Введение в нейронные сети. Урок 2. Keras

GeekBrains

План вебинара

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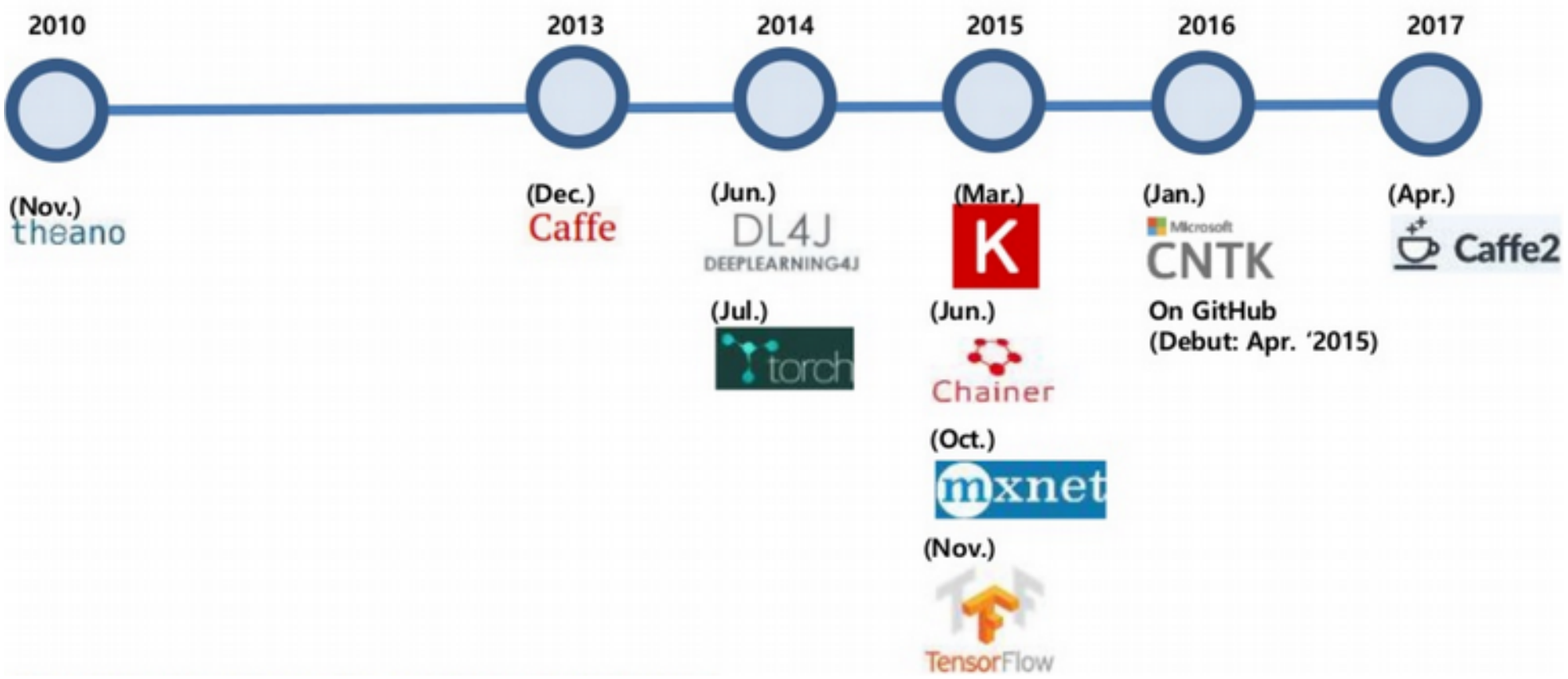
1. Инструменты для создания нейронных сетей.

2. Общие сведения о Keras

3. Синтаксис Keras

4. Практика

Инструменты для создания нейр. сетей



Общие сведения о Keras

Основы синтаксиса

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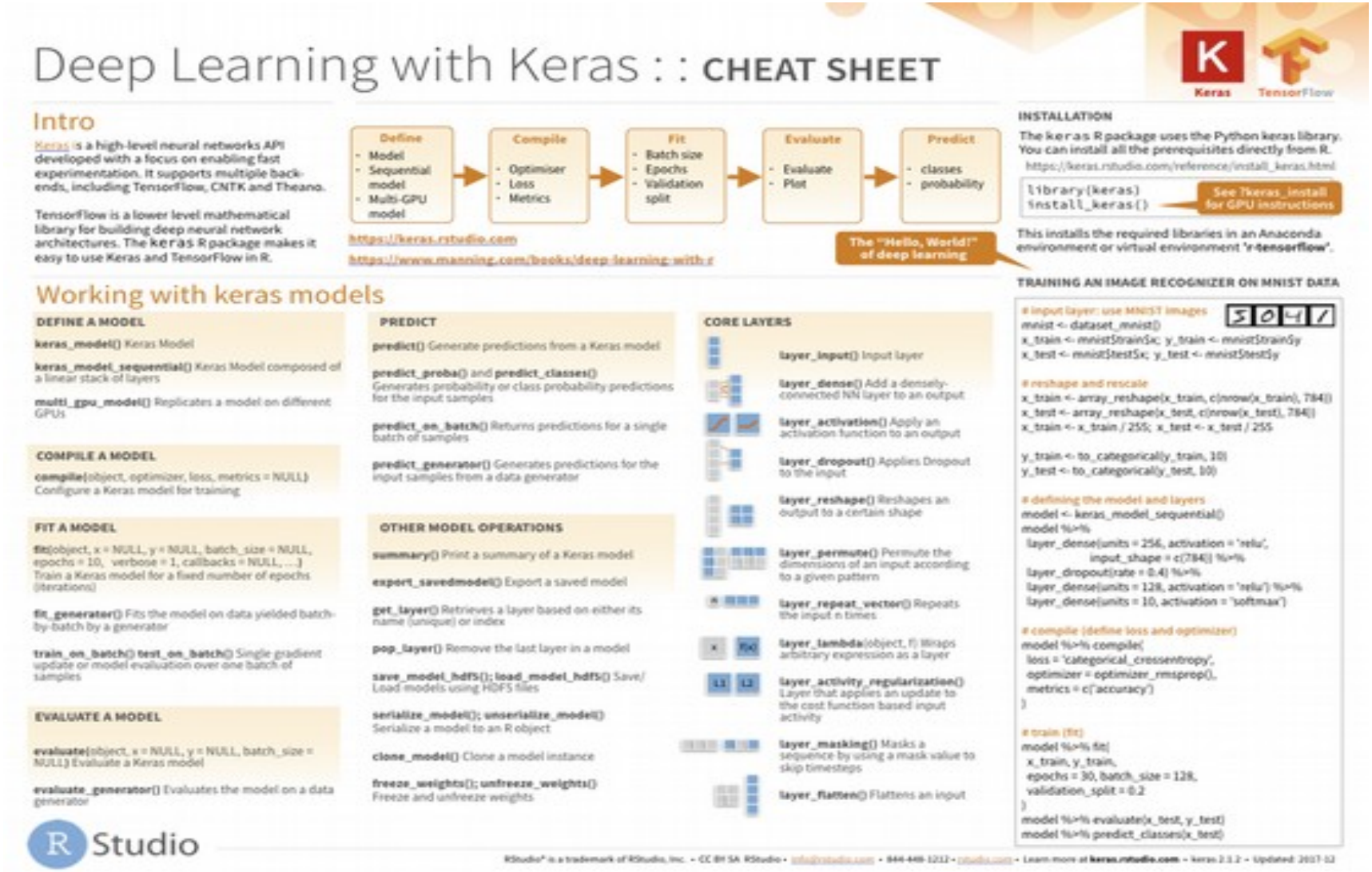


Deep Learning with Keras : : CHEAT SHEET

Keras TensorFlow

Intro

Keras is a high-level neural networks API developed with a focus on enabling fast experimentation. It supports multiple back-ends, including TensorFlow, CNTK and Theano.

TensorFlow is a lower level mathematical library for building deep neural network architectures. The keras R package makes it easy to use Keras and TensorFlow in R.

Define: Model; Sequential model; Multi-GPU model → Compile: Optimiser; Loss; Metrics → Fit: Batch size; Epochs; Validation split → Evaluate: Evaluate; Plot → Predict: classes; probability

https://keras.rstudio.com

https://www.manning.com/books/deep-learning-with-r

The "Hello, World!" of deep learning

INSTALLATION

The keras R package uses the Python keras library. You can install all the prerequisites directly from R.

https://keras.rstudio.com/reference/install_keras.html

```
library(keras)
install_keras()
```

See ?keras_install for GPU instructions

This installs the required libraries in an Anaconda environment or virtual environment 'r-tensorflow'.

Working with keras models

DEFINE A MODEL

keras_model() Keras Model

keras_model_sequential() Keras Model composed of a linear stack of layers

multi_gpu_model() Replicates a model on different GPUs

COMPILE A MODEL

compile(object, optimizer, loss, metrics = NULL) Configure a Keras model for training

FIT A MODEL

fit(object, x = NULL, y = NULL, batch_size = NULL, epochs = 10, verbose = 1, callbacks = NULL, ...) Train a Keras model for a fixed number of epochs (iterations)

fit_generator() Fits the model on data yielded batch-by-batch by a generator

train_on_batch() **test_on_batch()** Single gradient update or model evaluation over one batch of samples

EVALUATE A MODEL

evaluate(object, x = NULL, y = NULL, batch_size = NULL) Evaluate a Keras model

evaluate_generator() Evaluates the model on a data generator

PREDICT

predict() Generate predictions from a Keras model

predict_proba() and **predict_classes()** Generates probability or class probability predictions for the input samples

predict_on_batch() Returns predictions for a single batch of samples

predict_generator() Generates predictions for the input samples from a data generator

OTHER MODEL OPERATIONS

summary() Print a summary of a Keras model

export_savedmodel() Export a saved model

get_layer() Retrieves a layer based on either its name (unique) or index

pop_layer() Remove the last layer in a model

save_model_hdf5(); **load_model_hdf5()** Save/Load models using HDF5 files

serialize_model(); **unserialize_model()** Serialize a model to an R object

clone_model() Clone a model instance

freeze_weights(); **unfreeze_weights()** Freeze and unfreeze weights

CORE LAYERS

layer_input() Input layer

layer_dense() Add a densely-connected NN layer to an output

layer_activation() Apply an activation function to an output

layer_dropout() Applies Dropout to the input

layer_reshape() Reshapes an output to a certain shape

layer_permute() Permute the dimensions of an input according to a given pattern

layer_repeat_vector() Repeats the input n times

layer_lambda(object, f) Wraps arbitrary expression as a layer

layer_activity_regularization() Layer that applies an update to the cost function based input activity

layer_masking() Masks a sequence by using a mask value to skip timesteps

layer_flatten() Flattens an input

TRAINING AN IMAGE RECOGNIZER ON MNIST DATA

```
# input layer: use MNIST images
mnist <- dataset_mnist()
x_train <- mnist$train$x; y_train <- mnist$train$y
x_test <- mnist$test$x; y_test <- mnist$test$y

# reshape and rescale
x_train <- array_reshape(x_train, c(nrow(x_train), 784))
x_test <- array_reshape(x_test, c(nrow(x_test), 784))
x_train <- x_train / 255; x_test <- x_test / 255

y_train <- to_categorical(y_train, 10)
y_test <- to_categorical(y_test, 10)

# defining the model and layers
model <- keras_model_sequential()
model %>%
 layer_dense(units = 256, activation = 'relu',
             input_shape = c(784)) %>%
 layer_dropout(rate = 0.4) %>%
 layer_dense(units = 128, activation = 'relu') %>%
 layer_dense(units = 10, activation = 'softmax')

# compile (define loss and optimizer)
model %>% compile(
 loss = 'categorical_crossentropy',
 optimizer = optimizer_rmsprop(),
 metrics = c('accuracy')
)

# train (fit)
model %>% fit(
 x_train, y_train,
 epochs = 30, batch_size = 128,
 validation_split = 0.2
)
model %>% evaluate(x_test, y_test)
model %>% predict_classes(x_test)
```

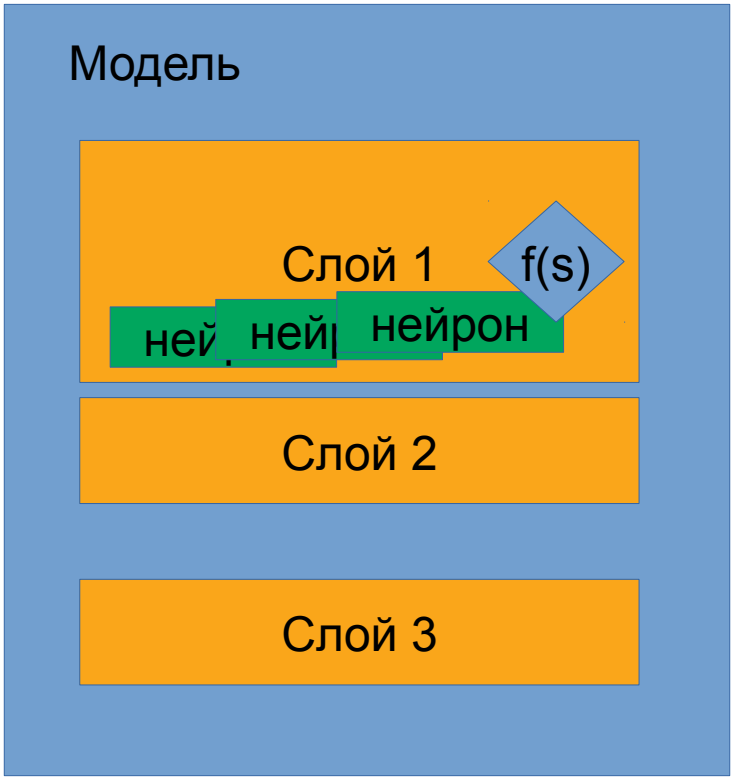
RStudio

Структура Keras

Models
- Sequential
- Model API

Layers
- сверточные
- рекуррентные
- полносвязные
- служебные

Preprocessing
- utils
- обработка изображений
- обработка текстов

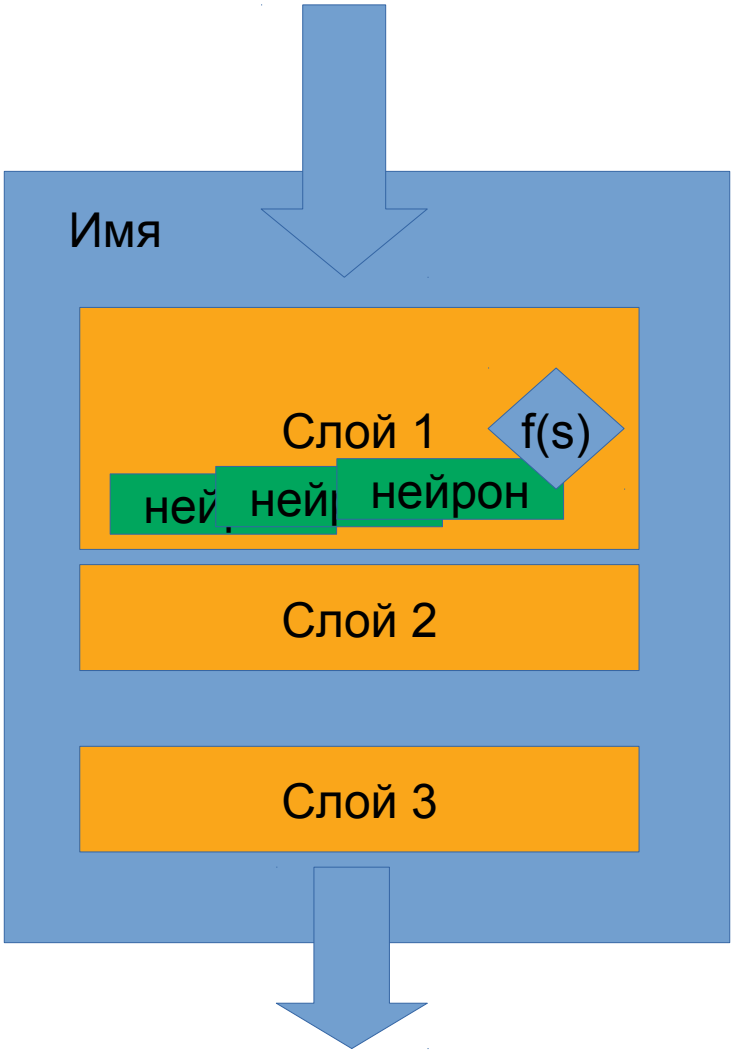


Models.Model

keras.Model()

inputs
outputs
name

Model.summary()

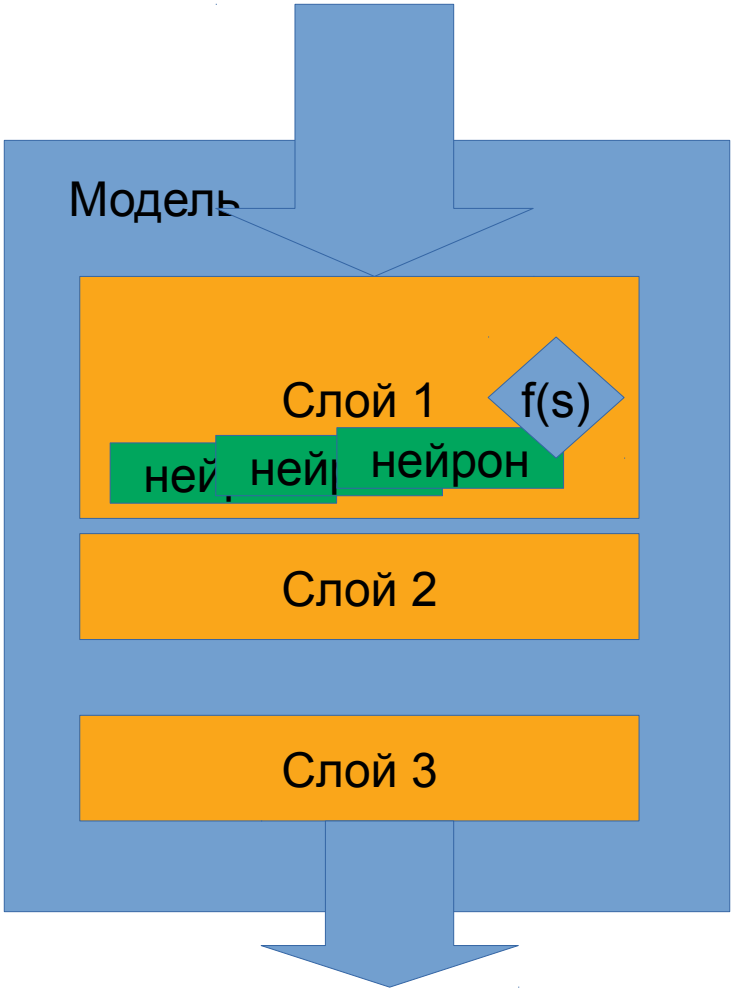


Models.Sequential

keras.Sequential()

layers,
Name

модель.add(<Слой>)

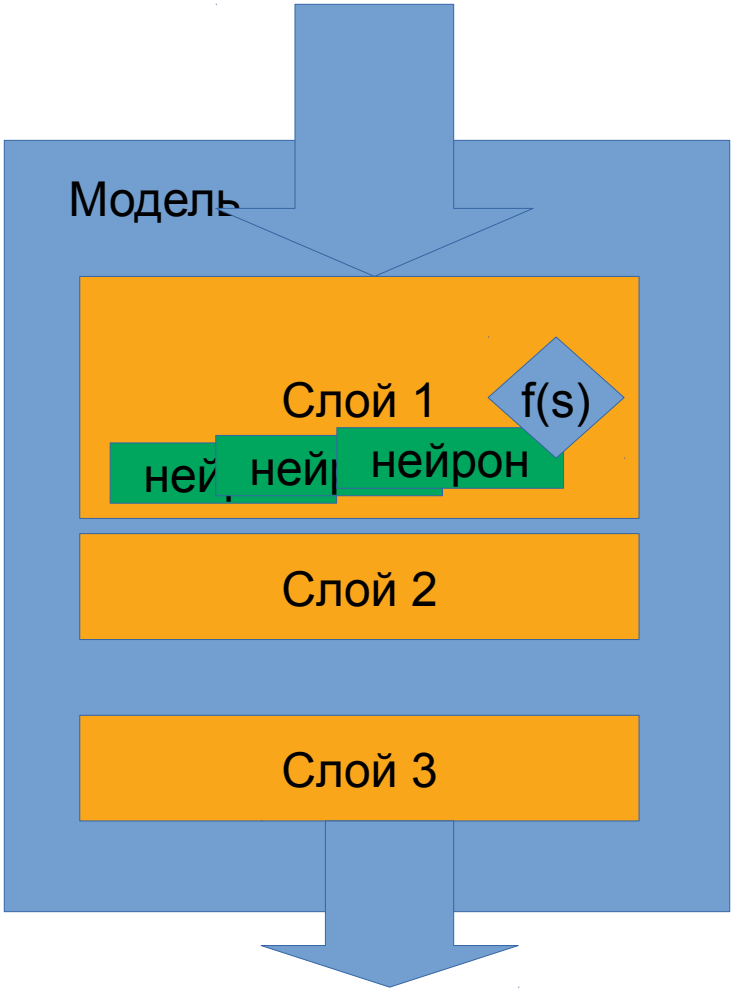


Models Model training API

```
Model.compile(
    optimizer="rmsprop",
    loss=None,
    metrics=None)

Model.fit(    x=None,   y=None,
    batch_size=None,
    epochs=1,
    validation_split=0.0)

Model.predict(  x)
```



Вопросы

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Практическое задание

1. Попробуйте обучить нейронную сеть на Keras на Fashion-MNIST датасете.
   Опишите в комментарии к уроку - какой результата вы добились от нейросети? Что помогло вам улучшить ее точность?

*2. Поработайте с документацией Keras. Найдите полезные команды не разобранные на уроке.