

TODAY

- ① RBM w/ GAUSSIAN HIDDEN NODES
- ② STACKED RBM / CONVOLUTIONAL NETS

THE BOLTZMANN LEARNING TRICK: SEPARATE

"ESTIMATED" VALUES

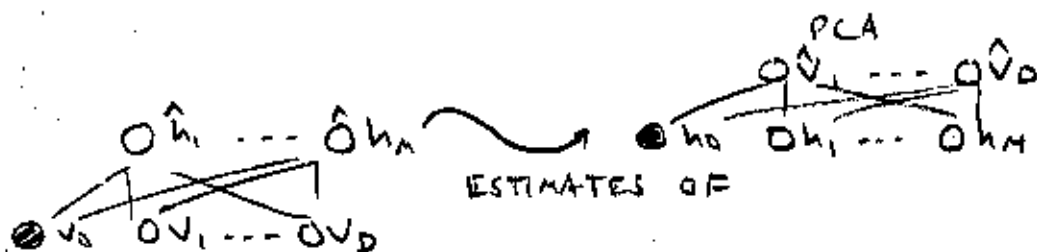
$$\hat{h} = \tilde{c} + W\tilde{v}^{(n)}$$

$$\hat{v}^{(n)} = \tilde{b} + W^T \hat{h}^{(n)}$$

"TRUE" VALUES

$$\tilde{v}^{(n)} = \text{OBSERVATION}$$

$$\tilde{h}^{(n)} = \text{TRUE } M\text{-DIMENSIONAL}$$



MMSE TRAINING:

$$E = \frac{1}{2} \sum_{n=1}^N \left\| \tilde{v}^{(n)} - (W^T \hat{h}^{(n)} + \tilde{b}) \right\|^2 + \frac{1}{2} \sum_{n=1}^N \left\| \tilde{h}^{(n)} - (W \tilde{v}^{(n)} + \tilde{c}) \right\|^2$$

TRUE VALUE
 $\hat{v}^{(n)}$

"TRUE" VALUE
 $\tilde{h}^{(n)}$

$$W = \text{argmin } E(W) = \text{argmax } \mathcal{L}(W)$$

$$\mathcal{L} = \sum_{n=1}^N \ln p(\tilde{v}^{(n)} | \hat{v}^{(n)}) + \sum_{n=1}^N \ln p(\tilde{h}^{(n)} | \hat{h}^{(n)})$$

$$p(\tilde{v}^{(n)} | \hat{v}^{(n)}) = \frac{1}{(2\pi)^{D/2}} e^{-\frac{1}{2} \|\tilde{v}^{(n)} - \hat{v}^{(n)}\|^2}$$

STOCHASTIC LEARNING

① Draw $\vec{h}^{(n)}$ at random from $\mathcal{N}(\vec{h}^{(n)}; W\vec{J}^{(n)} + \vec{c}, I)$

$$\textcircled{2} w_{ji} \leftarrow w_{ji} - \frac{\eta}{2} \frac{\partial}{\partial w_{ji}} \|\vec{h}^{(n)} - (W\vec{J}^{(n)} + \vec{c})\|^2 \\ - \frac{\eta}{2} \frac{\partial}{\partial w_{ji}} \|\vec{J}^{(n)} - (W^T \vec{h}^{(n)} + \vec{b})\|^2$$

THUS

$$\textcircled{a} w_{ji} \leftarrow w_{ji} - \eta v_i (\underbrace{\vec{m}_j^T \vec{J}^{(n)}}_{\hat{h}_j} + c_j - h_j^{(n)})$$

$$- \eta h_j (\underbrace{w_{ji}^T \vec{h}^{(n)}}_{\hat{J}_i} + b_i - v_i^{(n)})$$

$$\textcircled{b} c_j \leftarrow c_j - \eta (\underbrace{c_j + \vec{m}_j^T \vec{J}^{(n)}}_{\hat{h}_j} - h_j)$$

$$\textcircled{c} b_i \leftarrow b_i - \eta (\underbrace{b_i + \vec{w}_i^T \vec{h}^{(n)}}_{\hat{J}_i} - v_i)$$

DRAWBACK: $\vec{h}^{(n)}$ ALWAYS RANDOM \Rightarrow NO FINAL CONVERGENCE!

TEMPERATURE (BOLTZMANN SIMULATED ANNEALING)

INITIALIZE $\sigma^2 = \max_{i \in d \in D} \text{Variance}(V_d)$

ITERATE ① $\vec{h}^{(n)} \sim \mathcal{N}(\vec{h}^{(n)}; W\vec{J}^{(n)} + \vec{c}, \sigma^2 I)$

② a, b, c

③ $\sigma^2 \leftarrow \alpha \sigma^2$

FOR $\alpha \approx (0.9)^{1/N}$

TERMINATE e.g. WHEN $\sigma^2 < 10^{-4} \max_{i \in d \in D} \text{Variance}(V_d)$

GENERALIZATION: RBM TRAINING

W/ TEMPERATURES

INITIALIZE) T LARGE

ITERATE) ① $\vec{h}^{(n)} \sim \frac{1}{Z(\vec{v}^{(n)})} e^{-E(\vec{v}^{(n)}, \vec{h}^{(n)}) / T}$

② $w_{ji} \leftarrow w_{ji} - \eta \frac{\partial}{\partial w_{ji}} (\ln p(\vec{h}^{(n)} | \vec{v}^{(n)}) + \ln p(\vec{v}^{(n)} | \vec{h}^{(n)}))$

③ $T \leftarrow \alpha T$, $\alpha \approx (0.9)^{1/N}$

TERMINATE) WHEN T IS SMALL