

$$h_{ij} = \frac{\partial^2 E}{\partial w_i \partial w_j} = \langle x_i x_j \rangle = \frac{1}{m} \sum_p x_i^{(p)} x_j^{(p)}$$

$$\hat{X} = X + \begin{pmatrix} 1000 \\ \vdots \\ 1000 \end{pmatrix} \quad \begin{pmatrix} 1000 \\ \vdots \\ 1000 \end{pmatrix} = 1000$$

$$\begin{aligned} \tilde{H} &= \langle \tilde{X} \tilde{X}^T \rangle = \langle (X + 1000) (X + 1000)^T \rangle \\ &= \langle X X^T \rangle + \langle X \rangle 1000^T + 1000 \langle X^T \rangle + 1000 \\ &\text{assume } \langle X \rangle = 0 \\ &\text{as there is zero mean.} \end{aligned}$$

Make (S) input zero mean and make input unit variance

$$h_{ij} = \frac{\partial^2 E}{\partial w_i^2} = \langle x_i^2 \rangle = X^{-2} + \text{VAR}(x_i)$$

if we have n "experts"

~~$w_i(t)$ = weight for i for t^{th} game (± 1)~~

$w_i(t)$ = weight for expert (i) at time (t)

$x_i(t)$ = prediction for i for t^{th} game (± 1)

$$P(t) = \text{sign} \left(\sum_i w_i(t) x_i(t) \right)$$

$$z(t) = \pm 1 \quad \text{winner of game } t$$

$$v_i(t) = \frac{\# \text{ correct (on } t-1)}{t}$$

$$v_i(t+1) = \dots v_i(t) \quad [\text{continued next lecture}]$$

S = matrix of small numbers.