

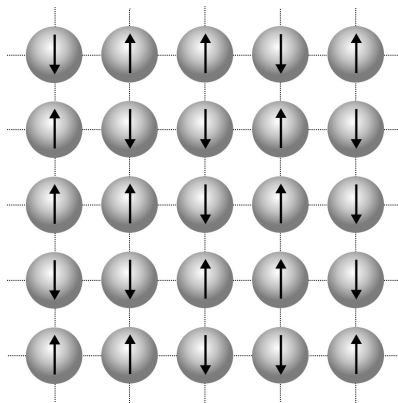
# Machine Learning reading group

## 4/24/18

An exact mapping between the Variational  
Renormalization Group and Deep Learning

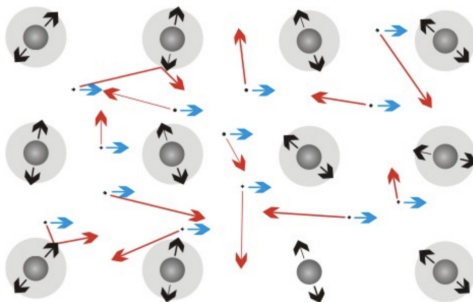
Mehta and Schwab, *arXiv* 2014

# Ising model of interacting spins



$$H(\mathbf{v}) = \sum_i K_i v_i + \sum_{ij} K_{ij} v_i v_j \quad \text{for } v_i = \pm 1$$

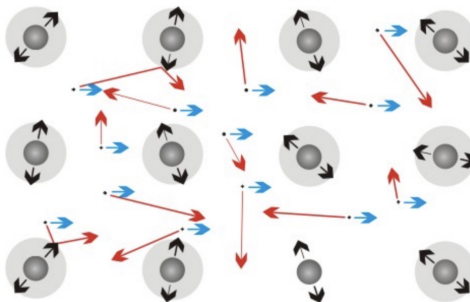
## Spins under thermal motion



$$P(\mathbf{v}) = \frac{e^{-H(\mathbf{v})}}{Z}$$

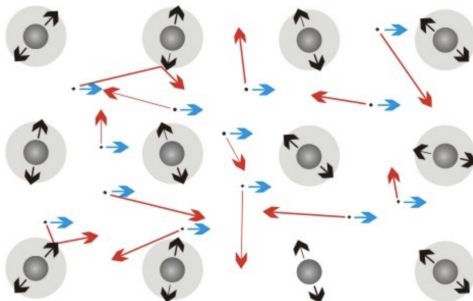
where  $Z = \sum_{\mathbf{v}'} e^{-H(\mathbf{v}')}$

## Thermal spin distribution



$$\begin{aligned}
 P(\uparrow / \downarrow | i) &= P(v_i = \pm 1) \\
 &= \sum_{v_{j \neq i} = \pm 1} \frac{e^{-H(\mathbf{v})}}{Z} \Bigg|_{v_i = \pm 1}
 \end{aligned}$$

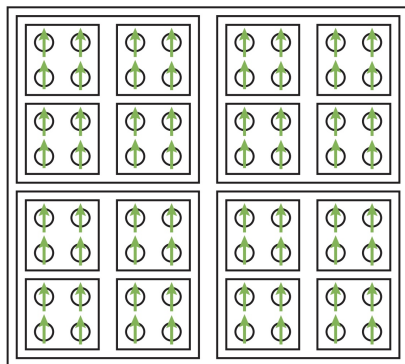
## Definition of *free energy*



$$F = -\log Z$$

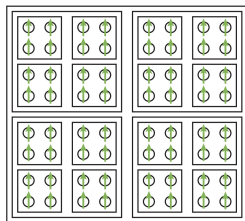
$F$  measures disequilibrium  
(e.g. polarization).

## Renormalized spins



- parameters  $\lambda$
- $\mathbf{v} \longrightarrow \mathbf{h}$
- $K \longrightarrow \tilde{K}$
- $H \longrightarrow H_{\lambda}^{RG}$
- $F^{\mathbf{v}} \longrightarrow F_{\lambda}^{\mathbf{h}}$

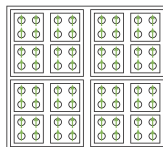
## Calculating effective spins $\mathbf{h}$



$\mathbf{h}$  is an average spin per block:

$$h_i = \frac{1}{A} \sum_j v_j$$

# Calculating effective coupling strengths $\tilde{K}$



Define  $T_\lambda$  as (negative) interaction energy between  $\mathbf{v}$  and  $\mathbf{h}$ , and integrate out  $\mathbf{v}$ :

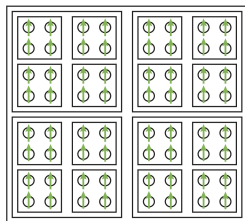
$$e^{-H^{RG}(\mathbf{h})} = \sum_{\mathbf{v}|\mathbf{h}} e^{T_\lambda(\mathbf{v},\mathbf{h})} \cdot e^{-H(\mathbf{v})}$$

This implies a renormalized free energy:

$$Z^{RG} = \sum_{\mathbf{h}} e^{-H^{RG}(\mathbf{h})}$$
$$F_\lambda = -\log Z^{RG}$$



# Objective of renormalization process

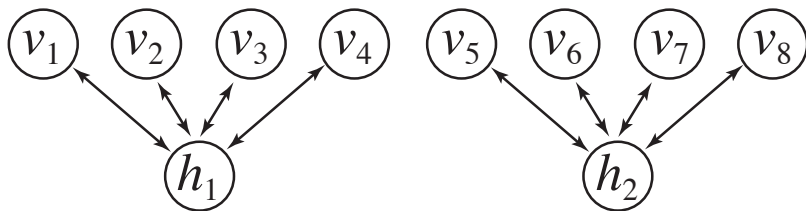


Adjust  $\lambda$  so that free energies match:

$$F_{\lambda}^{\mathbf{h}} = F^{\mathbf{v}}$$

Equivalently, we want  $e^{T\lambda} = 1$  for all states.

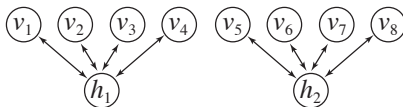
## RBM activity depends on an energy function



$$p_{\lambda}(\mathbf{v}, \mathbf{h}) = \frac{e^{-E_{\lambda}(\mathbf{v}, \mathbf{h})}}{Z_{\lambda}}$$

$E$  is typically 2nd order in  $\mathbf{v}, \mathbf{h}$

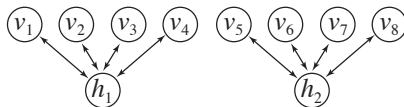
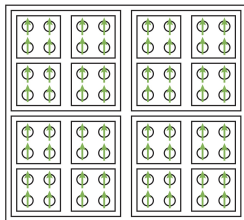
## Training objective



Adjust  $\lambda$  to reproduce input statistics:

$$\begin{aligned} 0 &= \text{KL divergence } (p(\mathbf{v}), p_{\lambda}(\mathbf{v})) \\ &= \sum_{\mathbf{v}} p(\mathbf{v}) \log \frac{p(\mathbf{v})}{p_{\lambda}(\mathbf{v})} \\ \longrightarrow p(\mathbf{v}) &= p_{\lambda}(\mathbf{v}) \end{aligned}$$

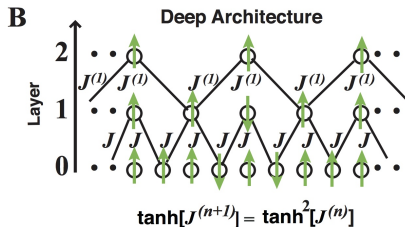
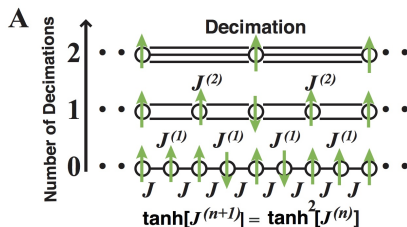
# Correspondence



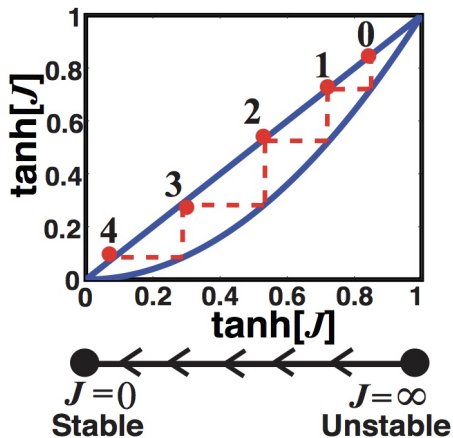
$$T - H \longleftrightarrow -E$$

$$F = F_{\lambda} \longleftrightarrow \frac{e^{-E(\mathbf{v}, \mathbf{h})}}{Z} = \frac{e^{-E_{\lambda}(\mathbf{v}, \mathbf{h})}}{Z_{\lambda}}$$

# Renormalizing by 'decimation' in 1D



## Renormalized 1D coupling strengths



## Renormalization in 2D

