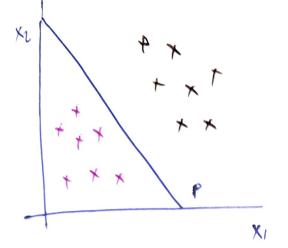
dependent on no. of sour

Linear Disuminant Function.



Our I draw a line to classify two classes,
hyperplane

Decision bounday.

$$2x_1 + 3x_2 - 7 = 0$$

$$\omega^2 \left[\begin{array}{c} \omega_1 \\ \omega_2 \end{array}\right], \quad \chi = \left[\begin{array}{c} \chi_1 \\ \chi_2 \end{array}\right]$$

Recision rule -

$$\hat{y} = \begin{cases} +1 & \text{if } w^{T}x + b > 0 \\ -1 & \text{if } w^{T}x + b < 0 \end{cases}$$

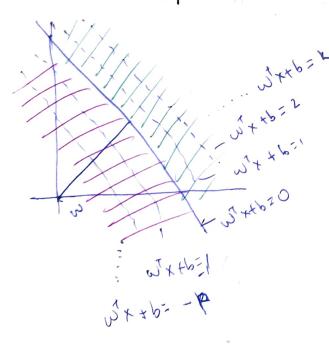
$$g(x) = y = \begin{cases} +1, & \omega^{\dagger}x + b > 0 \\ -1, & \omega^{\dagger}x + b < 0 \end{cases}$$

$$(\omega^{\dagger}x + b = 0)$$

$$(\omega^{\dagger}x + b = 0)$$

$$(\omega^{\dagger}x + b = 0)$$

Geometric Interpretation



Perceptron & Neuman → g(x) {-1,1} pecision role 6t moxotulat... (Adiation) Activation frs -1 Tanh : [-11] $f(n) = \frac{e^{x} - e^{-x}}{e^{x} + e^{-x}}$ 1 Symoid: [0,1] fine 1 Ite-x

Binary step
$$f^n$$

$$f(n \ge 1) \times > 0$$

$$-1, \times < 0$$

Generalized Discriminant for

I linear case -
$$g(x) = (w^{T}x + b_{x})$$

$$A = \begin{bmatrix} \omega_0 \\ \omega_1 \\ \vdots \\ \omega_N \end{bmatrix}, y = \begin{bmatrix} n_0 \\ n_1 \\ \vdots \\ n_n \end{bmatrix}$$

$$g(x) = A^T Y$$

2) Non linear case

2) Non linear case

Y:
$$\phi(x)$$

Y: $\phi(x)$

X': $\phi(x)$

X': $\phi(x)$

Y: $\phi(x)$

3(x)= ATY

In some higher dim space, non linear data will be linear.

why this ?