

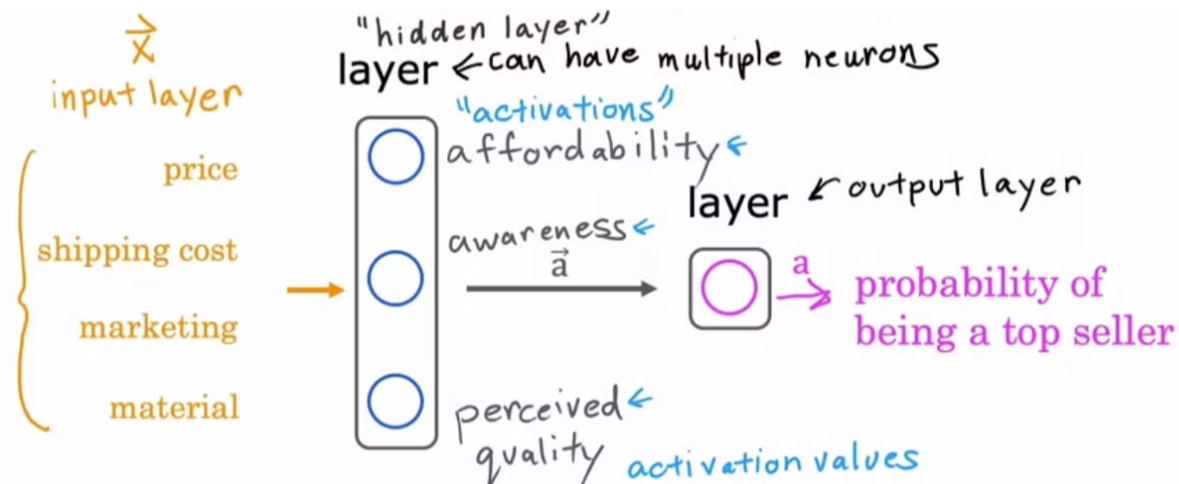
Advanced Learning Algorithms

Origins: algorithms that try to mimic the brain

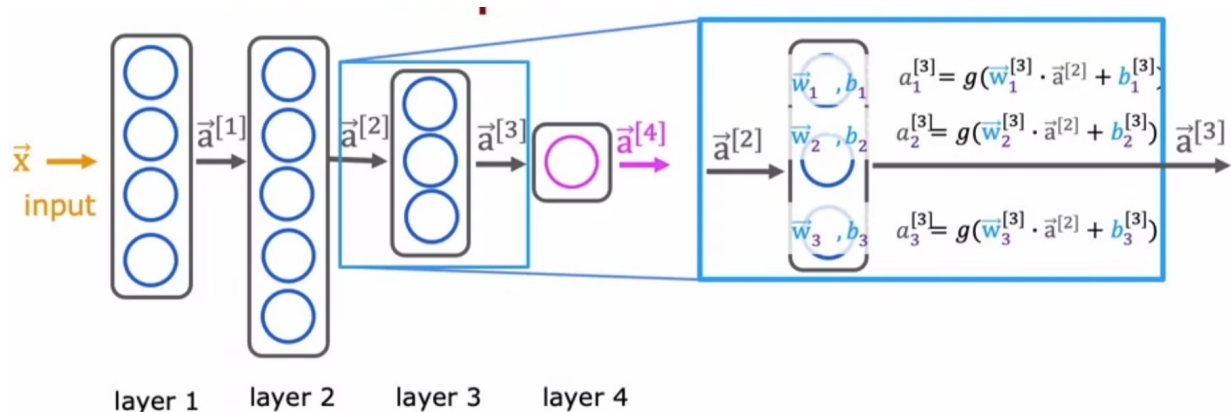
Ressurge from around 2005 under the term *deep learning*, as the amount of data rapidly increased. The rise of GPUs was also a major force.

Applications: speech recognition → computer vision → NLP ...

Example: predict if t-shirt will become a top-seller



It's like a version of logistic regression that learns its own features layer by layer. That replaces manual feature engineering.

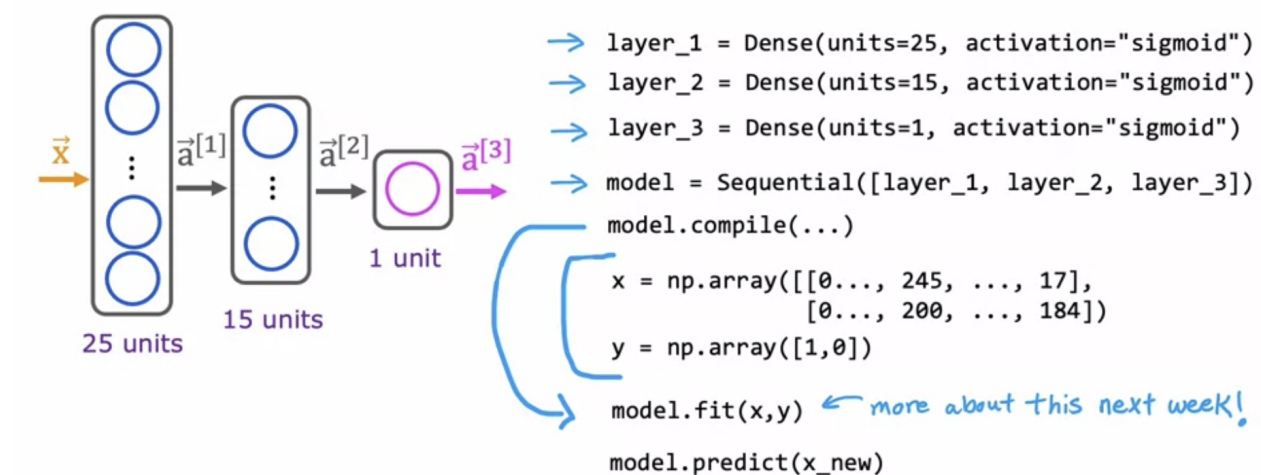
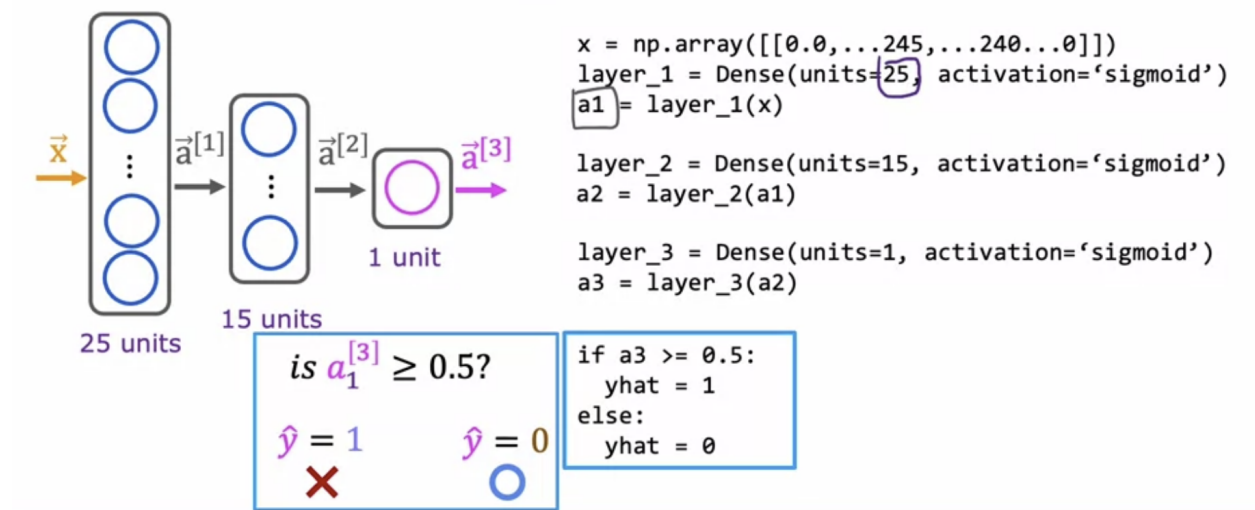


The activation function g could be the sigmoid function or other functions.

Forward propagation to make a prediction (inference).

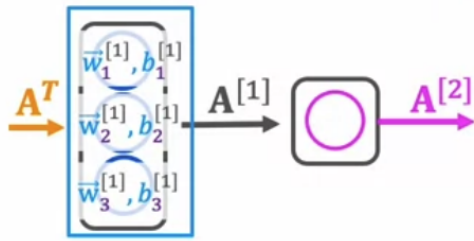
TensorFlow

Model for digit classification



Vectorized implementation makes neural networks run much faster!

Dense layer vectorized



$$A^T = \begin{bmatrix} 200 & 17 \end{bmatrix}$$

1×2

$$W = \begin{bmatrix} 1 & -3 & 5 \\ -2 & 4 & -6 \end{bmatrix}$$

2×3

$$B = \begin{bmatrix} -1 & 1 & 2 \end{bmatrix}$$

1×3

$$Z = A^T W + B$$

$$\begin{bmatrix} 165 & -531 & 900 \end{bmatrix}$$

$z_1^{[1]} \quad z_2^{[1]} \quad z_3^{[1]}$

$$A = g(Z)$$

$$\begin{bmatrix} 1 & 0 & 1 \end{bmatrix}$$

A

```
AT = np.array([[200, 17]])
W = np.array([[1, -3, 5],
              [-2, 4, -6]])
b = np.array([-1, 1, 2])
```

a_{in}

```
def dense(AT,W,b):
    z = np.matmul(AT,W) + b
    a_out = g(z)
    return a_out
```

$[[1,0,1]]$