

Deep Neural Nets and Keras

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Keras

Components

Keras hands–on Installation Fun with puppies, kitties, and DNNs

About

(Deep) artificial neural networks are among the most successful machine-learning models.

They are universal tools that can be used for supervised and/or unsupervised learning.

Artificial neural network

- a computational model evaluating a parametric function composed of many other parametric (sub)functions
- composed of many information processing units, organized into interconnected layers
- one unit solves a linearly separable problem, i.e. draws a hyperplane in an *n*-dimensional space

Keras

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- \cdot easy prototyping
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- \cdot accellerated by multicore and GPU

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Powered by a backend

- Tensorflow (default)
- \cdot Theano
- \cdot others (CNTK)

My favourite because

- sufficiently high-level (for my taste)
- allows mixing-in with the wonderfull Python ecosystem (scikit, matplotlib, ...)
- is programmer oriented
- \cdot well-documented, with lots of examples

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- one can cheat in it
 https://s3.amazonaws.com/assets.datacamp.com/
 blog_assets/Keras_Cheat_Sheet_Python.pdf

Keras components

Model

- THE (deep) neural network you want to use
- a stack of connected layers
- \cdot sequential API \times the bare MODEL class

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Layers

- individual levels that define the architecture and functionality of the Model
- different types, properties, params, functions
 - DENSE layers (this is the normal, fully-connected layer)
 - CONVOLUTIONAL layers (applies convolution operations on the previous layer)
 - POOLING layers (used after convolutional layers)
 - DROPOUT layers (regularization, prevent overfitting)

Loss functions

- compare the predicted output with the real output in each pass of the training algorithm
- tell the model how the weights should be updated
- mean-squared error, cross-entropy, ...

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Optimizers

- \cdot weight update strategies in the training process
- stochastic gradient descent, RMSProp, Adagrad

Keras hands-on

(Fairly) easy steps

- Get Python (Anaconda highly recommended: https://www.anaconda.com/download/)
- Get TensorFlow (https://www.tensorflow.org/install/)
- Get Keras (https://keras.io/)

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conda install keras conda install pydot

The mother of all classification demos: cats vs. dogs



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Published on Kaggle in 2014, contains 25,000 images of cats and dogs.

The mother of all classification demos: cats vs. dogs



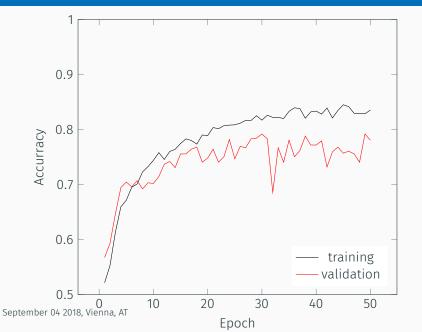
Published on Kaggle in 2014, contains 25,000 images of cats and dogs.

To make it a bit harder, we use only 1000 training images of each Septendersy. 2018, Vienna, AT

Computer demo ...

https://goo.gl/M5ShF3

From scratch[™]

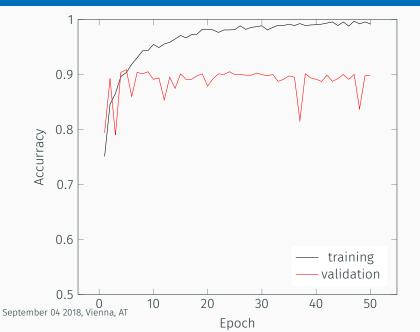


11

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)		
activation_1 (Activation)		
<pre>max_pooling2d_1 (MaxPooling2</pre>	(None, 74, 74, 32)	0
conv2d_2 (Conv2D)	(None, 72, 72, 32)	9248
activation_2 (Activation)	(None, 72, 72, 32)	
<pre>max_pooling2d_2 (MaxPooling2</pre>	(None, 36, 36, 32)	
conv2d_3 (Conv2D)	(None, 34, 34, 64)	
activation_3 (Activation)	(None, 34, 34, 64)	
max_pooling2d_3 (MaxPooling2	(None, 17, 17, 64)	
flatten_1 (Flatten)	(None, 18496)	
dense_1 (Dense)	(None, 64)	
activation_4 (Activation)	(None, 64)	0
dropout_1 (Dropout)	(None, 64)	
dense_2 (Dense)	(None, 1)	
activation_5 (Activation)	(None, 1)	
Total params: 1,212,513 Trainable params: 1,212,513 Non-trainable params: 0		

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VGG16 / ImageNet



What VGG16 dreams about?

Layer (type) input 1 (InputLayer)	Output		Paran #
terror a two costs was set			
input_1 (inputLayer)			
block1_conv1 (Conv2D)			
block1_conv2 (Conv2D)			
blockl_pool (MaxPooling2D)			
block2_conv1 (Conv2D)			
block2_conv2 (Conv2D)			
block2_pool (MaxPooling2D)			
block3_conv1 (Conv2D)			
block3_conv2 (Conv2D)			
block3_conv3 (Conv2D)			
block3_pool (MaxPooling2D)		28, 28, 256)	
block4_conv1 (Conv2D)			
block4_conv2 (Conv2D)			
block4_conv3 (Conv2D)			
block4_pool (MaxPooling2D)		14, 14, 512)	
block5_conv1 (Conv2D)			
block5_conv2 (Conv2D)			
block5_conv3 (Conv2D)			
block5_pool (MaxPooling2D)			
flatten (Flatten)			
fcl (Dense)			
fc2 (Dense)			
predictions (Dense)	(None,	1000)	4097000
Total params: 138,357,544 Trainable params: 138,357,544 Non-trainable params: 0			

What VGG16 dreams about?

Layer (type)	Output Shape	Paran 4
input 1 (InputLayer)	(None, 224, 224, 3)	0
block1 conv1 (Conv2D)		1792
block1 conv2 (Conv2D)		36928
block1 pool (MaxPooling2D)		0
block2 conv1 (Conv2D)		
block2_conv2 (Conv2D)		147584
block2_pool (MaxPooling2D)		0
block3_conv1 (Conv2D)		295168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590080
block3_conv3 (Conv2D)		590080
block3_pool (MaxPooling2D)		0
block4_conv1 (Conv2D)		1180160
block4_conv2 (Conv2D)		2359808
block4_conv3 (Conv2D)		2359808
block4_pool (MaxPooling2D)		0
block5_conv1 (Conv2D)		2359808
block5_conv2 (Conv2D)		2359808
block5_conv3 (Conv2D)		2359808
block5_pool (MaxPooling2D)		0
flatten (Flatten)		0
fcl (Dense)		102764544
fc2 (Dense)		16781312
predictions (Dense)		4097000
Total params: 138,357,544		
Trainable params: 138,357,54 Non-trainable params: 0		

