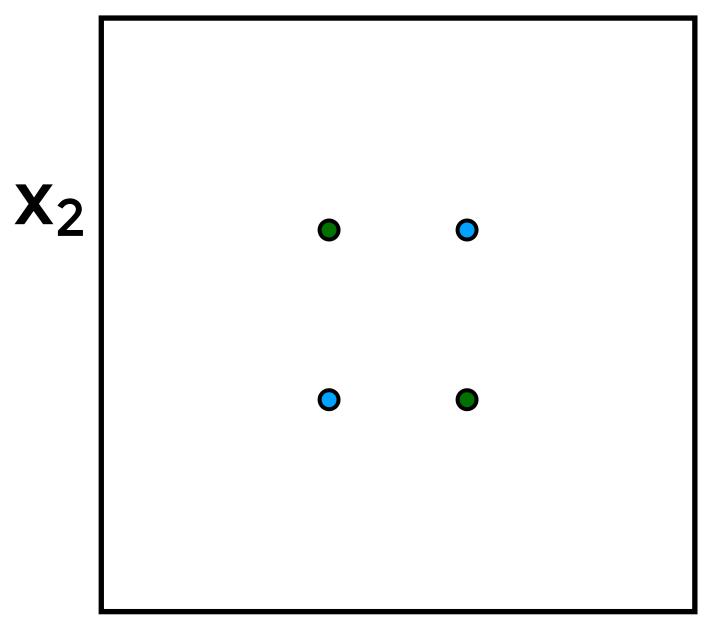
### Machine 60629A

Summary Neural Networks – Week #5

### Machine Learning I

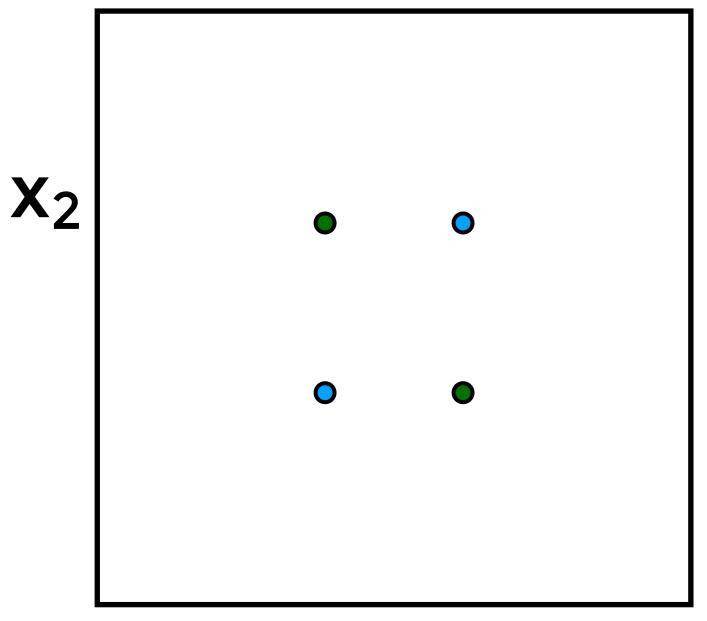
# What if data is not linearly separable?

Exclusive OR (XOR)



# What if data is not linearly separable?

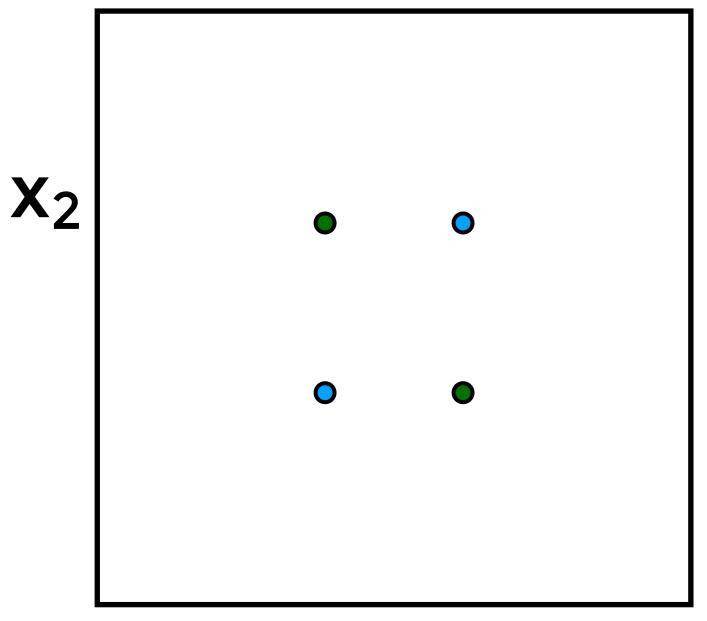
Exclusive OR (XOR)



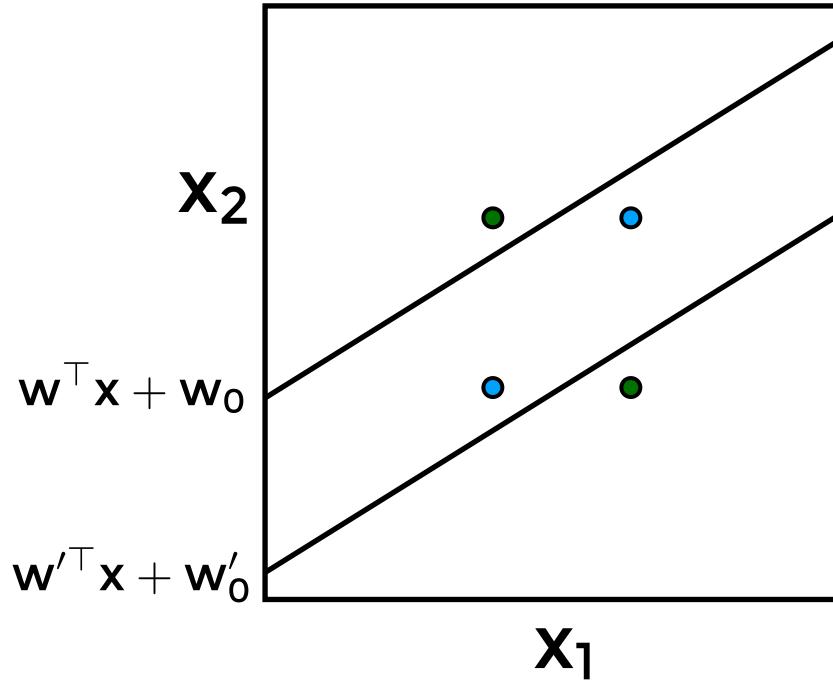
Use the joint decision of several linear classifier?

# What if data is not linearly separable?

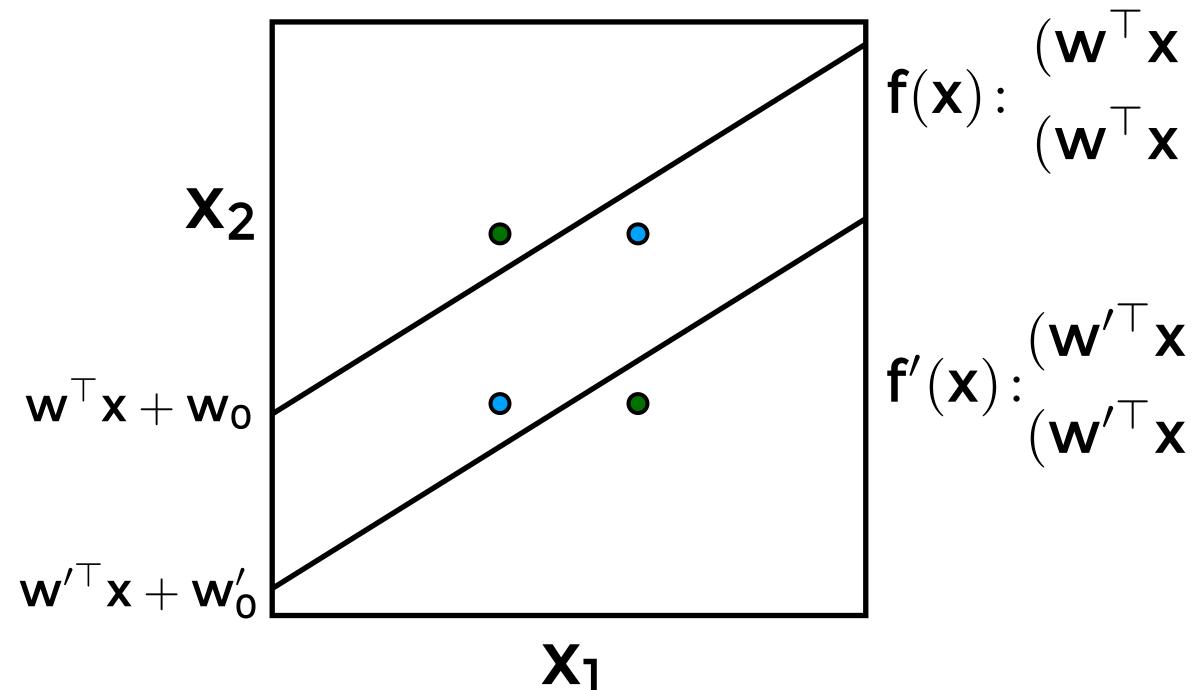
Exclusive OR (XOR)



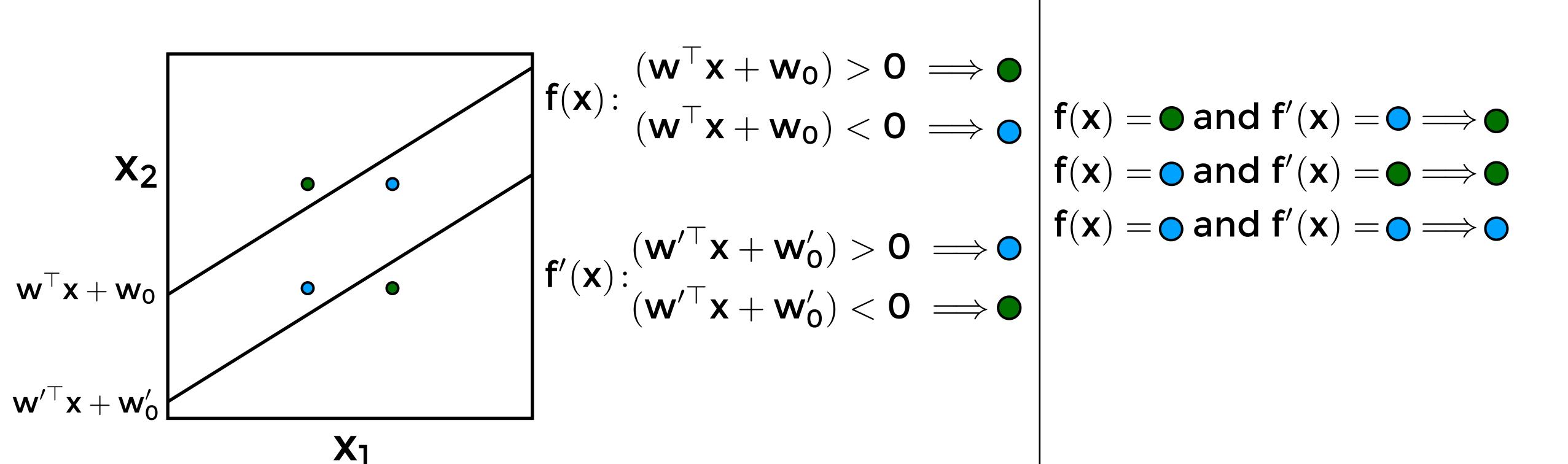
Use the joint decision of several linear classifier?



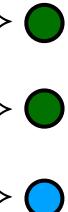


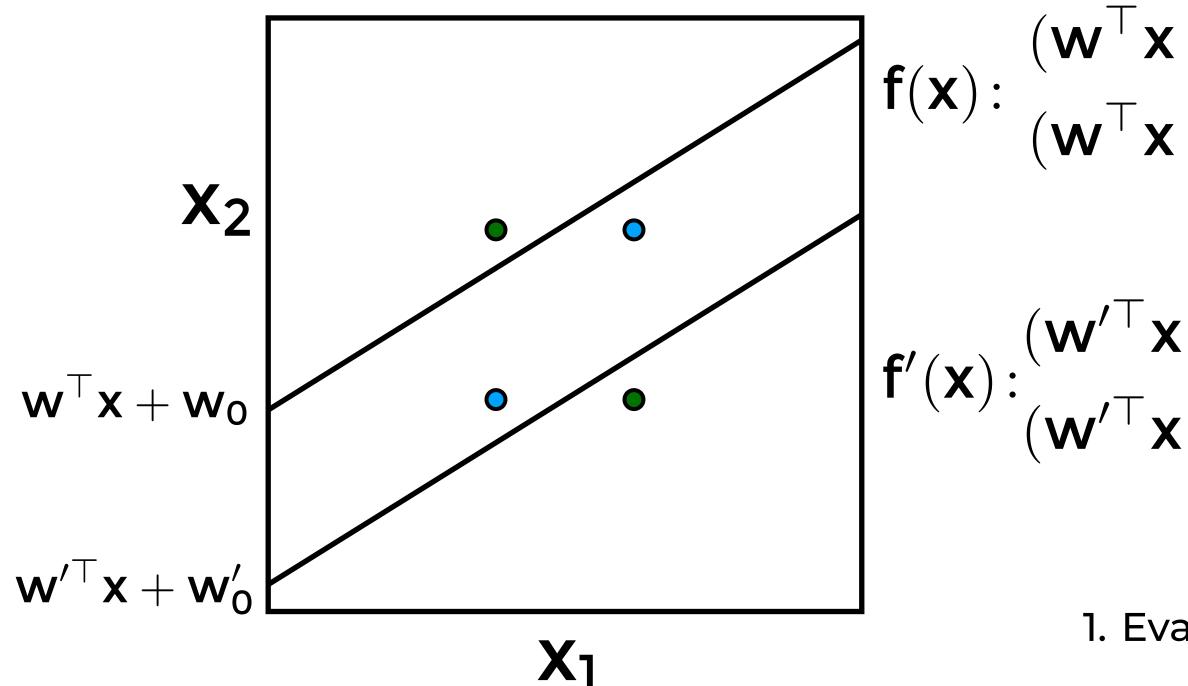


- $\begin{cases} f(\mathbf{x}): & (\mathbf{w}^{\top}\mathbf{x} + \mathbf{w}_{\mathbf{0}}) > \mathbf{0} \implies \mathbf{0} \\ (\mathbf{w}^{\top}\mathbf{x} + \mathbf{w}_{\mathbf{0}}) < \mathbf{0} \implies \mathbf{0} \end{cases}$
- $f'(\mathbf{x}): \begin{pmatrix} \mathbf{w}'^{\top}\mathbf{x} + \mathbf{w}'_{\mathbf{0}} \end{pmatrix} > \mathbf{0} \implies \mathbf{0} \\ (\mathbf{w}'^{\top}\mathbf{x} + \mathbf{w}'_{\mathbf{0}}) < \mathbf{0} \implies \mathbf{0} \end{pmatrix}$



 $\mathbf{f}'(\mathbf{x}): \begin{pmatrix} \mathbf{w}'^{\top}\mathbf{x} + \mathbf{w}'_{\mathbf{0}} \end{pmatrix} > \mathbf{0} \implies \mathbf{0} \\ (\mathbf{w}'^{\top}\mathbf{x} + \mathbf{w}'_{\mathbf{0}}) < \mathbf{0} \implies \mathbf{0} \end{pmatrix}$ 



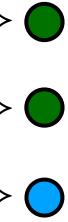


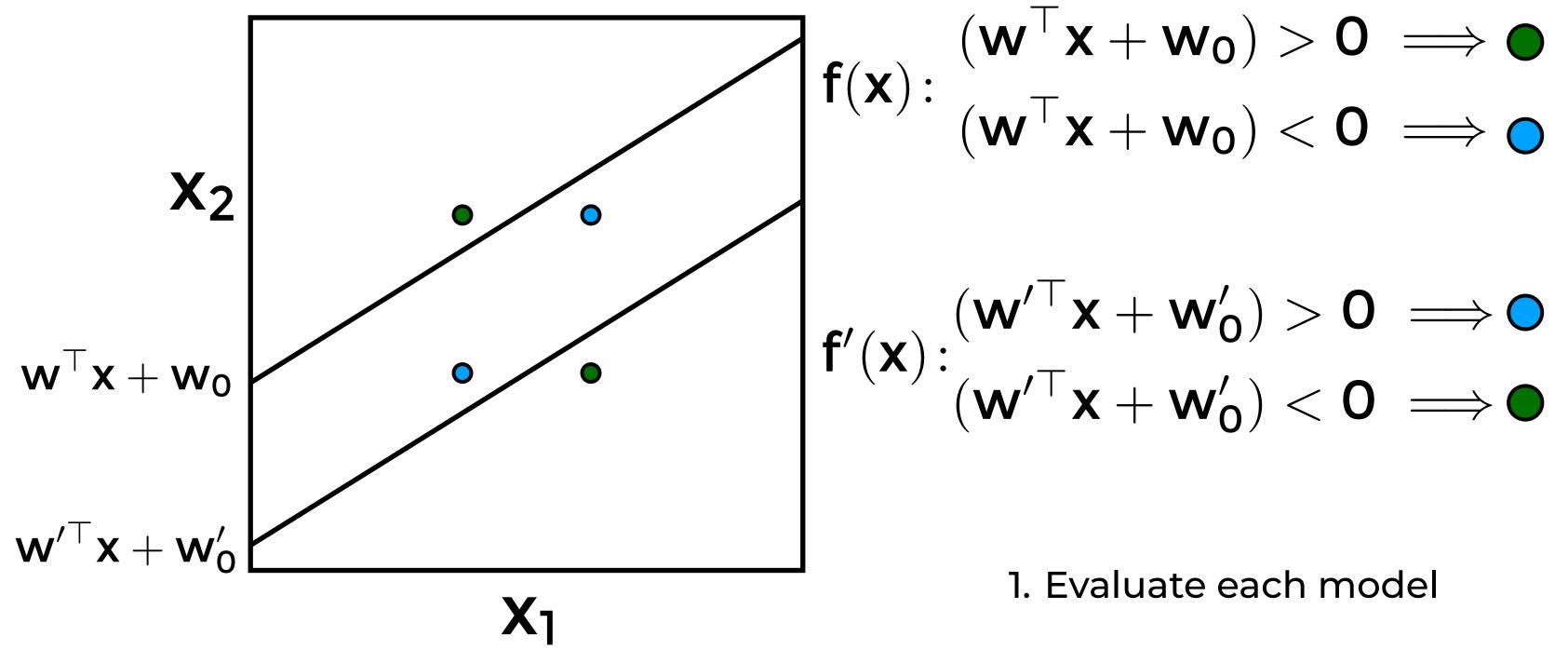
 $\begin{aligned} &| \mathbf{f}'(\mathbf{x}) : \\ & (\mathbf{w}'^{\top} \mathbf{x} + \mathbf{w}'_{\mathbf{0}}) > \mathbf{0} \implies \mathbf{0} \\ & (\mathbf{w}'^{\top} \mathbf{x} + \mathbf{w}'_{\mathbf{0}}) < \mathbf{0} \implies \mathbf{0} \end{aligned}$ 

1. Evaluate each model

 $\begin{aligned} f(\mathbf{x}) \colon & (\mathbf{w}^{\top}\mathbf{x} + \mathbf{w}_0) > \mathbf{0} \implies \mathbf{0} \\ (\mathbf{w}^{\top}\mathbf{x} + \mathbf{w}_0) < \mathbf{0} \implies \mathbf{0} \\ f(\mathbf{x}) = \mathbf{0} \text{ and } f'(\mathbf{x}) = \mathbf{0} \implies \mathbf{0} \\ f(\mathbf{x}) = \mathbf{0} \text{ and } f'(\mathbf{x}) = \mathbf{0} \implies \mathbf{0} \\ f(\mathbf{x}) = \mathbf{0} \text{ and } f'(\mathbf{x}) = \mathbf{0} \implies \mathbf{0} \end{aligned}$ 

2. Combine the output of models



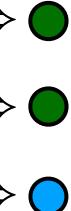


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1. Evaluate each model

$$f(\mathbf{x}) = \mathbf{O} \text{ and } f'(\mathbf{x}) = \mathbf{O} \Longrightarrow$$
$$f(\mathbf{x}) = \mathbf{O} \text{ and } f'(\mathbf{x}) = \mathbf{O} \Longrightarrow$$
$$f(\mathbf{x}) = \mathbf{O} \text{ and } f'(\mathbf{x}) = \mathbf{O} \Longrightarrow$$

2. Combine the output of models

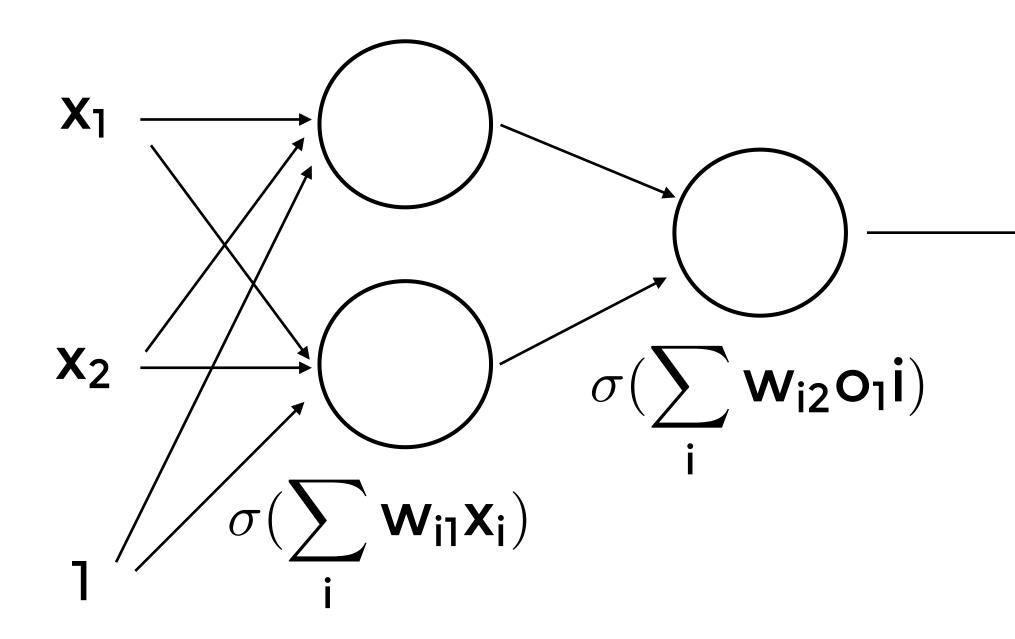




## Feed-forward neural network

Input Layer

Hidden Layer(s) Output Layer



- Each arrow denotes a connection
  - A signal associated with a weight
- Each node is the weighted sum of its input followed by a nonlinear activation
- Connections go left to right
  - No connections within a layer
  - No backward connections (recurrent)

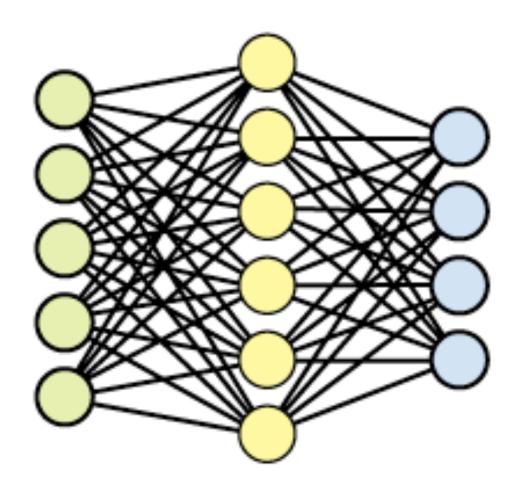
### Gradient descent

- No closed-form formula
- Repeat the following steps (for t=0,1,2,... until convergence):
  - 1. Calculate a gradient  $\nabla \mathbf{v}$
  - 2. Apply the update
- W<sub>ii</sub><sup>t-</sup>
- Stochastic gradient descent
  - One example at a time
- Batch gradient descent
  - All examples at a time

$$\mathbf{w}_{ij}^{\mathsf{t}}$$
  
 $^{+1} = \mathbf{w}_{ij}^{\mathsf{t}} - lpha 
abla \mathbf{w}_{ij}^{\mathsf{t}}$ 

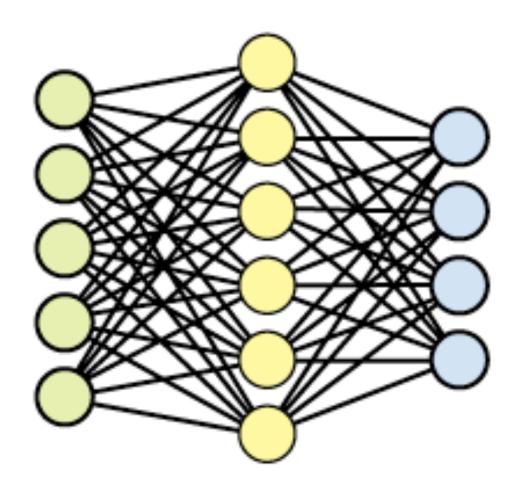
### From Neural Networks to Deep Neural Networks

#### A neural Network

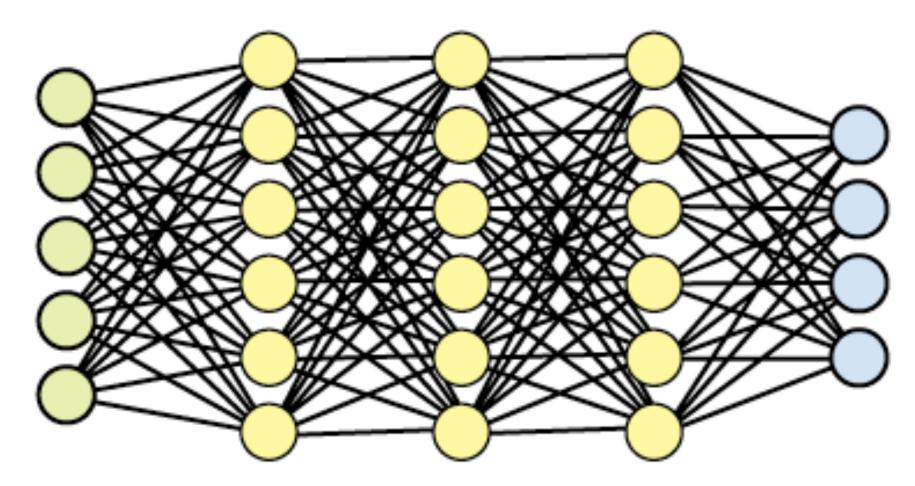


### From Neural Networks to Deep Neural Networks

#### A neural Network

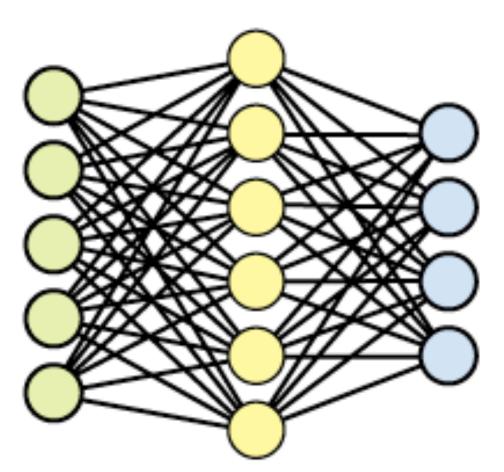


A deep neural Network



### From Neural Networks to Deep Neural Networks

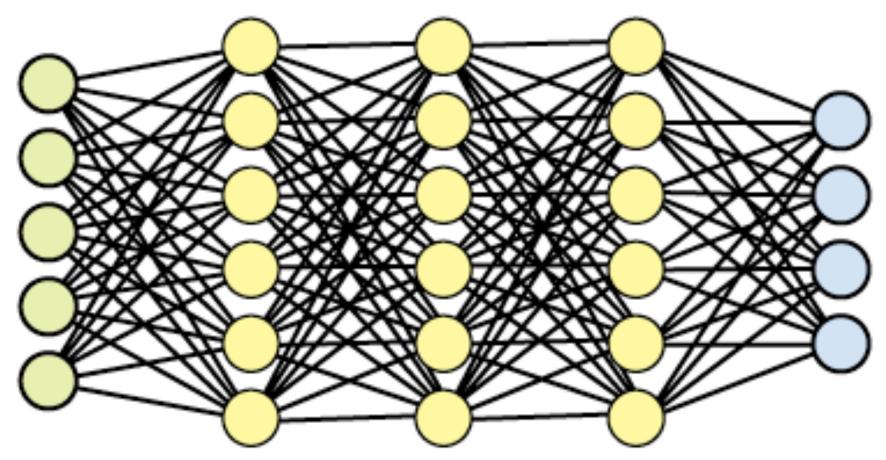
### A neural Network



Modern deep learning provides a powerful framework for supervised learning. By adding more layers and more units within a layer, a deep network can represent functions of increasing complexity.

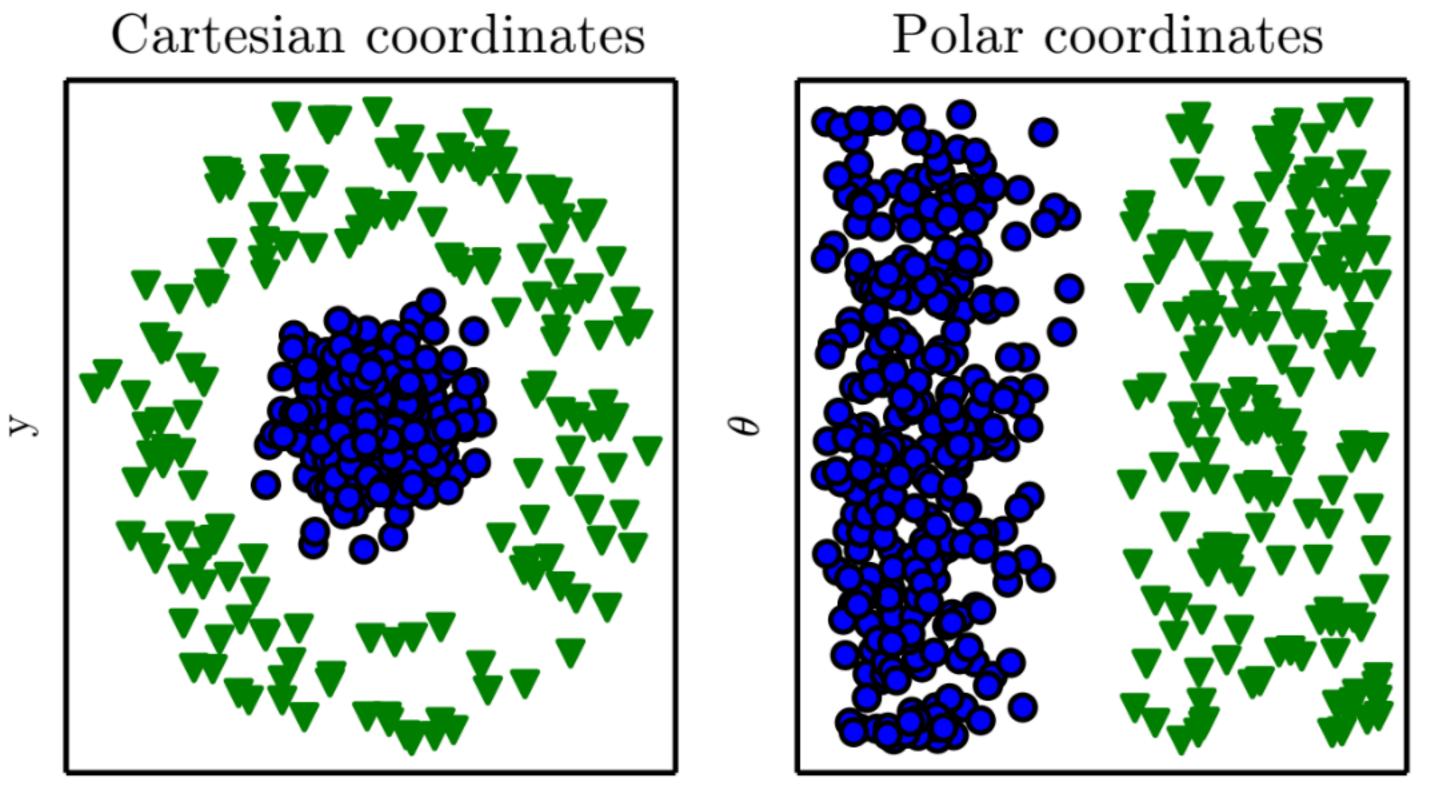
Deep Learning — Part II, p.163 http://www.deeplearningbook.org/contents/part\_practical.html

A deep neural Network



### Another View of deep learning

### • Representations are important





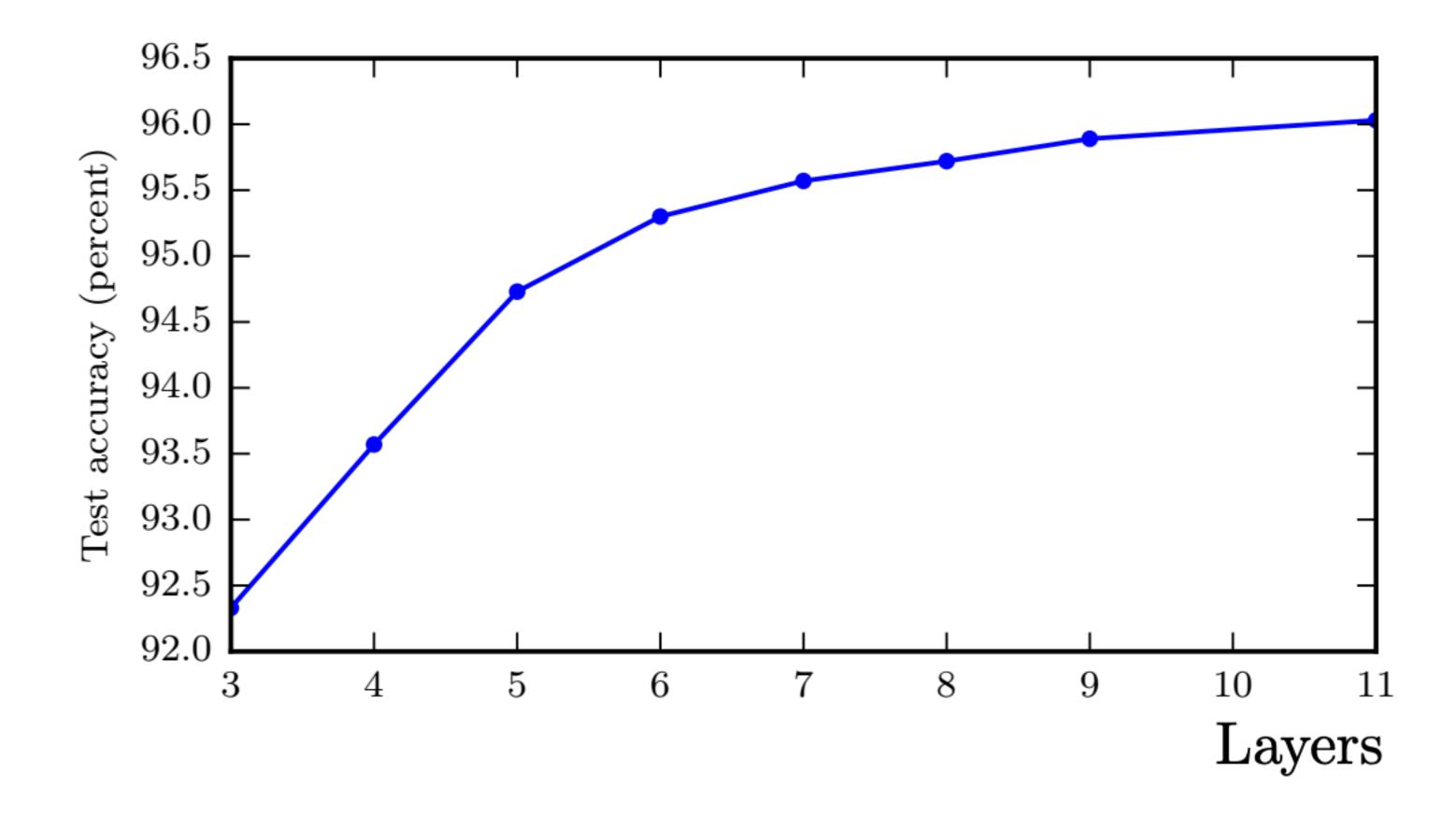
## Hyperparameters

- 1. Model specific
- 2. Optimisation Objective
  - Regularization, Early-stopping, Dropout
- 3. Optimization procedure
  - Momentum, Adaptive learning rates

Activation functions (output & hidden), Network size

### Wide or Deep?

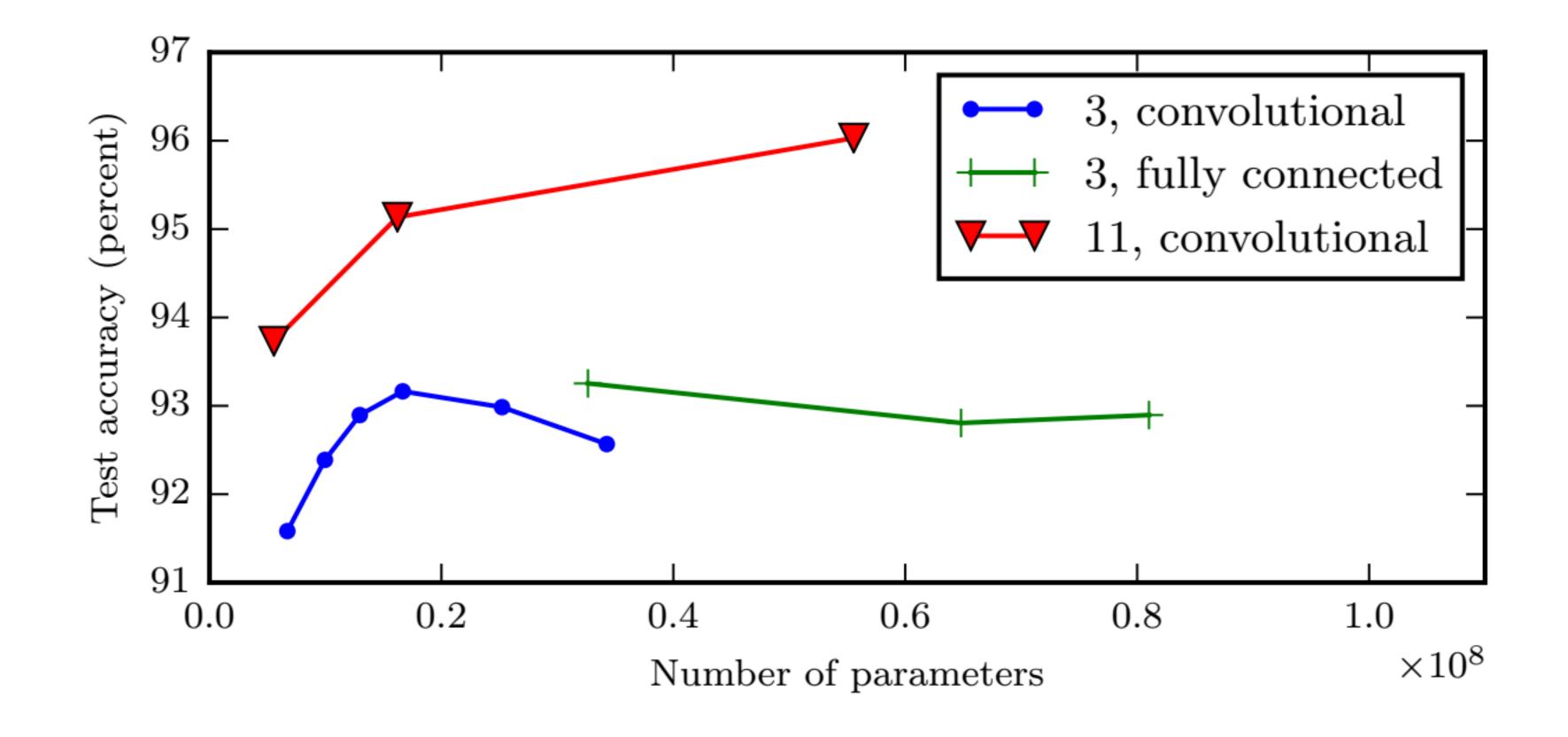
Laurent Charlin — 60629





[Figure 6.6, <u>Deep Learning</u>, book]

## Wide or Deep?



[Figure 6.7, <u>Deep Learning</u>, book]