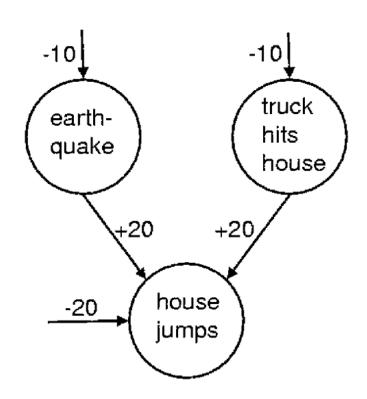
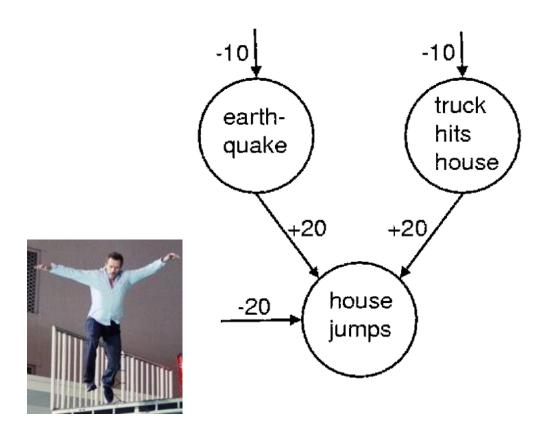
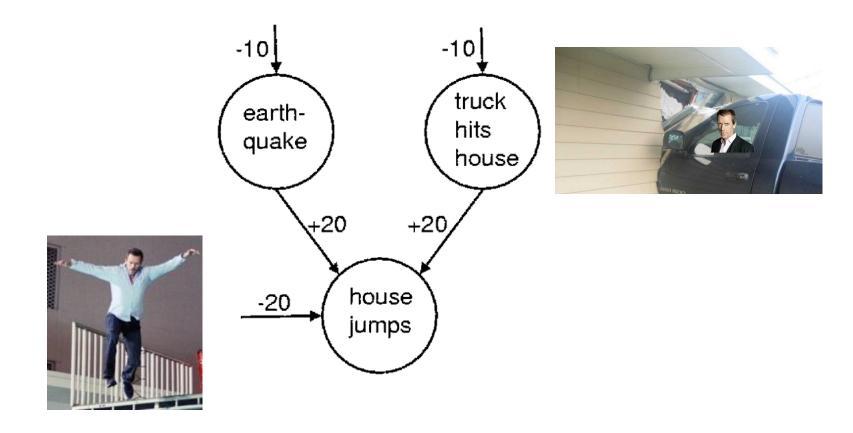
# A Fast Learning Algorithm for Deep Belief Nets

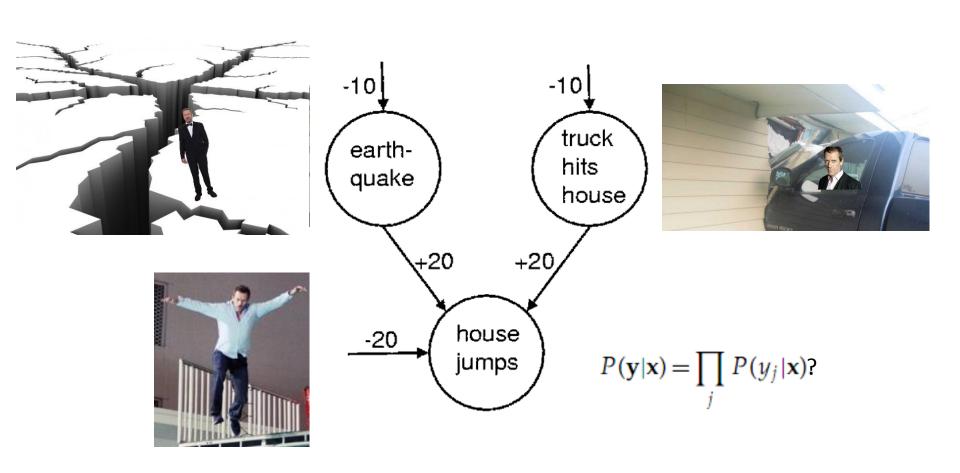
Hinton, Osindero, Teh

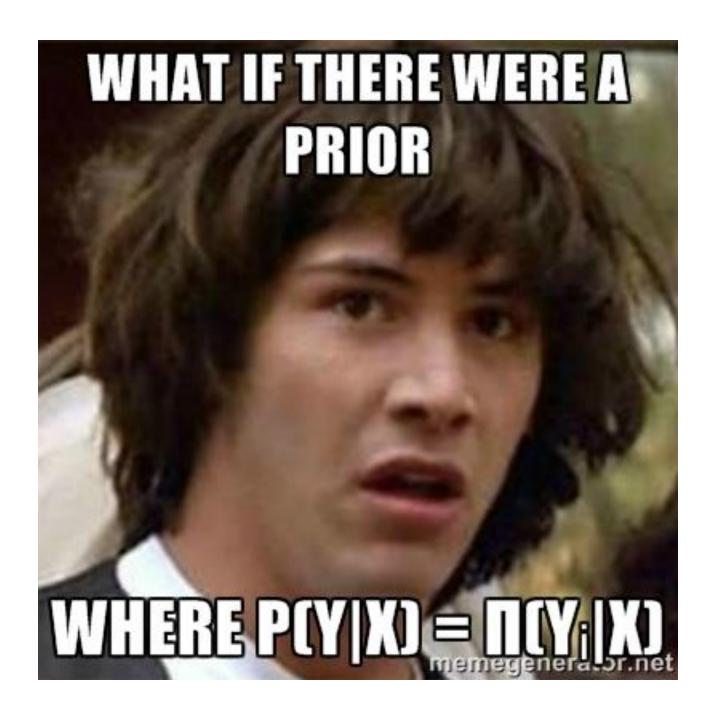












### **Complementary Priors**

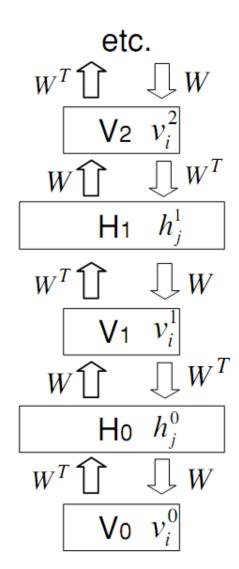
$$P(\mathbf{x}|\mathbf{y}) = \frac{1}{\Omega(\mathbf{y})} \exp\left(\sum_{j} \Phi_{j}(\mathbf{x}, y_{j}) + \beta(\mathbf{x})\right)$$

$$P(\mathbf{y}) = \frac{1}{C} \exp\bigg(\log \Omega(\mathbf{y}) + \sum_{j} \alpha_{j}(y_{j})\bigg),\,$$

$$P(\mathbf{x}, \mathbf{y}) = \frac{1}{C} \exp\bigg(\sum_{j} \Phi_{j}(\mathbf{x}, y_{j}) + \beta(\mathbf{x}) + \sum_{j} \alpha_{j}(y_{j})\bigg).$$



#### A specially structured deep network



#### Training our deep network

$$\frac{\partial \log p(\mathbf{v}^0)}{\partial w_{ij}^{00}} = \langle h_j^0 (v_i^0 - \hat{v}_i^0) \rangle,$$

$$\frac{\partial \log p(\mathbf{v}^0)}{\partial w_{ij}^{00}} = \langle h_j^0 (v_i^0 - v_i^1) \rangle.$$

$$\frac{\partial \log p(\mathbf{v}^0)}{\partial w_{ij}} = \langle h_j^0(v_i^0 - v_i^1) \rangle + \langle v_i^1(h_j^0 - h_j^1) \rangle + \langle h_j^1(v_i^1 - v_i^2) \rangle + \cdots$$

#### Training our deep network

$$\frac{\partial \log p(\mathbf{v}^0)}{\partial w_{ij}} = \langle h_j^0(v_i^0 - v_i^1) \rangle + \langle v_i^1(h_j^0 - h_j^1) \rangle + \langle h_j^1(v_i^1 - v_i^2) \rangle + \cdots$$

$$\frac{\partial \log p(\mathbf{v}^0)}{\partial w_{ij}} = \langle v_i^0 h_j^0 \rangle - \langle v_i^\infty h_j^\infty \rangle.$$

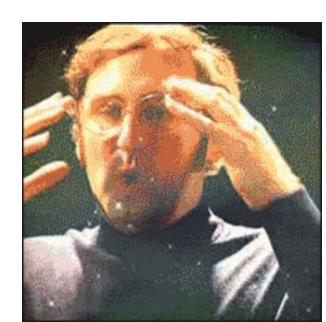
### Training our deep network

$$\frac{\partial \log p(\mathbf{v}^0)}{\partial w_{ij}} = \langle h_j^0(v_i^0 - v_i^1) \rangle + \langle v_i^1(h_j^0 - h_j^1) \rangle + \langle h_j^1(v_i^1 - v_i^2) \rangle + \cdots$$

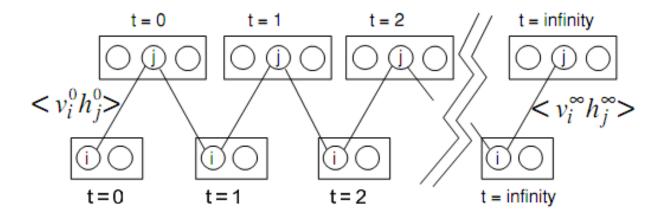
$$\frac{\partial \log p(\mathbf{v}^0)}{\partial w_{ij}} = \langle v_i^0 h_j^0 \rangle - \langle v_i^\infty h_j^\infty \rangle.$$

$$KL(P^0||P_\theta^\infty) - KL(P_\theta^n||P_\theta^\infty).$$

This is the update for a restricted Boltzmann Machine



### **RBM** training



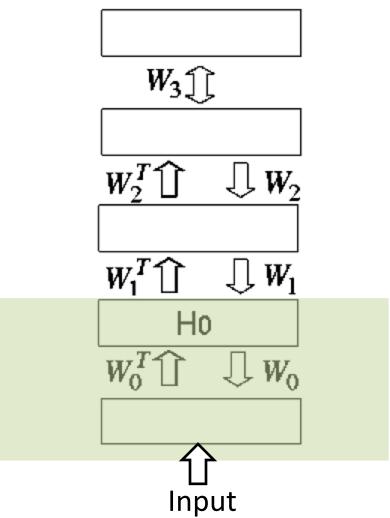
### Let's relax the assumptions



**RBM**  $W_3$  $\bigcup W_2$  $\bigcup W_1$ Hο  $\bigcup W_0$ 

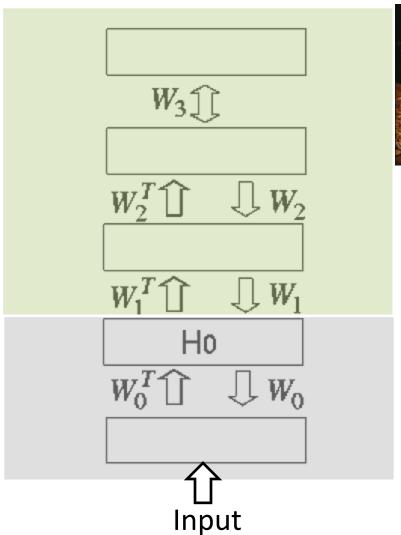
Input

## **Greedy Training**



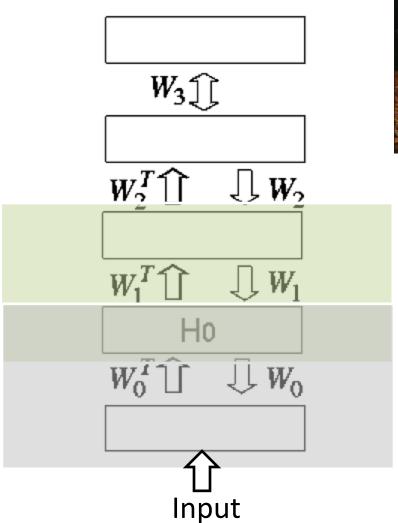


## **Greedy Training**

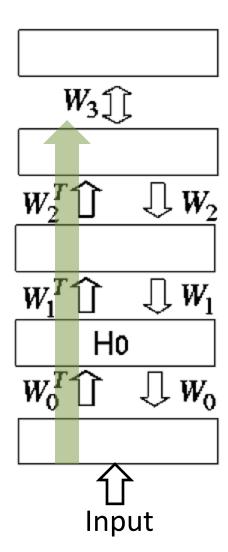


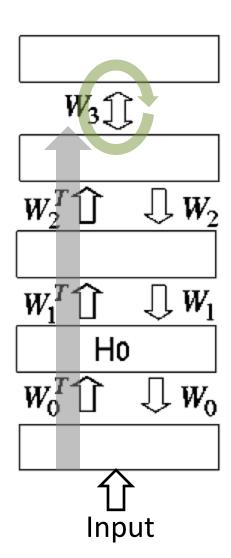


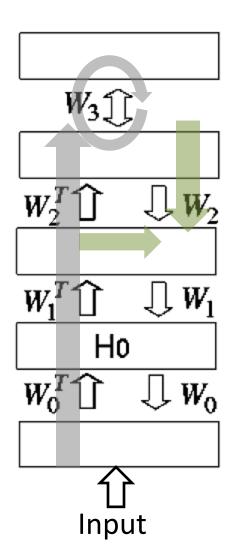
## **Greedy Training**

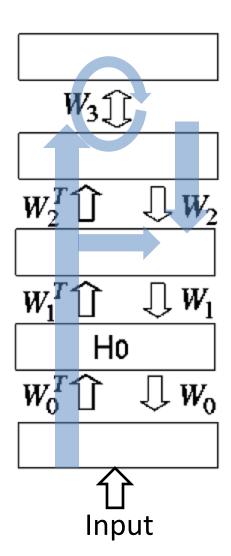


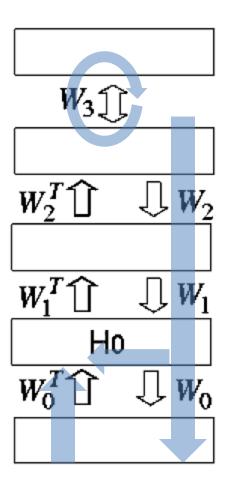












### Yes, this actually works



#### All Done!

