





## Machine learning classique

#### Intro to neural nets

## Comment adapter ce modèle à des données plus complexes?

# Linéarités

$$\det \mathbf{v}^{\top} \mathbf{a} \mathbf{1}$$

$$z_1 \stackrel{\text{def}}{=} \mathbf{x}^{\top} \boldsymbol{\beta}^{\mathbf{1}} + \alpha_1$$

$$\mathbf{x}^{\mathsf{T}} \boldsymbol{\beta}^{\mathbf{1}} +$$

$$\mathbf{x}^{\top} \boldsymbol{\beta}^{\mathbf{1}} +$$

$$\mathbf{x}^{\top} \boldsymbol{\beta}^{\mathbf{1}} +$$

$$\mathbf{x}^{\top} \boldsymbol{\beta}^{\mathbf{1}} +$$

$$z_2 \stackrel{\text{def}}{=} \mathbf{x}^{\top} \boldsymbol{\beta}^2 + \alpha_2$$

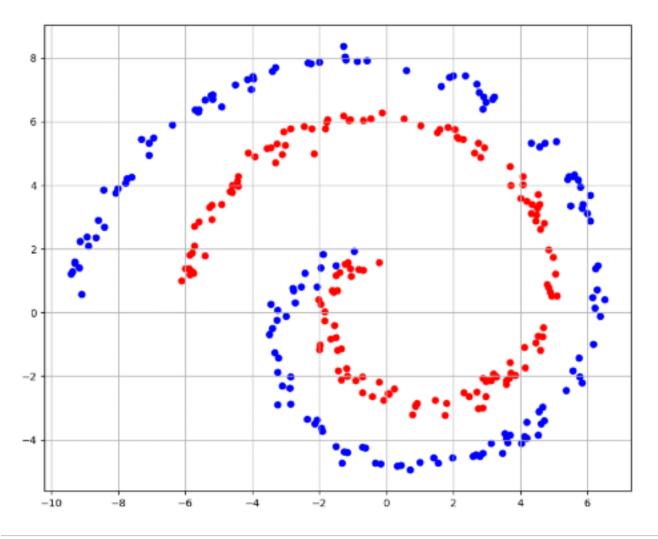
$$\vdots$$

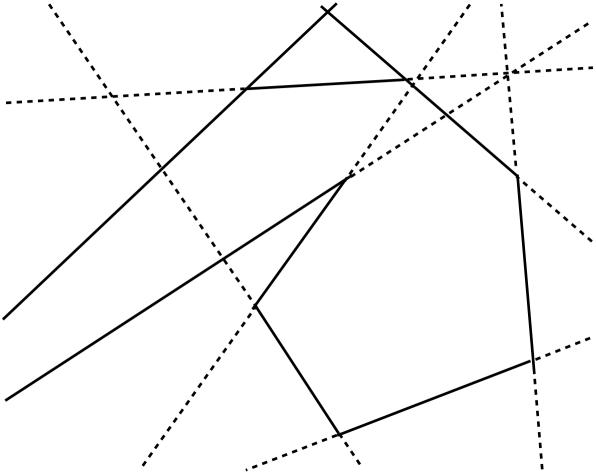
$$z_p \stackrel{\text{def}}{=} \mathbf{x}^{\top} \boldsymbol{\beta}^{\boldsymbol{p}} + \alpha_p .$$

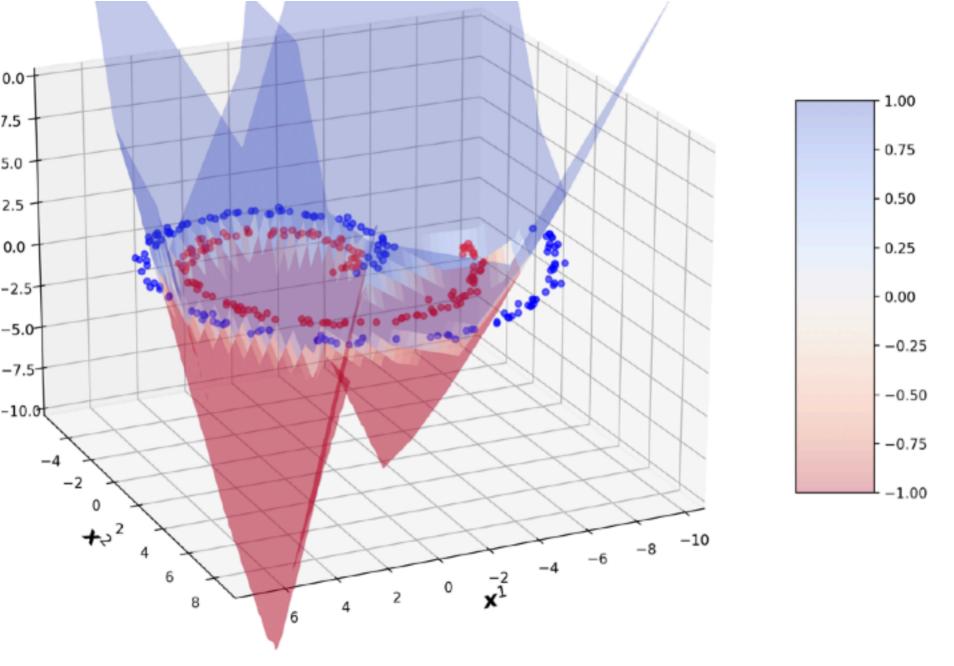
$$\max_{z}(z_1,\ldots,z_p)$$

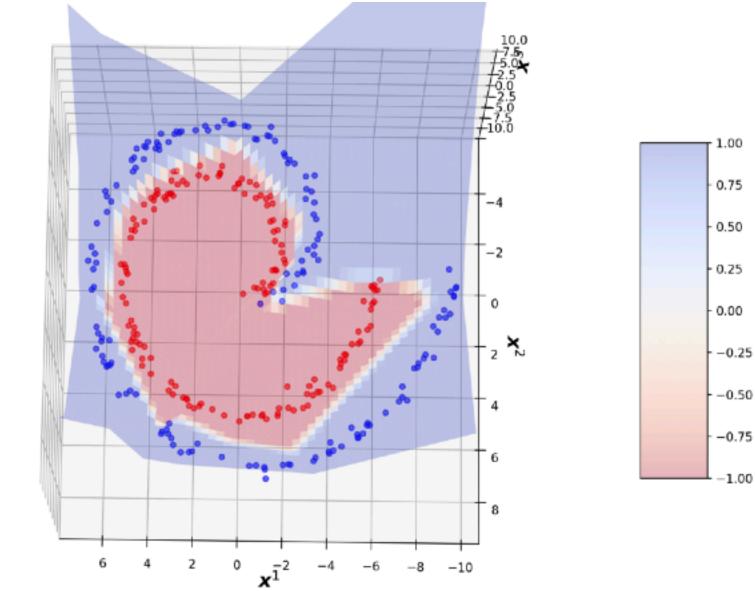
non-linéarité

$$oldsymbol{z}_1,\ldots, oldsymbol{z}_1$$





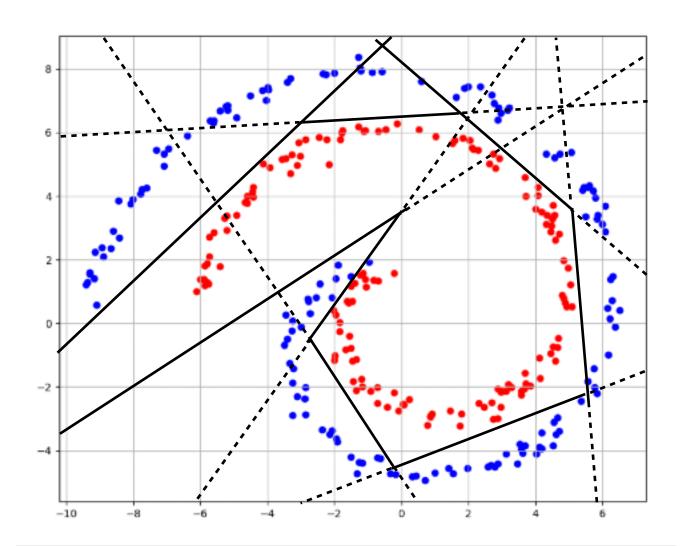




## Comment adapter ce modèle à des données plus complexes ?

-0.25

-0.50



#### Linéarités

$$z_1 \stackrel{\text{def}}{=} \mathbf{x}^{\top} \boldsymbol{\beta^1} + \alpha_1$$

$$z_2 \stackrel{\text{def}}{=} \mathbf{x}^{\top} \boldsymbol{\beta^2} + \alpha_2$$

$$\vdots$$

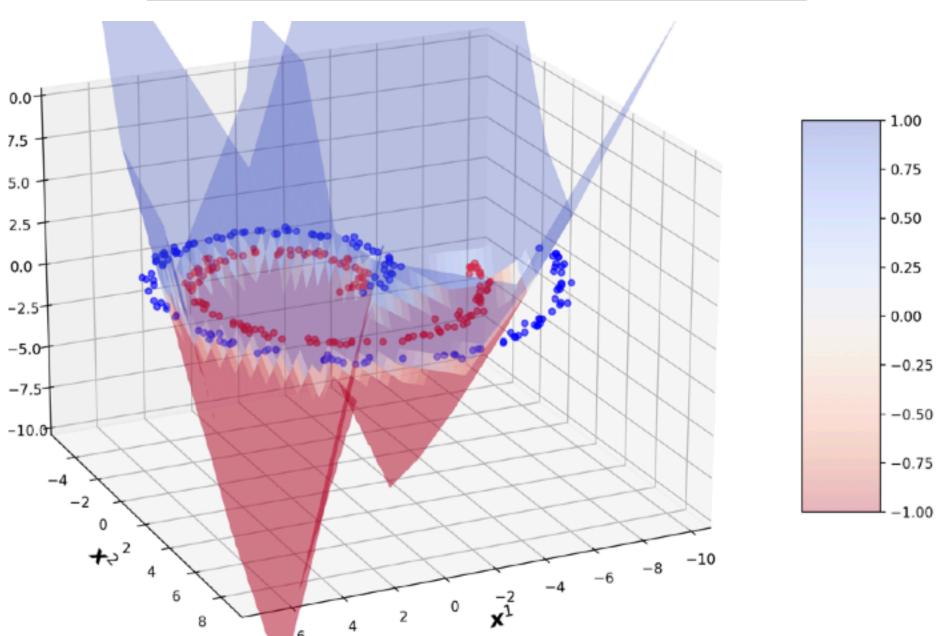
$$z_p \stackrel{\text{def}}{=} \mathbf{x}^{\top} \boldsymbol{\beta^p} + \alpha_p$$

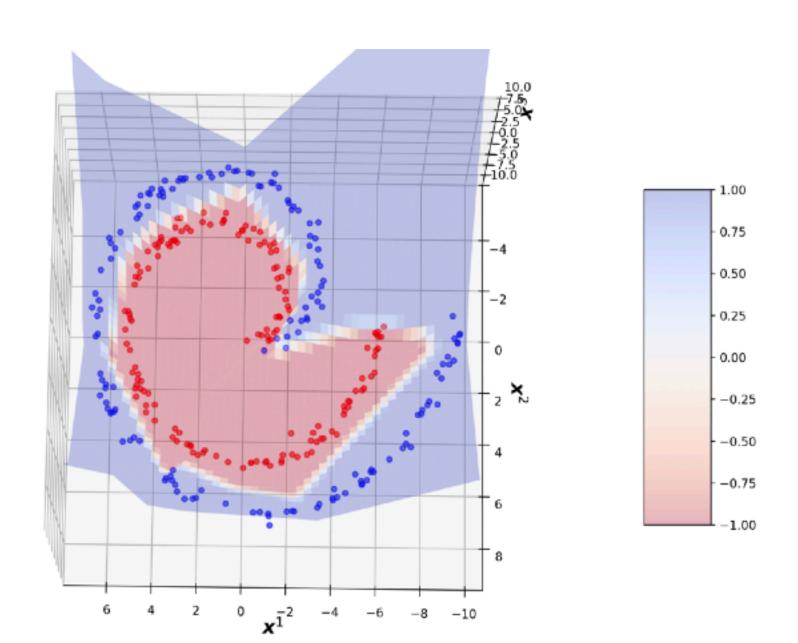
$$non-linéarité$$

$$\max(z_1, \dots, z_p)$$

$$\vdots$$

$$sigmoid$$









### Intro to neural nets

#### Linéarités

$$z_1 \stackrel{\text{def}}{=} \mathbf{x}^\top \boldsymbol{\beta^1} + \alpha_1$$

$$\vdots$$

$$z_p \stackrel{\text{def}}{=} \mathbf{x}^\top \boldsymbol{\beta^p} + \alpha_p$$

$$non-linéarité$$

$$\max(z_1, \dots, z_p) \longrightarrow \text{sigmoid}$$

En pratique, ce modèle ne fonctionne pas pour ces données complexes. Pourquoi à votre avis ?



