vec{\theta}; D)}{d\theta_k} = \frac{1}{N} \left( \sum_{n=1}^N \sum_c f_{c,k}(\vec{y}_c^{(n)}, \vec{x}^{(n)}) \right) - \frac{1}{N} \left( \sum_{n=1}^N \sum_c \sum_{y_c} \frac{p(\vec{y}_c | \vec{x}^{(n)})}{f_{c,k}(\vec{y}_c, \vec{x}^{(n)})} \right)$

MRF:  $\frac{d\tilde{\ell}(\vec{\theta}; D)}{d\theta_k} = \frac{1}{N} \left( \sum_n \sum_c f_{c,k}(\vec{y}_c^{(n)}) \right) - \frac{1}{N} \left( \sum_n \sum_c \sum_{y_c} \frac{p(\vec{y}_c)}{f_{c,k}(\vec{y}_c)} \right)$

*inference*

