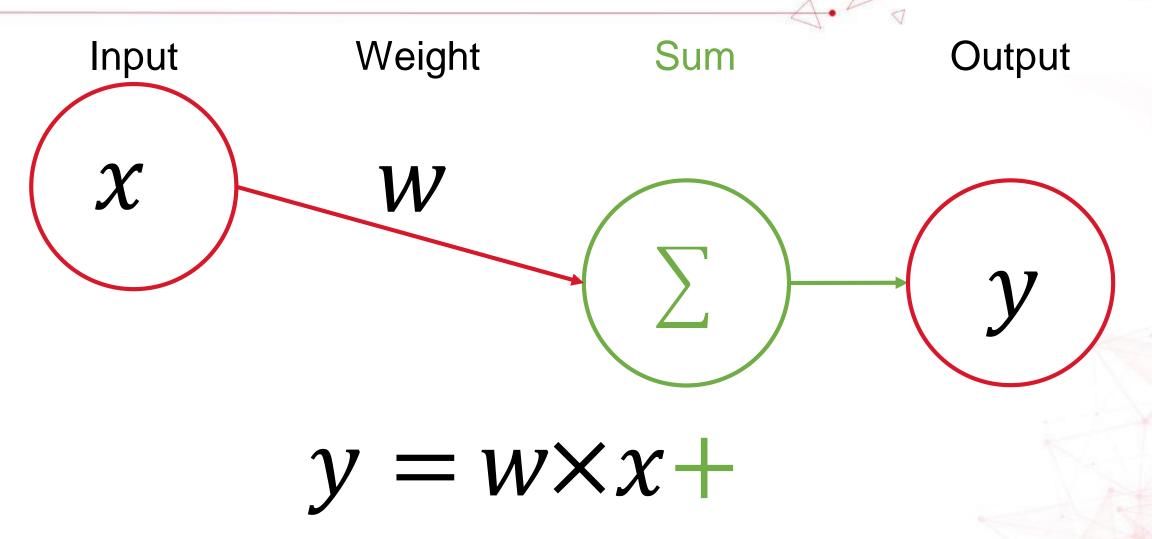


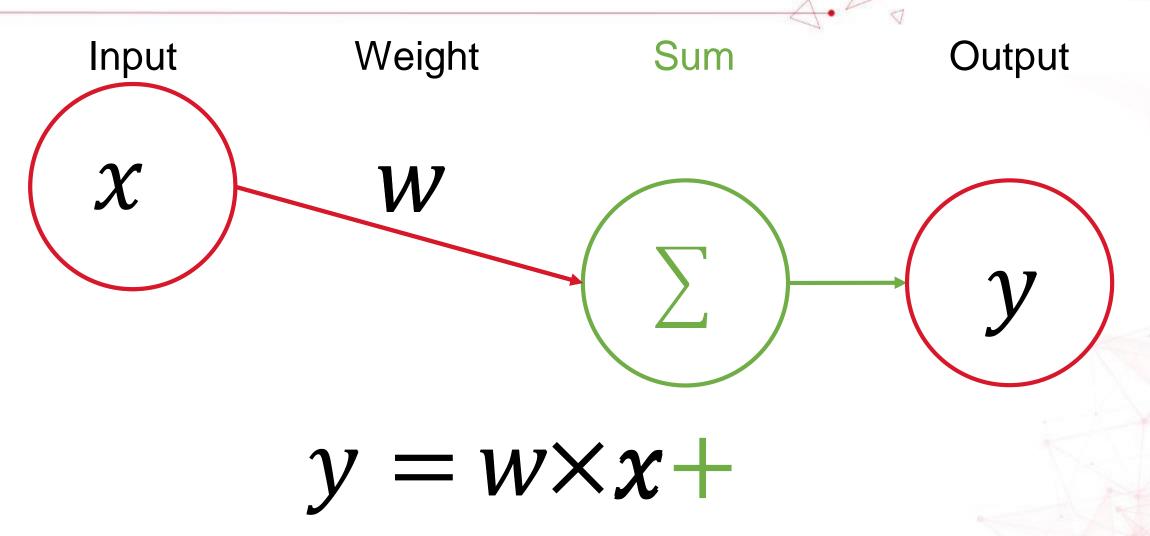
### 神經元 – Multiple Inputs





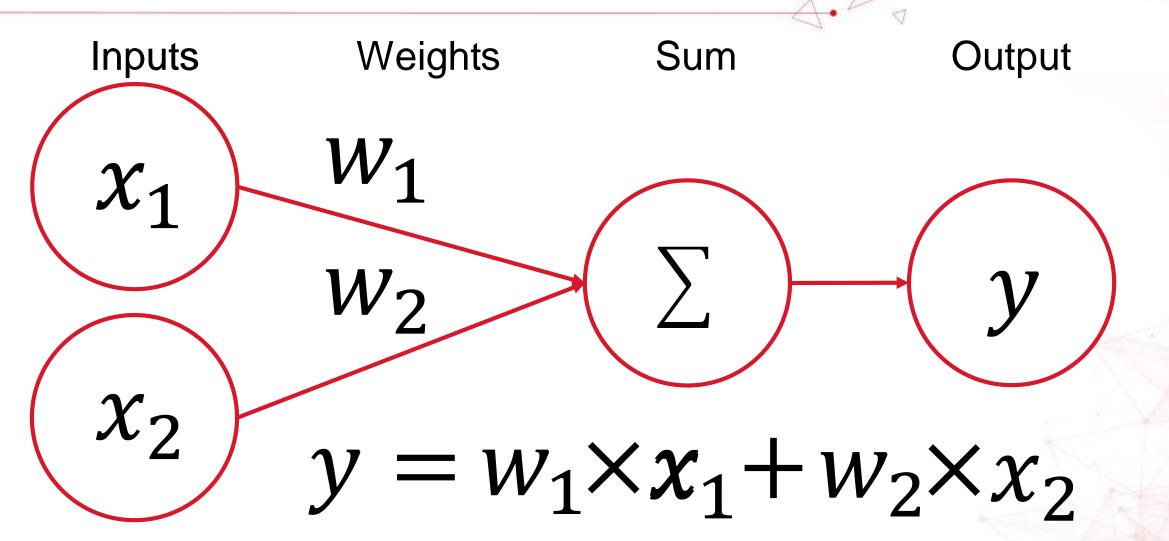
### 神經元 – Multiple Inputs



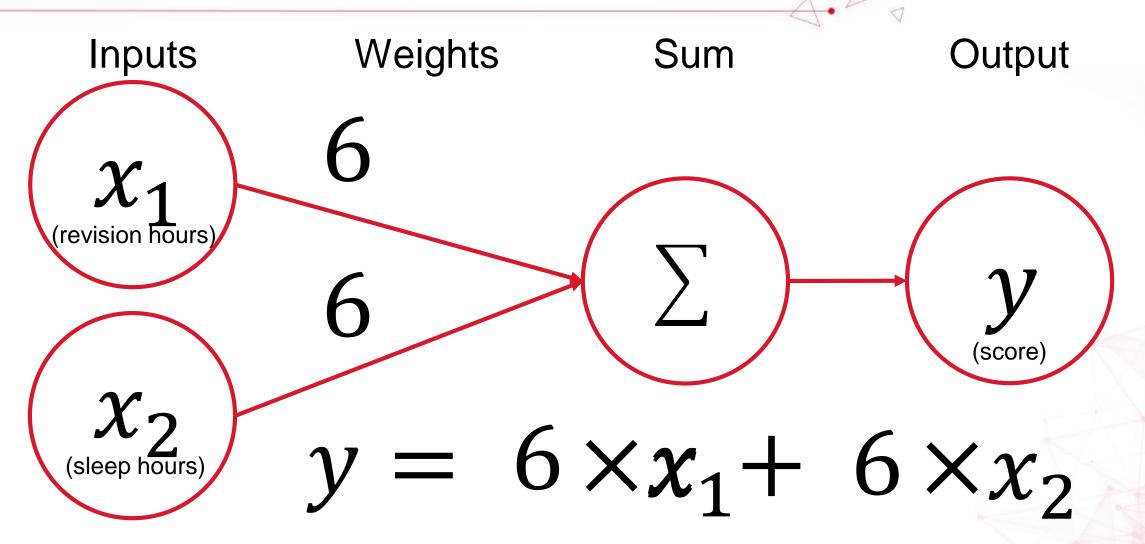


### 神經元 – Multiple Inputs

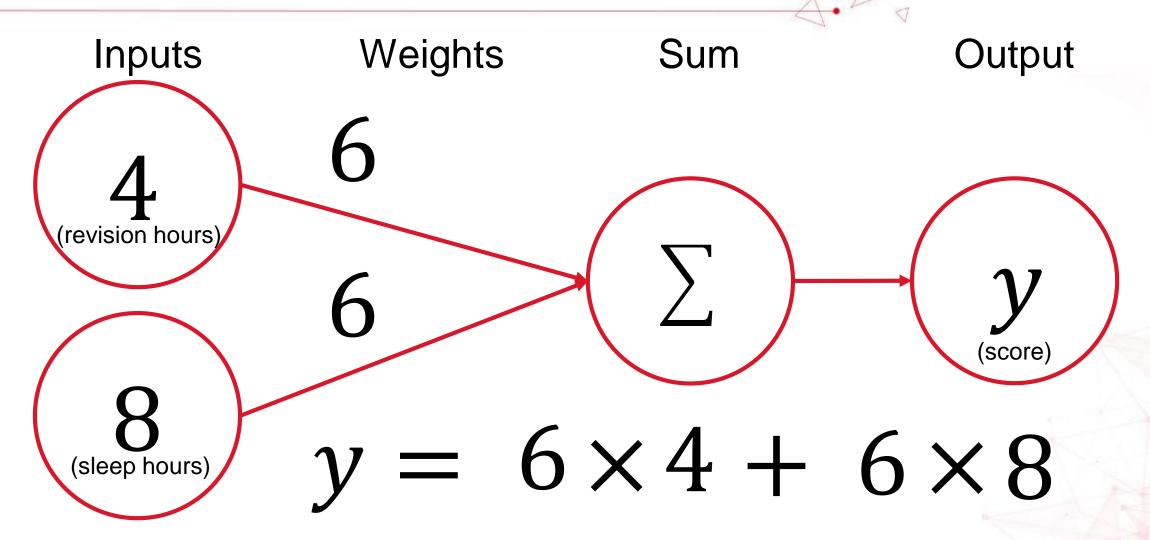




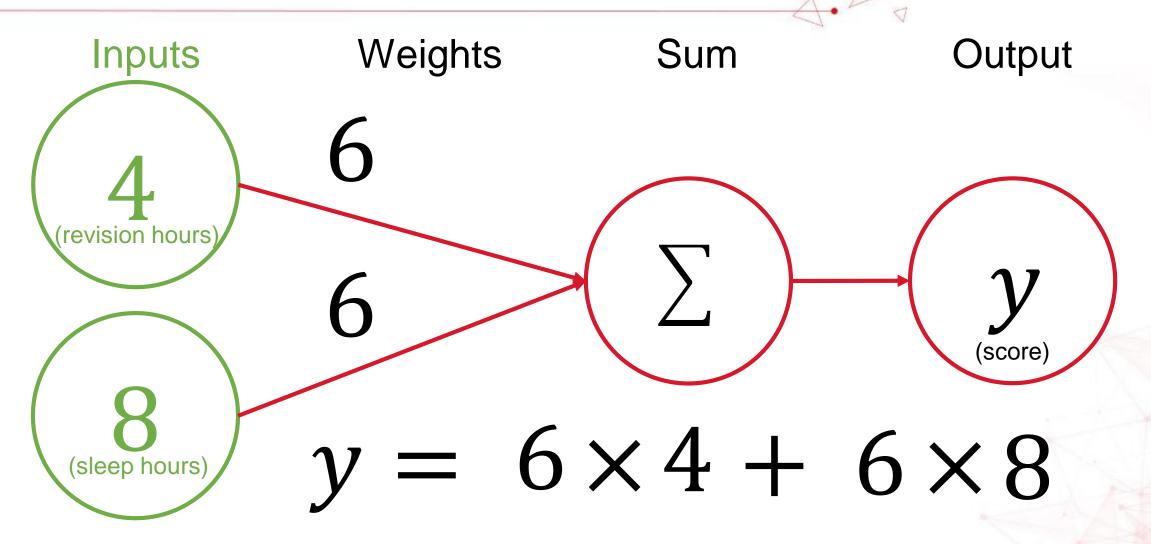




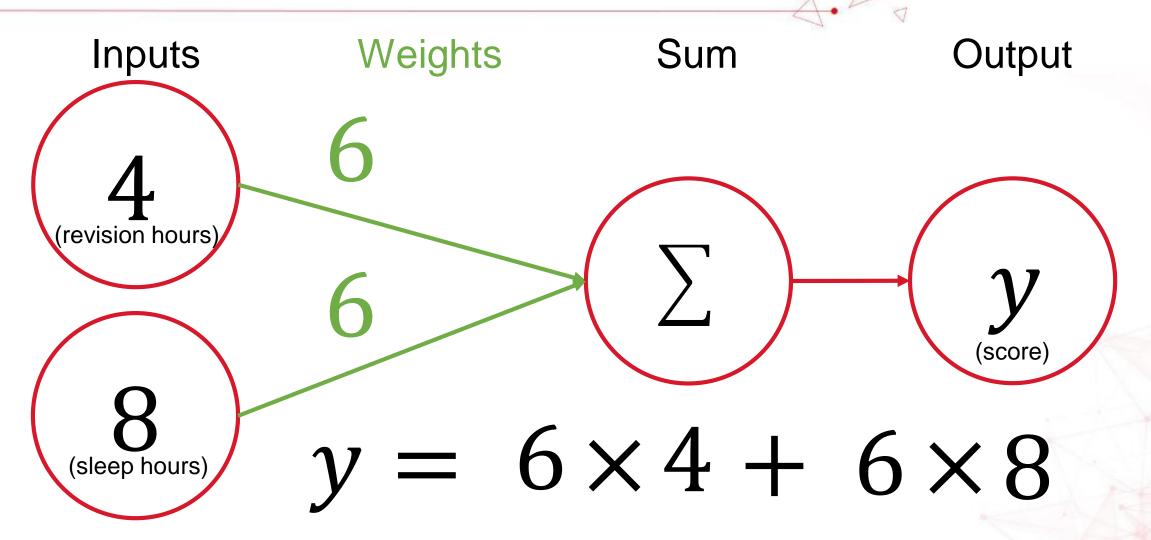




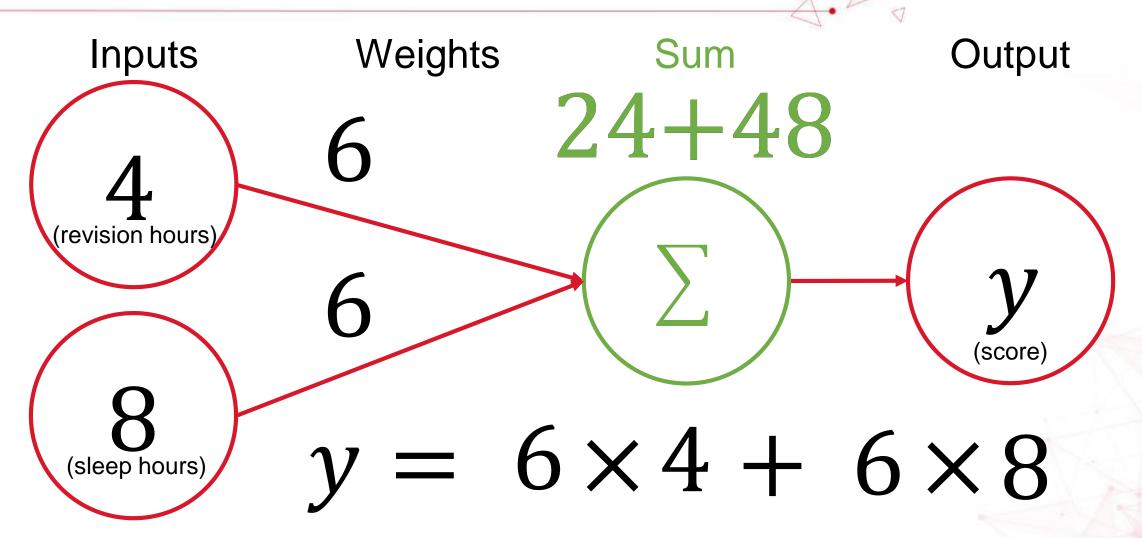




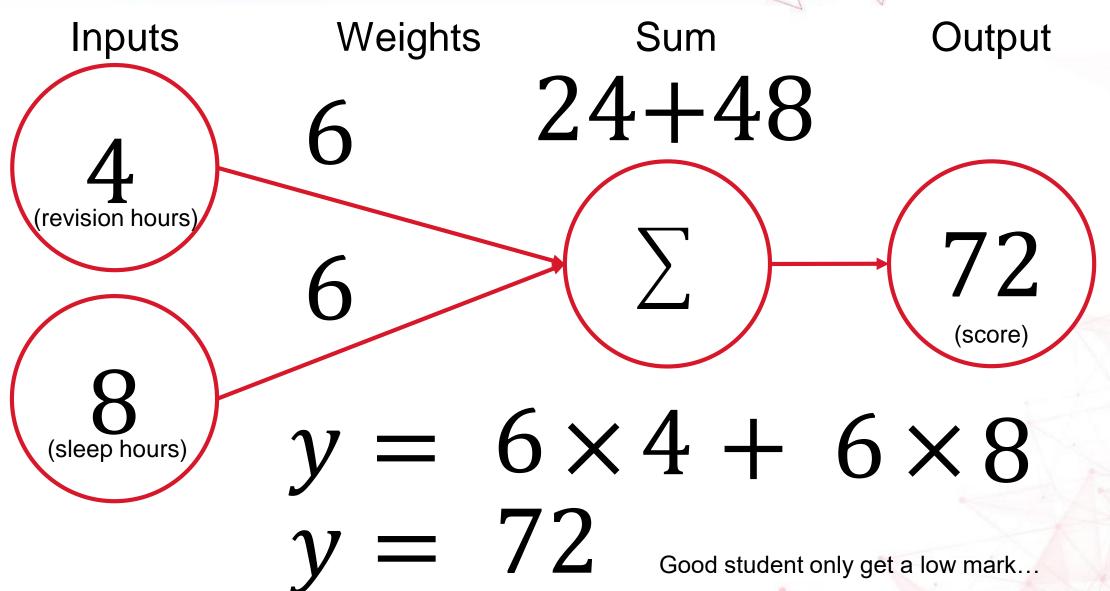




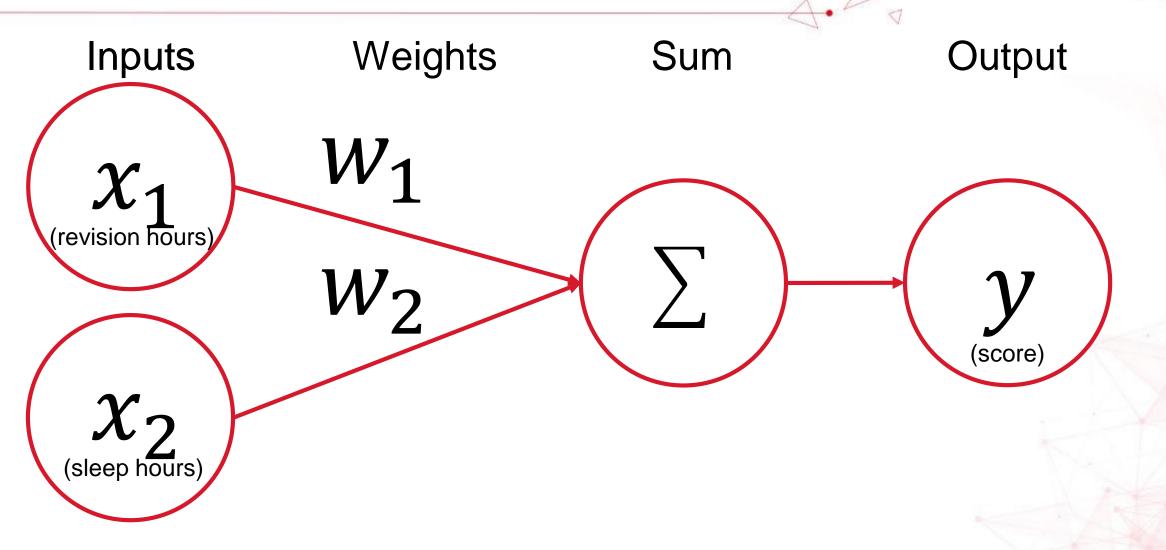






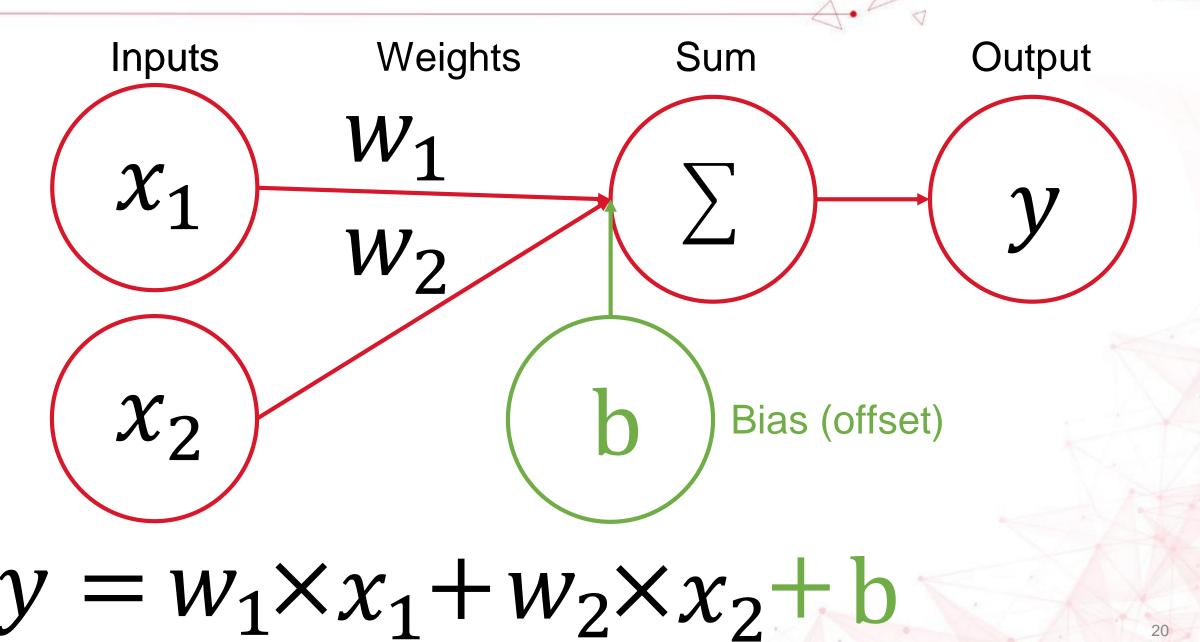




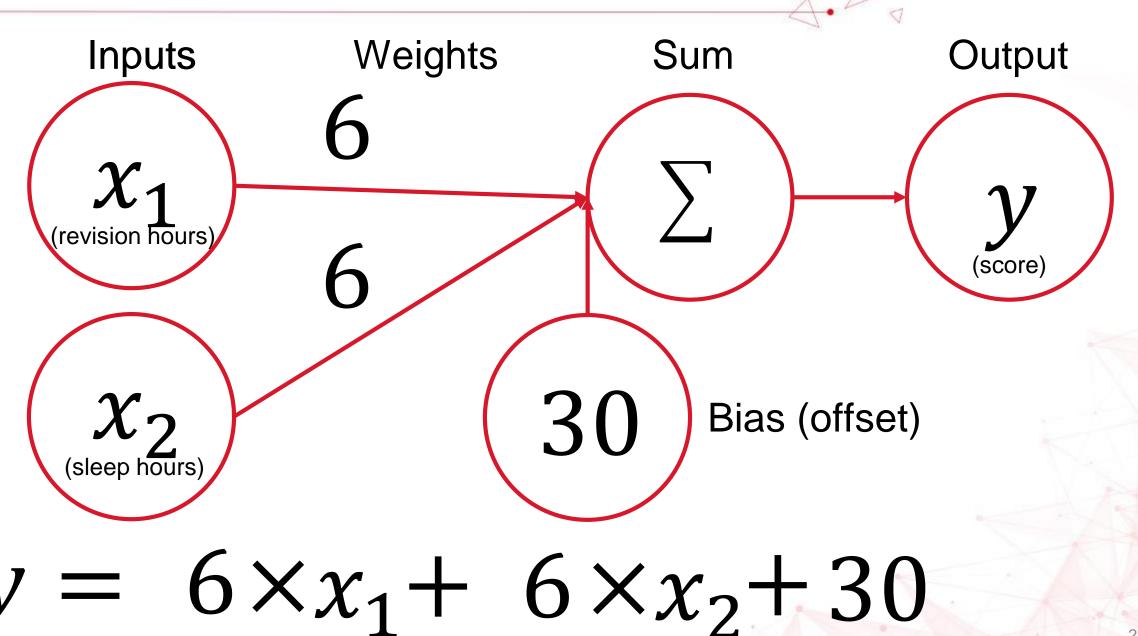


 $y = w_1 \times x_1 + w_2 \times x_2$ 

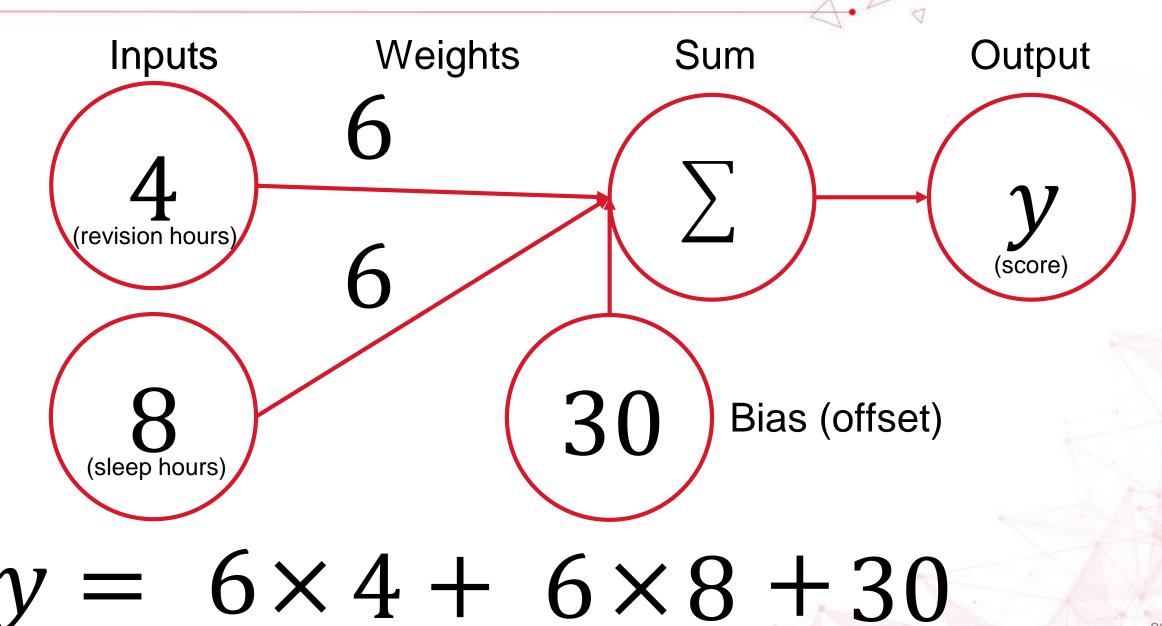




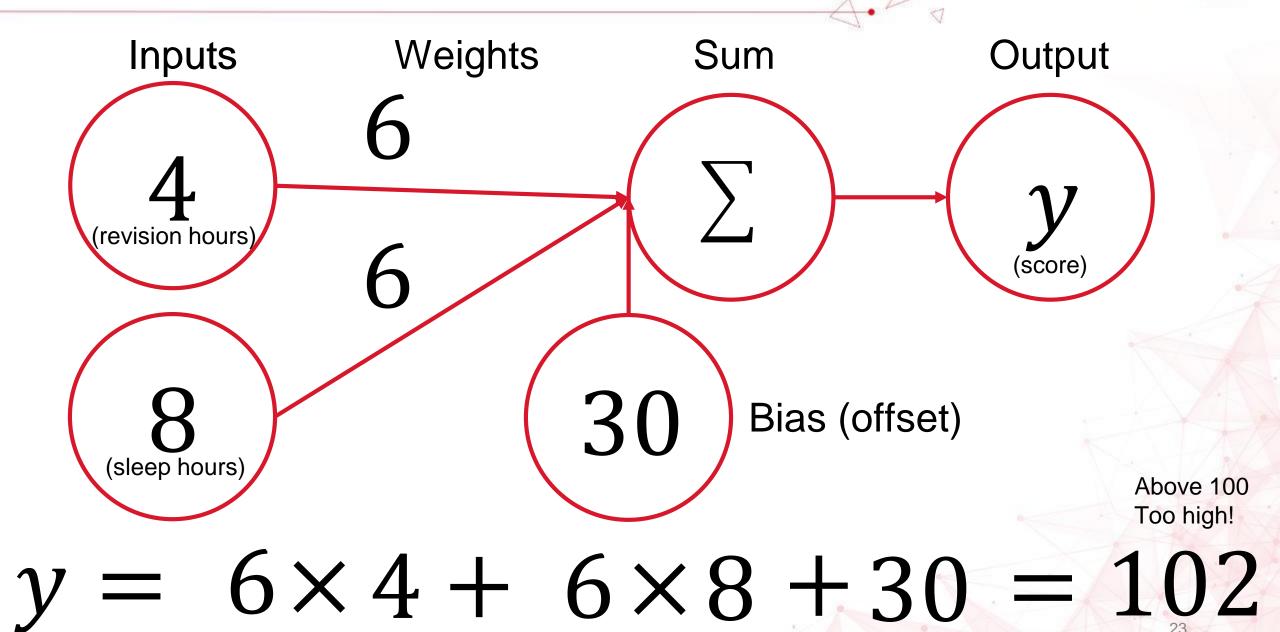




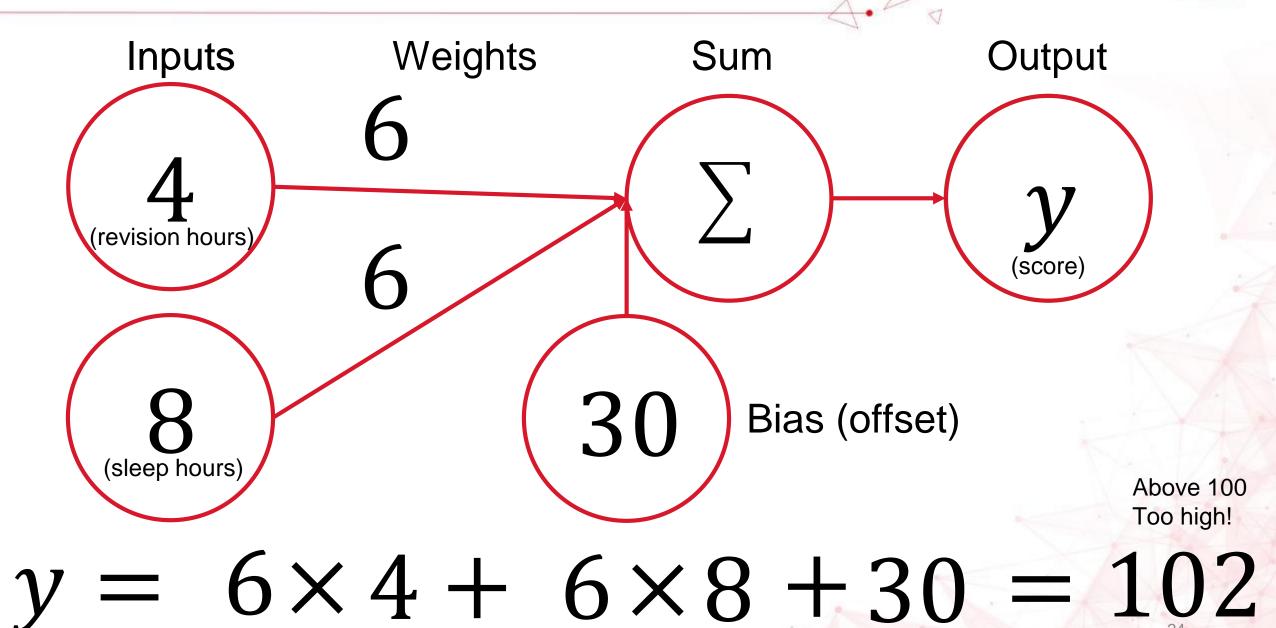






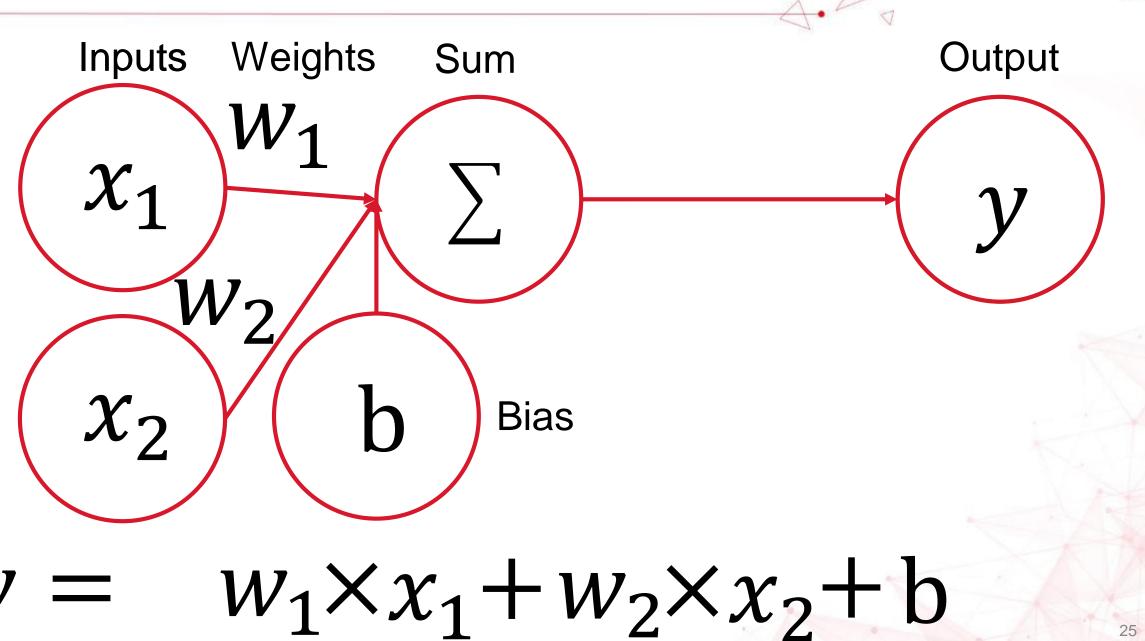






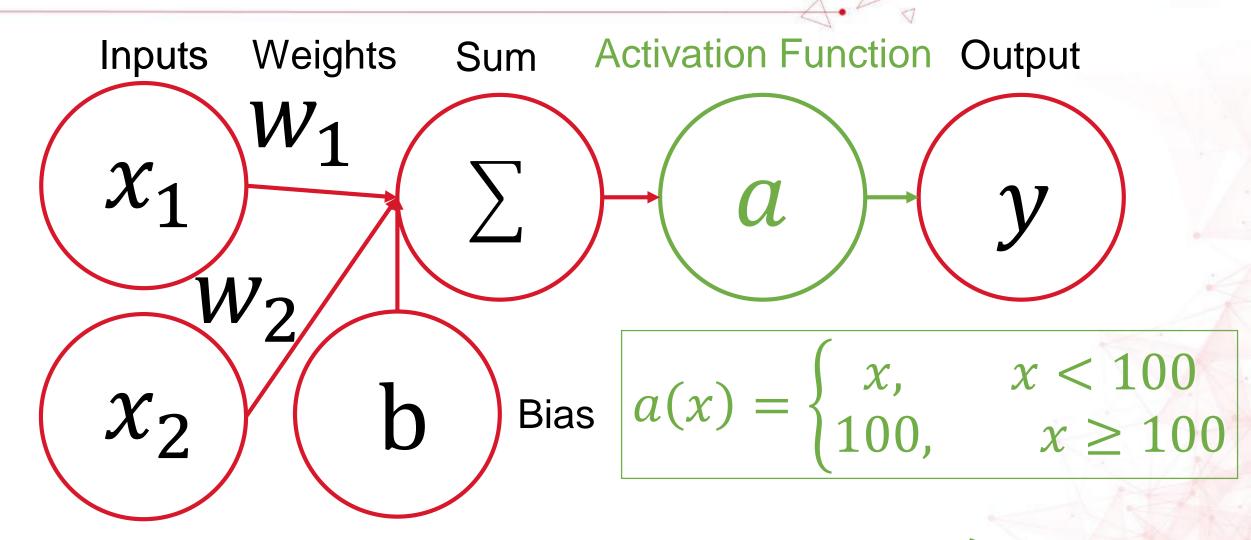
### 神經元 – Activation Function





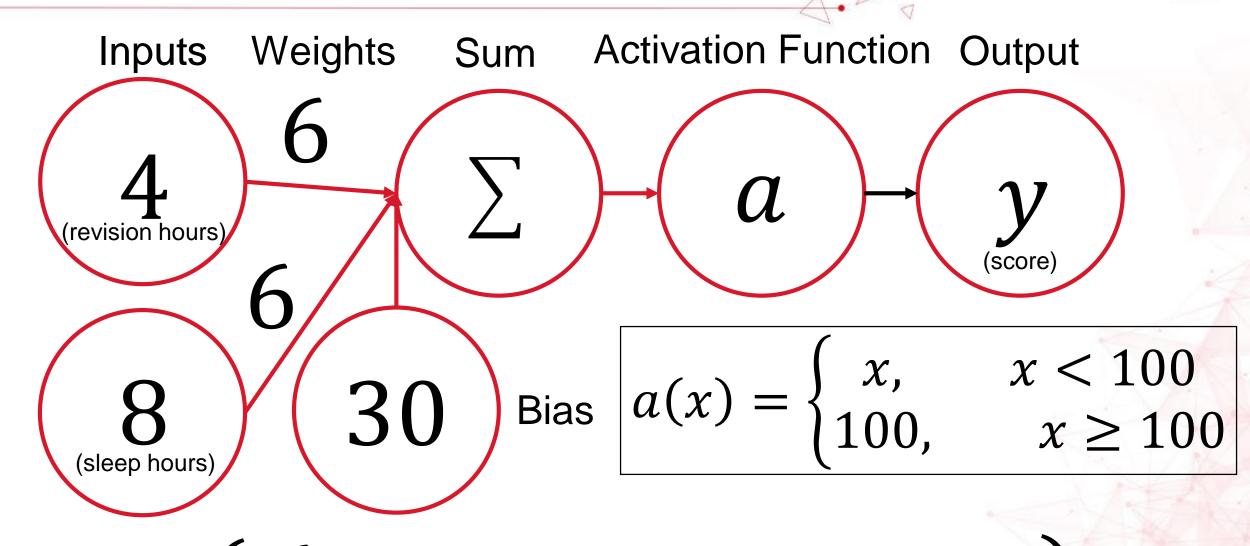
### 神經元 – Activation Function





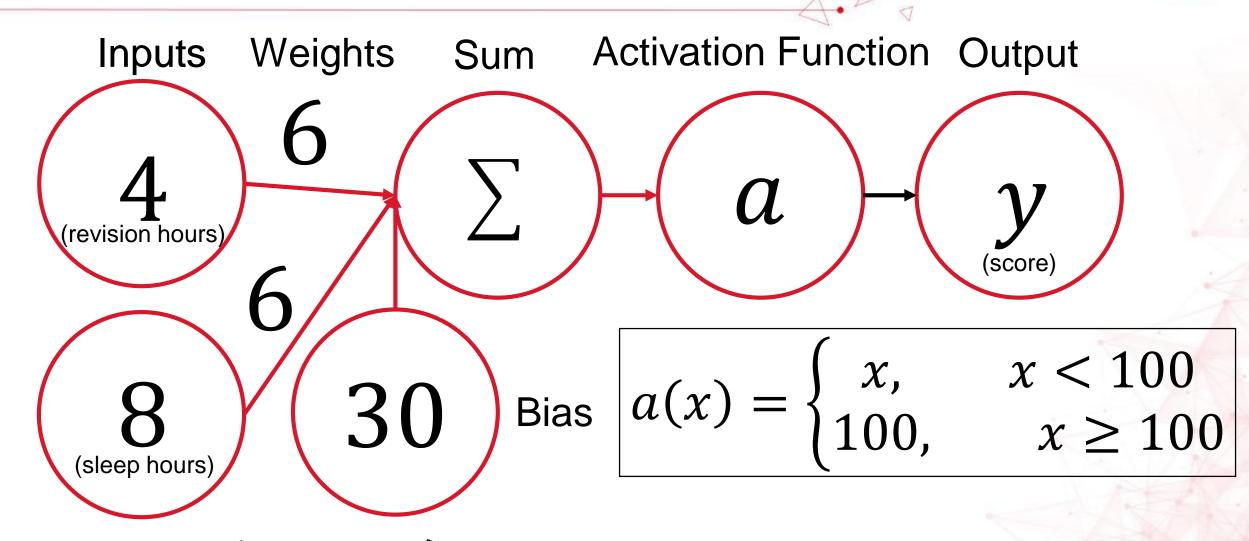
 $y = a(w_1 \times x_1 + w_2 \times x_2 + b)$ 





 $y = a(6 \times 4 + 6 \times 8 + 30)$ 

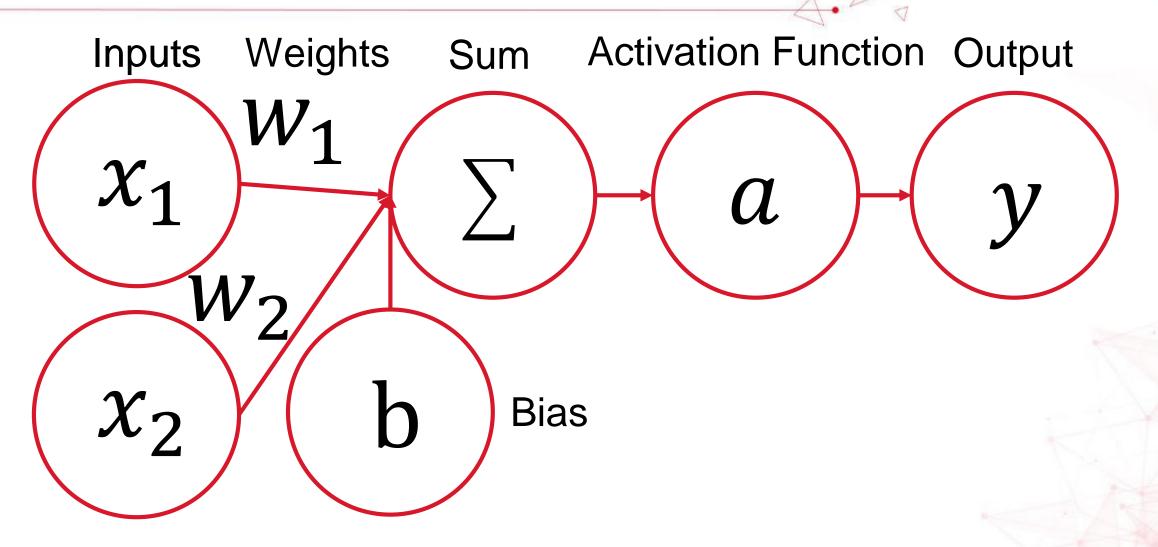




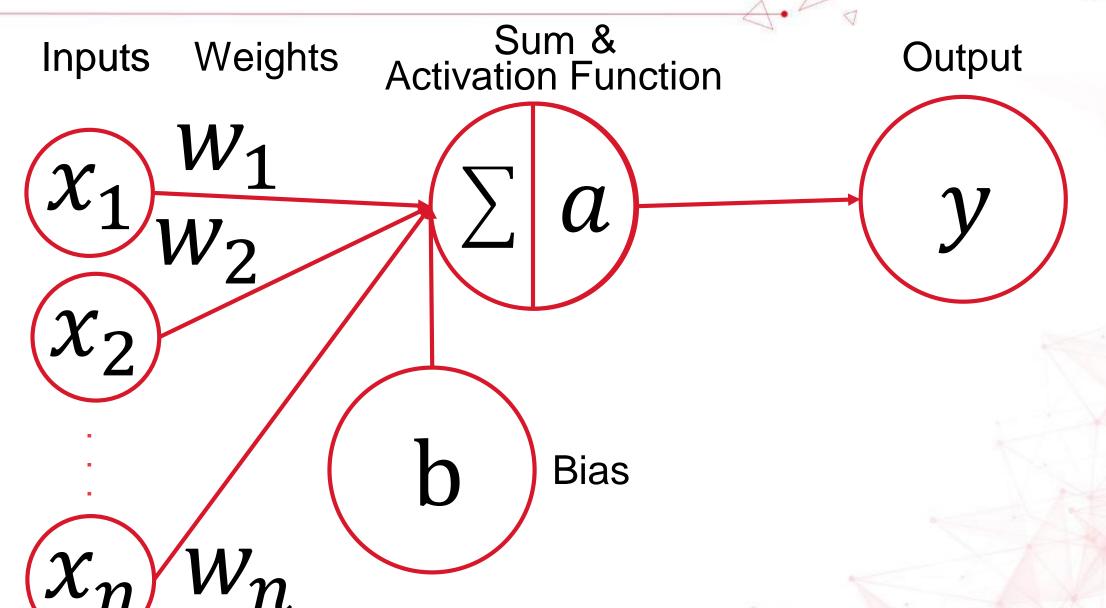
y = a(102) = 100

Yeah you got full marks!



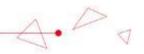


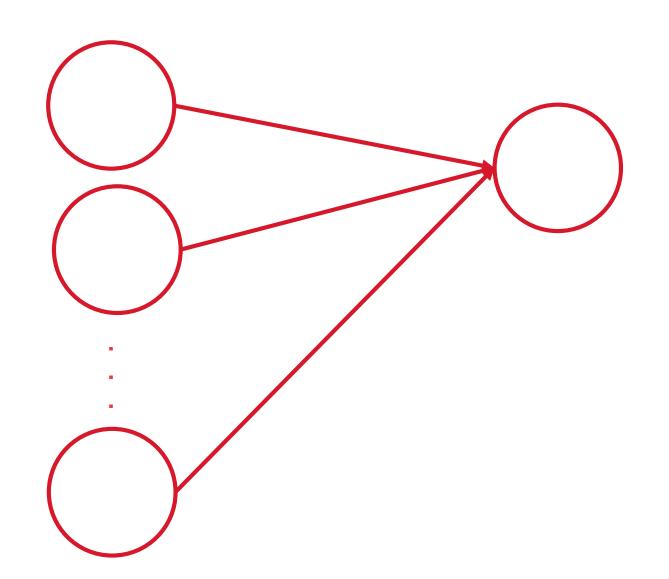




## From 神經元 To Neural Network

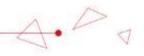


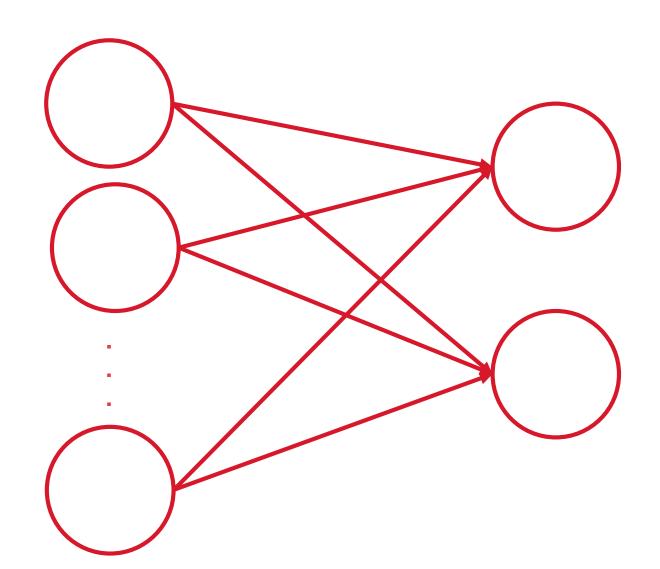




## From 神經元 To Neural Network

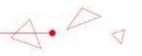


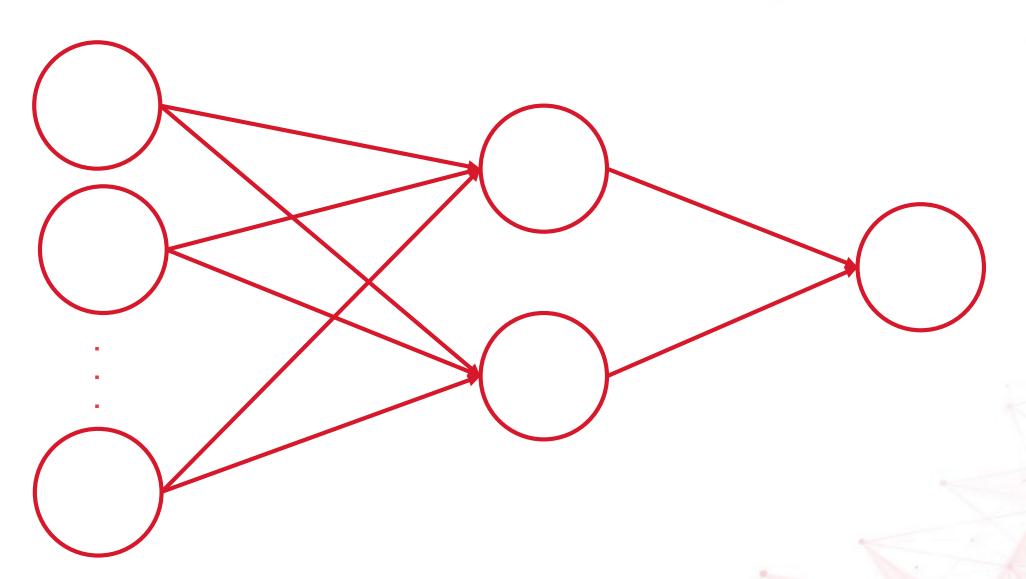




## From 神經元 To Neural Network

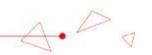


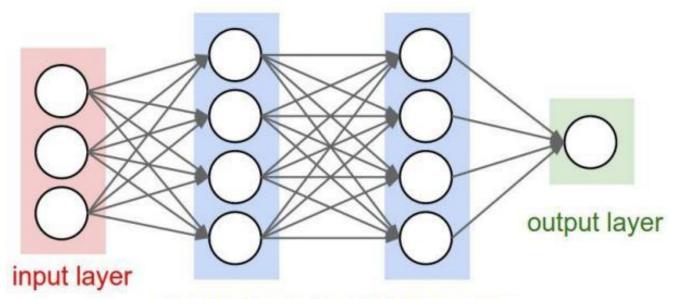


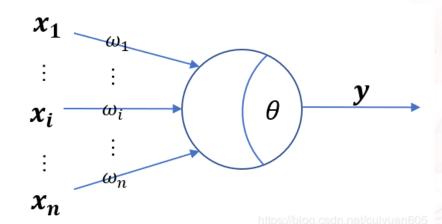


### 基本神經網路的結構



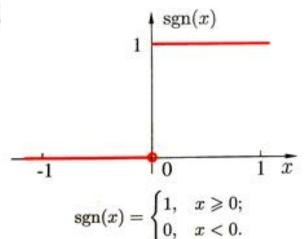


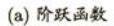


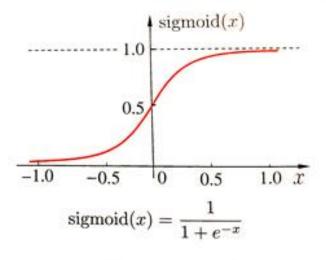


hidden layer 1 hidden layer

$$y = Act\left(\sum_{i=1}^{n} \omega_{i} x_{i} + \theta\right)$$

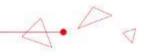


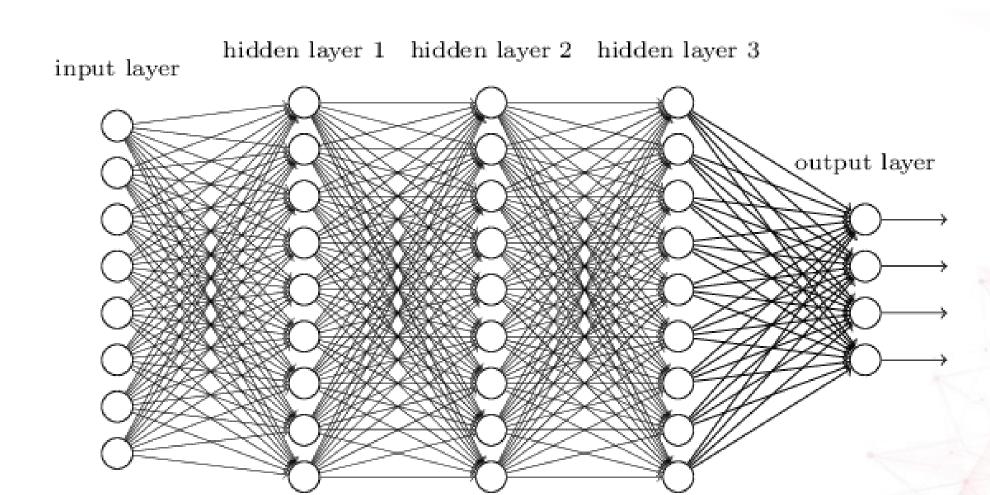




https:///(b) Sigmoid 函数 jyuan605







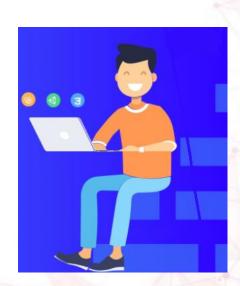
# SenseStudy實驗



# SenseStudy課程平臺"人工智能入門(下)"

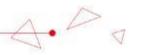
### 實驗7-2深度學習體驗:深度學習-人工神經網路

```
x train, y train, I train = load dataset("data/train.npz")
fig() + scatter(x train, y train, c=l train)
model = DeepLearning([2,4,4,1])
fig() + structure(model)
f_train = merge_features([x_train, y_train])
model.demo train('business')
model.show learning curve()
x test, y test, I test = load dataset("data/test.npz")
f test = merge features([x test, y test])
pred = model.predict(f test)
print(accuracy(pred > \overline{0.5}, I test))
```



# SenseStudy實驗





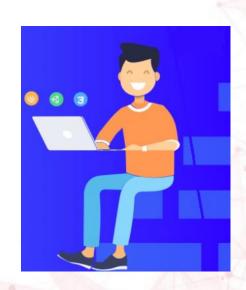
## SenseStudy課程平臺"人工智能入門(下)"

## 實驗7-3神經網路回歸-基於身高預測體重

train x, train y = load('hw.train')

```
test_x, test_y = load('hw.test')
fig() + scatter(train_x, train_y)

net=MLP([1,4,4,1])
fig()+structure(net)
net.train(train_x,train_y)
pred=net.predict(data=test_x)
fig()+scatter(test_x, test_y)+scatter(test_x, pred)
error=net.compute_error(pred,test_y)
print("test error = %f" % error)
```

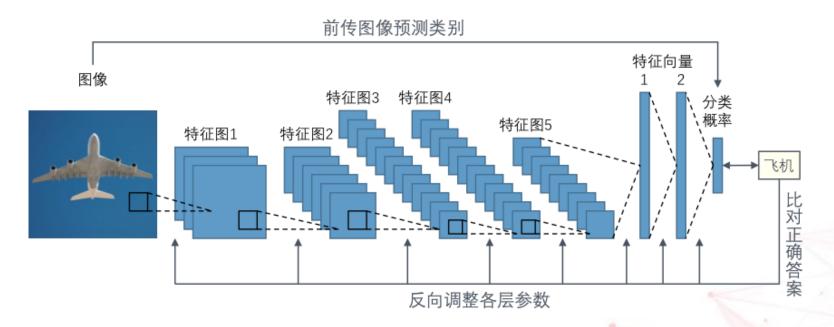


### 人工神經網路的訓練



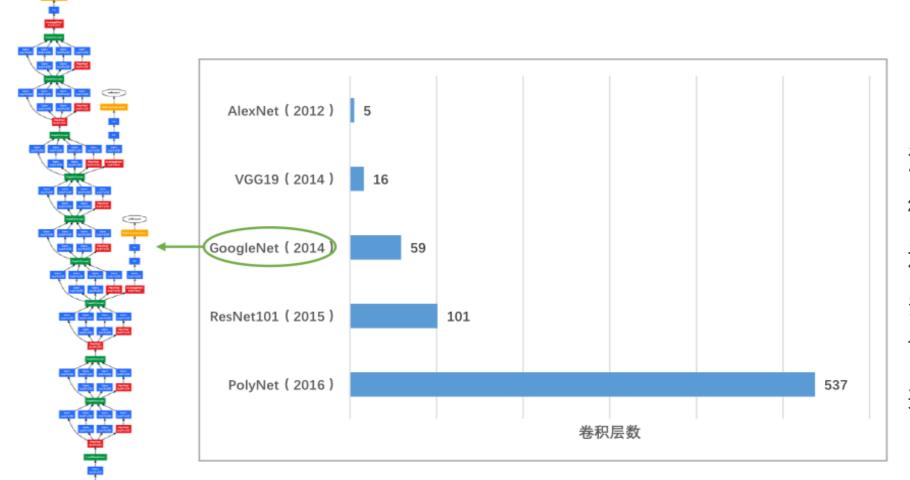


- 深度神經網路的訓練本質上就是尋找最佳參數的過程,參數包括:卷積層中所有卷積核的元素值、 全連接層中所有內積運算的係數等。
- 僅僅在非常簡單的深度網路AlexNet 中,需要學習的參數多達六千多萬個,其難度遠高於線性分類器的訓練。
- 科學家們提出了反向傳播演算法。直到今天為,仍是訓練神經網路最有效的手段之一。





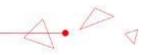




### Deeper & Better

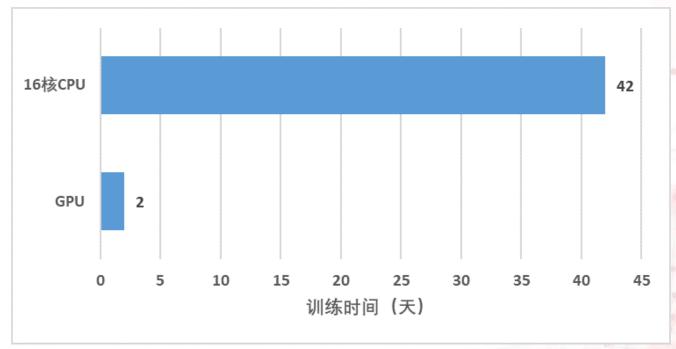
深度學習的"深"其實就表徵著神經網路的層數之多,更進一步就代表著模型參數之多。一個參數更多的模型,其可學習和調整的空間就更大,表達能力就更強。





### 以Alex Net為例,為完成ImageNet分類模型的訓練,CPU和GPU有顯著的區別

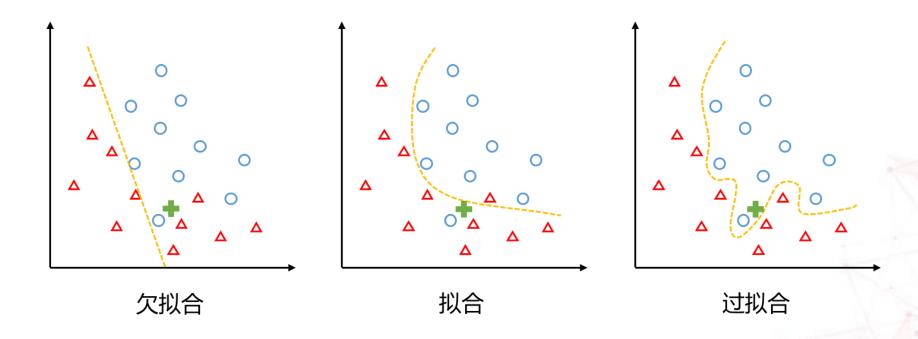






A. P.

欠擬合是指訓練資料集和測試資料集上表現都很差的現象。這種差主要來自 於網路本身的學習能力不足導致的。



過擬合是指在訓練資料集上表現很好,但是在測試資料集上,也就是新資料上面表現很差的現象。





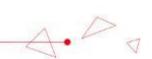
## **MNIST:** handwritten digits

MNIST: handwritten digits: 手寫數位圖集,最常用的完整的檢查。這個資料庫主要包含了60000張的訓練圖像和10000張的測試圖像,資料庫的裡的圖像都是28\*28大小的灰度圖像,每個圖元是一個八位元位元組(0~255)

MINST資料庫是由Yann提供的手寫數位資料庫檔







#### CIFAR10 / CIFAR100

CIFAR10:由10個類的60000個32x32彩色圖像組成,每個類有6000個圖像。有50000個訓練圖像和10000個測試圖像。

飞机

汽车

鸟

鹿

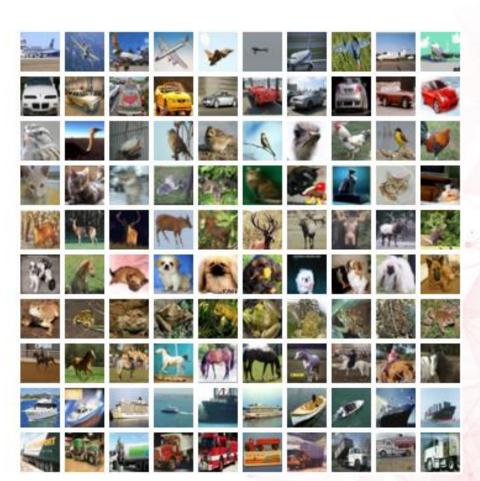
狗

马

船

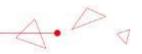
卡车

青蛙



他們由Alex Krizhevsky, Vinod Nair和Geoffrey Hinton收集。





### **ImageNet**

ImageNet是一個電腦視覺系統識別專案,是目前世界上圖像識別最大的資料庫。 是美國斯坦福的電腦科學家李飛飛類比人類的識別系統建立的。



ImageNet資料集有1500多萬幅圖片,涵蓋2萬多個類別;其中有超過百萬的圖片有明確的類別標注和圖像中物體位置的標注。