

**STRUCTURED MODEL LEARNING (SS2015)**  
**2. SEMINAR**

**Assignment 1.** Consider a collection  $\{S_i \mid i = 1, 2, 3\}$  of three binary valued random variables, i.e.,  $s_i \in \{0, 1\}$  for  $i = 1, 2, 3$ . Suppose we fix all three pairwise marginal distributions  $p_{ij}(s_i, s_j) = \mu(s_i, s_j)$ , where

$$\mu(s_i, s_j) = \begin{cases} \frac{1}{2}\alpha & \text{if } s_i = s_j, \\ \frac{1}{2}(1 - \alpha) & \text{otherwise} \end{cases}$$

and  $\alpha$  is a fixed real number from the interval  $[0, 1]$ . We seek a simple joint distribution  $p(s_1, s_2, s_3)$  that has the given marginals. Someone proposes the properly normalised product of  $\mu$ -s as an answer

$$\bar{p}(s_1, s_2, s_3) = \frac{1}{Z(\alpha)} \mu(s_1, s_2) \mu(s_2, s_3) \mu(s_1, s_3).$$

Prove that the pairwise marginal distributions of  $\bar{p}$  do not(!) coincide with the function  $\mu$ .

**Assignment 2.** Consider a collection  $\{S_i \mid i = 1, 2, 3\}$  of three binary valued random variables as in the previous assignment. Let us fix the following pairwise marginal distributions

$$p(s_1, s_2) = \mu(s_1, s_2), \quad p(s_1, s_3) = \mu(s_1, s_3), \quad p(s_2, s_3) = \tilde{\mu}(s_2, s_3)$$

where

$$\mu(s_i, s_j) = \begin{cases} 0.5 & \text{if } s_i = s_j, \\ 0 & \text{otherwise,} \end{cases} \quad \tilde{\mu}(s_i, s_j) = \begin{cases} 0.5 & \text{if } s_i \neq s_j, \\ 0 & \text{otherwise,} \end{cases}$$

Do the  $\mu$ -s represent a valid system of pairwise marginal distributions? I.e., does there exist a joint distribution  $p(s_1, s_2, s_3)$  that has pairwise marginals coinciding with the  $\mu$ -s?

**Assignment 3.** Let  $S = \{S_i \mid i \in V\}$  be a  $K$ -valued random field and let  $\mathcal{P}$  denote the set of all possible joint probability distributions  $p: K^{|V|} = \mathcal{S} \rightarrow \mathbb{R}_+$ , s.t.  $\sum_{s \in \mathcal{S}} p(s) = 1$ .

**a)** Prove that the distribution  $p \in \mathcal{P}$  with highest entropy is the uniform distribution. Prove that it factorises into the product of its unary marginal distributions.

**b)** Let us fix unary marginal distributions for each  $S_i, i \in V$  by  $p(s_i) = \mu_i(s_i)$ . We assume that the functions  $\mu_i: K \rightarrow \mathbb{R}_{++}$  fulfil  $\sum_{k \in K} \mu_i(k) = 1$  for all  $i \in V$ .

Prove that the distribution

$$p(s) = \prod_{i \in V} \mu_i(s_i)$$

has the highest entropy among all joint distributions  $p \in \mathcal{P}$  which have the given unary marginals. What happens if the functions  $\mu_i$  are not necessarily strictly positive?

**c)** We equip  $V$  with the structure of an undirected graph  $(V, E)$ . Let us fix pairwise marginal distributions for each pair of variables  $S_i, S_j$  where  $\{i, j\} \in E$  by setting  $p(s_i, s_j) =$

$\mu_{ij}(s_i, s_j)$ . All functions  $\mu_{ij}: K^2 \rightarrow \mathbb{R}_{++}$  fulfil

$$\sum_{k, k' \in K} \mu_{ij}(k, k') = 1.$$

Furthermore, we assume that the system of  $\mu$ -s represents a valid system of pairwise marginals, i.e. there exists at least one strictly positive joint distribution  $\bar{p} \in \mathcal{P}$  whose pairwise marginal distributions coincide with  $\mu$ -s.

Fill in details for the derivation in Section 2.B of the lecture and prove that the distribution  $p \in \mathcal{P}$  with highest entropy (among all those that have the given fixed pairwise marginals) has the form

$$p_u(s) = \frac{1}{Z(u)} \exp \left[ \sum_{ij \in E} u_{ij}(s_i, s_j) \right],$$

where  $u$ -s are Lagrange multipliers which have to be determined such that  $p$  has the required pairwise marginals.

**Assignment 4.** Consider a  $K$ -valued Gibbs random field  $\{S_i \mid i \in V\}$ ,  $s_i \in K$  on an undirected graph  $(V, E)$  with joint probability distribution

$$p_u(s) = \frac{1}{Z(u)} \exp \left[ \sum_{ij \in E} u_{ij}(s_i, s_j) \right].$$

Try to find all possible re-parametrisations of the  $u$ -s.

*Reminder:* Simply put, re-parametrisation means that different potentials  $u \neq \tilde{u}$  may define the same distribution  $p_u = p_{\tilde{u}}$ .