

Centro Brasileiro de Pesquisas Físicas



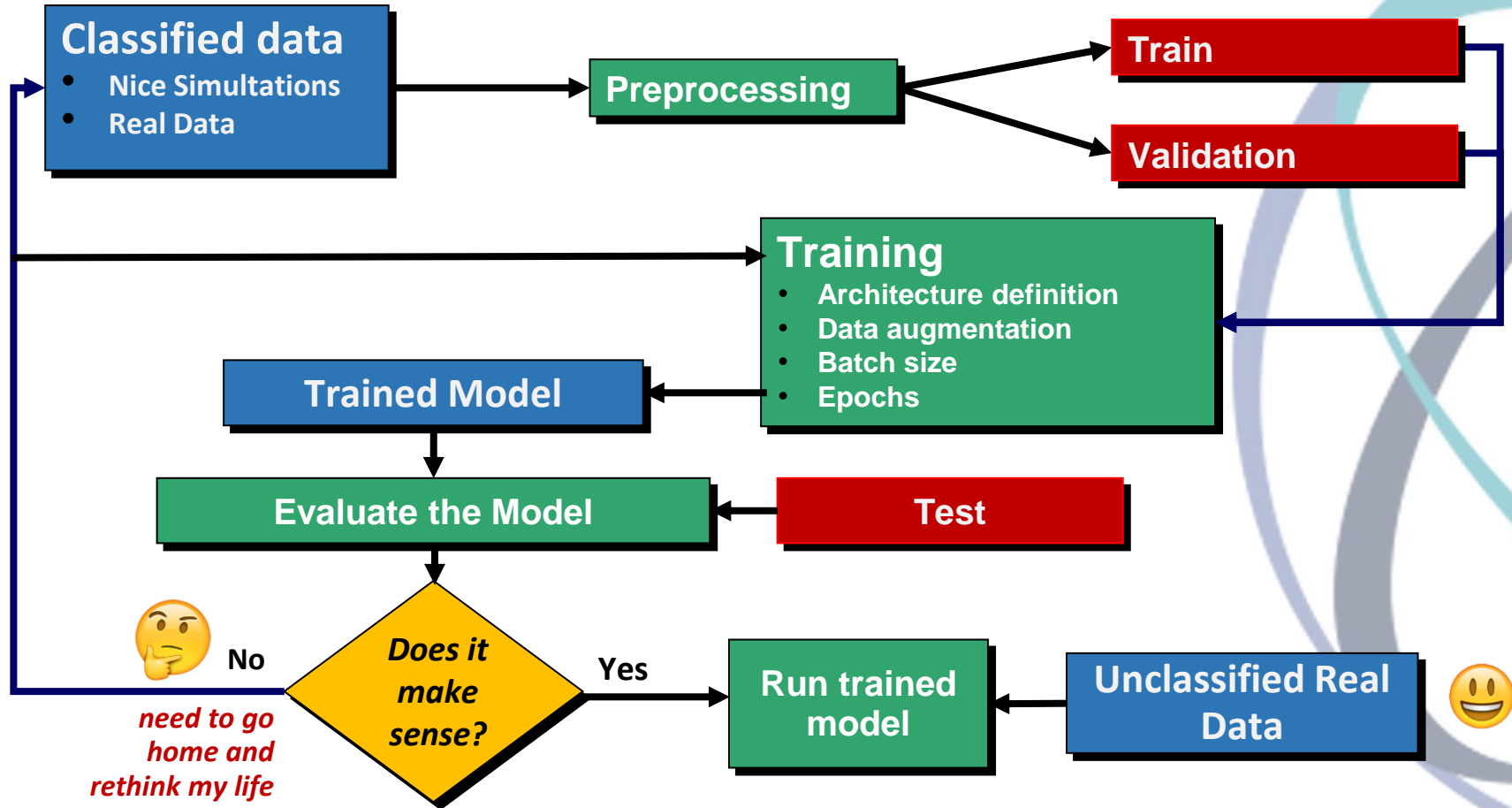
Métodos para Análise de grande volume de dados e Astroinformática

Clécio Roque De Bom – debom@cbpf.br

clearnightsrthebest.com

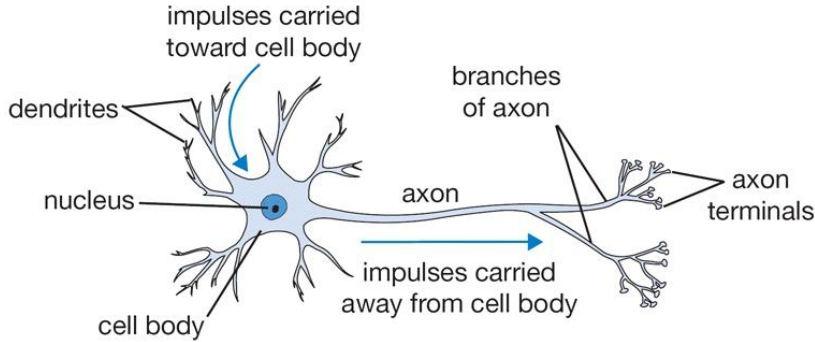


ML: Workflow Supervised – Again!



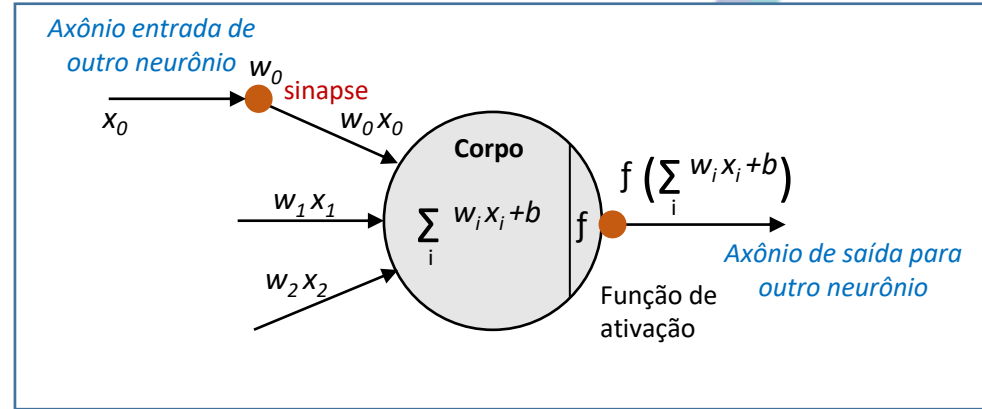
Neurônio

Computação (Neurônio Artificial) → Inspiração na Biologia



Neurônio Biológico: bloco computacional de processamento do cérebro.

Cérebro Humano: ~100 – 1.000 trilhões de sinapses



Neurônio Artificial: bloco computacional de processamento das Redes Neurais Artificiais.

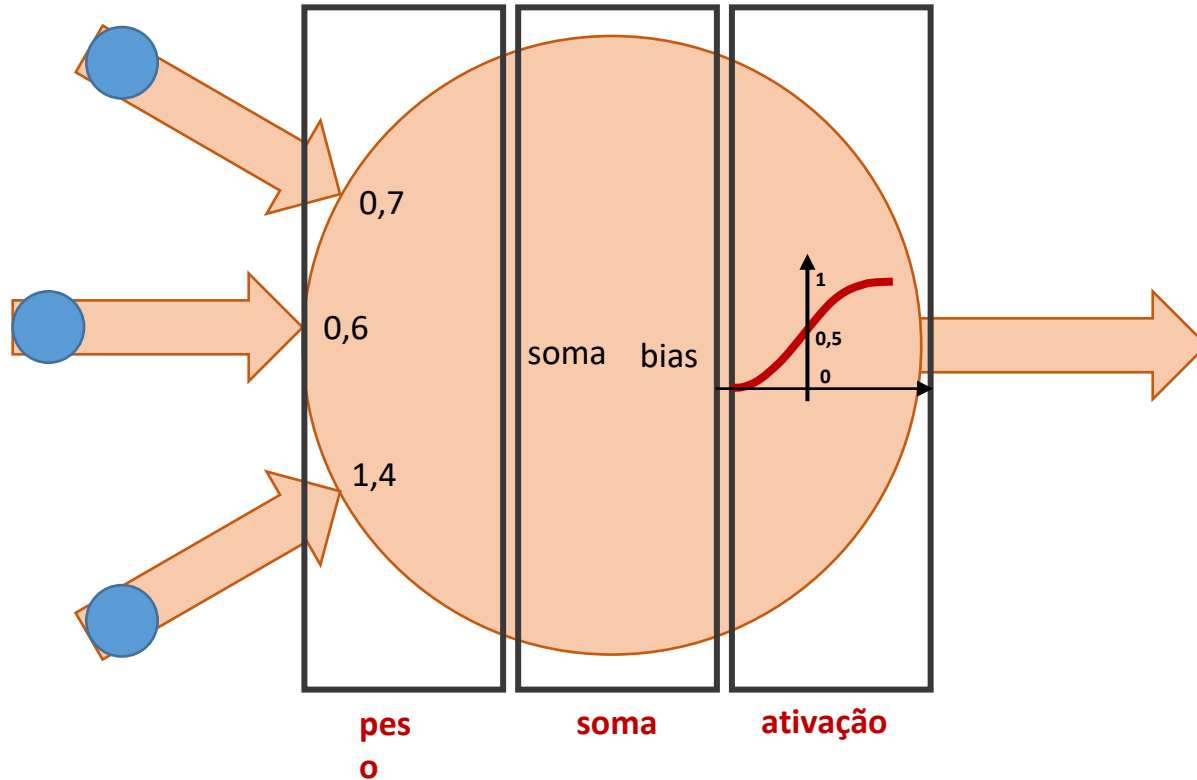
Rede Neural Artificial : ~1 – 10 bilhões de sinapses.

10.000 x

Rede Neural Artificial

Perceptron

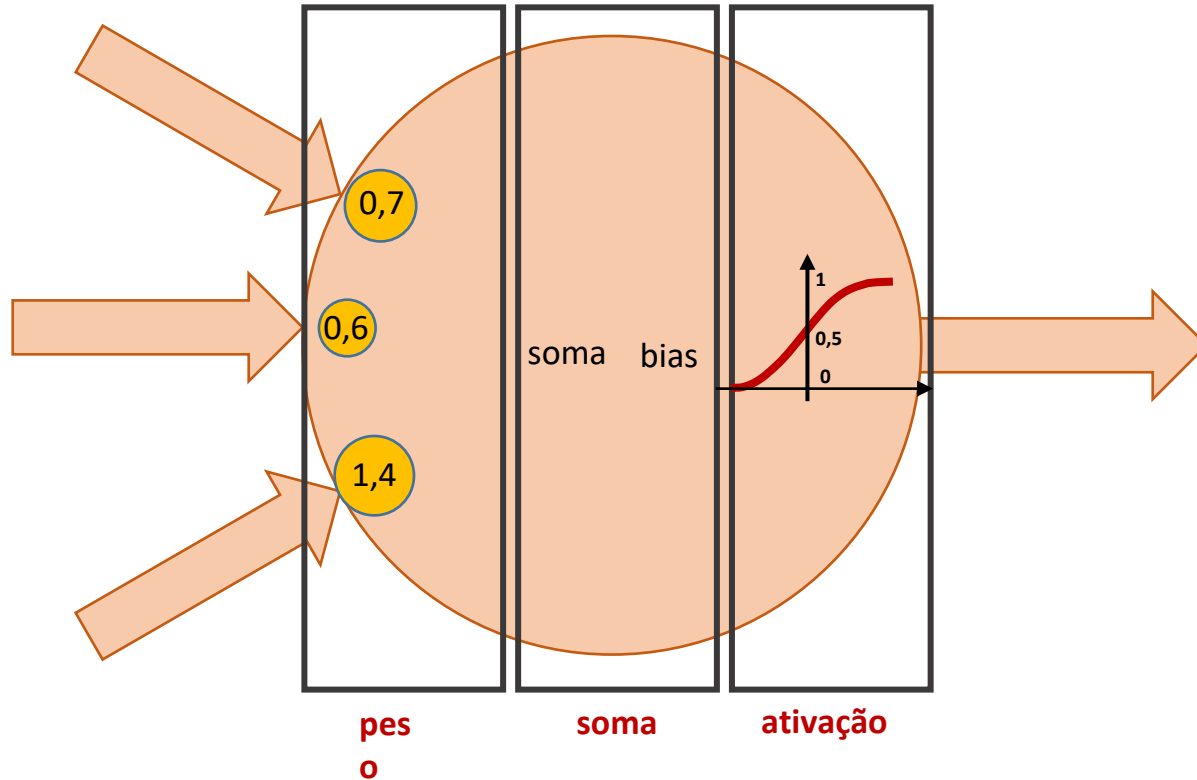
Frank Rosenblatt (1957)



Rede Neural Artificial

Perceptron

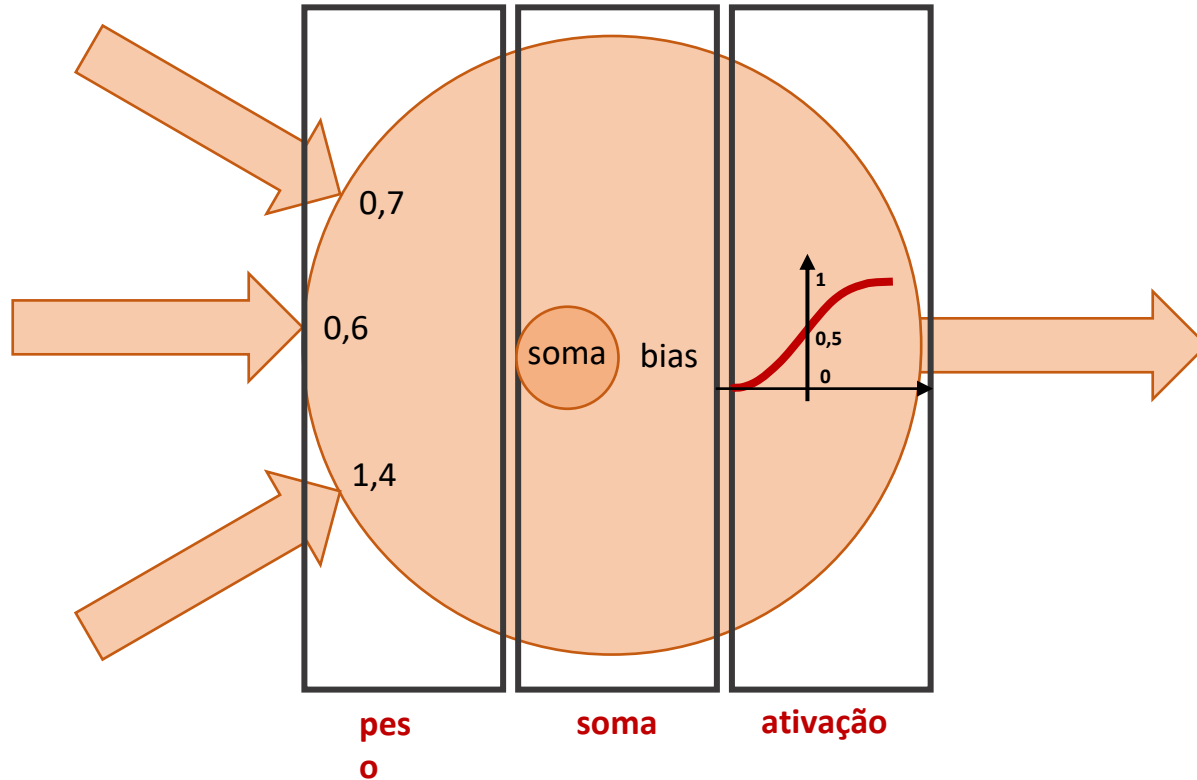
Frank Rosenblatt (1957)



Rede Neural Artificial

Perceptron

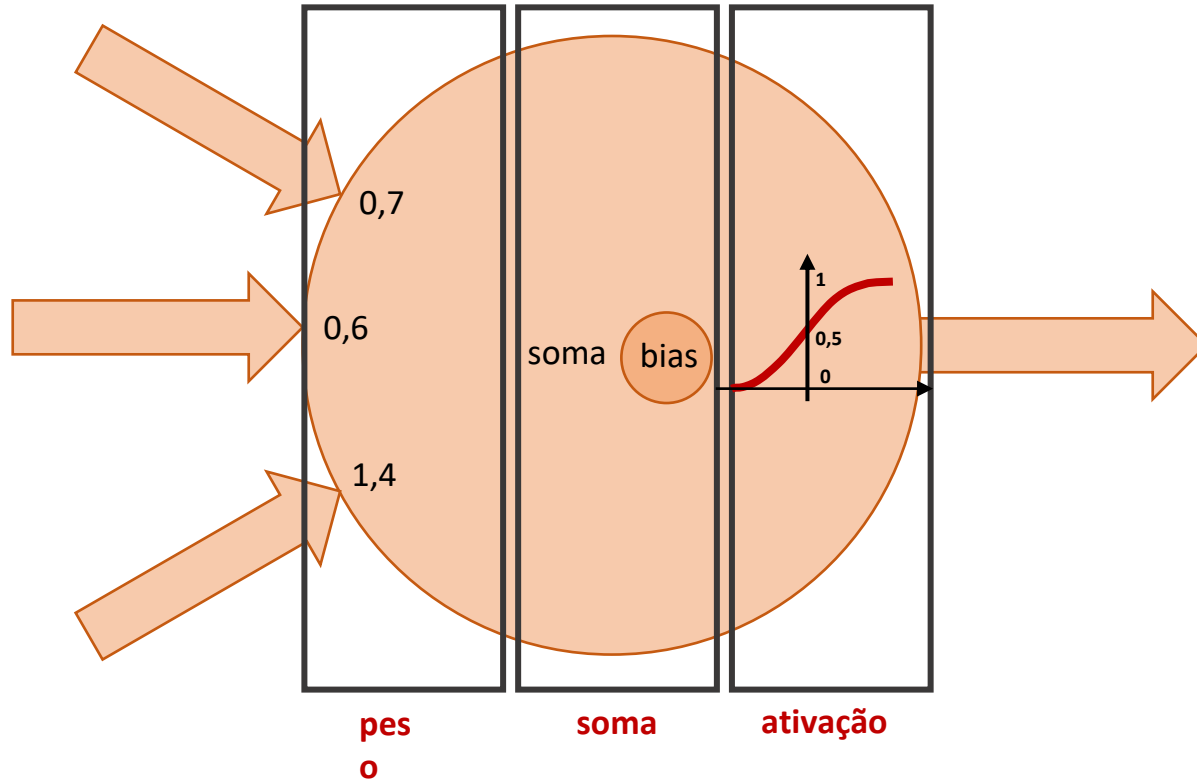
Frank Rosenblatt (1957)



Rede Neural Artificial

Perceptron

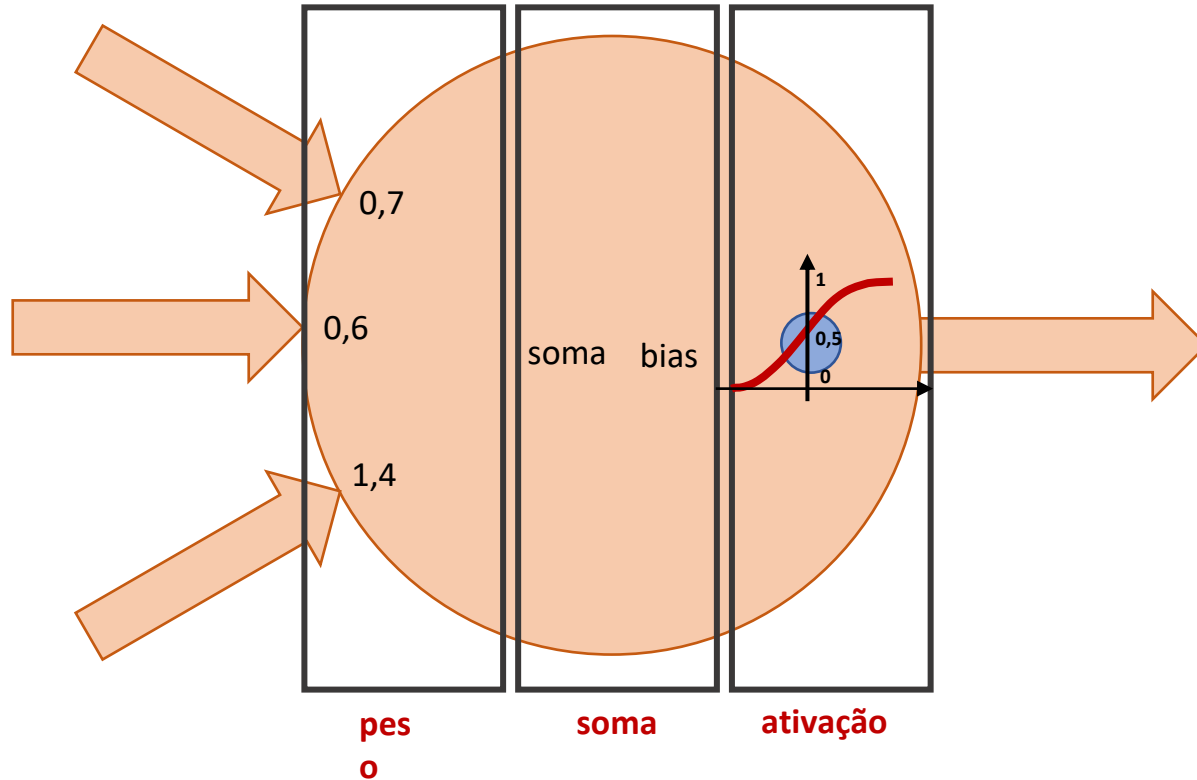
Frank Rosenblatt (1957)



Rede Neural Artificial

Perceptron

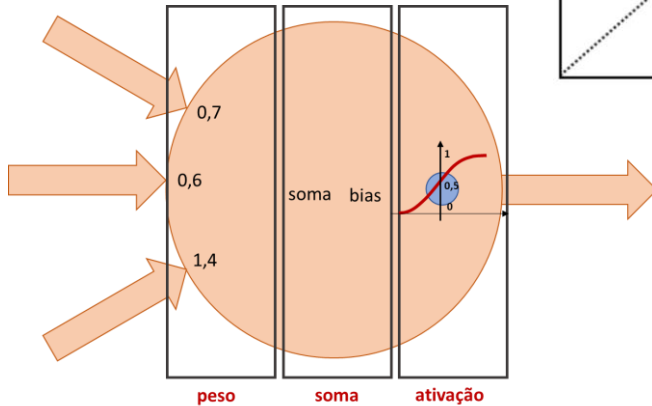
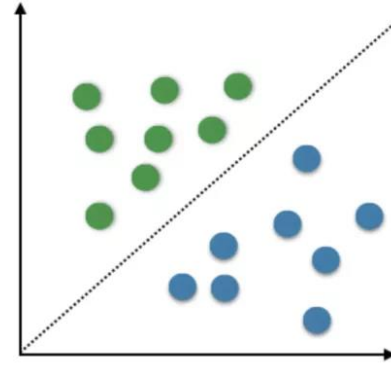
Frank Rosenblatt (1957)



Rede Neural Artificial

Perceptron

Frank Rosenblatt (1957)



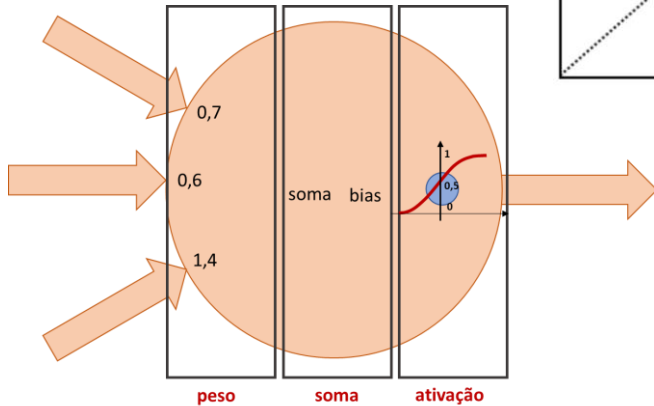
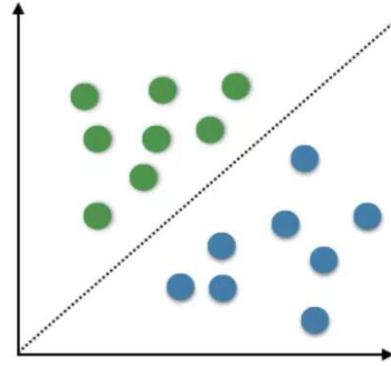
Algoritmo

- Inicialize a rede Perceptron com pesos (w) aleatórios;
- Para uma data entrada, processe a saída da rede;
- Se a saída da rede não for igual a saída desejada, então a rede deve ser alterada, trocando os valores dos pesos (w) das sinapses;
- Repita esse procedimento com todos os dados de treinamento até a rede Perceptron não apresentar mais erros.

Rede Neural Artificial

Perceptron

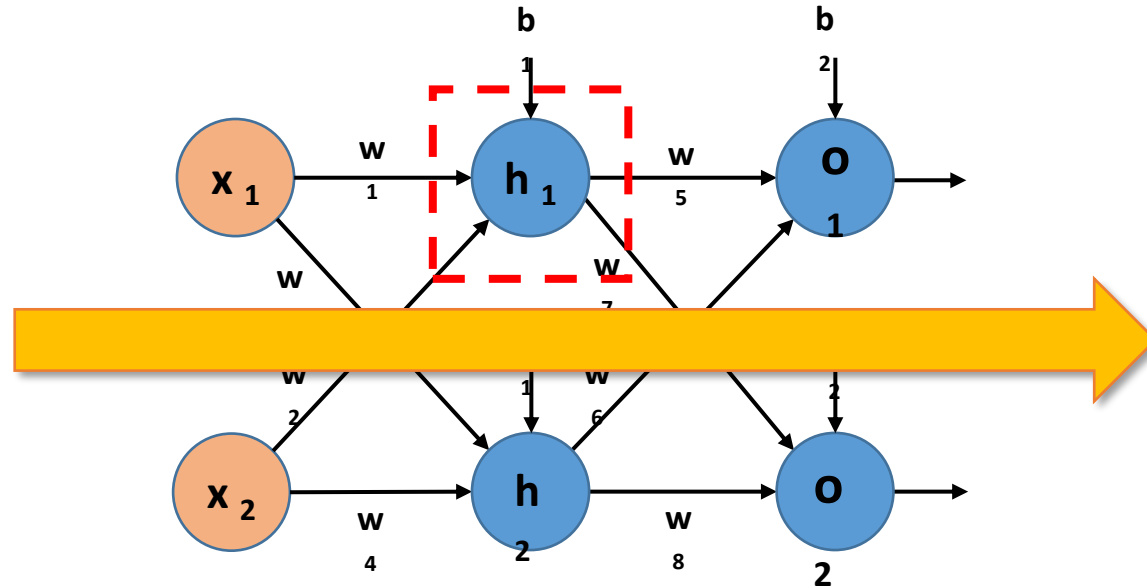
Frank Rosenblatt (1957)



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Feedforward and Backpropagation

Fase 1 Propagação

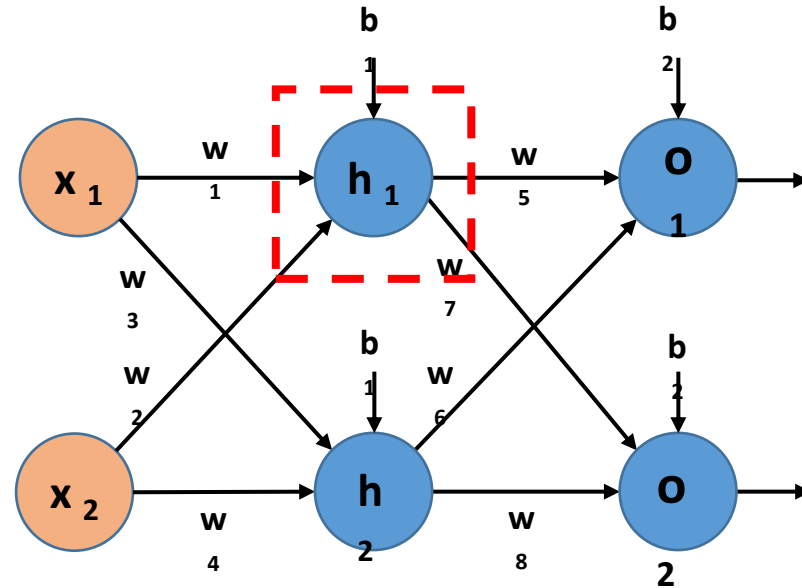


$$uh_1 = x_1 * w_1 + x_2 * w_2 + b_1 * 1$$

$$g(h_1) = g(uh_1) = \frac{1}{1+e^{-uh_1}}$$

Feedforward and Backpropagation

Fase 1 Propagação

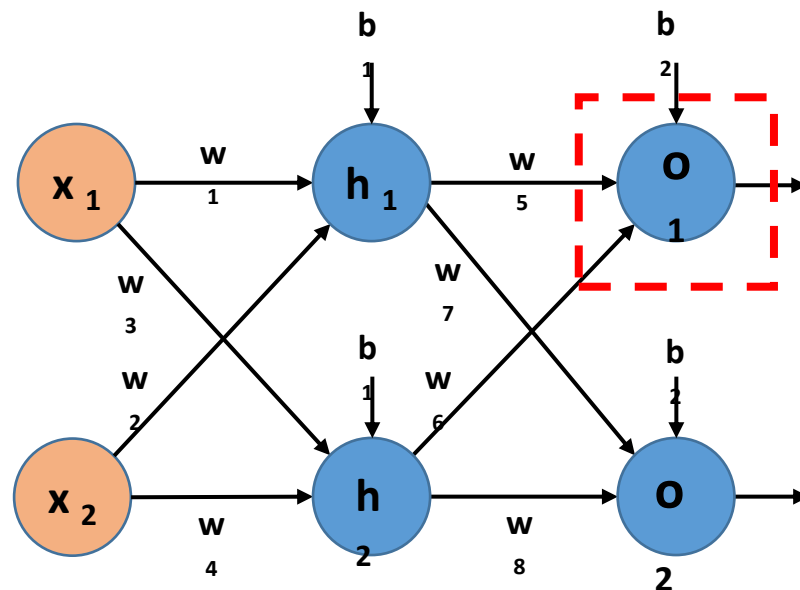


$$uh_2 = x_1 * w_3 + x_2 * w_4 + b_1 * 1$$

$$g(h_2) = g(uh_2) = \frac{1}{1 + e^{-uh_2}}$$

Feedforward and Backpropagation

Fase 1 Propagação

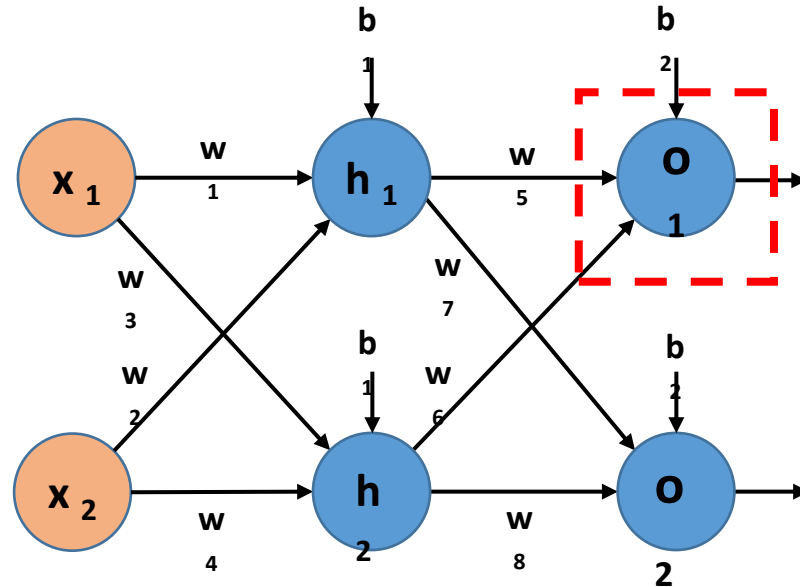


$$uo_1 = g(h_1) * w_5 + g(h_2) * w_6 + b_2 * 1$$

$$\hat{y}_1 = g(o_1) = g(uo_1) = \frac{1}{1+e^{uo_1}}$$

Feedforward and Backpropagation

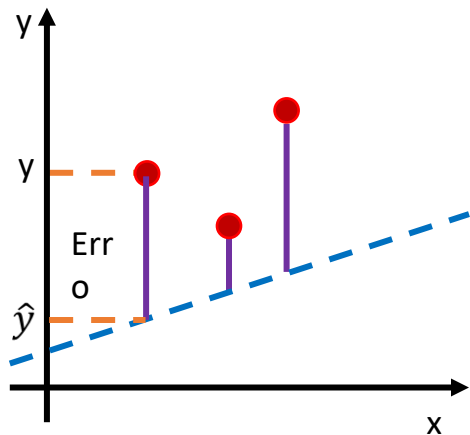
Fase 1 Propagação



$$uo_2 = g(h_1) * w_7 + g(h_2) * w_8 + b_2 * 1$$

$$\hat{y}_2 = g(o_2) = g(uo_2) = \frac{1}{1+e^{uo_2}}$$

Rede Neural Artificial: Erro/Custo



y = valor original

\hat{y} = valor predito

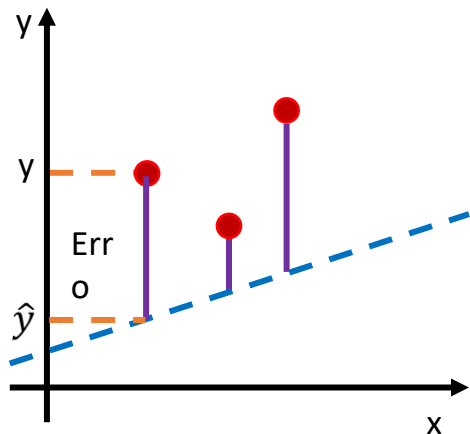
$$\hat{y} = w_0 + w_1 x$$

$$J(w_0, w_1) = \frac{\sum_{i=1}^m (\hat{y}_i - y_i)^2}{m \text{ (média)}}$$

CUST
COST

MSE (*Mean Square Error* – Erro quadrático médio)

Rede Neural Artificial: Erro/Custo



y = valor original

\hat{y} = valor predito

$$\hat{y} = w_0 + w_1 x$$

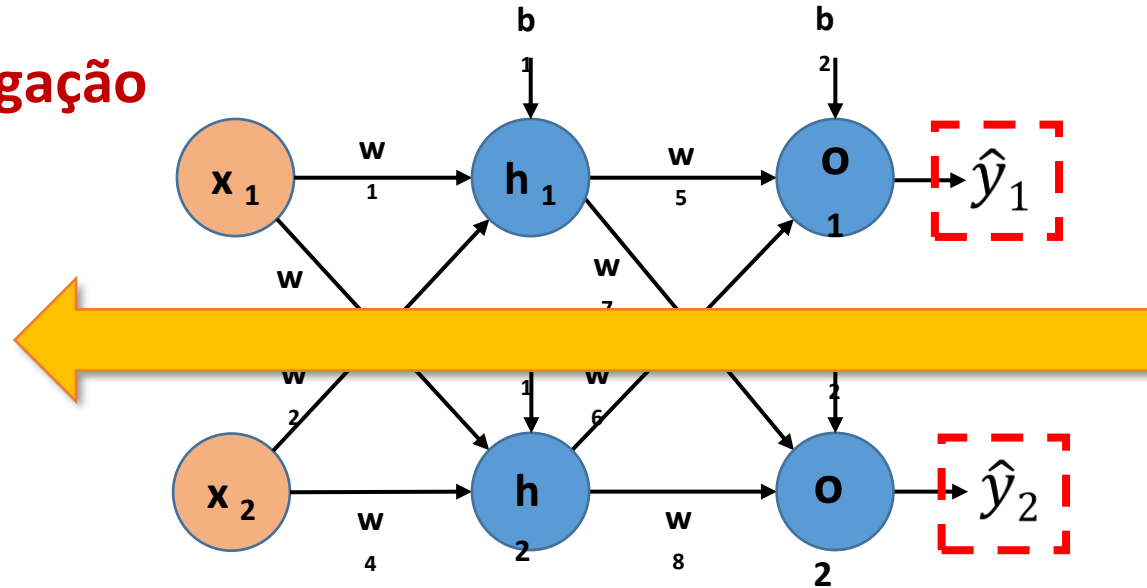
$$J(w_0, w_1) = \frac{1}{2m} \sum_{i=1}^m (\hat{y}_i - y_i)^2$$

MSE (*Mean Square Error* – Erro quadrático médio)

Como reduzir o custo? $\min_{(w_0, w_1)} J(w_0, w_1)$

Feedforward and Backpropagation

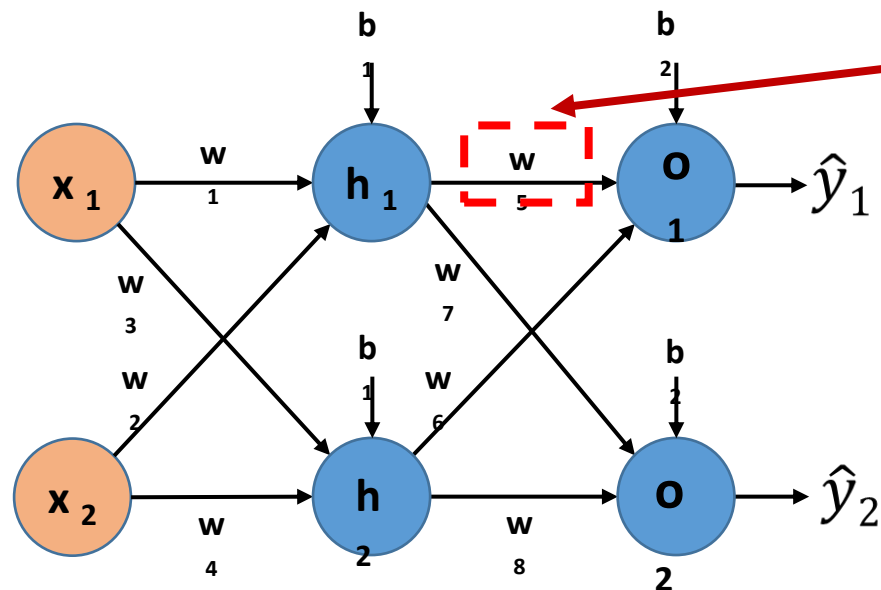
Fase 1 Retropropagação



$$E_{total} = \frac{1}{2} \sum_{k=1}^N (\hat{y}_k - y_k)^2 = E_{o1} + E_{o2}$$

Feedforward and Backpropagation

Fase 1 Retropropagação

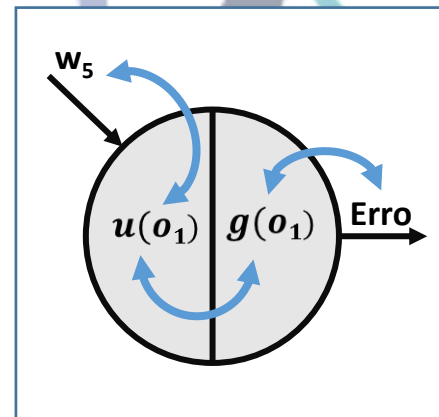


Correção de w_5 :
Queremos estimar quanto w_5 afeta o Erro total

E_{total}

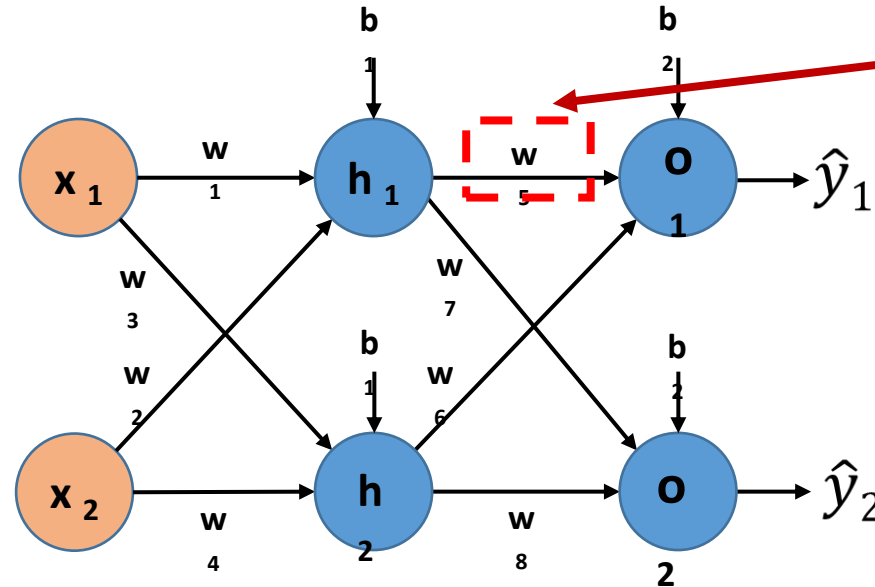
$$\frac{\partial E_{total}}{\partial w_5} = \text{gradiente em relação a } w_5$$

$$\frac{\partial E_{total}}{\partial w_5} = \frac{\partial E_{total}}{\partial g_{o_1}} * \frac{\partial g_{o_1}}{\partial u_{o_1}} * \frac{\partial u_{o_1}}{\partial w_5}$$



Feedforward and Backpropagation

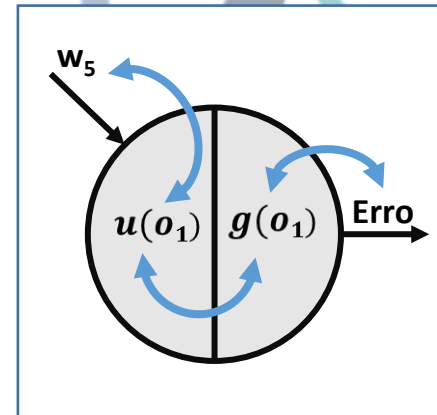
Fase 1 Retropropagação



Correção de w_5 :
Queremos estimar
quanto w_5 afeta o
Erro total

E_{total}

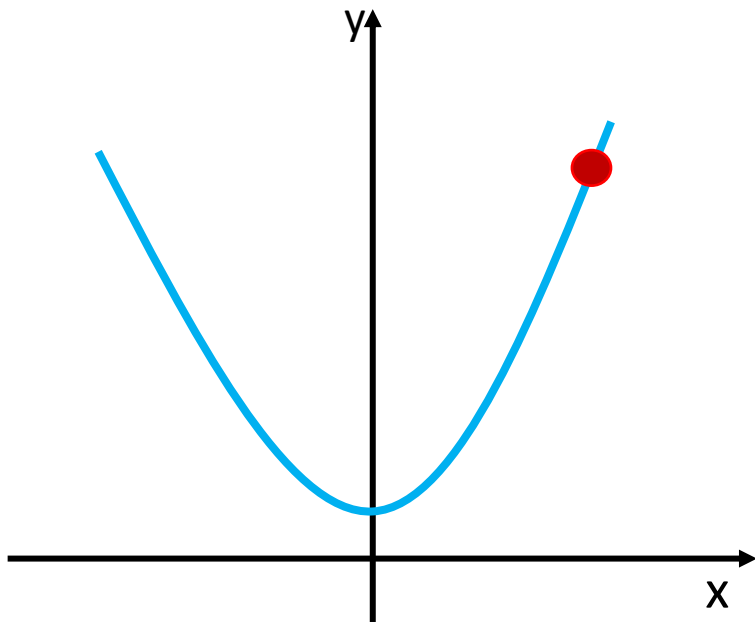
$$w_5^+ = w_5 - \eta * \frac{\partial E_{total}}{\partial w_5} :$$



Rede Neural Artificial: Otimização

Função quadrática

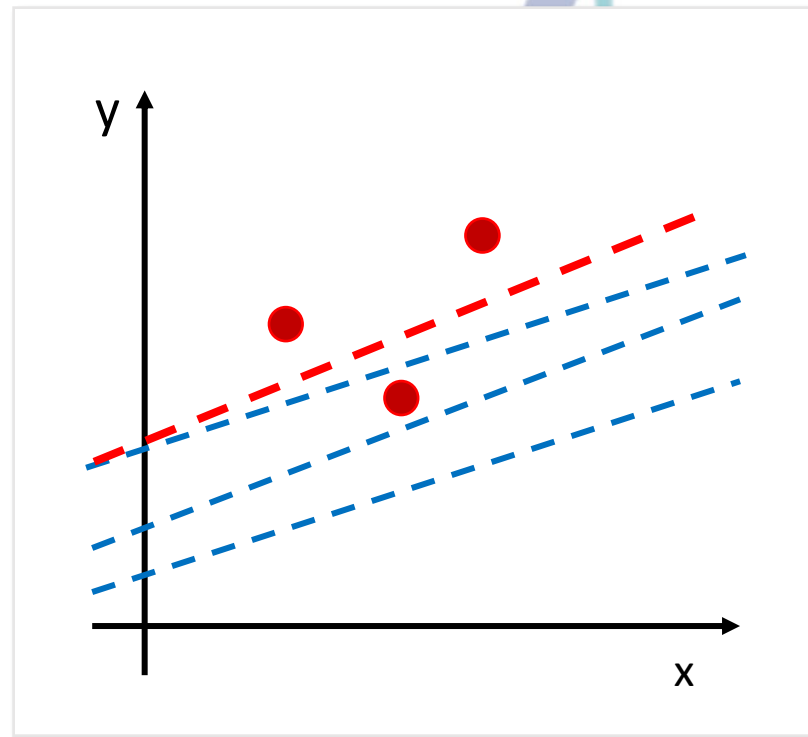
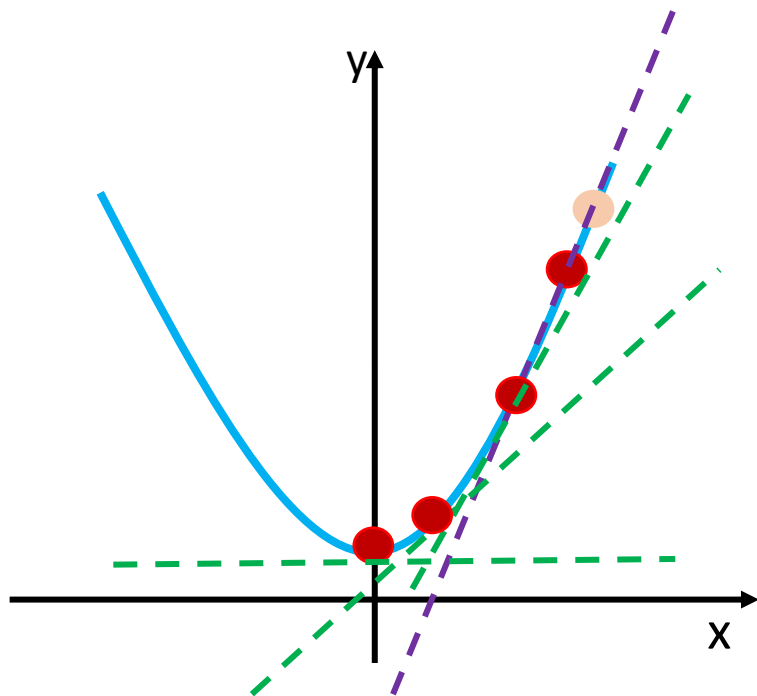
$$(\hat{y}_i - y_i)^2$$



Rede Neural Artificial: Otimização

Função quadrática

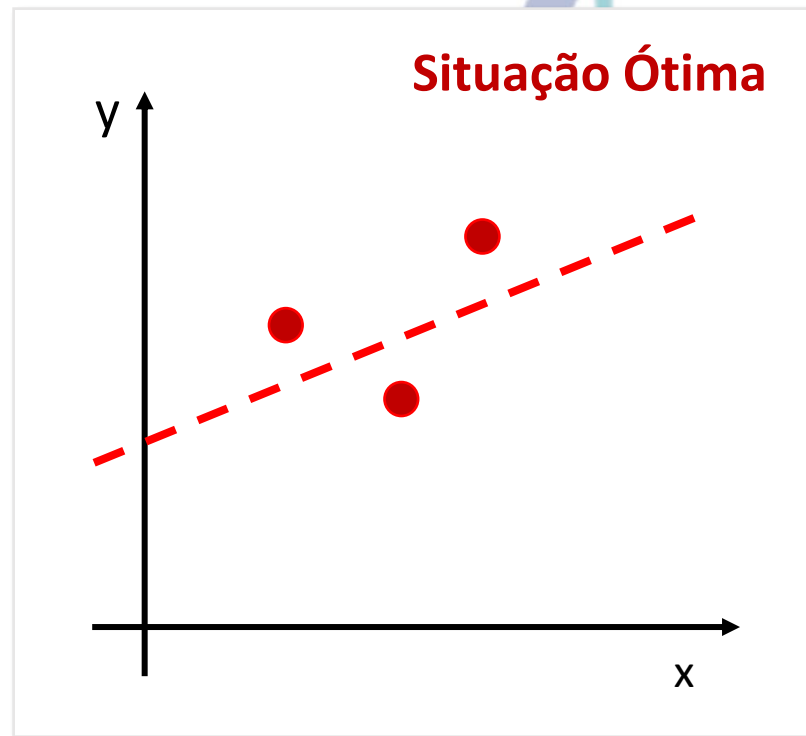
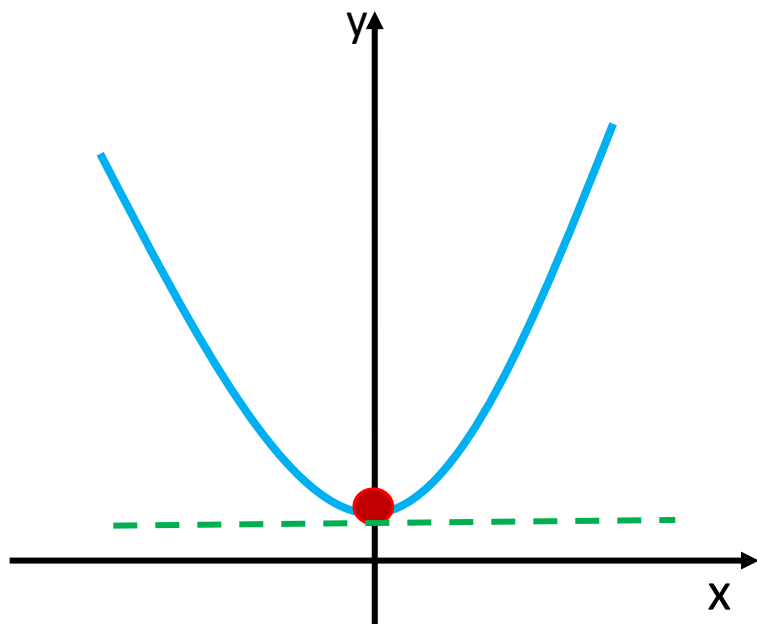
$$(\hat{y}_i - y_i)^2$$



Rede Neural Artificial: Otimização

Função quadrática

$$(\hat{y}_i - y_i)^2$$



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