

Softmax

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My opinion on how and why softmax materializes.
Consider the transformation

$$\mathbf{p} = \operatorname{argmin}_{\mathbf{w} \in \mathbf{S}} \langle \mathbf{w}, \mathbf{x} \rangle \quad (1)$$

where $\mathbf{x} \in \mathbf{R}^d$, $\mathbf{w} \in \mathbf{S} \subset [0, 1]^d$ such that $\mathbf{S} = \left\{ \mathbf{w} \mid \sum_i^d \mathbf{w}_i = 1 \right\}$, $\mathbf{p} \in [0, 1]^d$ and $\sum_{i=1}^d \mathbf{p}_i = 1$.
 $\hat{\mathbf{p}}$ with entropy regularizer

$$\hat{\mathbf{p}} = \operatorname{argmin}_{\mathbf{w} \in \mathbf{S}} \langle \mathbf{w}, \mathbf{x} \rangle + \sum_{i=1}^d \mathbf{w}_i \log(\mathbf{w}_i) \quad (2)$$

Since $\mathbf{w} \in \mathbf{S}$, add a Lagrange multiplier $\lambda (\langle \mathbf{w}, \mathbf{1} \rangle - 1)$ to the objective function.

$$\hat{\mathbf{p}} = \operatorname{argmin}_{\mathbf{w} \in \mathbf{S}} \langle \mathbf{w}, \mathbf{x} \rangle + \sum_{i=1}^d \mathbf{w}_i \log(\mathbf{w}_i) + \lambda (\langle \mathbf{w}, \mathbf{1} \rangle - 1) \quad (3)$$

$$= \operatorname{argmin}_{\mathbf{w} \in \mathbf{S}} \sum_{i=1}^d \mathbf{w}_i \mathbf{x}_i + \sum_{i=1}^d \mathbf{w}_i \log(\mathbf{w}_i) + \lambda \left(\sum_{i=1}^d \mathbf{w}_i - 1 \right) \quad (4)$$

$$(5)$$

Differentiate the objective function with respect to \mathbf{w}_i and equate to 0

$$\mathbf{x}_i + 1 + \log(\mathbf{w}_i) + \lambda = 0 \quad (6)$$

$$\mathbf{w}_i^* = \exp(-\mathbf{x}_i) \exp(-1 - \lambda) \quad (7)$$

$$= \frac{\exp(-\mathbf{x}_i)}{\exp(1 + \lambda)} \quad (8)$$

Set λ such that $\sum_i^d \mathbf{w}_i^* = 1$

$$\mathbf{w}_i^* = \frac{\exp(-\mathbf{x}_i)}{\sum_{i=1}^d \exp(-\mathbf{x}_i)} \quad (9)$$

$$\hat{\mathbf{p}} = \mathbf{w}^* \quad (10)$$

Reference

Luca Trevison. The “Follow-the-Regularized-Leader” algorithm. Topics in computer science and optimization (Fall 2019).