

## OPTIMAL TREE DENSITY APPROXIMATORS

## Why

Which is the optimal tree to use for tree density approximation?

## **Definition**

We want to choose the tree whose corresponding approximator for the given density achieves minimum relative entropy with the given density among all tree density approximators. We call such a density an *optimal tree approximator* of the given density. We call a tree according to which an optimal tree approximator factors and optimal approximator tree.

## Result

**Prop.** 1. Let  $g : \mathbb{R}^n \to [0,1]$  be a density. A tree T on  $\{1,\ldots,n\}$  is an optimal approximator tree if and only if it is a maximal spanning tree of the differential mutual information graph of q.

*Proof.* First, denote the optimal approximator of g for tree T by  $f_T^*$ . Recall

$$f_T^* = f_1 \prod_{i 
eq 1} f_{i|\mathsf{pa}\,i|}$$

Second, recall d(g, f) = H(g, f) - H(g). Since H(g) does not depend on f, f is a minimizer of d(g, f) if and only if it is a minimizer of H(g, f).

Third, express the cross entropy of  $f_T^*$  relative to g as

$$H(q, p_T^*) = h(q_1) - \sum_{j \neq i} \left( \int_{\mathbf{R}^d} g(x) \log g_{i|pai}(x_i, x_{\mathbf{pa}i}) dx \right)$$

$$= H(q_1) - \sum_{i \neq 1} \sum_{a \in A} q(a) (\log q_{i,\mathbf{pa}i}(a_i, a_{\mathbf{pa}i}) - \log q_{\mathbf{pa}i}(a_{\mathbf{pa}i}))$$

$$= H(q_1) - \sum_{i \neq 1} \sum_{a \in A} q(a) (\log q_{i,\mathbf{pa}i}(a_i, a_{\mathbf{pa}i}) - \log q_{\mathbf{pa}i}(a_{\mathbf{pa}i}) - \log q_{\mathbf{pa}i}(a_{\mathbf{pa}i}) - \log q_{\mathbf{pa}i}(a_{\mathbf{pa}i}) - \log q_{\mathbf{pa}i}(a_{\mathbf{pa}i})$$

$$= \sum_{i=1}^n H(q_i) - \sum_{i \neq 1} I(q_i, q_{\mathbf{pa}i})$$

$$= \sum_{i=1}^n H(q_i) - \sum_{i \neq 1} I(q_i, q_j)$$

where  $\mathbf{pa}i$  denotes the parent of vertex i in T (i = 2, ..., n).  $H(g_i)$  does not depend on the choice of tree. Choosing a tree to minimize the second term in the final expression above is equivalent to choosing a maximal spanning tree from the weighted graph with differential mutual information edge weights; namely, the mutual information graph of g.

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