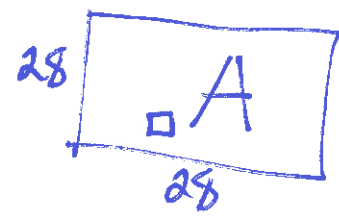
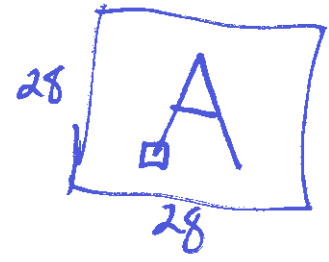
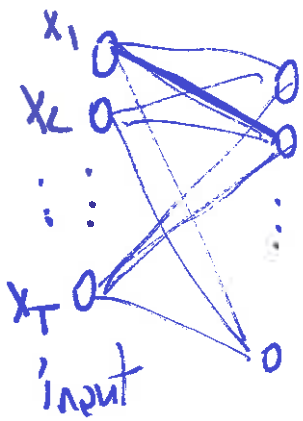


Input

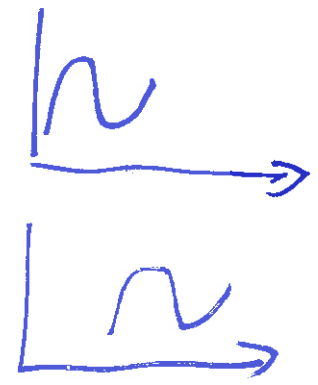
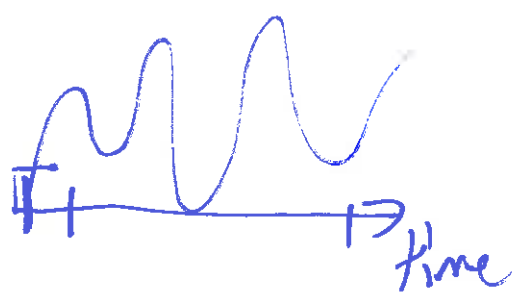
$$q_{ij}^{(2)} = \left[\sum_{i,j} w_{ij}^{(1)} x_{ij} \right] + w_0^{(1)}$$

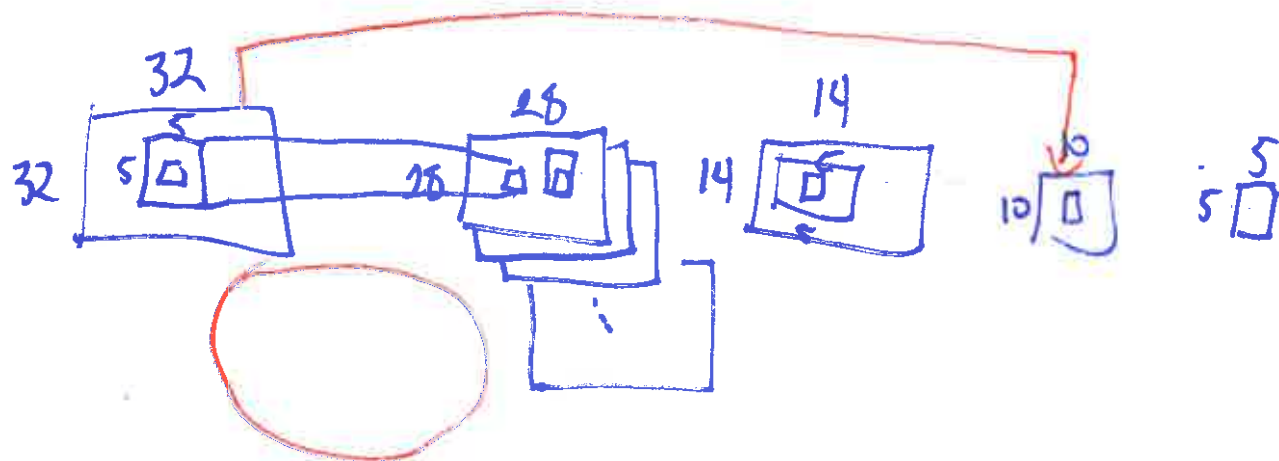
25 + 1



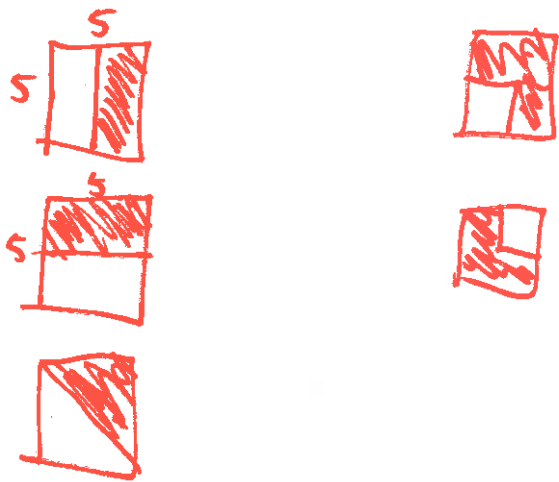


$O(D^2L)$

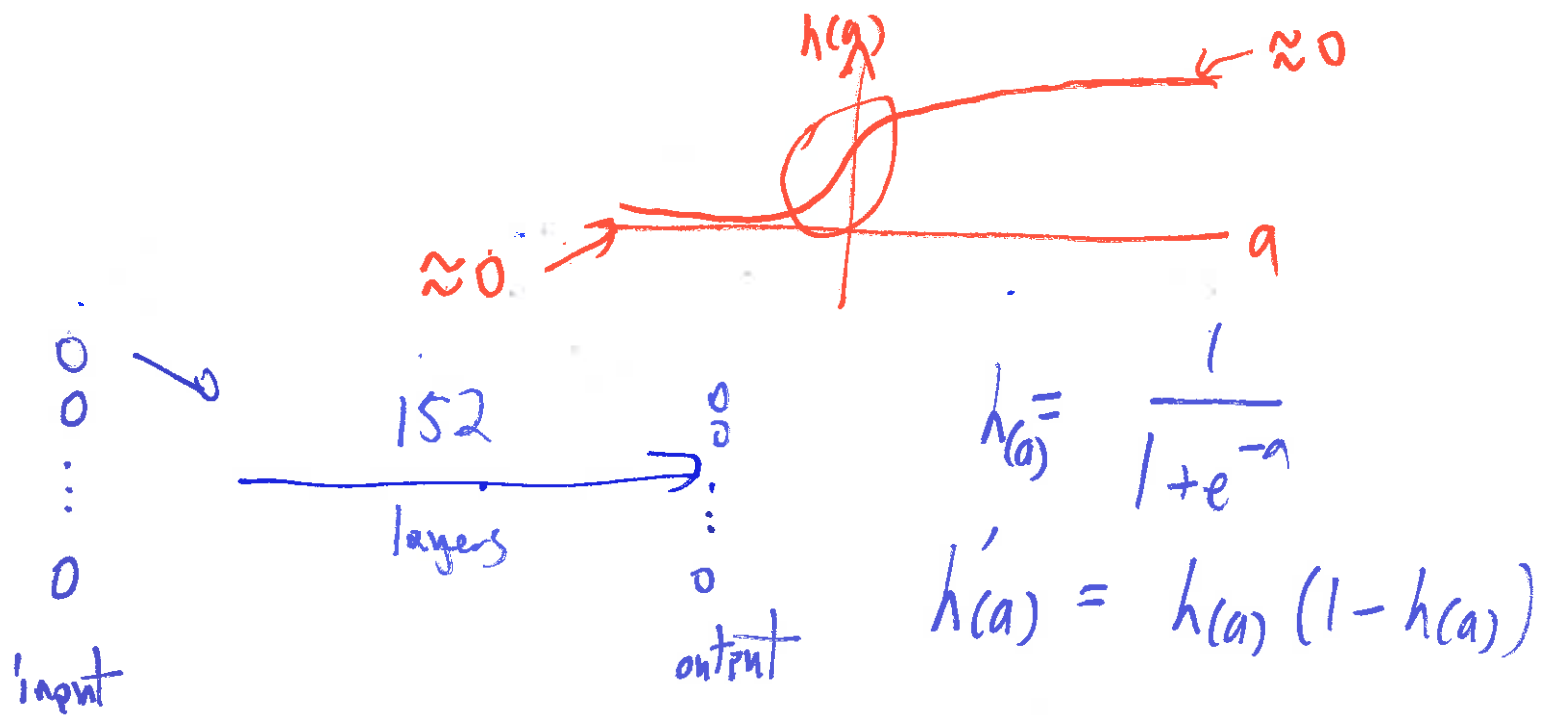




$$(5 \times 5 + 1) \times 6$$



$$w^{(\tau+1)} = w^{(\tau)} + \eta \nabla E_n(w^{(\tau)})$$



$$\frac{\partial E_n}{\partial w_{ij}^{(2)}} = h'(a^{(2)}) \cdot h'(a^{(3)}) \cdot \dots \cdot h'(a^{(151)})$$

≈ 0

$$h(a) = \begin{cases} 0 & \text{if } a < 0 \\ a & \text{o.w.} \end{cases}$$

$$\vec{X} = (x_1, x_2, \dots, x_M)$$

$$w^T \phi(x)$$

$$\vec{Z} = (z_1, z_2, \dots, z_M)$$

$$\vec{X}, \vec{Z} \in \mathbb{R}^2 \quad M=2$$

$$\begin{aligned}
 k(x, z) &= (1 + x^T z)^2 && \text{2 mults, 1 add} \\
 &= (1 + x_1 z_1 + x_2 z_2)^2 && \text{1 add} \\
 &= 1 + 2x_1 z_1 + 2x_2 z_2 + x_1^2 z_1^2 + 2x_1 x_2 z_1 z_2 + x_2^2 z_2^2 && \text{1 mult} \\
 &= (1, \sqrt{2} x_1, \sqrt{2} x_2, x_1^2, \sqrt{2} x_1 x_2, x_2^2)^T (1, \sqrt{2} z_1, \sqrt{2} z_2, z_1^2, \sqrt{2} z_1 z_2, z_2^2) && \text{3 mult., 2 add} \\
 &= \phi(x)^T \phi(z) && \text{6 mult. + 5 add.}
 \end{aligned}$$

$$\phi(x) = (1, \sqrt{2} x_1, \sqrt{2} x_2, x_1^2, \sqrt{2} x_1 x_2, x_2^2)^T \leftarrow 6 \text{ mults}$$

For all training examples x_i
compute $\|x_i - x_j\|$
return label of x_i with min distance

$$\begin{aligned} & \phi(x_i)^T \phi(x_j) \\ & \parallel \\ & k(x_i, x_j) \end{aligned}$$

For all training examples x_i
compute $\|\phi(x_i) - \phi(x_j)\|^2$
return label of x_i with min distance²

For all training examples x_i
compute $\phi(x_i)^T \phi(x_i) - 2\phi(x_i)^T \phi(x_j) + \phi(x_j)^T \phi(x_j) \equiv f_{ij}$
return label of x_i with minimum f_{ij}

$$\text{direct} \left\{ \begin{array}{l} \vec{x}_i \mapsto \vec{\phi}(\vec{x}_i) \\ \vec{x}_j \mapsto \vec{\phi}(\vec{x}_j) \\ \phi(x_i)^T \phi(x_j) \end{array} \right.$$

$$k(x_i, x_j) = \phi(x_i)^T \phi(x_j)$$

$$\Phi = N \begin{bmatrix} \phi_j(x_i) \end{bmatrix}^M$$

$$\vec{x} \in \mathbb{R}^D$$

$$\phi_0(\vec{x}) = 1$$

$$\phi_1(\vec{x}) = x_1$$

$$\phi_2(\vec{x}) = x_2$$

⋮

$$\phi_m(\vec{x}) = x_m$$

$$\phi_{m+1}(\vec{x}) = x_1^2$$

⋮

w^*

$$\phi_0(\vec{x}) = 1$$

$$\phi_1(\vec{x}) = \frac{1}{\sqrt{2}} x_1$$

⋮

$$\phi_{m+1}(\vec{x}) = x_1^2$$

⋮

$$\tilde{w}^* = \begin{pmatrix} w_1^* \\ \vdots \\ w_{m+1}^* \\ \frac{w_{m+1}^*}{\sqrt{2}} \end{pmatrix}$$