

* Neural Networks.

① A Graphical Presentation of Linear models

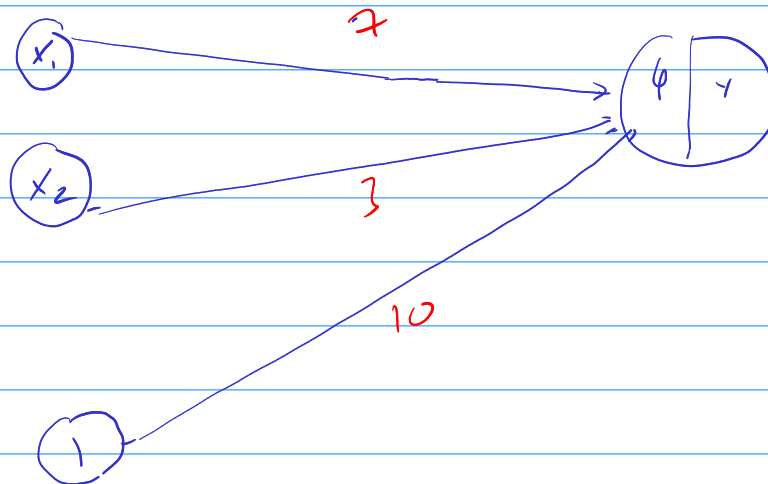
$$y = 7x_1 + 3x_2 + 10$$

Data

x_1	x_2	y
...

Input layer

output layer



$$\phi(t) = t$$

$$\Leftrightarrow x_1 \cdot 7 + x_2 \cdot 3 + 1 \cdot 10 = y$$

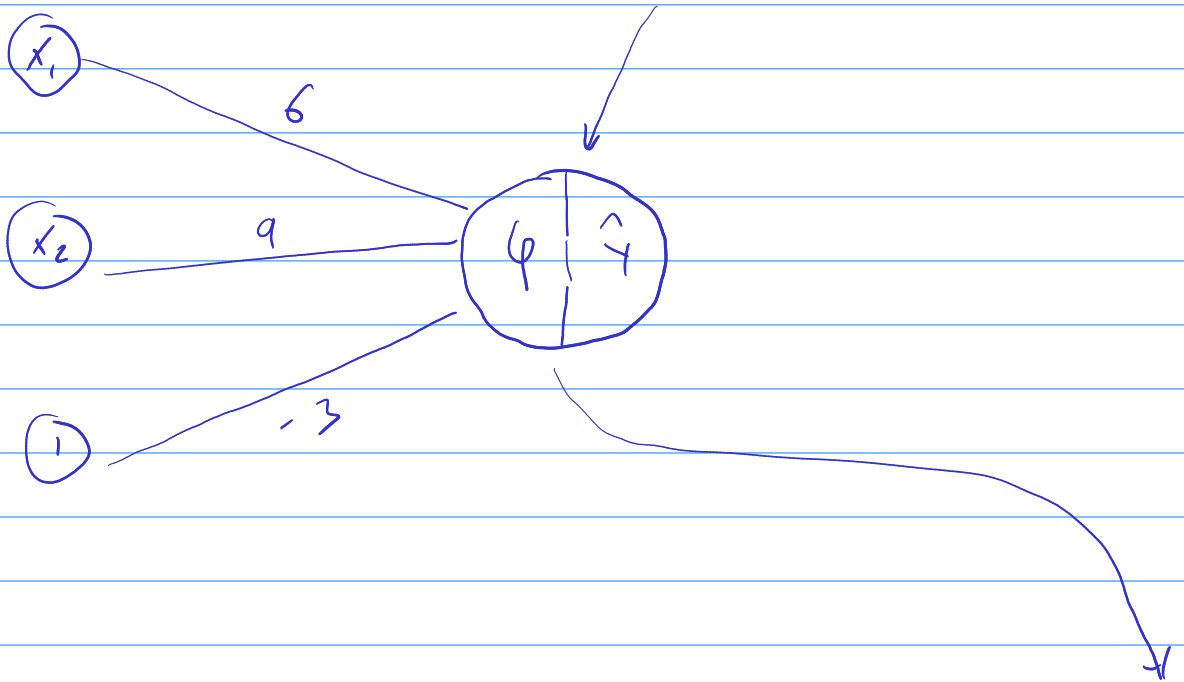
② A Graphical presentation of Logistic Regression

x_1	x_2	y

$$\hat{y} = \frac{1}{1 + e^{-[6x_1 + 9x_2 - 3]}}$$

Input layer

output layer

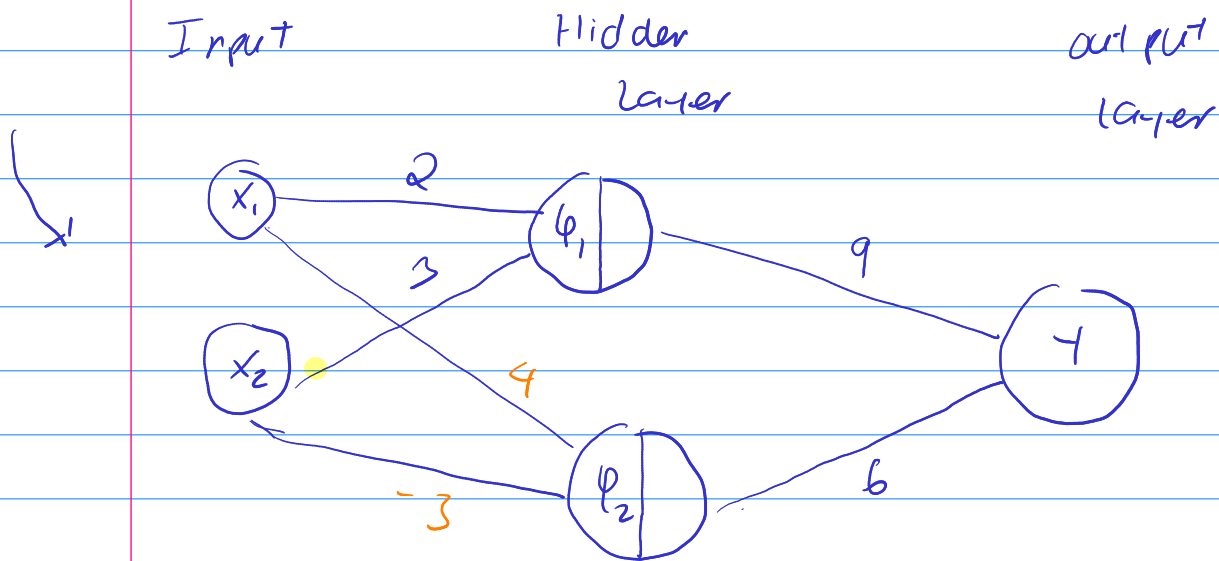


$$\left(x_1 \cdot 6 + x_2 \cdot 9 + 1 \cdot (-3) \right) \cdot \frac{1}{1 + e^{-(\dots)}} = 4$$

$$\phi(t) = \frac{1}{1 + e^{-t}}$$

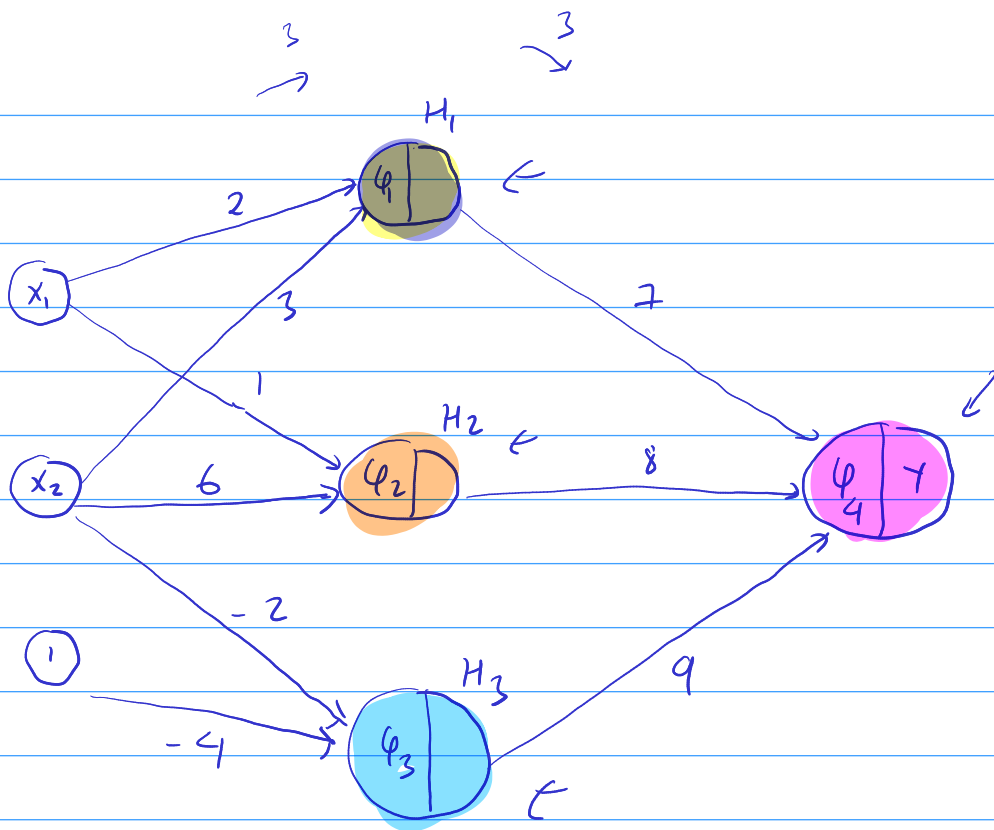
$$y = 9 \cdot (2x_1 + 3x_2)^2 + 6 \cdot \cos(4x_1 - 3x_2)$$

x_1	x_2	y



$$\phi_1(t) = t^2 ; \quad \phi_2(t) = \cos(t)$$

ϕ_1 and ϕ_2 are called activation functions



$$\varphi_1(t) = \left(\frac{t^3}{6}\right)$$

$$\varphi_2(t) = \log(t)$$

$$\varphi_3(t) = \sqrt{t}$$

$$\varphi_4 = \sin(t)$$

$$\sin \left[7 \cdot \underbrace{(2x_1 + 3x_2)}_{H_1} + 8 \cdot \underbrace{\log(x_1 + 6x_2)}_{H_2} + 9 \cdot \underbrace{\sqrt{(-2x_2 - 4)}}_{H_3} \right]$$

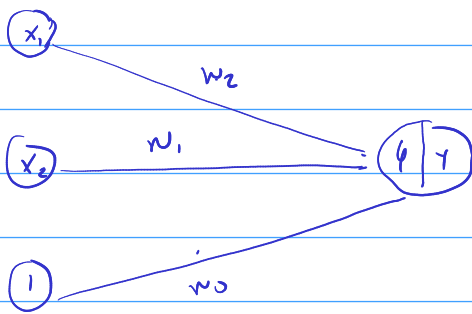
$$= 7$$

$$(7 = \sin(7 \cdot H_1 + 8 \cdot H_2 + 9 \cdot H_3))$$

(*) $\varphi_1, \varphi_2, \varphi_3$ and φ_4 are called activation functions

(*) Parameters in NN.

input layer



$$\phi(t) = +$$

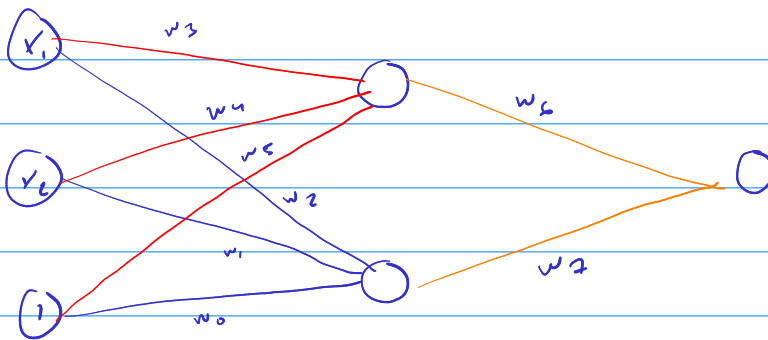
$$\gamma = \underline{w_0} + \underline{w_1} \cdot \underline{x_1} + \underline{w_2} \cdot \underline{x_2}$$

(untrained network) : 3 parameters . training this network means

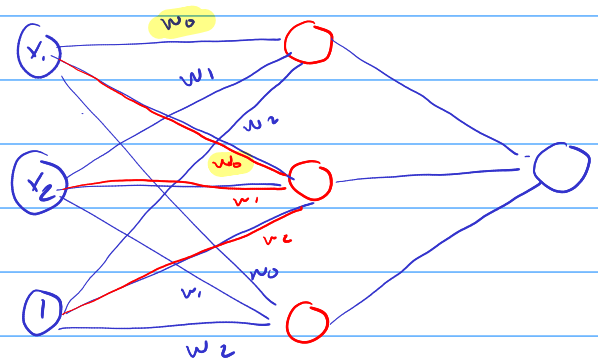
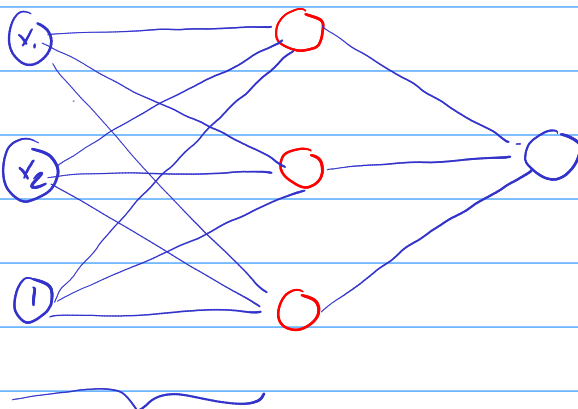
finding these 3 parameters.

Input layer

8 parameters



(*)



Fully connected layer
($3 \times 3 = 9$ parameters)

we can set conditions on
parameters to reduce the number
of parameters.

Path

u_1	u_2	u_3
x_1	$\ln x_2$	$x_1 \cdot \ln x_2$

x_1	x_2	y
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$$y = x_1^{x_2}$$

\Rightarrow

$\ln y$

$=$

$$x_2 \cdot \ln x_1$$

$$\phi_1(t) = \ln t$$

x_1

?

\ln

ϕ_1

x_2

x_1, x_2	x_1	x_2	y
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$$y = x_1 + x_2 + x_1 x_2$$

$$y = (\alpha x_1)^{\beta x_2}$$

\Rightarrow

$$\ln y = \beta x_2 \cdot \ln(\alpha x_1)$$

$$\ln(\ln y) = \ln(\beta x_2) +$$

$$\ln(\ln(\alpha x_1))$$

