

## 1. Random fields (undirected graphical models)

- A  $K$ -valued random field is a collection  $\{S_i \mid i \in V\}$  of  $K$ -valued random variables
- $i \in V$  can be: pixels, object parts, etc.
- $K \in K$  can be: colours, segment labels, depth values, poses of object parts, etc.
- $S \in K^V \cong \mathcal{S}$  denotes realisations  $S_i \in K, i \in V$
- $p \in \mathcal{P}$  denotes a p.d. or p.m.  $p: \mathcal{S} \rightarrow \mathbb{R}_+$

### 1A. Exponential families

- $S = \{S_i \mid i \in V\}$  is a  $K$ -valued random field
- $\Phi: \mathcal{S} = K^V \rightarrow \mathbb{R}^n$  is a random vector

Consider the task

$$\inf_{p \in \mathcal{P}} \left\{ \sum_{S \in \mathcal{S}} p(S) \log p(S) \mid \mathbb{E}_p(\Phi) = \mu, \sum_{S \in \mathcal{S}} p(S) = 1 \right\}$$

Lagrange function

$$L(p) = \sum_{S \in \mathcal{S}} p(S) \log p(S) - \langle u, \mathbb{E}_p(\Phi) - \mu \rangle - \lambda \left[ \sum_{S \in \mathcal{S}} p(S) - 1 \right]$$

$$\dots \rightarrow p_u(S) = \exp[\langle u, \Phi(S) \rangle - \log Z(u)],$$

where  $u \in \mathbb{R}^n$ ,  $\lambda \in \mathbb{R}$  are Lagrange ~~mul~~ multipliers and  $Z(u)$  is the normalisation constant

$$Z(u) = \sum_{S \in \mathcal{S}} e^{\langle u, \Phi(S) \rangle}$$

### Questions

(1) Which  $\mu \in \mathbb{R}^n$  qualify?

$$\mathbb{E}_p(\Phi) = \sum_{S \in \mathcal{S}} p(S) \Phi(S) = \mu \quad p(S) \geq 0 \quad \forall S \in \mathcal{S}$$

$$\sum_{S \in \mathcal{S}} p(S) = 1$$

$\Rightarrow \mu$  is in the relative interior of  $\text{conv } \Phi(\mathcal{S})$ , where

$$\Phi(\mathcal{S}) = \{\Phi(s) \mid s \in \mathcal{S}\} \subset \mathbb{R}^n$$

(2)  $u \mapsto \mu$ ? Yes, because

$$\begin{aligned} \nabla \log Z(u) &= \nabla \log \sum_{s \in \mathcal{S}} e^{\langle u, \Phi(s) \rangle} \\ &= \frac{1}{Z(u)} \sum_{s \in \mathcal{S}} e^{\langle u, \Phi(s) \rangle} \Phi(s) = \mathbb{E}_u(\Phi) = \mu \end{aligned}$$

(3)  $\mu \mapsto u$ ? (provided that  $\mu \in \text{ri}(\text{conv } \Phi(\mathcal{S}))$ )

No, not always.  $\mu$  is unique, but not  $u$ !

$$\langle u, \Phi(s) \rangle - \log Z(u) = \langle \tilde{u}, \Phi(s) \rangle - \log Z(\tilde{u}) \quad \forall s \in \mathcal{S}$$

i.e.

$$\langle u - \tilde{u}, \Phi(s) \rangle = \text{const}_s \quad \forall s \in \mathcal{S}$$

$$\{u \in \mathbb{R}^n \mid \langle u, \Phi(s) \rangle = \text{const}_s, \forall s \in \mathcal{S}\} = L^\perp$$

$$\text{aff } \Phi(\mathcal{S}) = \Phi(s_0) - \underbrace{\text{span}(\Phi(\mathcal{S}) - \Phi(s_0))}_L \quad \text{for any } s_0 \in \mathcal{S}$$

Suppose we write

$$\text{aff } \Phi(\mathcal{S}) = \{u \in \mathbb{R}^n \mid Au = b\} \quad \text{with some } A: \mathbb{R}^n \rightarrow \mathbb{R}^m,$$

then  $L = \text{Ker } A$  and  $L^\perp = \text{Im } A^T$ .

Answer:

• if  $\text{aff } \Phi(\mathcal{S}) = \mathbb{R}^n \Rightarrow \mu \mapsto u$  is a mapping

• if  $\text{aff } \Phi(\mathcal{S}) = \{u \in \mathbb{R}^n \mid Au = b\} \neq \mathbb{R}^n \Rightarrow$

we can re-parametrise  $u$  by

$$u \rightarrow u + A^T \psi, \quad \psi \in \mathbb{R}^m$$

and have

$$\langle u + A^T \psi, \Phi(s) \rangle = \langle u, \Phi(s) \rangle + \langle \psi, b \rangle$$

$$\log Z(u + A^T \psi) = \log Z(u) + \langle \psi, b \rangle$$

### B. Graphical models on undirected graphs

- $S = \{S_i \mid i \in V\}$  is a  $K$ -valued random field
- $(V, E)$  is an undirected graph
- Let us fix marginal distributions  $p(s_i, s_j) = \mu_{ij}(s_i, s_j)$  for all edges  $\{i, j\} \in E$  and all  $s_i, s_j \in K$

We search the p.d. with maximal entropy (i.e. minimal information) under the above constraints. Everything said in A. applies because any marginal prob.  $p(s_i = k, s_j = k')$  can be seen as expectation of a random variable  $\Phi_{ij, kk'}(s) = \delta_{s_i, k} \delta_{s_j, k'}$ .

Hence, if  $\mu_{ij}(s_i, s_j)$  define a valid set of marginal distributions, then the task has a unique solution

$$p_u(s) = \frac{1}{Z(u)} \exp \sum_{j \in E} u_{ij}(s_i, s_j) = \frac{1}{Z(u)} \exp \langle u, \Phi(s) \rangle,$$

where

$$\Phi(s) = (\Phi_{ij, kk'}(s))_{j \in E, k, k' \in K} \in \mathbb{R}^{|\mathcal{E}| |K|^2}$$

However, the potentials  $u_{ij}: K^2 \rightarrow \mathbb{R}$  are defined up to re-parametrisations only.

The distribution  $p_u$  is a Gibbs random field and factorises over the edges of the graph  $(V, E)$ .

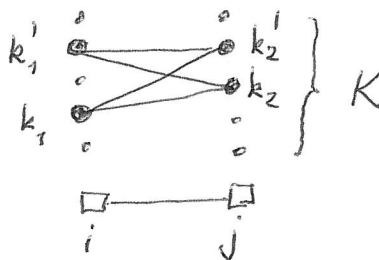
Remark 1 All this can be generalised to (undirected) hypergraphs

Let us now analyse the possible re-parametrisations

$$u \rightarrow u + v \quad \text{s.t.} \quad \langle v, \Phi(s) \rangle = \text{const}_s$$

a) Fix an edge  $\{ij\}$  and consider four labellings  $s^k$ ,  $k=1,2,3,4$  which coincide on  $V \setminus \{ij\}$  and differ on  $\{ij\}$  as follows

$$\begin{aligned} (s_i^1, s_j^1) &= (k_1, k_2) \\ (s_i^2, s_j^2) &= (k_1, k_2') \\ (s_i^3, s_j^3) &= (k_1', k_2) \\ (s_i^4, s_j^4) &= (k_1', k_2') \end{aligned}$$



We have

$$\begin{aligned} \langle v, \Phi(s^1) + \Phi(s^4) - \Phi(s^2) - \Phi(s^3) \rangle &= \\ &= v_{ij} \cdot (k_1, k_2) + v_{ij} \cdot (k_1', k_2') - v_{ij} \cdot (k_1, k_2') - v_{ij} \cdot (k_1', k_2) \stackrel{!}{=} 0 \end{aligned}$$

This holds for any edge  $\{ij\}$  and any  $k_1, k_1', k_2, k_2'$ .  
It follows that

$$v_{ij}(s_i, s_j) = \psi_{ij}^-(s_i) + \psi_{ji}^+(s_j)$$

with some functions  $\psi_{ij}^-(s_i)$  for directed edges

b) Fix a node  $i \in V$  and consider two labellings  $s^1, s^2$  which differ on  $i$  only:  $s_i^1 = k$ ,  $s_i^2 = k'$ .

We have

$$\begin{aligned} \langle v, \Phi(s^1) \rangle - \langle v, \Phi(s^2) \rangle &= \\ &= \sum_{j \in V_i} \psi_{ij}^-(s_j = k) - \sum_{j \in V_i} \psi_{ij}^-(s_j = k') \stackrel{!}{=} 0 \end{aligned}$$

This holds for any node  $i \in V$  and any  $k, k' \in K$ .

Hence, it follows that

$$\sum_{j \in W_i} \psi_{ij}^-(s_i) = \text{const}_i$$

Finally, we have all possible re-parametrisations given by

$$u_{ij}(s_i, s_j) \rightarrow \psi_{ij}^-(s_i) + u_{ij}(s_i, s_j) + \psi_{ji}^-(s_j)$$

$$\text{s.t.} \quad \sum_{j \in W_i} \psi_{ij}^-(s_i) = c_i$$

□

Problems: All tasks

- given potentials  $u_{ij}: K^2 \rightarrow \mathbb{R} \forall \{ij\} \in E$ , compute  $Z(u)$  and/or marginal probabilities  $p_u(s_i)$ ,  $p_u(s_i, s_j)$
- check whether  $\mu_{ij}: K^2 \rightarrow \mathbb{R}_+ \forall \{ij\} \in E$  represent a valid system of pairwise marginal prob's
- given a valid system of pairwise marginals  $\mu_{ij}: K^2 \rightarrow \mathbb{R}_+ \forall \{ij\} \in E$ , compute the potentials  $u_{ij}$
- given the potentials  $u_{ij}: K^2 \rightarrow \mathbb{R} \forall \{ij\} \in E$ , find the most probable realisations (labellings)

$$\begin{aligned} \operatorname{argmax}_{s \in \mathcal{S}} p_u(s) &= \\ &= \operatorname{argmax}_{s \in \mathcal{S}} \sum_{ij \in E} u_{ij}(s_i, s_j) \end{aligned}$$

are NP-hard. Polynomial time complexity algorithms exist if  $(V, E)$  is acyclic.