

$$\text{Eigenvector } (A) \rightarrow [\vec{v}_1 : \vec{v}_2]$$

$$\rightarrow \{\lambda_1, \lambda_2\}$$

$$A \vec{v}_1 = \lambda_1 \vec{v}_1$$

$$A \vec{v}_2 = \lambda_2 \vec{v}_2$$

$$A [\vec{v}_1, \vec{v}_2] = [\vec{v}_1, \vec{v}_2] \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

$$A V = V \Delta \Rightarrow A = V \Delta V^{-1}$$

Lecture cont. (Recall!)

$$y = \{0, 1\}$$

$x_1, \dots, x_p$  feature

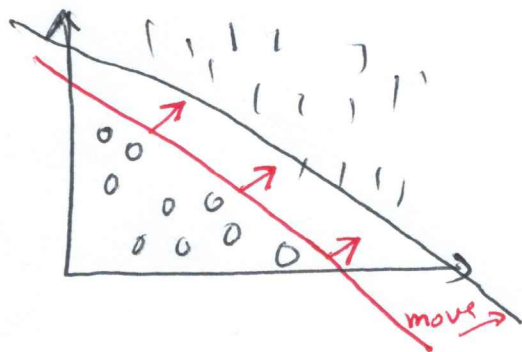
need  $\hat{y} = g(\vec{x})$

$$\mathbb{D} = \left\{ \begin{bmatrix} \leftarrow \vec{x}_1 \rightarrow \\ \leftarrow \vec{x}_2 \rightarrow \\ \vdots \\ \leftarrow \vec{x}_n \rightarrow \end{bmatrix} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \right\}$$

$$\mathcal{A}(\mathcal{H}, \mathbb{D})$$

$$\mathcal{H} = \left\{ \mathbb{1}_{\vec{w} \cdot \vec{x} > 0}, \vec{w} \in \mathbb{R}^{p+1} \right\}$$

# ➤ Perceptron Learning Algorithm



①  $\vec{w} = \vec{0} \rightarrow \vec{w}^{t=0}$   
or random.

② Calculate  $\hat{y}_i = 1$

③ update all  $j = 1 \dots p+1$

$$w_1^{t=1} = w_1^{t=0} + (y_i - \hat{y}_i)$$

$$w_2^{t=1} = w_2^{t=0} + (y_i - \hat{y}_i) x_{i,2}$$

$$\vdots$$

$$w_{p+1}^{t=1} = w_{p+1}^{t=0} + (y_i - \hat{y}_i) x_{i,p+1}$$

④ Repeat for  $i = 1 \dots n$

⑤ Repeat steps 2  $\rightarrow$  4 until error is reached.  
or a max # of iterations.

If  $\mathbb{D}$  is "linearly separable" i.e.  $\exists \vec{w}$   
s.t.  $\forall \vec{x} \in \mathbb{D}, \vec{w} \cdot \vec{x} > 0$  yielded an error in  $\mathbb{D}$ , then  
the alg will find it.

$$\text{BAE}(g) = \sum_{i=1}^n \mathbb{1}_{\hat{y} \neq y}$$

# of error.

$$\mathbb{1}_{\vec{w} \cdot \vec{r} \geq 0}$$

