

# Multi Layer Perceptron (MLP)

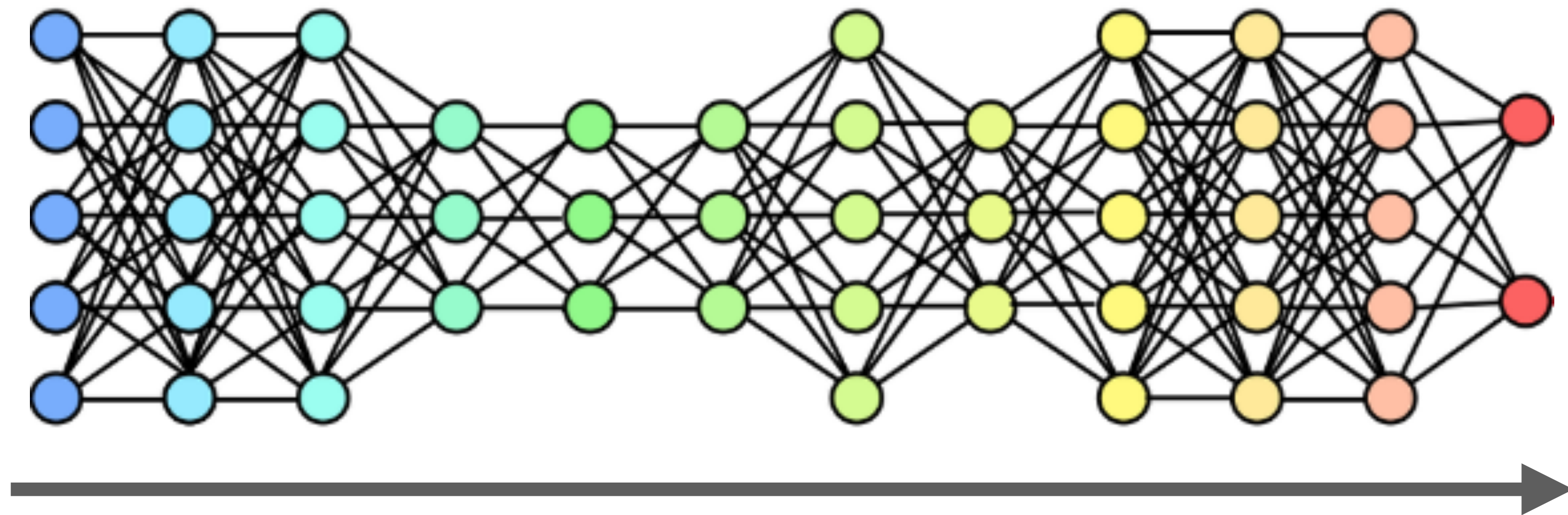
**Reti Neurali Totalmente Connesse**

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**UNIVERSITÀ  
DEGLI STUDI  
DI TRIESTE**

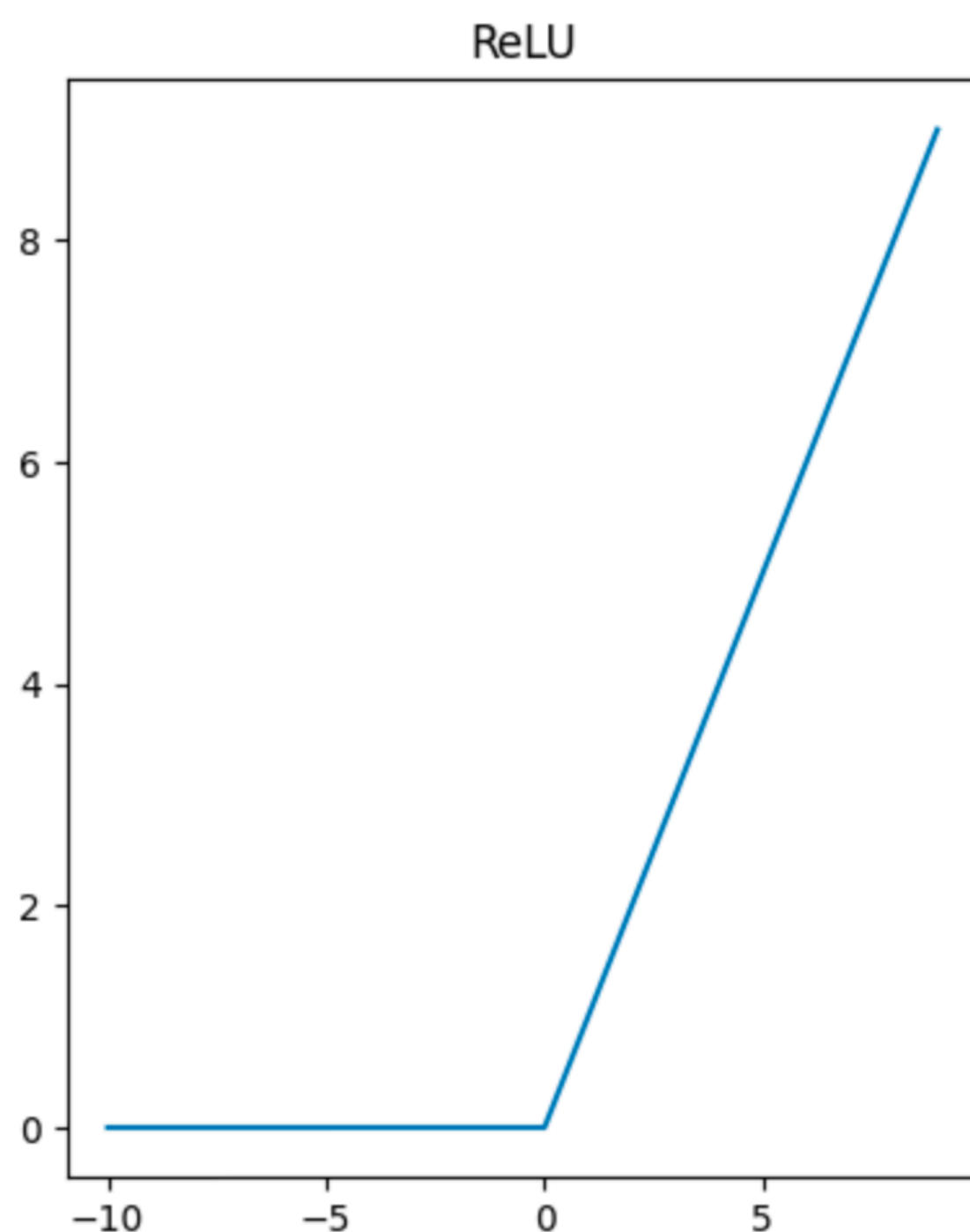
# Architettura MLP



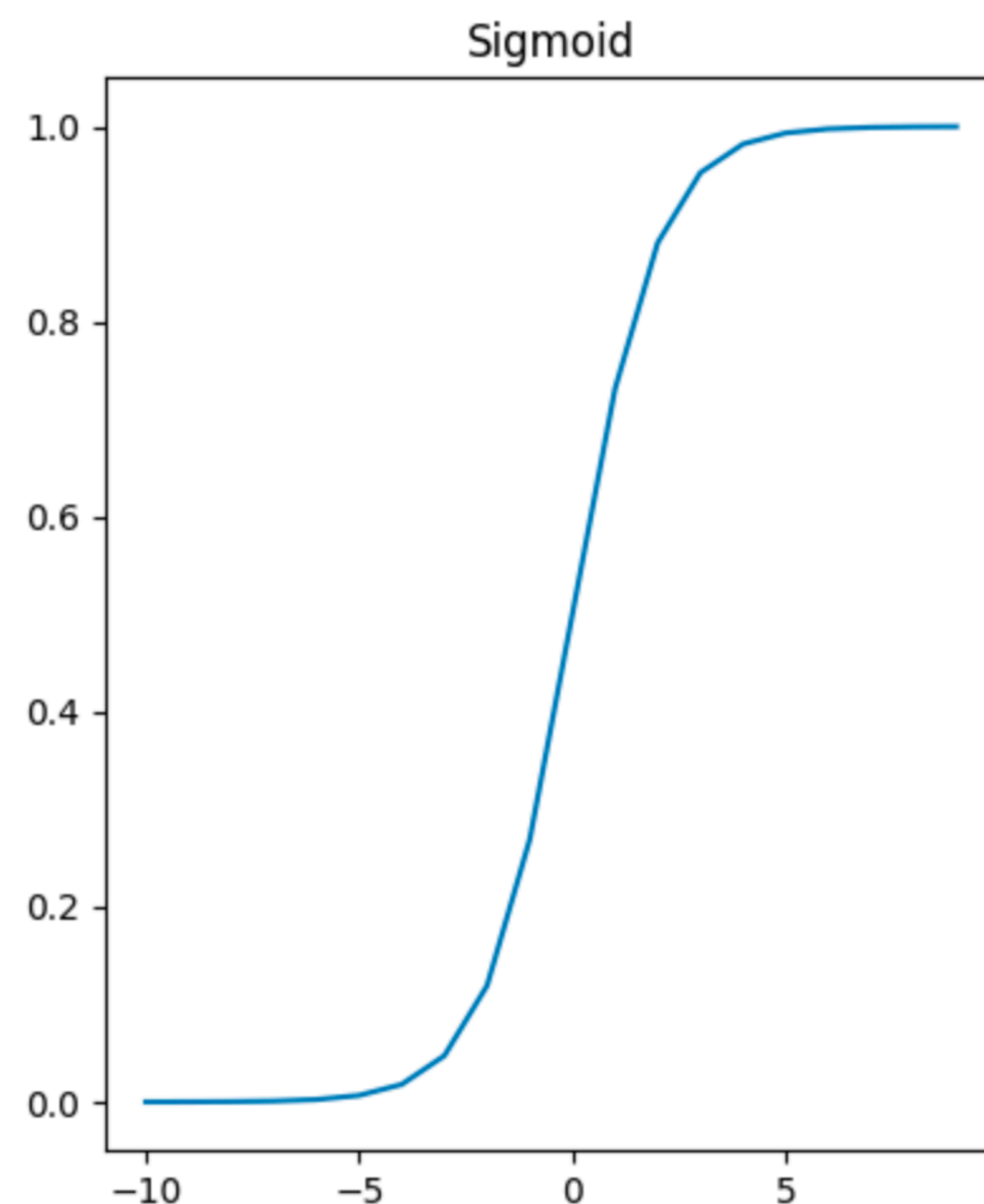
I calcoli per produrre un output dato un input vanno  
in questa direzione

L'errore (in funzione di  $w$  e  $b$ ), ovvero la funzione di  
costo o perdita, viene calcolato in questa direzione

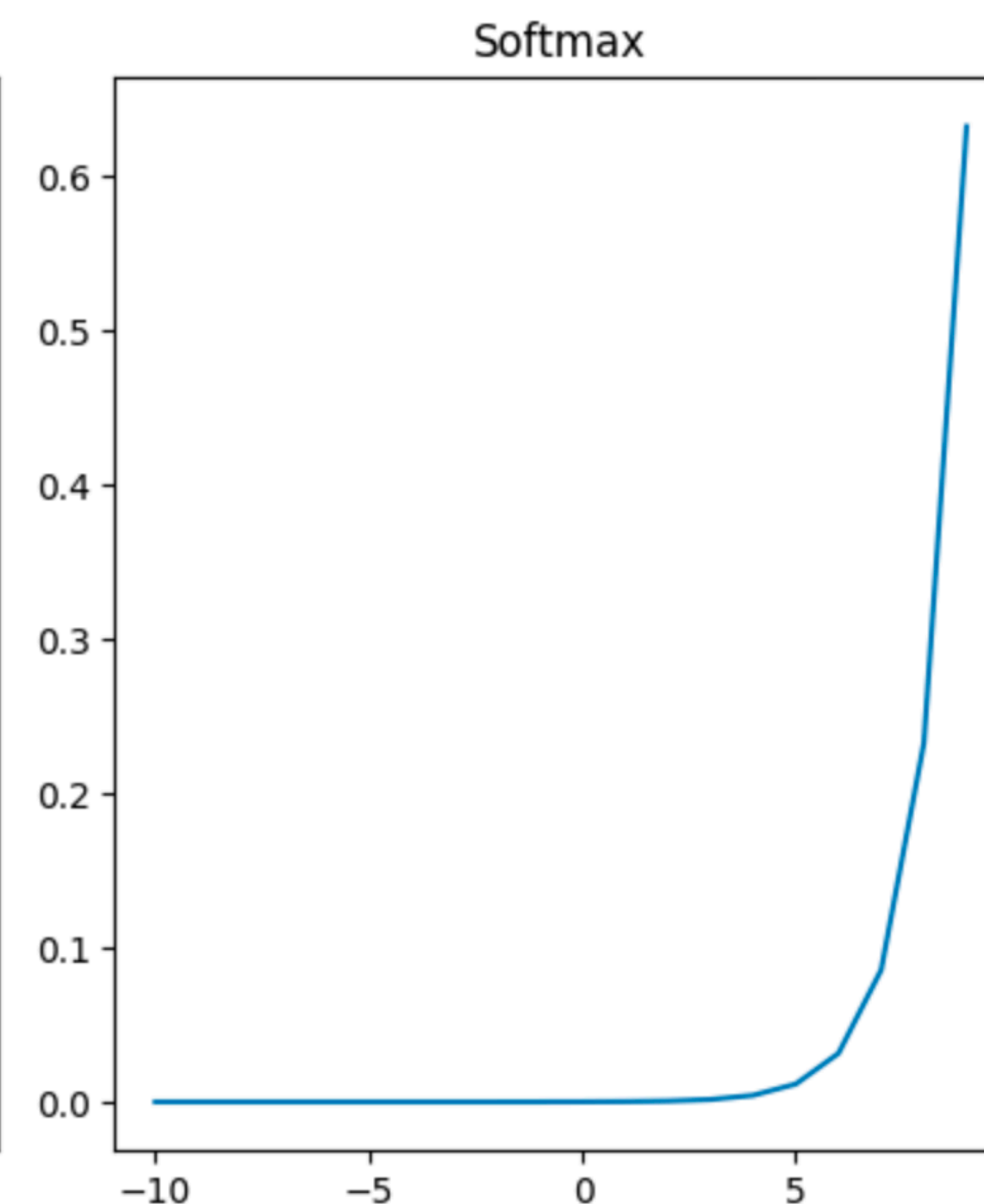
# Funzioni di Attivazione



Strati  
Intermedi



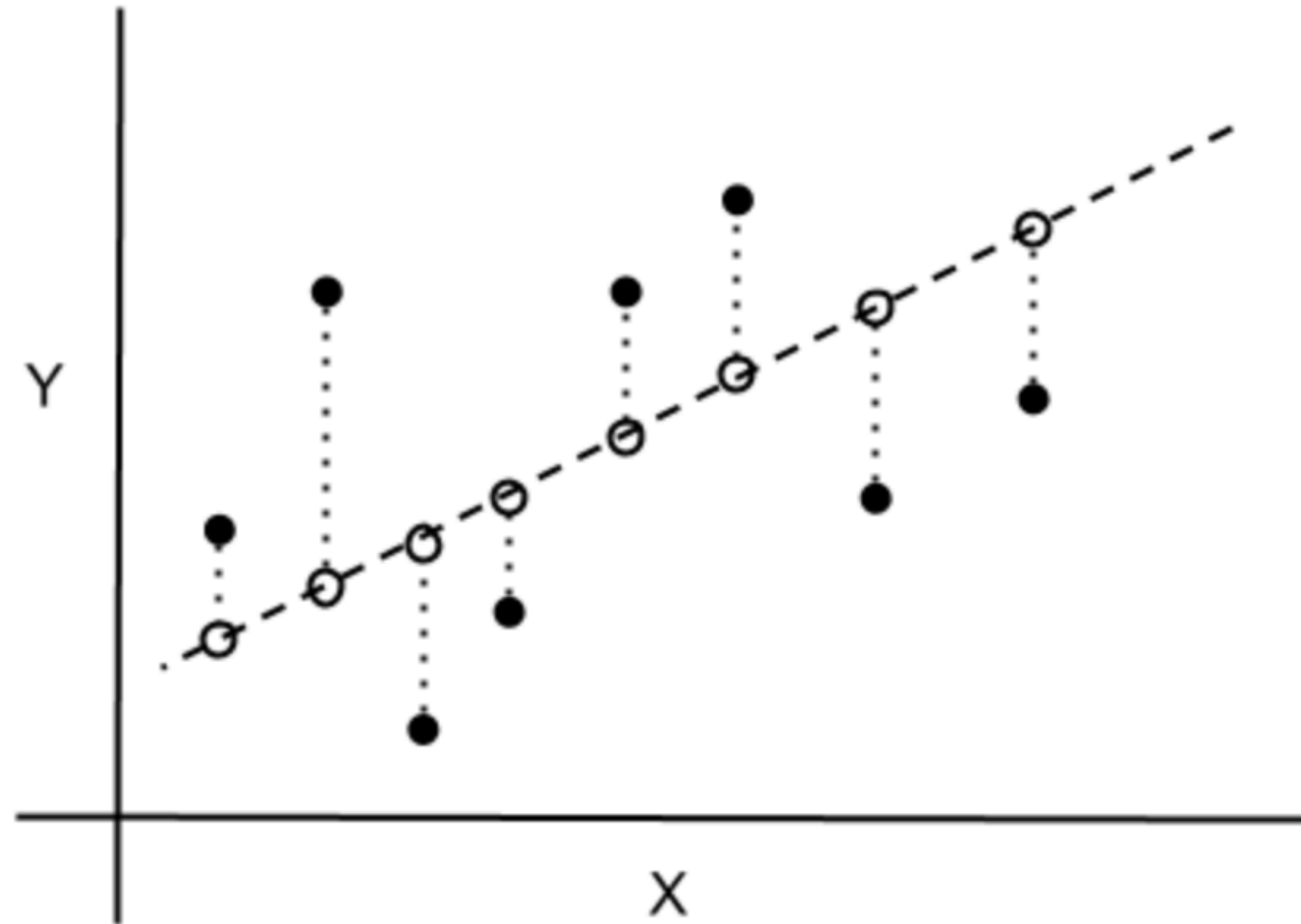
Output  
classificazione  
Binaria



Output  
classificazione  
Multi-classe

# Metriche di Valutazione Modelli

(Root) Mean Square Error - (R)MSE

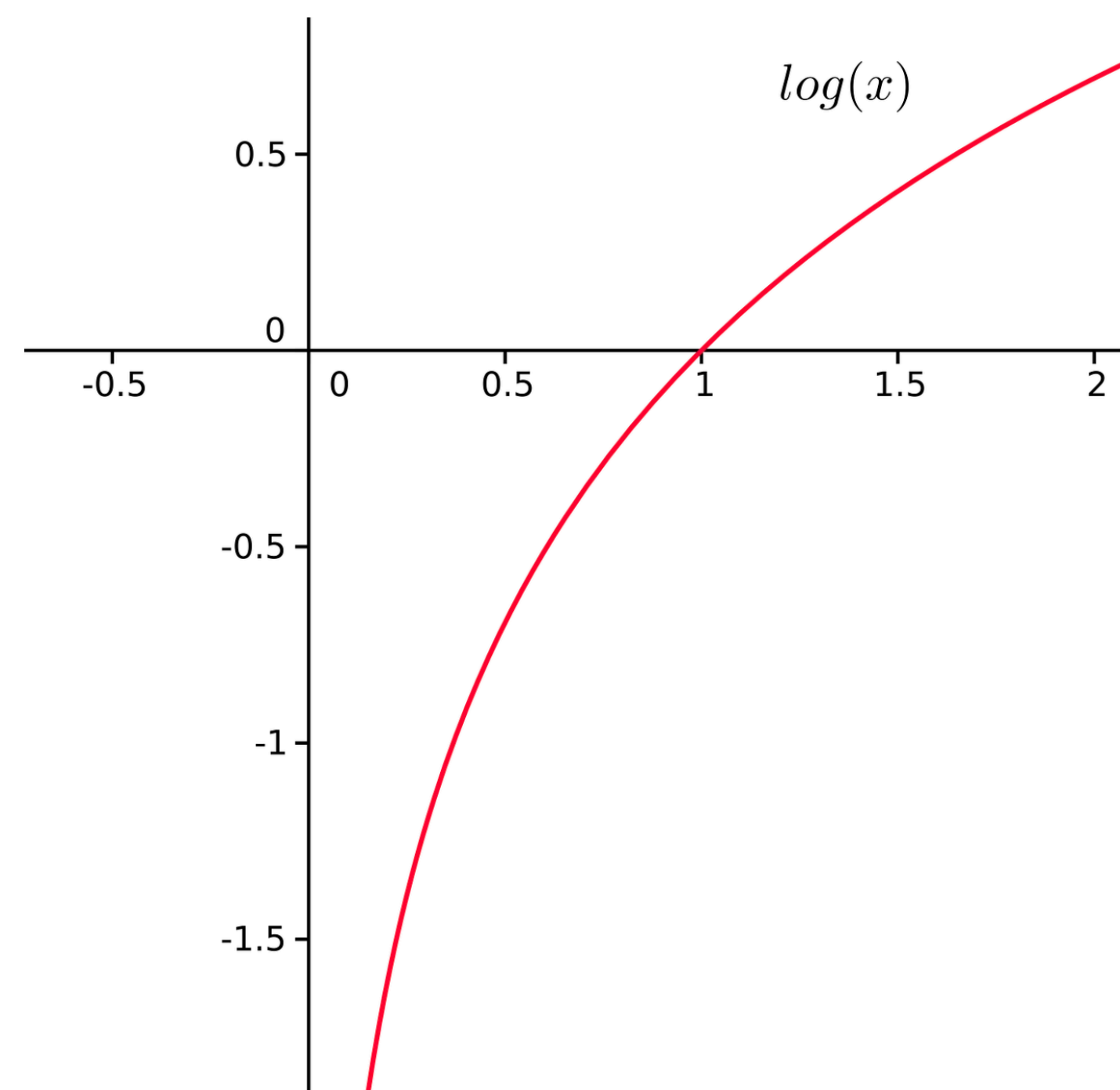


$$MSE = \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

# Metriche di Valutazione Modelli

(Binary) Cross Entropy - (B)CE



$$CE_x = - \sum_{i=1}^c t_i \log(p_i)$$

$$BCE_x = - (t \log(p) + (1 - t) \log(1 - p))$$

# Metriche di Valutazione Modelli

## Accuratezza e Confusion Matrix

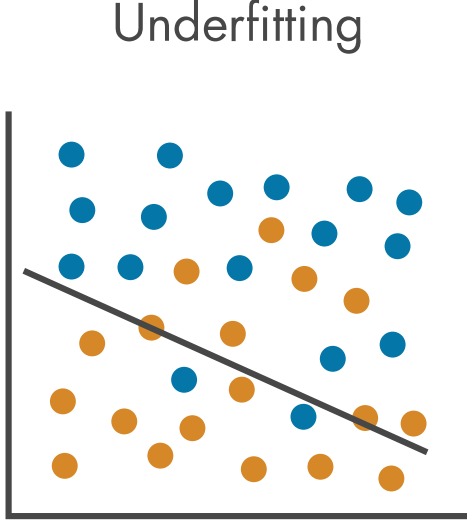
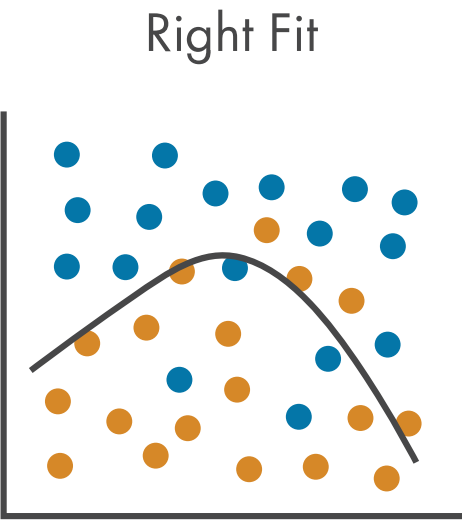
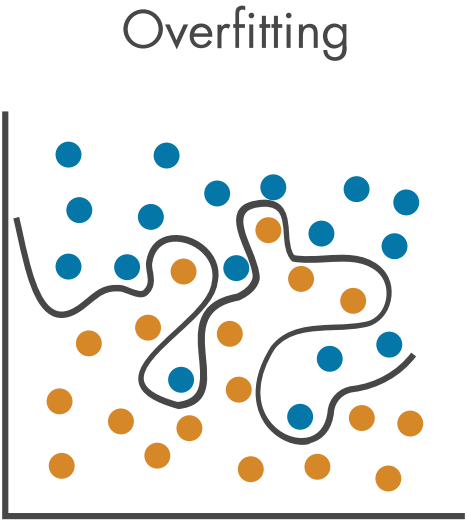
		PREDICTED	
		NEGATIVE	POSITIVE
ACTUAL	NEGATIVE	TRUE NEGATIVES (TN)	FALSE POSITIVES (FP)
	POSITIVE	FALSE NEGATIVES (FN)	TRUE POSITIVES (TP)

$$ACC = \frac{1}{n} \sum_{i=0}^n \mathbb{I}(y_i = \hat{y}_i) = \frac{\#pred\_corrette}{\#pred\_totali}$$

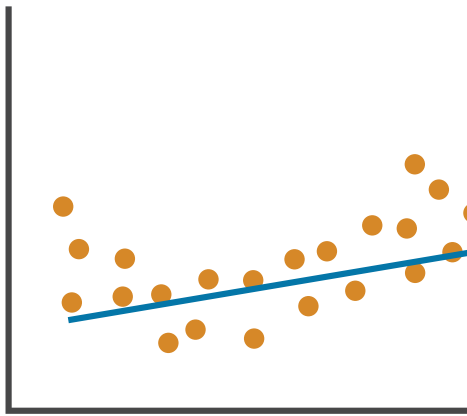
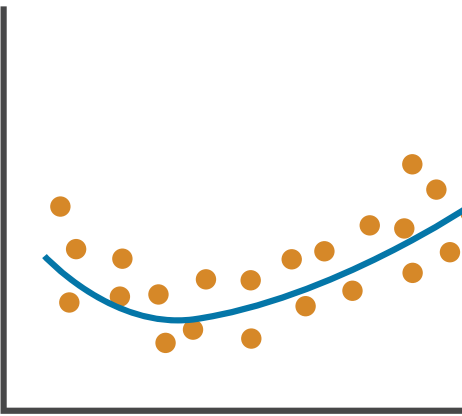
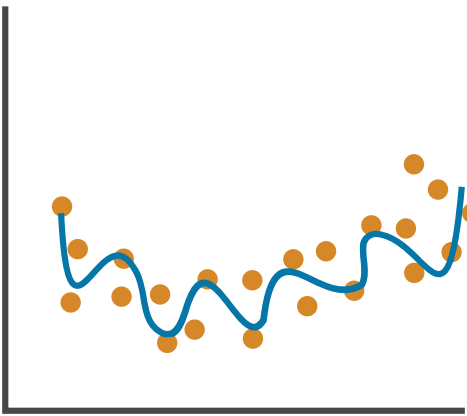


# Overfitting e Dropout

Classification

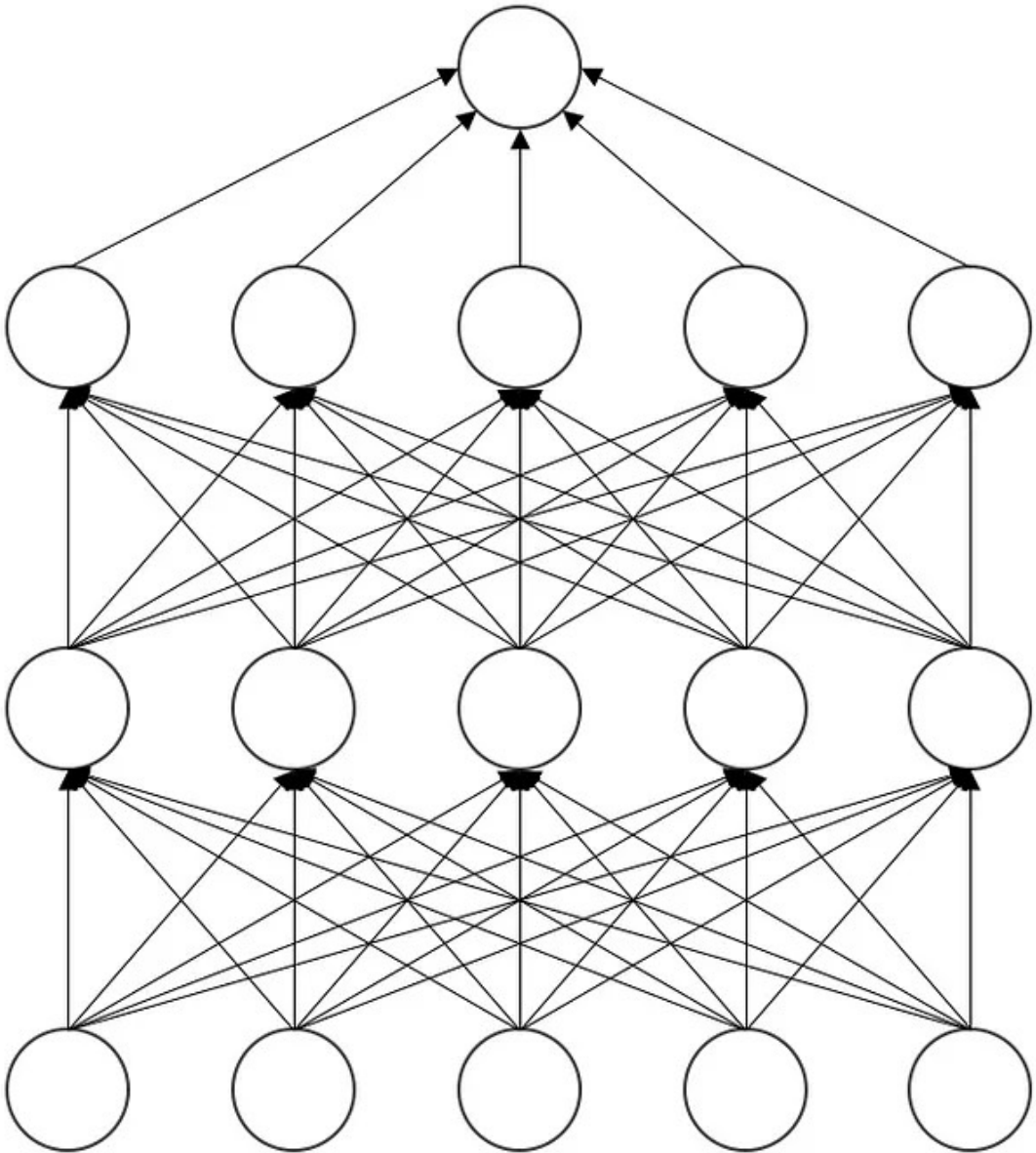


Regression

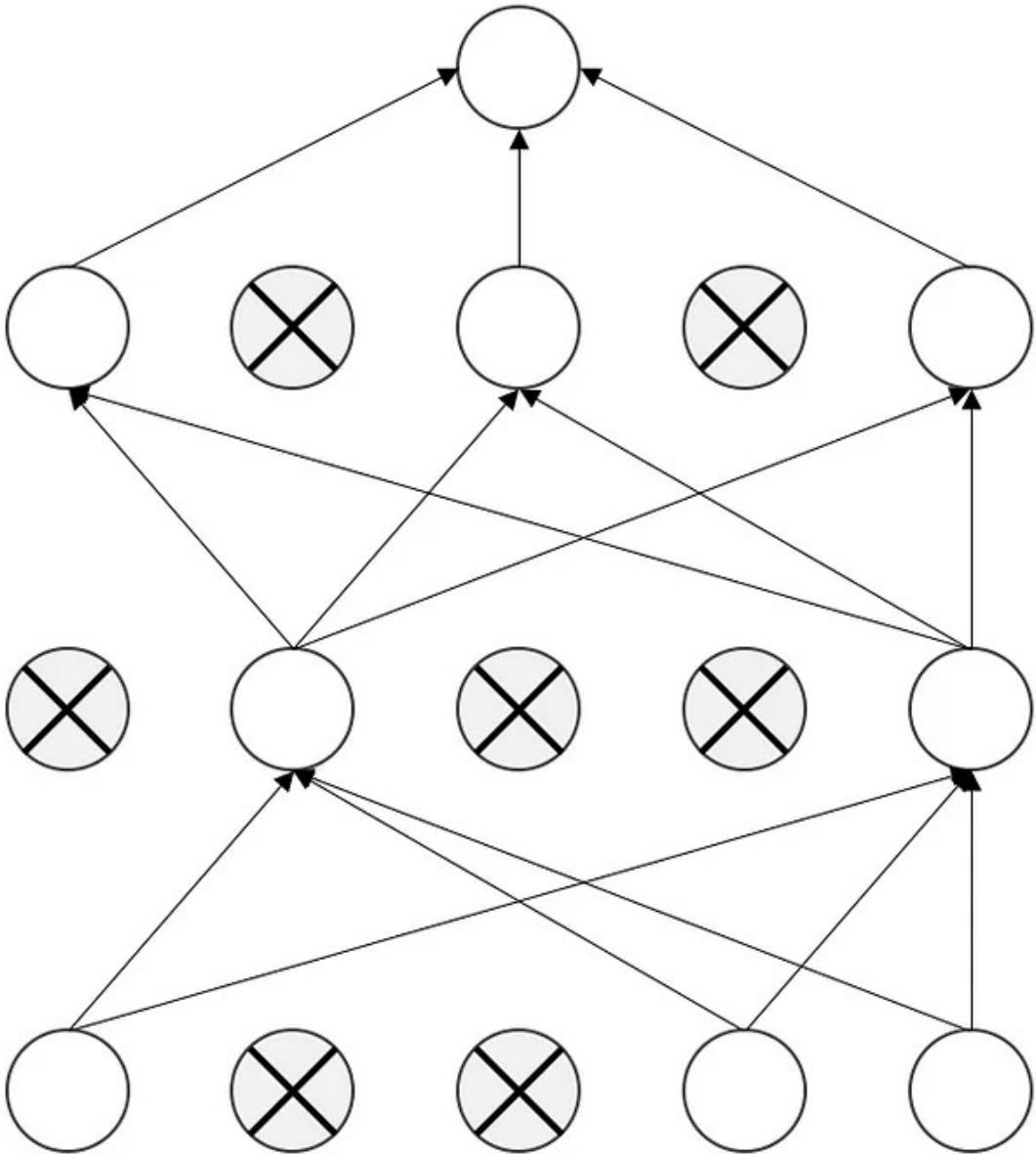


Problema

(Possibile) soluzione



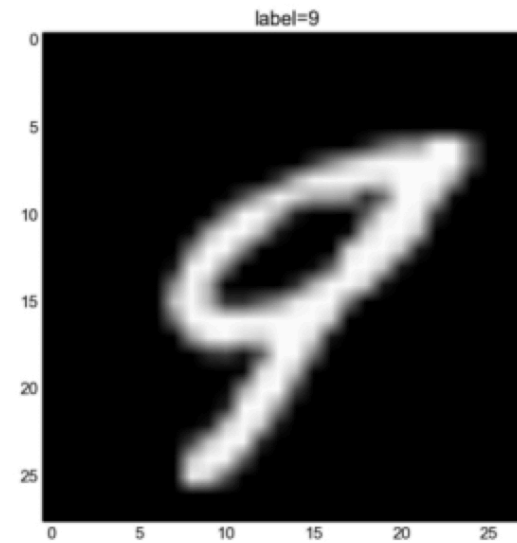
Standard Neural Net



After applying dropout

# One-Hot-Encoding

Trasformare Variabili Categorie in Vettori



0, 0, 0, 0, 0, 0, 0, 0, 0, 1

0 1 2 3 4 5 6 7 8 9