

Applied Deep Learning

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Learning objectives of today

Goals: Creating neural networks with Keras and TensorFlow

- Understanding the role that frameworks such as TensorFlow play in enabling the design, training, and use of arbitrarily complex neural networks
- Getting started on applying TensorFlow to create different types of feed-forward networks

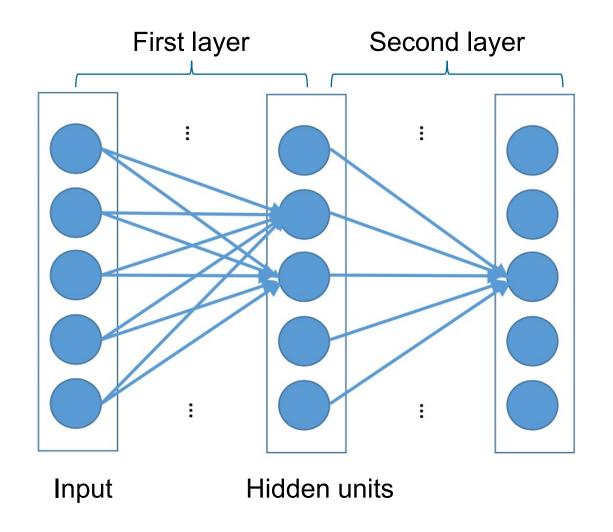
How will we do this?

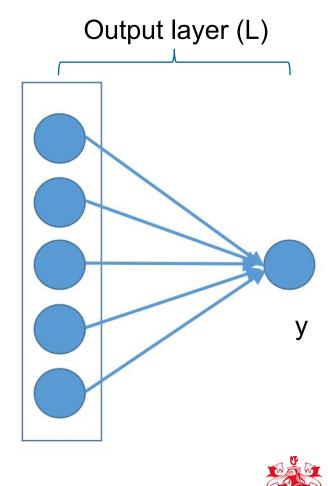
- We start with a quick recap of the functioning of neural networks
- We then introduce TensorFlow and Keras as programming frameworks for neural networks
- We create a number of neural networks, using TensorFlow's Sequential API





Components



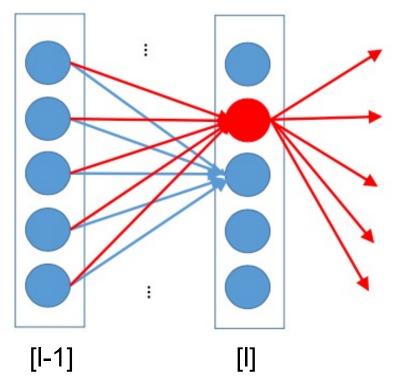


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Source: Liang

Hidden layers

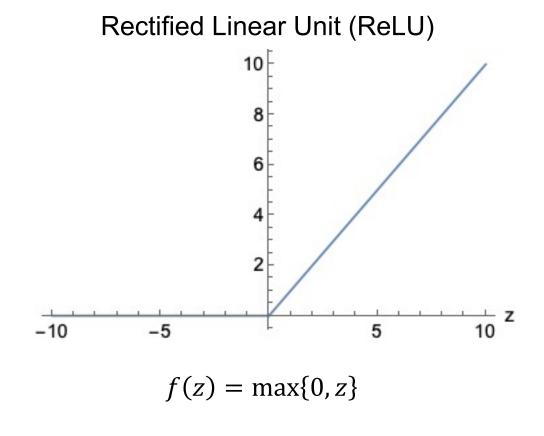
"
$$x$$
" = $a^{[l-1]}$ $z = a^{[l-1]}w + b$ $f(z)$

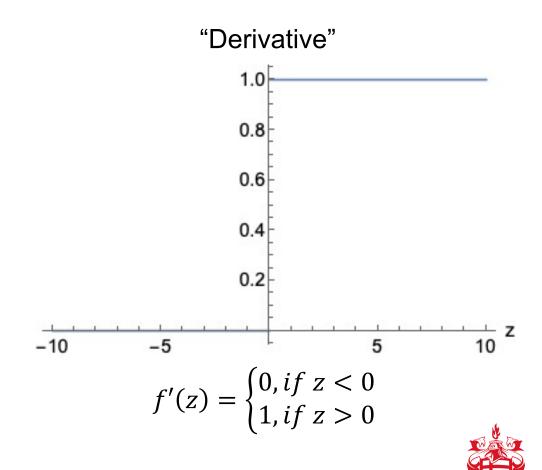


- *f* is what we call an "activation function"
- There are many activation functions, and new ones are invented all the time
- Many of these functions do just fine, or slightly better than existing ones



Typical activation functions: Rectified Linear Unit

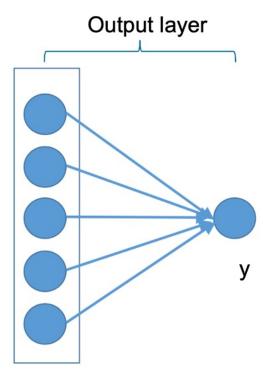




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Binary classification

- Input: $a^{[L-1](i)}$
- As usual, we make a linear transformation: $z^{[L](i)} = a^{[L-1](i)} w^{[L]} + b^{[L]}$
- We then use the logistic sigmoid function $\hat{y}^{(i)} = f(z^{[L](i)}) = \sigma(z^{[L](i)}) = \frac{1}{1 + e^{-z^{[L](i)}}}$
- We can interpret the output as the probability of $y^{(i)} = 1$



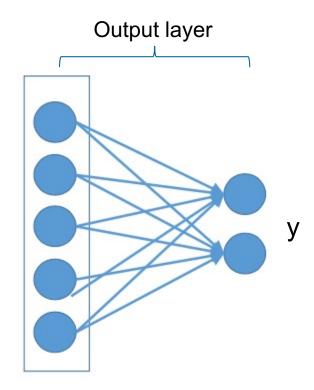


Multi-class classification

- We again make a linear transformation with a matrix of weights: $\mathbf{z}^{[L](i)} = \mathbf{a}^{[L-1](i)} \mathbf{W}^{[L]} + \mathbf{b}^{[L]}$
- Note that $\mathbf{z}^{[L](i)} = \begin{pmatrix} z_1^{[L](i)} & z_2^{[L](i)} & \cdots & z_K^{[L](i)} \end{pmatrix}$
- We then use the softmax function on each of the outputs:

$$\hat{y}_k^{(i)} = f(\mathbf{z}^{[L](i)}) = \frac{e^{-z_k^{[L](i)}}}{\sum_{k=1}^K e^{-z_k^{[L](i)}}}$$

- This implies that $\hat{y}_k^{(i)} \in (0,1)$ and $\sum_{k=1}^K \hat{y}_k^{(i)} = 1$
- Hence, we can interpret $\hat{y}_k^{(i)}$ as the probability that $y^{(i)} = k$ ("belongs to class k")





Generalizing our optimization algorithm

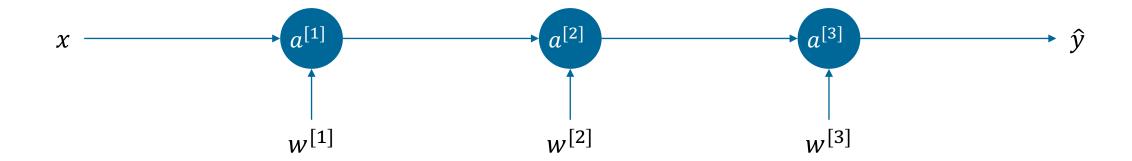
- 1. Decide a "learning rate" α
- 2. Start with some parameters θ
- 3. For a certain number of iterations
 - Compute $J(\theta)$
 - Let $\theta := \theta \alpha \nabla_{\theta} J(\theta)$

(initialization)

(forward propagation) (back-propagation)

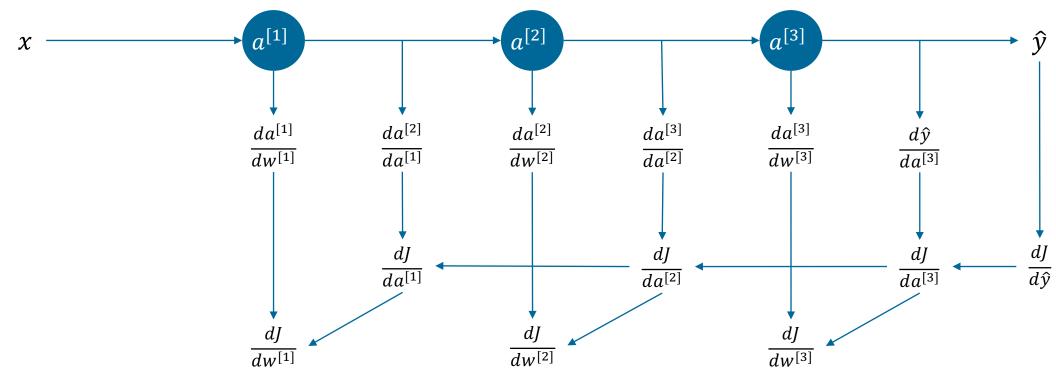


Step 1: Forward propagation through the computational graph





Step 2: Back-propagation through the computational graph





Deep learning programming frameworks

Deep learning in Python: the two main rivals

TensorFlow

- Developed by Google
- Python API based on C engine
- Largest community
- Parallelization of models and data
- Visualization (TensorBoard)
- Relatively intuitive when familiar with numpy
- Runs on most systems (e.g., TensorFlow Lite)
- Broader framework (e.g., reinforcement learning)
- Can be slower than other frameworks

Pytorch

- Developed by Facebook
- Python API based on LUA engine (Torch)
- Many pre-trained models
- Highly modular
- Easy to run on GPUs
- Runs on most systems (e.g., PyTorch Mobile)



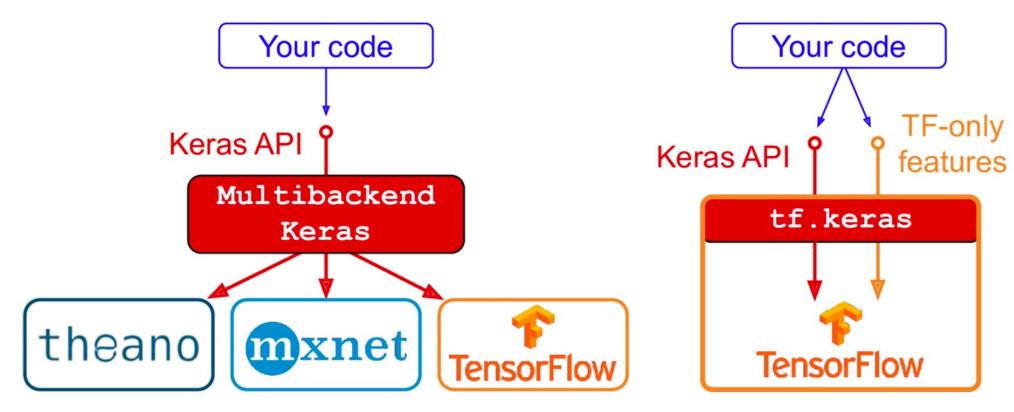


Wasn't there also this thing called Keras?

- High-level deep learning API to build, train, evaluate, execute neural networks
 - Documentation: https://keras.io
- Requires computation backend:
 - TensorFlow
 - Microsoft Cognitive Toolkit
 - Theano
 - Apache MXNEt
 - Core ML (Apple)
 - JavaScript & TypeScrit (for web browsers)
 - PlaidML
 - •
- TensorFlow now comes with its own Keras implementation tf.keras (with some added features)
 - E.g., use Data API from TF



Keras in general and Keras using TensorFlow





Source: Geron

TensorFlow in practice







Sources

- DeepLearning.AI, n.d.: <u>deeplearning.ai</u>
- Geron, 2019, Hands-on Machine Learning with Scikit-Learn, Keras, and TensorFlow
- Goodfellow, Bengio, Courville, 2016, The Deep Learning Book: http://www.deeplearningbook.org
- Liang, 2016, Introduction to Deep Learning: https://www.cs.princeton.edu/courses/archive/spring16/cos495/

