

# Introduction to neural networks

## Deep Neural Networks

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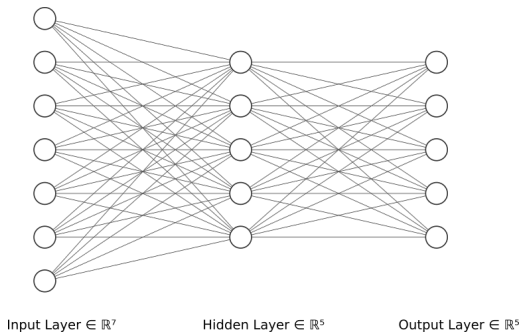


## 1 Deep fully connected networks

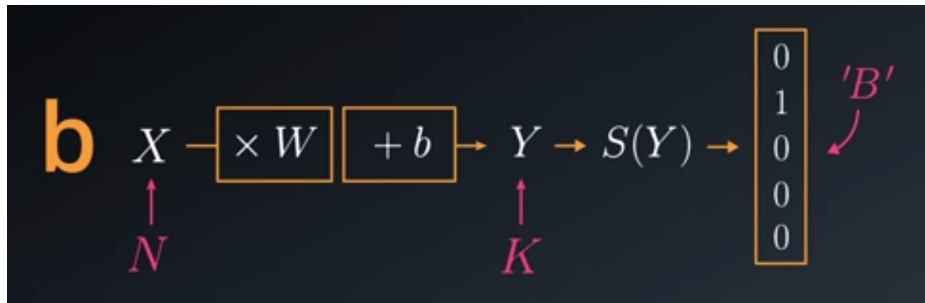
## 1 Deep fully connected networks

- Multinomial regression model

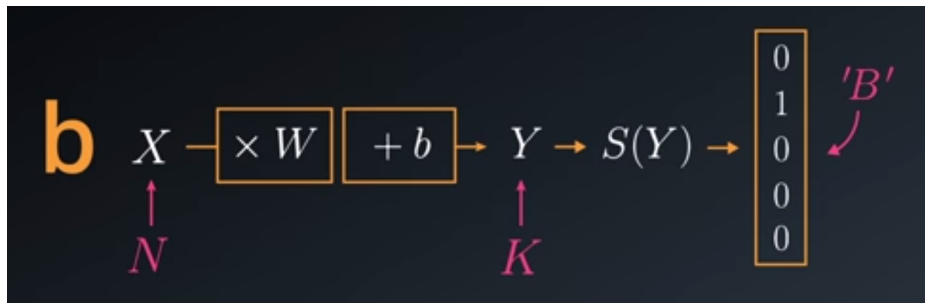
$$X \rightarrow XW + b \rightarrow f(h) \rightarrow Y \rightarrow S(Y) \rightarrow P \quad (1)$$



# Linear model complexity



# Linear model complexity



- Number of parameters in each layer:

$$f(n, k) = (n + 1)k \quad (2)$$

where  $n$  is the number of inputs and  $k$  the number of outputs.

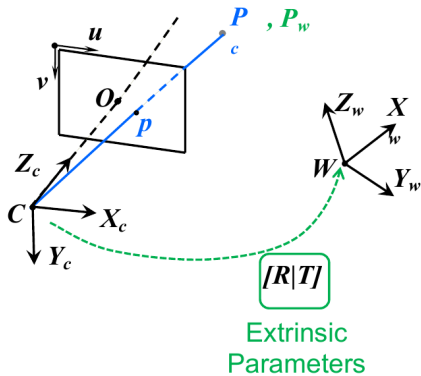
# Linear models advantages

- Efficiency
- Hardware available



# GPU

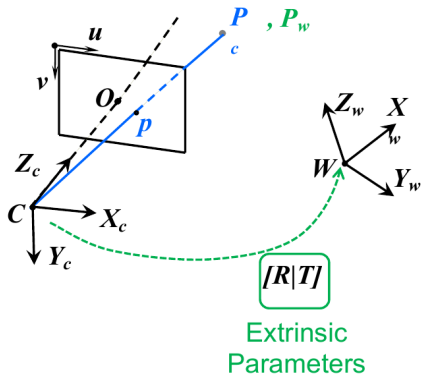
Why are so effective?





# GPU

Why are so effective?



Projection Matrix

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K [R|T] \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

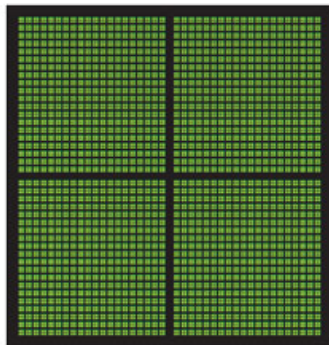
$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

# GPU

Why are so effective?



CPU  
MULTIPLE CORES



GPU  
THOUSANDS OF CORES



# Linear models disadvantages

- Linear models are restricted to linear combinations!

$$Y = x_1 + x_2 \quad (3)$$

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- How do we improve them?

# Linear models disadvantages

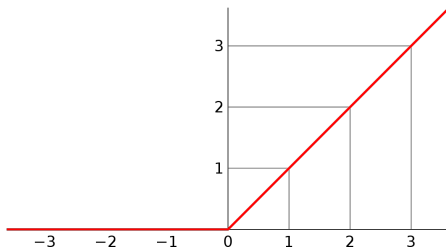
- Linear models are restricted to linear combinations!

$$Y = x_1 + x_2 \quad (3)$$

- How do we improve them?  
Activation functions!

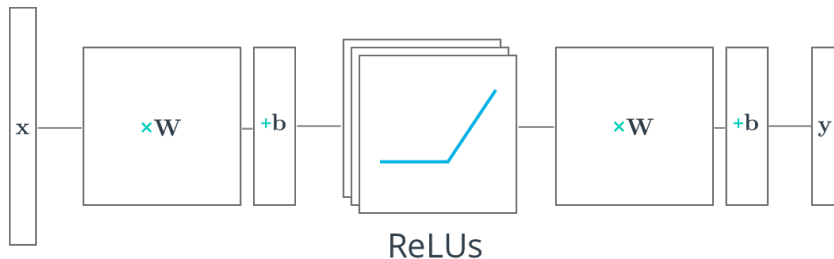
- Rectified linear unit

$$y = \max(0, x)$$



- Derivative?

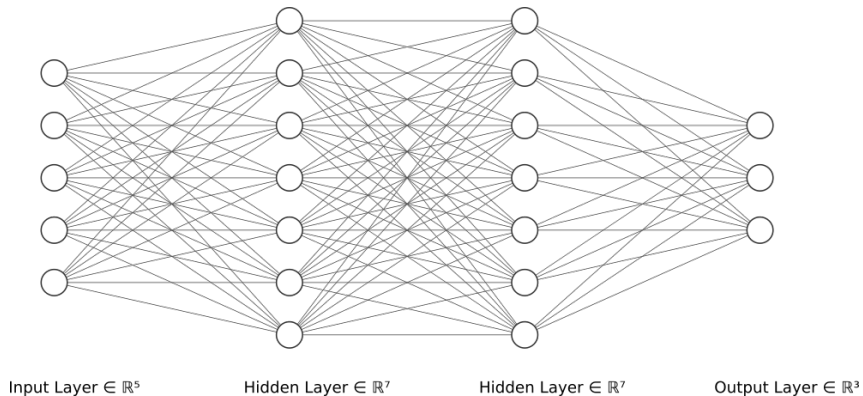
# Multilayer Neural Network



- Layer 1: the set of  $W$  and  $b$  applied to  $X$  and passed through ReLUs. The output is fed to the next one, but is not observable outside the network (hidden layer).
- Layer 2: the weights and biases applied to these intermediate outputs, followed by  $S(y)$ .



- Determine the amount of weights required in the following network:

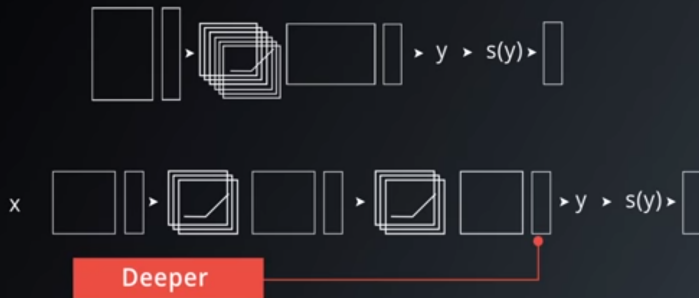


# Training Deep Neural Network (DNN)

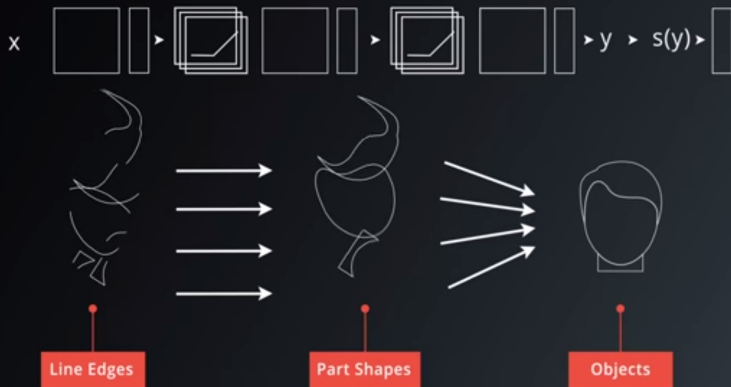
- The chain rule

$$\frac{\partial f(g(x))}{\partial x} = \frac{\partial f}{\partial g} \frac{\partial g}{\partial x} \quad (4)$$

- Training a Deep Learning Network



## Training a Deep Learning Network



# References

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