

FoDA L28

Classification : KNNs, Decision Trees,
and Neural Nets ... oh my!

Dec 6, 2022

Classification

Input $(x_i, y_i) \in \mathbb{R}^d \times \{-1, +1\}$

$\hookrightarrow (x_i, y_i) \in \mathbb{R}^d \times \{-1, +1\}$

x_i y_i

$\xrightarrow{\quad}$ Train
 $\xrightarrow{\quad}$ Test

$$(x_i, y_i) \sim u_{\text{iid}}$$

Goal Build function $f: \mathbb{R}^d \rightarrow \{-1, +1\}$

$f(x) = \text{sign}(g(x))$

$g(x): \mathbb{R}^d \rightarrow \mathbb{R}$

\hookrightarrow on new data $(x_i, y_i) \sim u$

$f(x') = y'$ w/ high probability

~~K - Nearest Neighbor~~

Classifier

(KNN)

$$|z = 5|$$

$$S_k(x') \subset X$$

Vote on
signs
in $S_k(x')$

$$4+$$

$$1-$$

$$f(x') = +$$

$$\partial x$$

x' among

$$+$$

$$-$$

$$+$$

$$-$$

choice of
distance

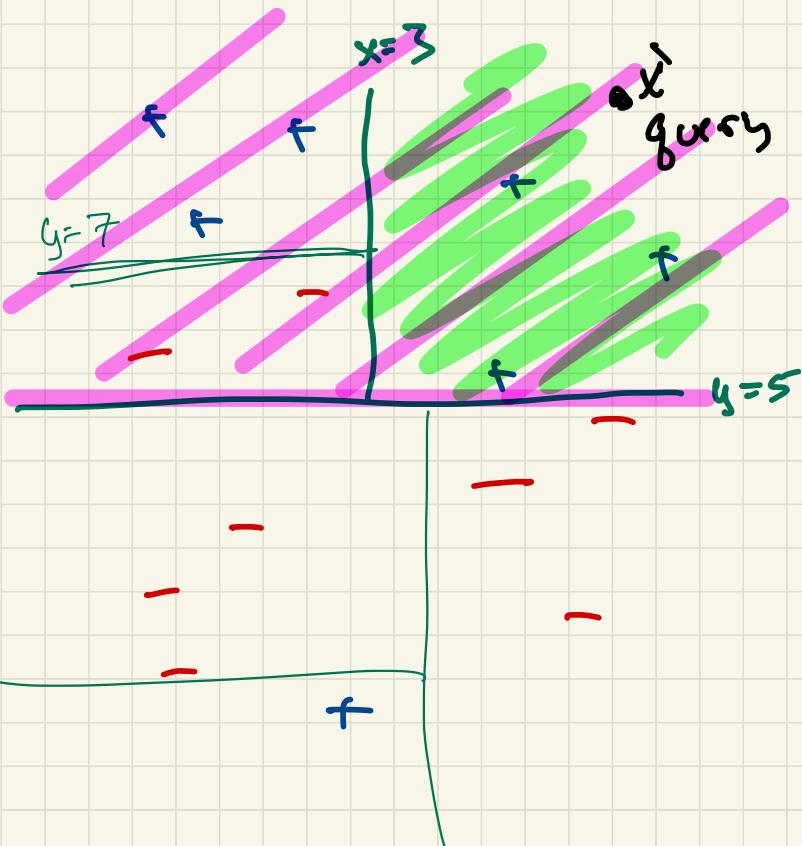
$$X \subset \mathbb{R}^d$$

$$x_i$$

$$:$$

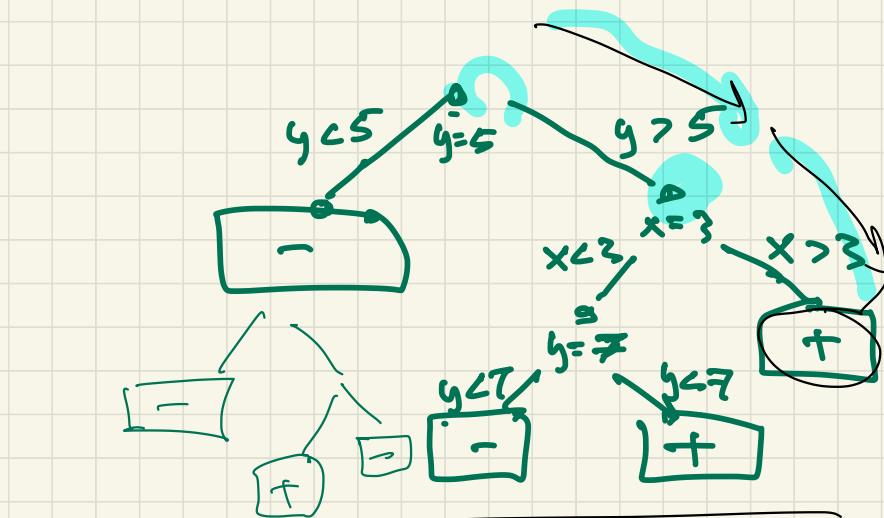
$$:$$

Decision Trees



$X \in \mathbb{R}^d$

$$f(x_i) = +$$



How to split?

- choose best

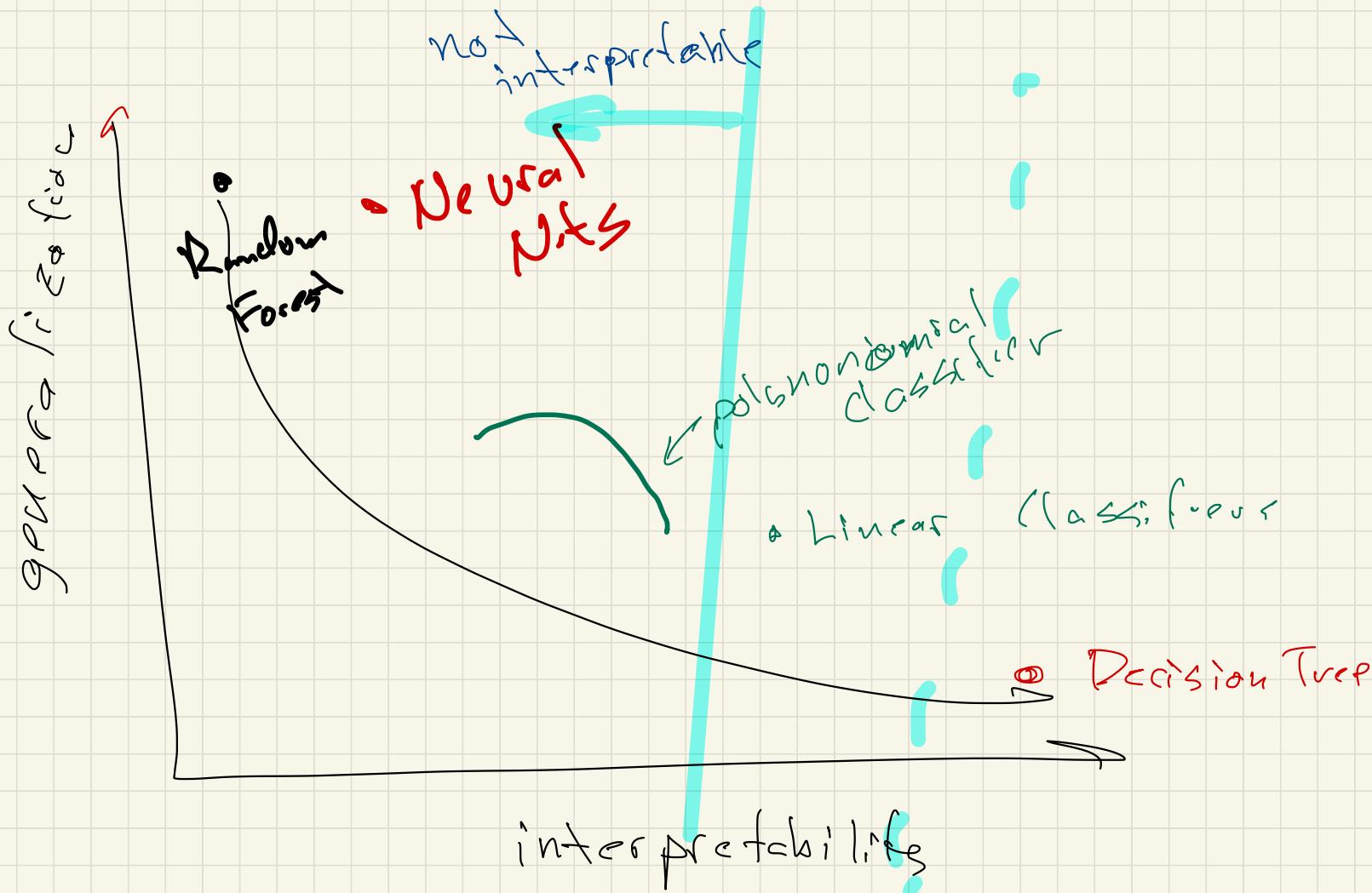
Gini Impurity Index

$$P_+ = \frac{\# \text{ + pts}}{\text{total \# pts}} = \frac{6}{8} = \frac{3}{4}$$

$$P_- = \frac{\# \text{ - pts}}{\text{total \# pts}} = \frac{2}{8} = \frac{1}{4}$$

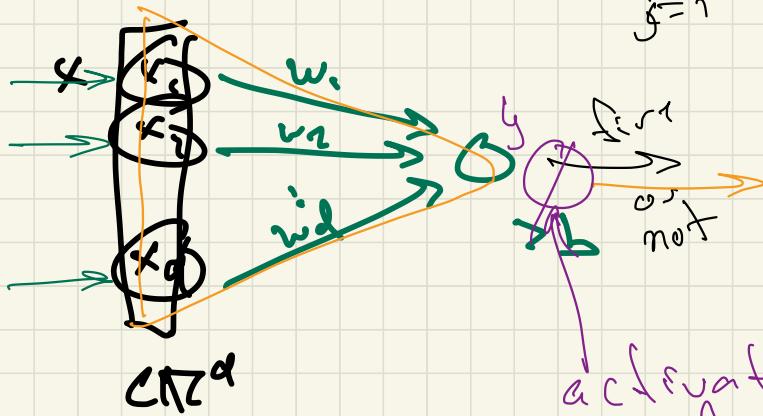
$$G = P_+ (1 - P_+) + P_- (1 - P_-)$$

$$G = \frac{3}{4} \left(1 - \frac{3}{4}\right) + \frac{1}{4} \left(1 - \frac{1}{4}\right) = \frac{3}{16} + \frac{3}{16} = \frac{6}{16} = \frac{3}{8}$$



Neural Nets

Neuron



$$\sum_{j=1}^d x_j w_j + b \geq 0$$

\Leftrightarrow

$$\underbrace{\langle w, x \rangle - b}_{\geq 0}$$

$$y = g_{w,b}(x)$$

linear classifier

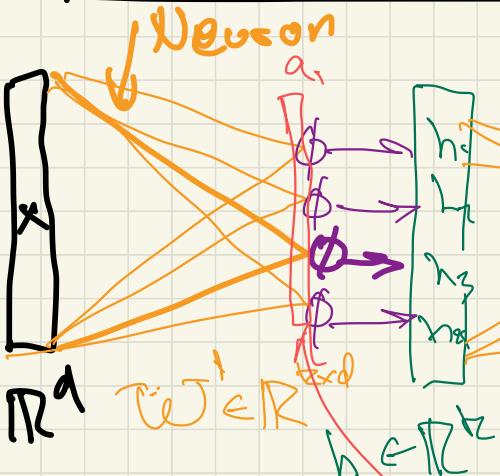
activation
function

$$\phi(y) = \text{ReLU}(y) = \max\{0, y\}$$

$$\phi(y) = \text{Sigmoid}(y) = \frac{1}{1 + \exp(-y)} \in (0, 1)$$



Neural Network



$$W = \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_k \end{bmatrix} \quad w_i \in \mathbb{R}^d$$

$$a = W^1 x$$

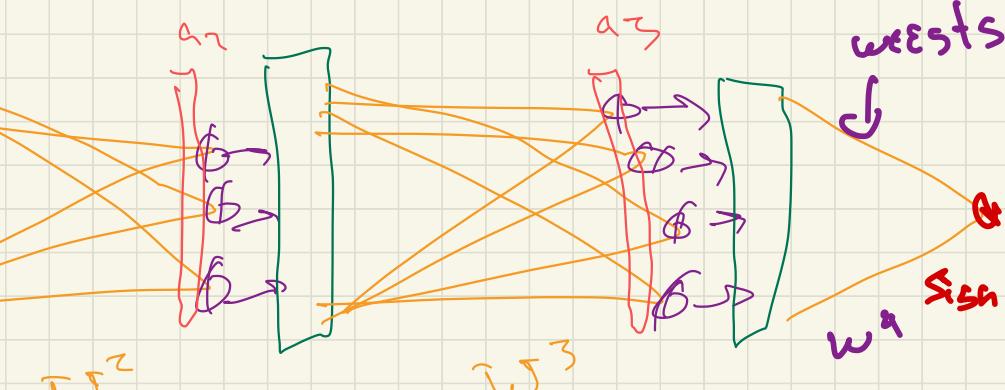
w/ no activation

$$a_2 = W^2 W^1 x$$

$$a_3 = W^3 W^2 W^1 x$$

$$B = W^3 W^2 W^1$$

back propagation



$$g(x) = w^3 \phi(w^2 \phi(w^1 x))$$

میں تھے