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Exercise Sheet 15

Exercise 1: RBM with Ternary Hidden Units (20+10 P)

We consider a variant of the restricted Boltzmann machine with ternary hidden units $h \in \{-1,0,1\}^H$. Input features remain binary, i.e. $x \in \{0,1\}^d$, like for the classical RBM. The probability model is given by:

$$p(\boldsymbol{x}, \boldsymbol{h} | \boldsymbol{\theta}) = \frac{1}{\mathcal{Z}} \exp \left(\sum_{j=1}^{H} \boldsymbol{w}_{j}^{\top} \boldsymbol{x} \cdot h_{j} + \sum_{j=1}^{H} h_{j} b_{j} \right)$$

where $\theta = (\boldsymbol{w}_j, b_j)_{j=1}^H$ are the parameters of the model, and where and \mathcal{Z} is the partition function that normalizes the probability distribution to 1.

(a) Show that this modified RBM can also be expressed as a product of experts

$$p(\boldsymbol{x}|\theta) = \frac{1}{\mathcal{Z}} \prod_{j=1}^{H} g_j(\boldsymbol{x}, \theta_j),$$

with

$$g_j(\boldsymbol{x}, \theta_j) = 1 + 2 \cosh(\boldsymbol{w}_j^{\top} \boldsymbol{x} + b_j),$$

where cosh is the hyperbolic cosine function.

(b) Show that the gradient of the log-likelihood assigned to some data point x_n by the modified RBM has the form

$$\forall_{j=1}^{H}: \frac{\partial \log p(\boldsymbol{x}_{n}|\boldsymbol{\theta})}{\partial \boldsymbol{w}_{j}} = \boldsymbol{x}_{n} \cdot \sigma(\boldsymbol{w}_{j}^{\top} \boldsymbol{x}_{n} + b_{j}) - \mathbb{E}_{\boldsymbol{x} \sim p(\boldsymbol{x}|\boldsymbol{\theta})} [\boldsymbol{x} \cdot \sigma(\boldsymbol{w}_{j}^{\top} \boldsymbol{x} + b_{j})]$$

$$\forall_{j=1}^{H}: \frac{\partial \log p(\boldsymbol{x}_{n}|\boldsymbol{\theta})}{\partial b_{i}} = \sigma(\boldsymbol{w}_{j}^{\top} \boldsymbol{x}_{n} + b_{j}) - \mathbb{E}_{\boldsymbol{x} \sim p(\boldsymbol{x}|\boldsymbol{\theta})} [\sigma(\boldsymbol{w}_{j}^{\top} \boldsymbol{x} + b_{j})]$$

where
$$\sigma(t) = \frac{\sinh(t)}{0.5 + \cosh(t)}$$
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Exercise 2: Product of Gaussian Mixture Models (20 + 10 P)

Consider the product of experts:

$$p(\boldsymbol{x}|\theta) = \frac{1}{\mathcal{Z}} \prod_{j=1}^{H} g_j(\boldsymbol{x}, \theta_j)$$

where each expert is a Gaussian mixture model in d-dimensions, and where each element of the mixture is Gaussian with identity covariance:

$$\forall_{j=1}^{H}: g_{j}(\boldsymbol{x}, \theta_{j}) = \sum_{k=1}^{C} \alpha_{jk} \frac{1}{(2\pi)^{d/2}} \exp\left(-\frac{1}{2} \|\boldsymbol{x} - \boldsymbol{\mu}_{jk}\|^{2}\right).$$

(a) Show that $p(\mathbf{x}|\theta)$ can be rewritten as a mixture of C^H elements, where each mixture element (indexed by the vector $\mathbf{k} \in \{1, \dots, C\}^H$) has center

$$m_k = \frac{1}{H} \sum_{j=1}^{H} \mu_{jk_j}.$$

(b) Give the centers m_k of the mixture model equivalent to a product of two mixture models, where each mixture model in the product has 2 elements, where the first mixture has the two-dimensional centers $\mu_{11} = \binom{2}{0}$ and $\mu_{12} = \binom{4}{0}$, and where the second mixture has the two-dimensional centers $\mu_{21} = \binom{0}{2}$ and $\mu_{22} = \binom{4}{0}$.

Exercise 3: Programming (40 P)

Download the programming files on ISIS and follow the instructions.