

**GRAPHICAL MARKOV MODELS (WS2016)**  
**4. SEMINAR: ADDITIONAL HINTS**

**Assignment 1.** None

**Assignment 2.** The non-trivial direction is Definition 1b  $\Rightarrow$  Definition 1a. Assume that

$$p(s) = \prod_{\{i,j\} \in E} g_{ij}(s_i, s_j)$$

where  $T = (V, E)$  is an undirected tree and  $s_i \in K, i \in V$  are random variables.

Choose any vertex  $r \in V$  as root and denote the corresponding edge orientations by  $\vec{E}_r$ . Choose a leaf  $l \in V$  and denote by  $m$  the only vertex connected to  $l$  by an edge. Prove that the conditional distribution  $p(s_l | s_m)$  can be written as

$$p(s_l | s_m) = g_{ml}(s_m, s_l) b(s_m),$$

where  $b$  is some function which depends only on the value of  $s_m$ . Denote the sub-tree obtained by removing the node  $l$  by  $T' = (V', E')$  and let  $s'$  be a realisation on this sub-tree. Deduce that

$$p(s') = a(s_m) \prod_{\{i,j\} \in E'} g_{ij}(s_i, s_j),$$

where  $a$  is some function which depends only on the value of  $s_m$ .

Apply this recursively and conclude that Definition 1a follows from Definition 1b.

**Assignment 3.** Notice that the function  $x \log x$  is defined for non-negative  $x \geq 0$  and is convex. Notice that there is a constraint for every edge  $\{i, j\} \in E$  and any pair  $k, k' \in K$ . Consequently, there will be a Lagrange multiplier  $\lambda_{ij}(k, k')$  for each of these constraints. Consider the minimisation of the Lagrange function w.r.t. the variables  $p(s)$  and conclude that the optimal joint distribution is a product of factors, each on depending only on a pair of random variables (for edges of  $T$ ).