More layers

CONVOLUTIONAL LAYERS



layer_conv_1d() 1D, e.g. temporal convolution



Transposed 2D (deconvolution) layer_conv_2d_transpose()

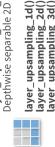
layer_conv_2d() 2D, e.g. spatial convolution over images



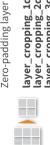
layer_conv_3d_transpose()
Transposed 3D (deconvolution)
layer_conv_3d() 3D, e.g. spatial convolution over volumes

layer_conv_lstm_2d() Convolutional LSTM

layer_separable_conv_2d()







layer_cropping_1d()
layer_cropping_2d()
layer_cropping_3d()

POOLING LAYERS

Cropping layer



layer_max_pooling_1d()
layer_max_pooling_2d()
layer_max_pooling_3d()
Maximum pooling for 1D to 3D



layer_average_pooling_1d()
layer_average_pooling_2d()
layer_average_pooling_3d()
Average pooling for 1D to 3D

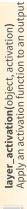


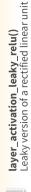
layer_global_average_pooling_1d() layer_global_average_pooling_2d() layer_global_average_pooling_3d() Global average pooling

Global maximum pooling



ACTIVATION LAYERS





layer_activation_parametric_relu() Parametric rectified linear unit

p

layer_activation_thresholded_relu()
Thresholded rectified linear unit 1

layer_activation_elu()
Exponential linear unit

DROPOUT LAYERS



Applies dropout to the input layer_dropout()



layer_spatial_dropout_2d()
layer_spatial_dropout_3d()
Spatial_1D to 3D version of dropout

RECURRENT LAYERS



layer_simple_rnn()
Fully-connected RNN where the output is to be fed back to input

layer_gru() Gated recurrent unit - Cho et al

Fast GRU implementation backed layer_cudnn_gru() by CuDNN

layer_Istm() Long-Short Term Memory unit -Hochreiter 1997

layer_cudnn_lstm()
Fast LSTM implementation backed
by CuDNN

LOCALLY CONNECTED LAYERS

layer_locally_connected_1d()
layer_locally_connected_2d()
Similar to convolution, but weights are not
shared, i.e. different filters for each patch

Preprocessing

SEQUENCE PREPROCESSING

pad_sequences()

Pads each sequence to the same length (length of :he longest sequence)

Generates skipgram word pairs skipgrams()

Generates word rank-based probabilistic sampling make_sampling_table() table

TEXT PREPROCESSING

:ext_tokenizer() Text tokenization utility

fit_text_tokenizer() Update tokenizer internal vocabulary save_text_tokenizer(); load_text_tokenizer() Save a text tokenizer to an external file

texts_to_sequences_generator() texts_to_sequences();

ransforms each text in texts to sequence of integers

texts_to_matrix(); sequences_to_matrix() Convert a list of sequences into a matrix text_one_hot() One-hot encode text to word indices

text_hashing_trick()

Converts a text to a sequence of indexes in a fixedsize hashing space

text_to_word_sequence()

Convert text to a sequence of words (or tokens)

IMAGE PREPROCESSING

mage_load() Loads an image into PIL format.

!low_images_from_directory() !low_images_from_data()

Generates batches of augmented/normalized data from images and labels, or a directory image_data_generator() Generate minibatches of image data with real-time data augmentation.

generator internal statistics to some sample data fit_image_data_generator() Fit image data

generator_next() Retrieve the next item

image_array_save() 3D array representation image_to_array(); image_array_resize()





TensorFlow Keras

Pre-trained models

prediction, feature extraction, and fine-tuning. that are made available alongside pre-trained weights. These models can be used for Keras applications are deep learning models

xception_preprocess_input()
Xception v1 model application_xception()

Inception v3 model, with weights pre-trained inception_v3_preprocess_input() application_inception_v3() on ImageNet

inception_resnet_v2_preprocess_input() Inception-ResNet v2 model, with weights application_inception_resnet_v2() trained on ImageNet application_vgg16(); application_vgg19() VGG16 and VGG19 models

application_resnet50() ResNet50 model

mobilenet_preprocess_input() mobilenet_decode_predictions() mobilenet_load_model_hdf5() MobileNet model architecture application_mobilenet()

IM GENET

geNet is a large database of images with labels, extensively used for deep learning

imagenet_decode_predictions() imagenet_preprocess_input()

images for ImageNet, and decodes predictions Preprocesses a tensor encoding a batch of

Callbacks

given stages of the training procedure. You can use callbacks to get a view on internal states A callback is a set of functions to be applied at and statistics of the model during training.

callback_learning_rate_scheduler() Learning callback_early_stopping() Stop training when a monitored quantity has stopped improving

callback tensorboard() TensorBoard basic rate scheduler visualizations RStudio® is a trademark of RStudio, Inc. • CC BY SA RStudio• info@rstudio.com • 844-448-1212• rstudio.com • Learn more at keras.rstudio.com • keras 2.1.2 • Updated: 2017-12