

**GRAPHICAL MARKOV MODELS (WS2017)**  
**4. SEMINAR**

**Assignment 1.** Consider the following probabilistic model for real valued sequences  $\mathbf{x} = (x_1, \dots, x_n)$ ,  $x_i \in \mathbb{R}$  of fixed length  $n$ . Each sequence is a combination of a leading part  $i \leq k$  and a trailing part  $i > k$ . The boundary  $k = 1, \dots, n$  is random with some categorical distribution  $\boldsymbol{\pi} \in \mathbb{R}_+^n$ ,  $\sum_k \pi_k = 1$ . The values  $x_i$ , in the leading and trailing part are statistically independent and distributed with some probability density function  $p_1(x)$  and  $p_2(x)$  respectively. Altogether the distribution for pairs  $(\mathbf{x}, k)$  reads

$$p(\mathbf{x}, k) = \pi_k \prod_{i=1}^k p_1(x_i) \prod_{j=k+1}^n p_2(x_j). \quad (1)$$

The densities  $p_1$  and  $p_2$  are known. Given an i.i.d. sample of sequences  $\mathcal{T}^m = \{\mathbf{x}^\ell \in \mathbb{R}^n \mid \ell = 1, \dots, m\}$ , the task is to estimate the unknown boundary distribution  $\boldsymbol{\pi}$  by the EM-algorithm.

**a)** The E-step of the algorithm requires to compute the values of auxiliary variables  $\alpha_\ell^{(t)}(k) = p(k \mid \mathbf{x}^\ell)$  for each example  $\mathbf{x}^\ell$  given the current estimate  $\boldsymbol{\pi}^{(t)}$  of the boundary distribution. Give a formula for computing these values from model (1).

**b)** The M-step requires to solve the optimisation problem

$$\frac{1}{m} \sum_{\ell=1}^m \sum_{k=1}^n \alpha_\ell^{(t)}(k) \log p(\mathbf{x}^\ell, k) \rightarrow \max_{\boldsymbol{\pi}}$$

Substitute the model (1) and solve the optimisation task.

**Assignment 2.** Prove equivalence of Definitions 1a and 1b given in the lecture for a Markov model on a tree (see Sec. 10).

**Assignment 3.** Let  $T = (V, E)$  be an undirected tree and  $s_i, i \in V$  be a field of  $K$ -valued random variables. Suppose that  $v_{ij}(k, k'), k, k' \in K$  is a system of pairwise probabilities associated with the edges  $\{i, j\} \in E$  of the tree. Consider the set  $\mathcal{P}(\mathbf{v})$  of all joint probability distributions  $p(s)$ , which have  $\mathbf{v}$  as pairwise marginals, i.e.

$$\sum_{s \in K^{|V|}} p(s) \delta_{s_i k} \delta_{s_j k'} = v_{ij}(k, k') \quad \forall \{i, j\} \in E, \forall k, k' \in K.$$

We want to find the distribution with highest entropy

$$H(p) = - \sum_{s \in K^{|V|}} p(s) \log p(s)$$

in  $\mathcal{P}(\mathbf{v})$ . Prove that the unique maximiser is the Markov model on the tree  $T$  defined by the edge marginals  $\mathbf{v}$ .

*Hint:* Formulate and solve the constrained optimisation task by using its Lagrange function.