

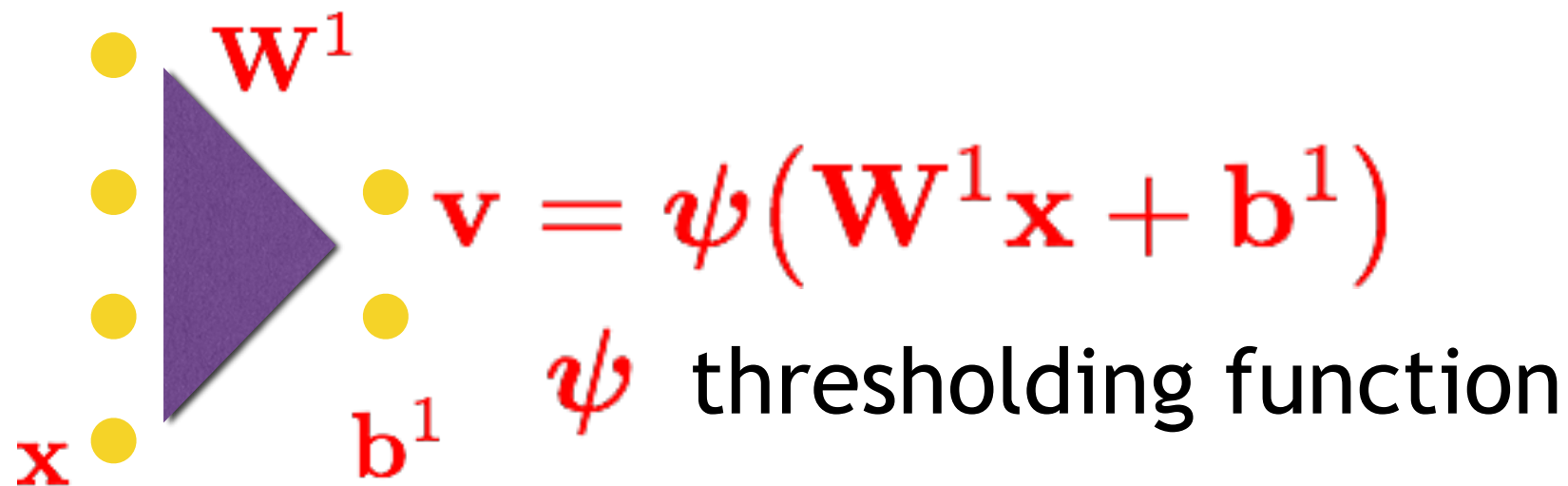
Machine Learning for Sensory Signals

Neural Networks and Deep Learning

13-04-2017

Neural Networks

Perceptron Model [McCulloch, 1943, Rosenblatt, 1957]

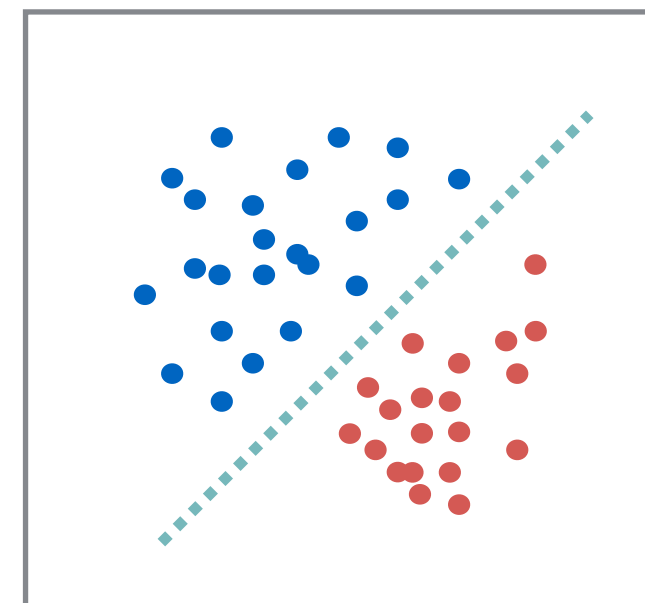


Targets are binary classes [0,1]

Useful for linearly separable data

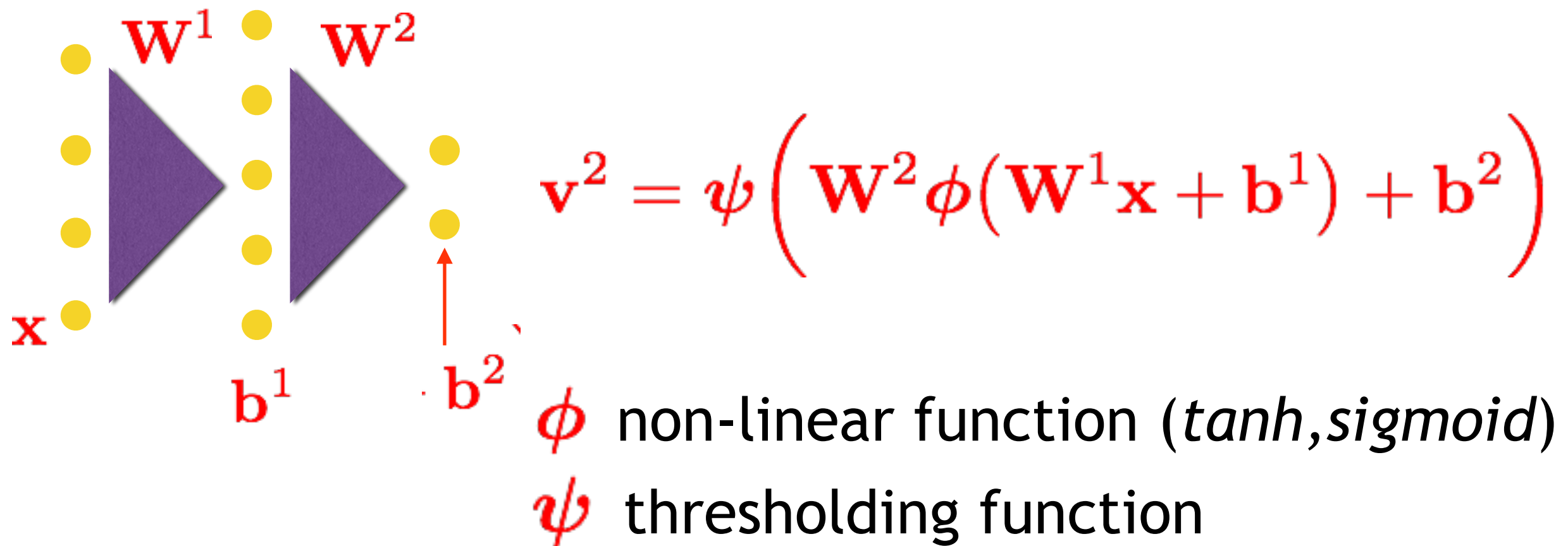
Early Application for simulating the brain

Data - not linearly separable often



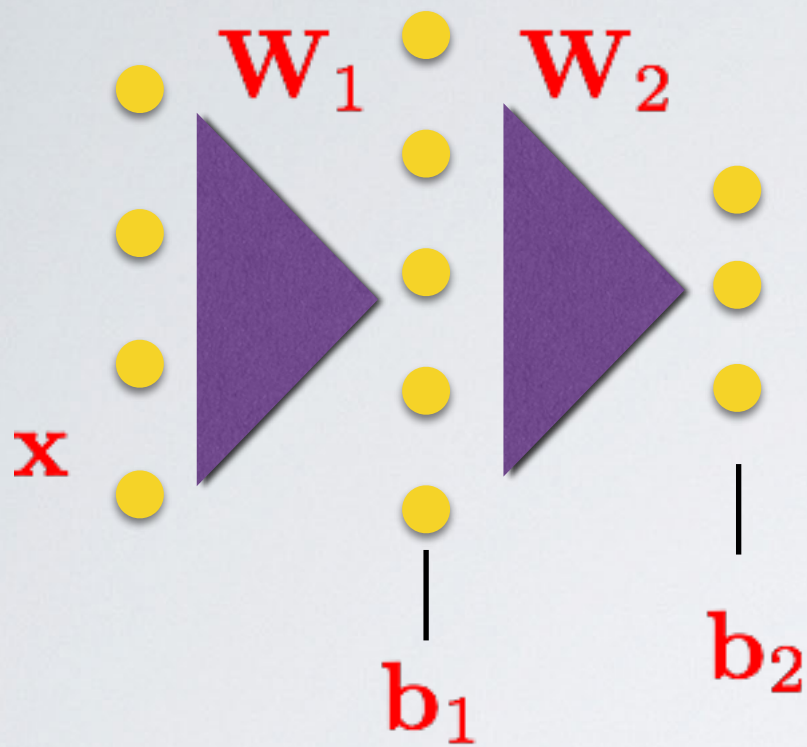
Neural Networks

Multi-layer Perceptron [Hopfield, 1982]



- Useful for classifying non-linear data.
- Modeling of complex real world data like speech and images is some-what limited.

Parameter Learning

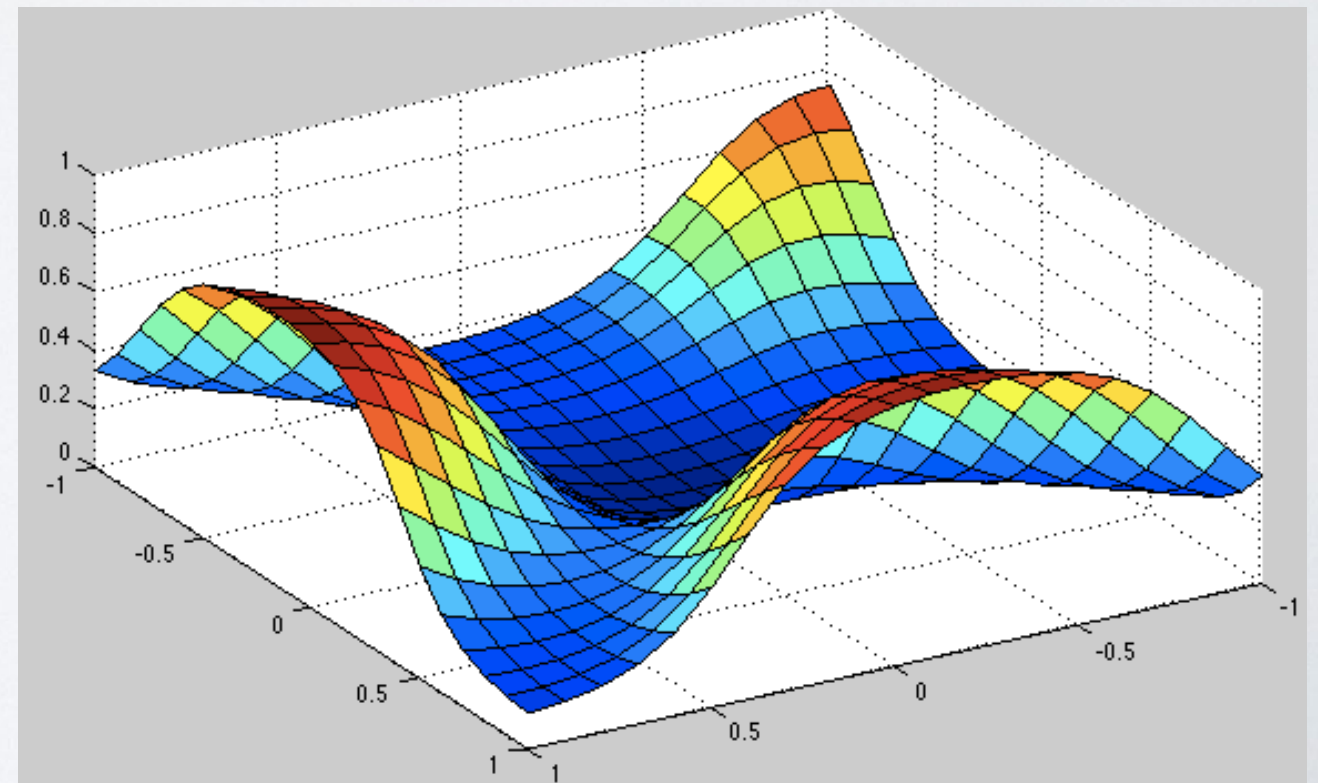


$$\mathbf{v}^2 = \psi \left(\mathbf{W}^2 \phi(\mathbf{W}^1 \mathbf{x} + \mathbf{b}^1) + \mathbf{b}^2 \right)$$

Error function for entire data

$$J_{MSE} = \sum_{i=1}^M ||\mathbf{v}_i - \mathbf{y}_i||^2$$

Typical Error Surface as a function of parameters (weights and biases)



Parameter Learning

Error surface close to a local optima

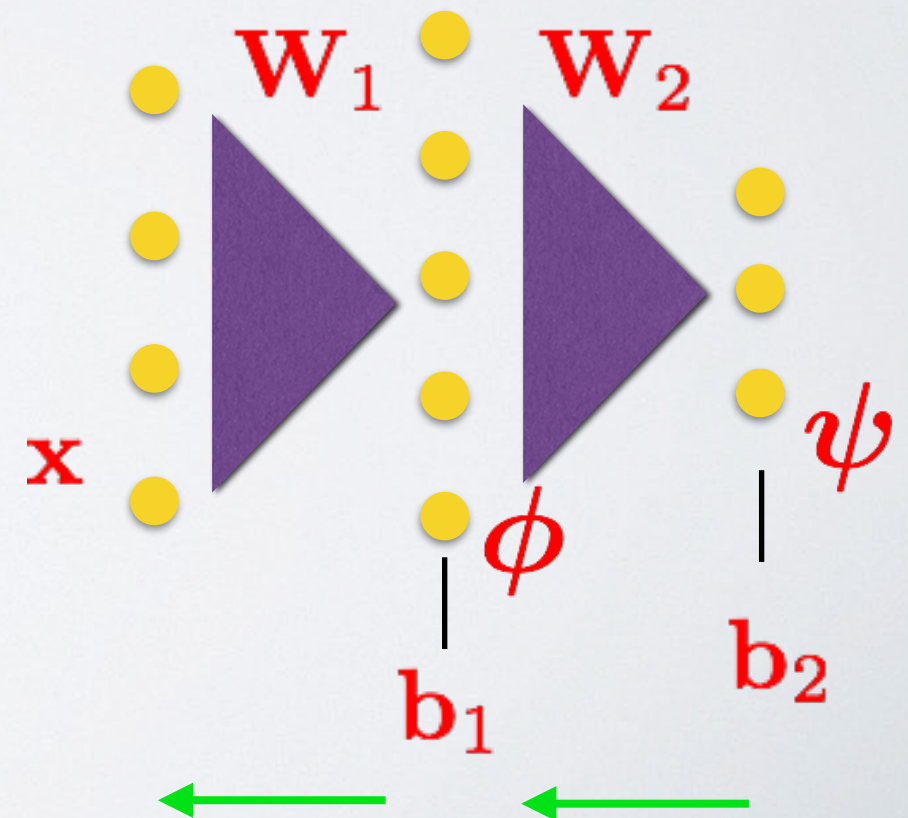
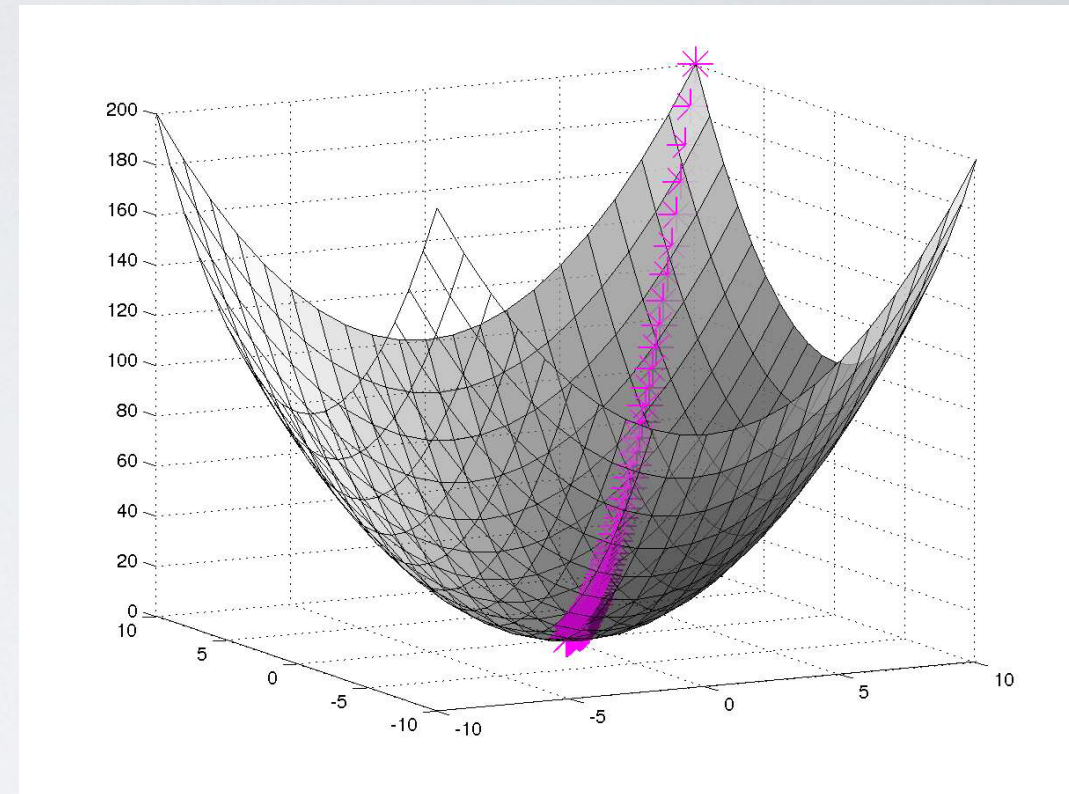
Non-linear nature of error function

- Move in the reverse direction of the gradient

$$\mathbf{W}_1^t = \mathbf{W}_1^{t-1} - \eta \frac{\partial J}{\partial \mathbf{W}_1}$$

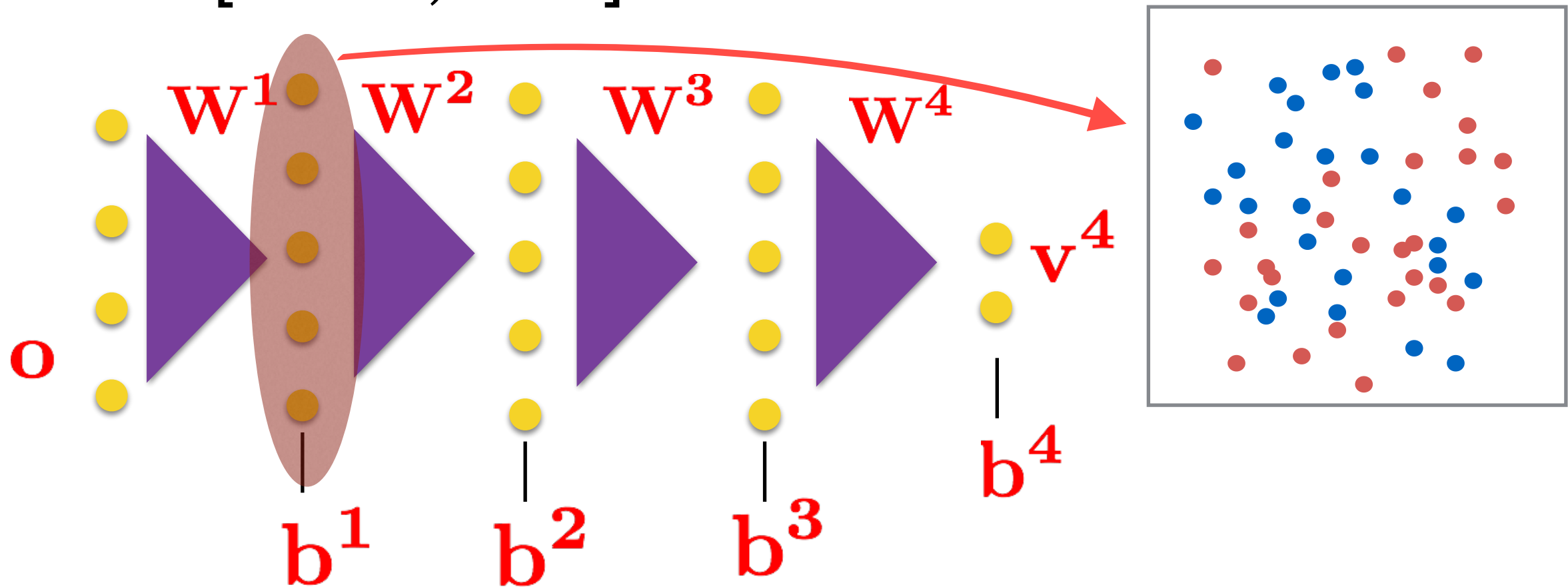
Error back propagation

$$\frac{\partial J}{\partial \mathbf{W}_1} = \frac{\partial J}{\partial \psi} \times \frac{\partial \psi}{\partial \phi} \times \frac{\partial \phi}{\partial \mathbf{W}_1}$$



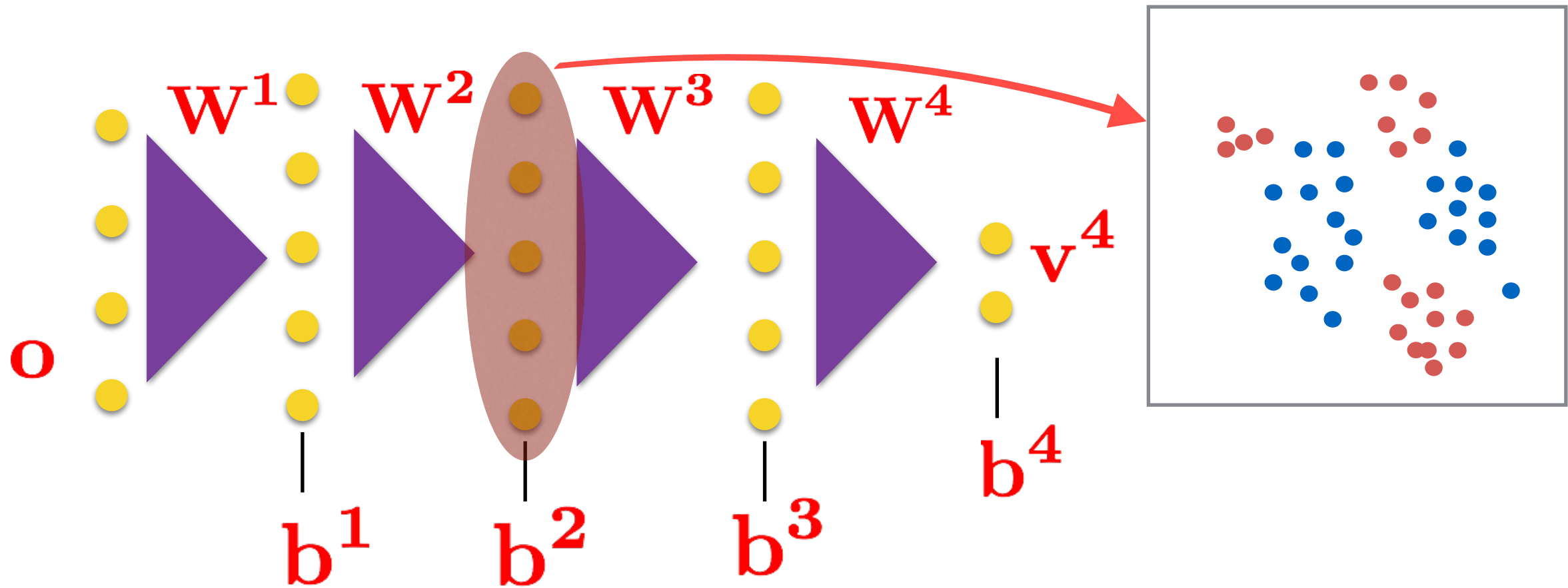
Deep Networks

Neural networks with multiple hidden layers - Deep networks [Hinton, 2006]



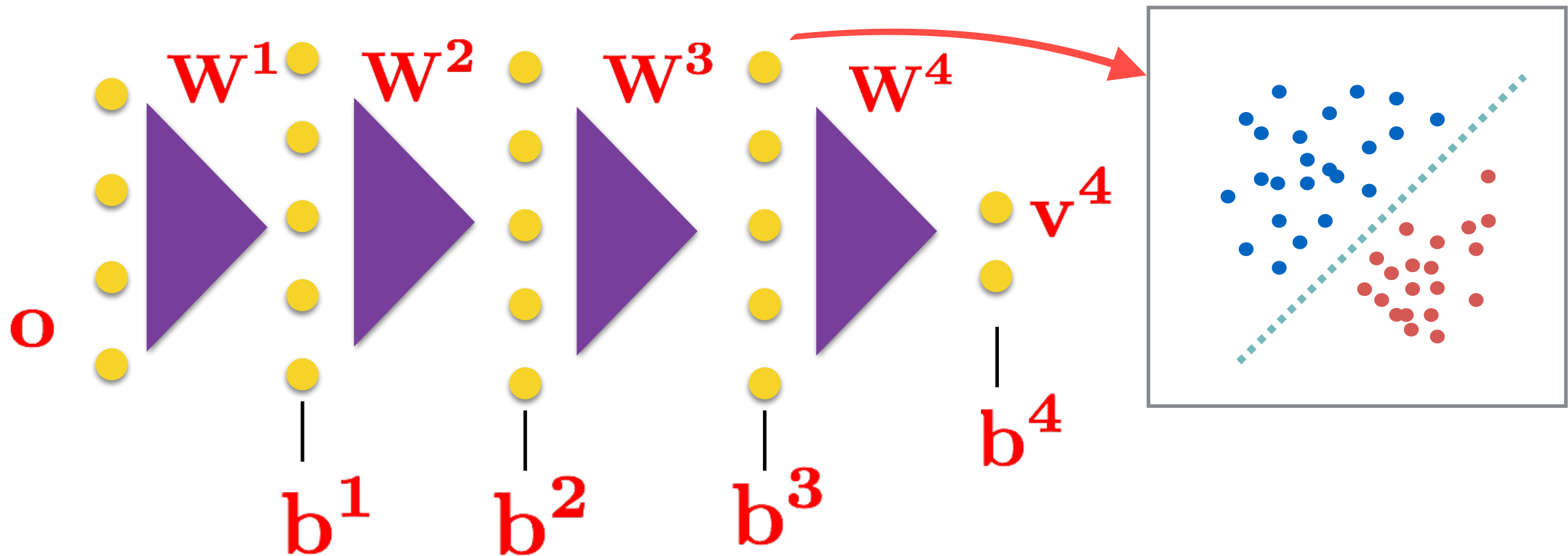
Deep Networks

Neural networks with multiple hidden layers - Deep networks



Deep Networks

Neural networks with multiple hidden layers - Deep networks



Deep networks perform **hierarchical data abstractions** which enable the non-linear separation of complex data samples.