Feed-forward Neural Networks (Part 2: learning)



Outline (part 2)

- Learning feed-forward neural networks
- SGD and back-propagation







Simple example

A long chain like neural network







2 hidden units: training





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 \bullet After ${\sim}10$ passes through the data





10 hidden units

 Randomly initialized weights (zero offset) for the hidden units





10 hidden units

 After ~ 10 epochs the hidden units are arranged in a manner sufficient for the task (but not otherwise perfect)







• 2 hidden units can no longer solve this task





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10 hidden units







Decision boundaries

Symmetries introduced in initialization can persist...



100 hidden units (random offset initialization)





Size, optimization

- Many recent architectures use ReLU units (cheap to evaluate, sparsity)
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500 hidden units





Summary (part 2)

- Neural networks can be learned with SGD similarly to linear classifiers
- The derivatives necessary for SGD can be evaluated effectively via back-propagation
- Multi-layer neural network models are complicated... we are no longer guaranteed to reach global (only local) optimum with SGD
- Larger models tend to be easier to learn ... units only need to be adjusted so that they are, collectively, sufficient to solve the task