Keras: Theory and Examples

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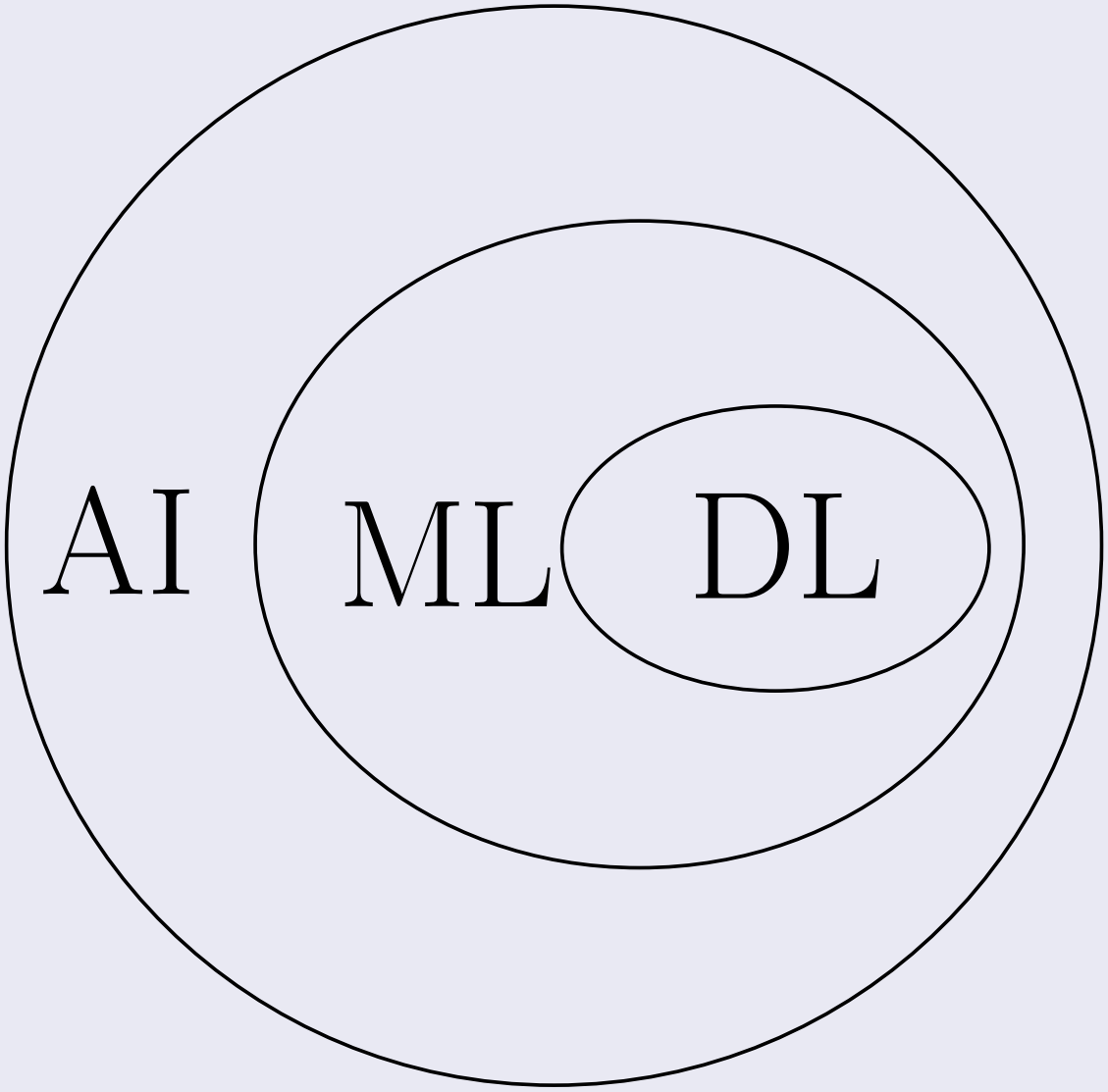
OUTLINE

Introduction



Architecture of a Neural Network

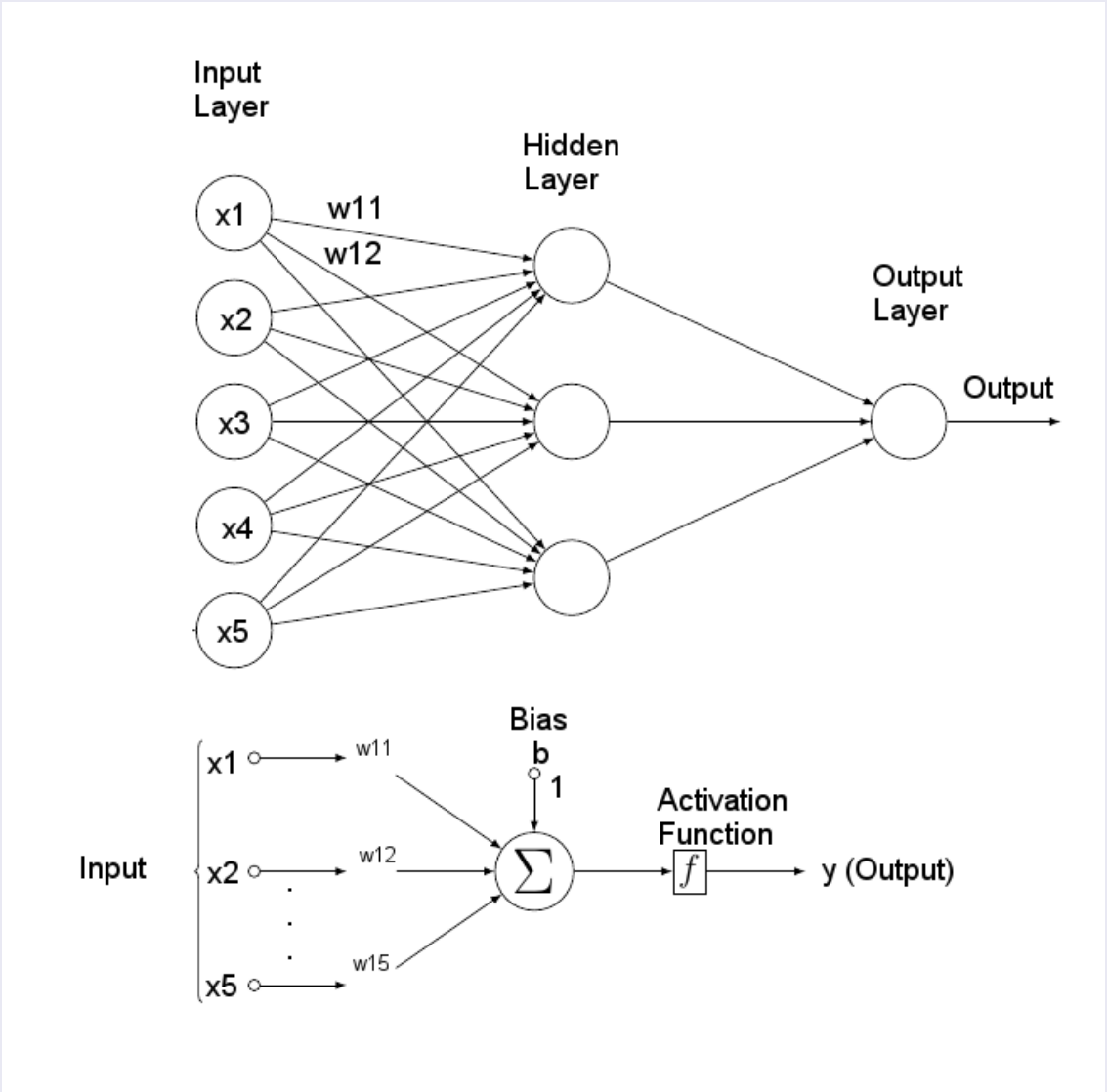


Figure: A Neural Network

Architecture of a Neural Network

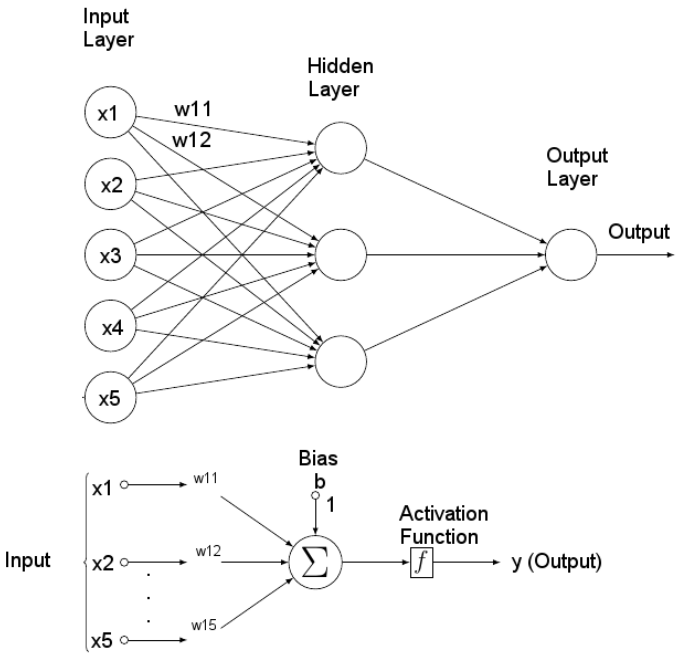


Figure: A Neural Network

Learning Steps (Decisions to be made):

1. Application (Problem)
2. Type of model
3. No. of layers
4. No. of nodes
5. Initialization of weights
6. Activation Function
7. Optimization Function
8. Evaluation Metrics
9. Dataset
10. Testing and Training Data
11. Batch size
12. Epoch

Keras

- NN: development (implementation and experimentation) is difficult.

Keras is

- high-level neural networks library
- written in Python
- capable of running on top of
  - TensorFlow (open source software library for numerical computation)
  - Theano (numerical computation library for Python)
  - CNTK (Microsoft Cognitive Toolkit): Deep learning framework
- developed with a focus on enabling fast experimentation (through user friendliness, modularity, and extensibility)
- and much more

Guiding principles

- Modularity
  - configurable modules
    - neural layers, cost functions, optimizers, initialization schemes, activation functions, regularization schemes are all standalone modules that you can combine to create new models
- Minimalism
  - Each module should be kept short and simple
- Easy extensibility
  - New modules are simple to add (as new classes and functions)
  - suitable for advanced research
- Work with Python
  - Models are described in Python code, which is compact, easier to debug, and allows for ease of extensibility
- User friendliness

Dependencies

- No need to worry

- Python 2.7+
- numpy: fundamental package for scientific computing with Python
- scipy: library used for scientific computing and technical computing
- Matplotlib (Optional, recommended for exploratory analysis)
- HDF5 and h5py (Optional, required if you use model saving/loading functions)
- Theano

Installation

- Once again no need to worry
- Follow instructions provided in "keras installation" file
- Alternatively you may visit Keras Installation Page

Keras provides

What is in the toolbox ?

- Models
- Layers
- Preprocessing
- Metrics
- Optimizers
- Activations
- Datasets
- Constraints
- Initializers
- Loss (Objecitve) Function
- and many more...

Model

- Model
  - core data structure of Keras
  - a way to organize layers
- Two types:
  - Sequential
  - Model class API
- Sequential Model: a linear stack of layers
- functional API: for defining complex models, such as models with shared layers

Layers

- Core Layers
  - Dense
  - Activation
  - Dropout
  - Flatten
  - many more ...
- Convolutional Layers
- Pooling Layers
- Recurrent Layers
- Your own Keras layers
- and many more ...

Core Layers

Dense

- fully connected NN layer: connection to all activation in previous layer

Activation

- Applies an activation function
  - detailed next

Dropout

- Applies Dropout to the input
- randomly setting a fraction p of input units to 0
- prevent overfitting

Flatten

- Flattens the input

- many more

Activation Function: Sigmoid

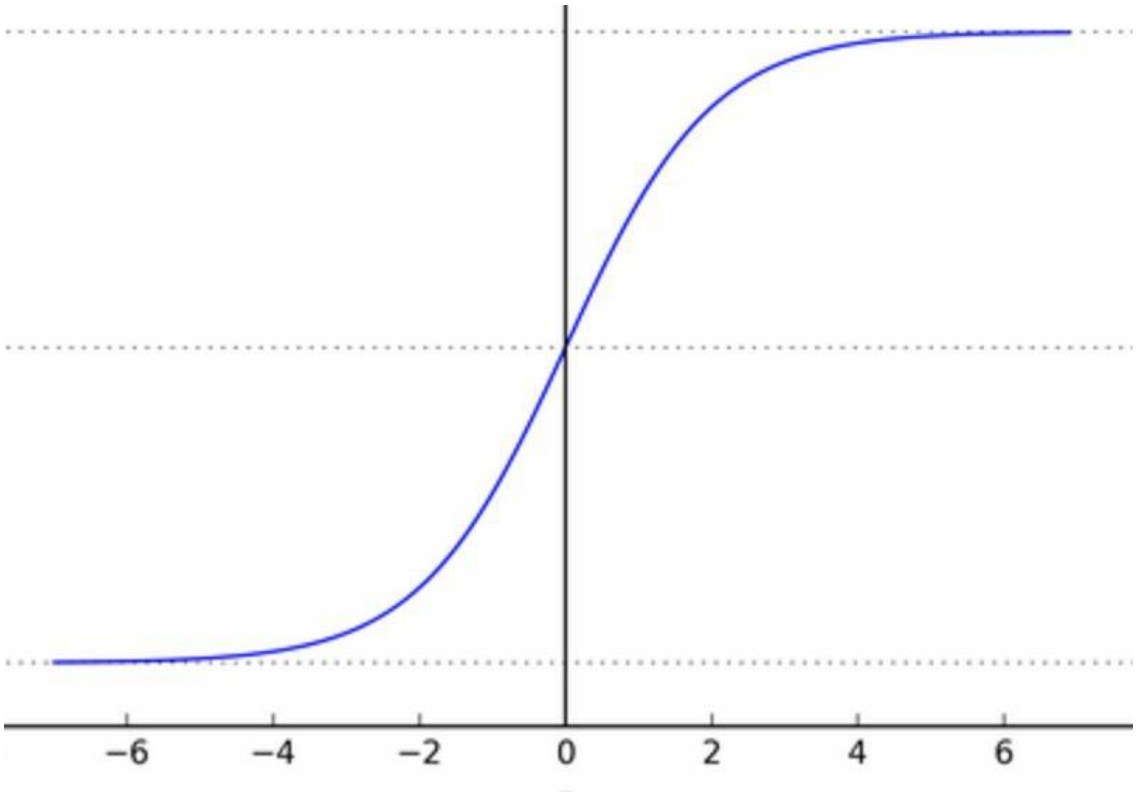


Figure: Sigmoid Function

$$\sigma(x) = \frac{1}{1 + e^{-x}} \qquad (1)$$

Activation Function: ReLU (rectified linear unit)

Figure: ReLU

$$f(x) = max(0, x) \quad (2)$$

Activation Function: softmax

- usually used on the output layer to turn the outputs into probability-like values
- Sigmoid: two class
- softmax: multiclass

$$\sigma(z)_i = \frac{e^{z_i}}{\sum_{j=1}^{K} e^{z_j}} \tag{3}$$

for i=1 to K and K is number of output units in output layer

Activation Function

linear

$$f(x) = x \tag{4}$$

- and many more...

Keras provides

Optimizer

- the specific algorithm used to update weights while we train our model
- such as *sgd* (Stochastic gradient descent optimizer)

objective function or loss function

- used by the optimizer to navigate the space of weights
- such as *mse* (mean squared error)

metrics

- used to judge the performance of your model
- such as *accuracy*

API

- Keras provides nice API
- documentation
- few with small example

Building a Simple Deep Learning Network Using Keras

Steps

- Import libraries and modules
- Load image data
- Preprocess data
- Define model architecture
- Compile model
- Fit and evaluate Model
- Improvements

Keras examples

- mnist

Building a Simple Deep Learning Network Using Keras

Import libraries and modules

- as in example file

Dataset

- Keras provides in-built support to many datasets
- such as MNIST
  - database of handwritten digits
  - used extensively in optical character recognition and machine learning research
  - training set of 60,000 examples, and a test set of 10,000 examples
  - digits have been size-normalized and centered in a fixed-size image
  - black and white digits
  - 28×28 pixels
  - Keras provides method to load MNIST data set (example file)

Building a Simple Deep Learning Network Using Keras

Preprocessing input data for Keras

- With Theano backend, the depth of the input image must be declared explicitly
- MNIST images have a depth of 1
- Also, convert data type to *float32* and normalize values
- as in example

Preprocessing class labels for Keras

- 10 different classes, one for each digit
- as in example

Building a Simple Deep Learning Network Using Keras

Model Architecture

- Usually most time consuming
- Use sequential model
- a Sequential model is declared as
  model = Sequential()
- Adding layers
  - The model needs to know what input shape it should expect
  - first layer in a Sequential model (and only the first, because following layers can do automatic shape inference) needs to receive information about its input shape
  - Dense(32, input_dim=784) specifies that it is
    - first (input) layer
    - output dimension is 32 (1^{st} argument
    - input dimension is 784
    - If no activation function specified, no activation is applied (ie. "linear" activation: a(x) = x).

Model architecture

- one hidden layer with the same number of neurons as there are inputs (784)
- init: name of initialization function for the weights of the layer. normal for values randomly drawn from normal distribution.
- there are many other initializations available in Keras
- rectifier activation function is used for the neurons in the hidden layer
- softmax activation function is used on the output layer to turn the outputs into probability-like values and allow one class of the 10 to be selected as the models output prediction

Compile model

- Before training, configure the learning process, using compile() method. Three argements:
  - optimizer: ANN training process is an optimization task with the aim of finding a set of weights to minimize some objective function
  - loss function: the objective function that model try to minimize
  - list of metrics: used to judge performance of model, similar to objective function however not used for training purpose
- Logarithmic loss is used as the loss function
- ADAM gradient descent algorithm is used to learn the weights

Train and Evaluate model

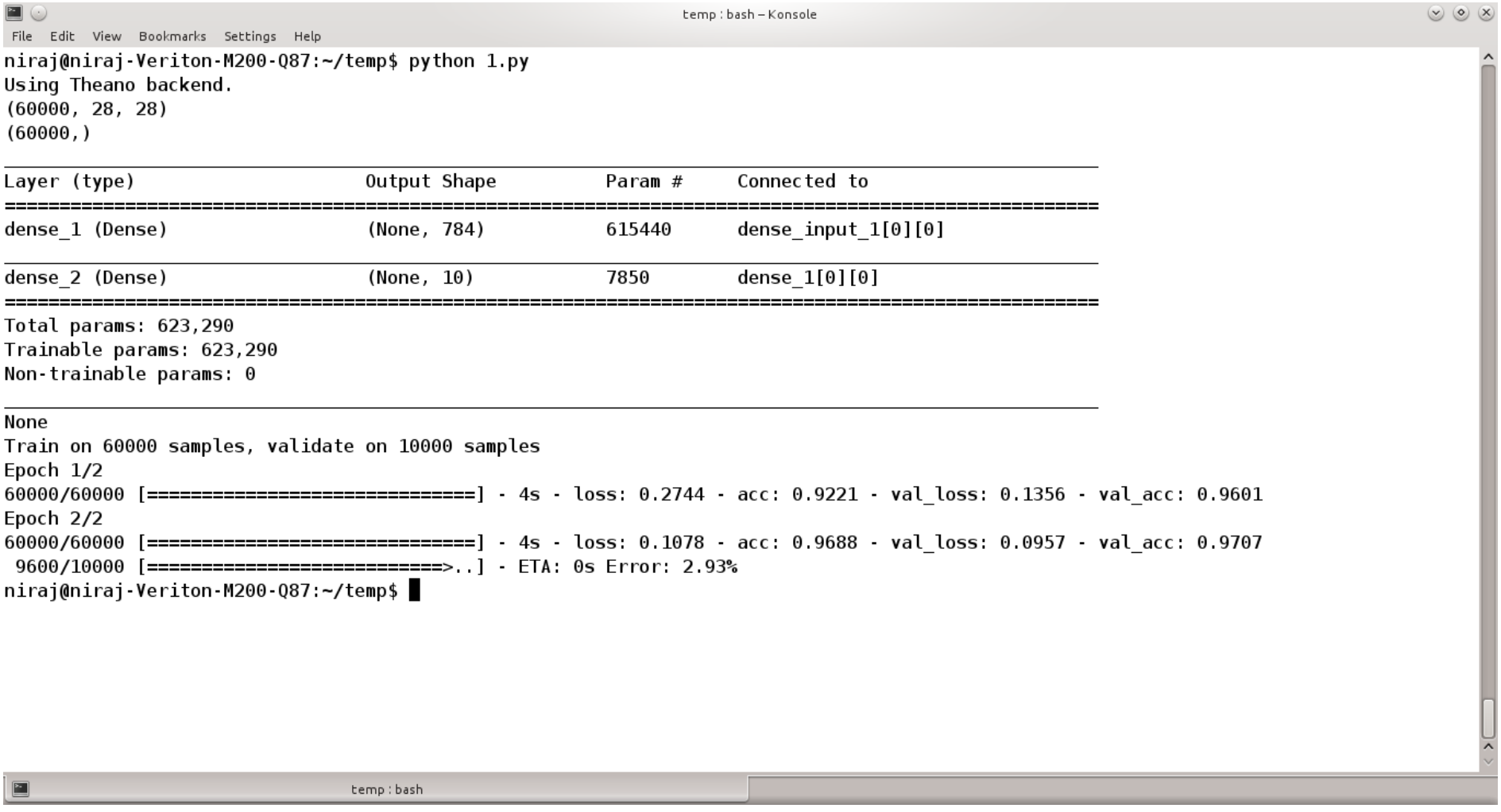
Train model

using fit() function

Evaluate model on test data

using evaluate() function

Performance of Simple Network



Improving Performance of Simple Network: additional hidden layers

temp : bash – Konsole

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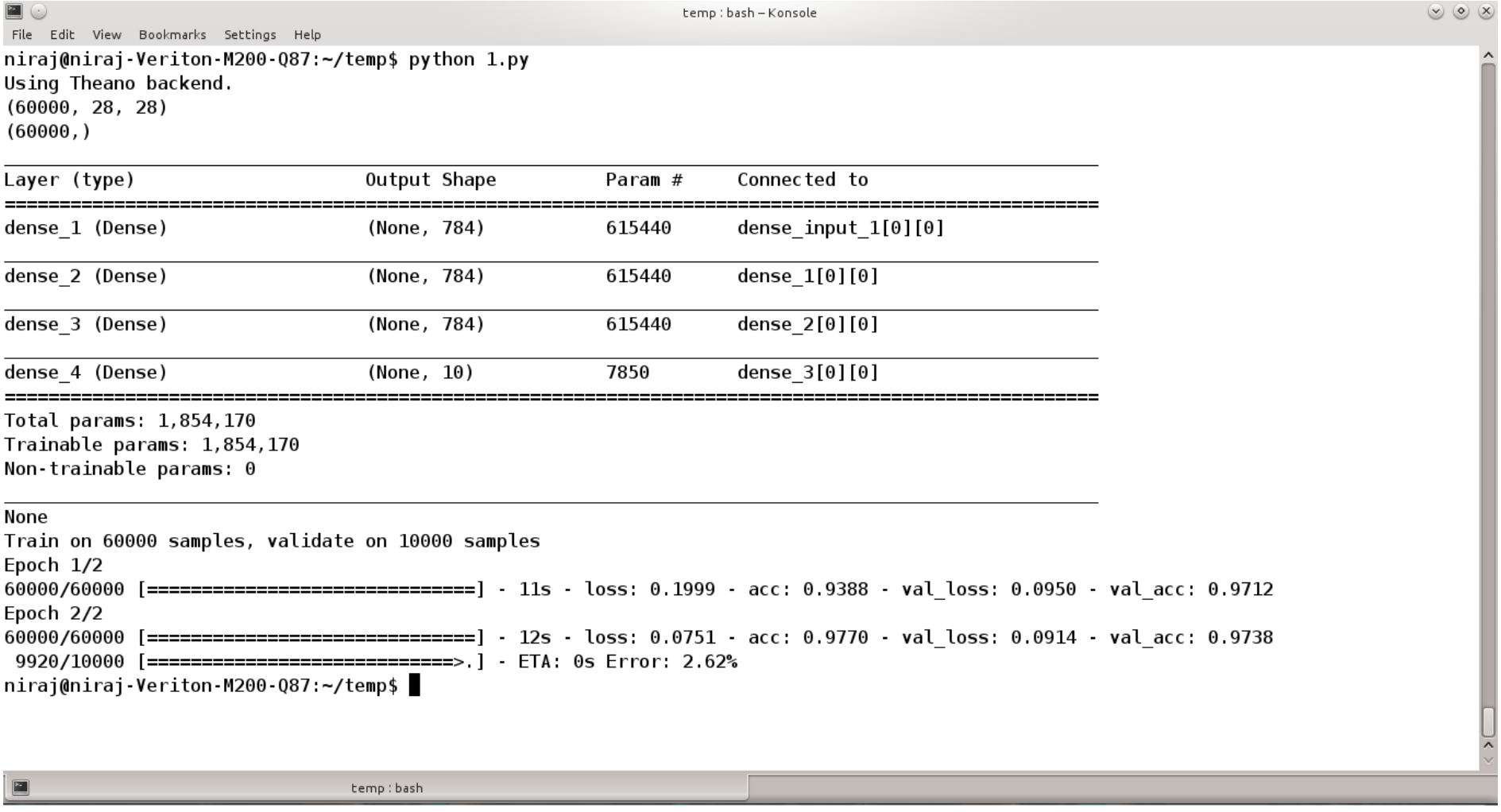
```
niraj@niraj-Veriton-M200-Q87:~/temp$ python 1.py
Using Theano backend.
(60000, 28, 28)
(60000,)
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
dense_1 (Dense)              (None, 784)               615440
_________________________________________________________________
dense_2 (Dense)              (None, 784)               615440
_________________________________________________________________
dense_3 (Dense)              (None, 10)                7850
=================================================================
Total params: 1,238,730
Trainable params: 1,238,730
Non-trainable params: 0
_________________________________________________________________
None
Train on 60000 samples, validate on 10000 samples
Epoch 1/2
60000/60000 [==============================] - 8s - loss: 0.2184 - acc: 0.9354 - val_loss: 0.1094 - val_acc: 0.9639
Epoch 2/2
60000/60000 [==============================] - 8s - loss: 0.0755 - acc: 0.9767 - val_loss: 0.0852 - val_acc: 0.9720
 9824/10000 [============================>.] - ETA: 0s Error: 2.80%
niraj@niraj-Veriton-M200-Q87:~/temp$
```

temp : bash

Improving Performance of Simple Network: additional hidden layers



Improving Performance of Simple Network: introducing dropout layer

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```
niraj@niraj-Veriton-M200-Q87:~/temp$ python 1.py
Using Theano backend.
(60000, 28, 28)
(60000,)

_________________________________________________________________
Layer (type)                 Output Shape              Param #
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_________________________________________________________________
dense_2 (Dense)              (None, 784)               615440
_________________________________________________________________
dense_3 (Dense)              (None, 784)               615440
_________________________________________________________________
dropout_1 (Dropout)          (None, 784)               0
_________________________________________________________________
dense_4 (Dense)              (None, 10)                7850
=================================================================
Total params: 1,854,170
Trainable params: 1,854,170
Non-trainable params: 0
_________________________________________________________________
None
Train on 60000 samples, validate on 10000 samples
Epoch 1/2
60000/60000 [==============================] - 13s - loss: 0.2014 - acc: 0.9386 - val_loss: 0.1017 - val_acc: 0.9697
Epoch 2/2
60000/60000 [==============================] - 14s - loss: 0.0771 - acc: 0.9760 - val_loss: 0.0811 - val_acc: 0.9740
10000/10000 [==============================] - 1s
 Error: 2.60%
niraj@niraj-Veriton-M200-Q87:~/temp$
```
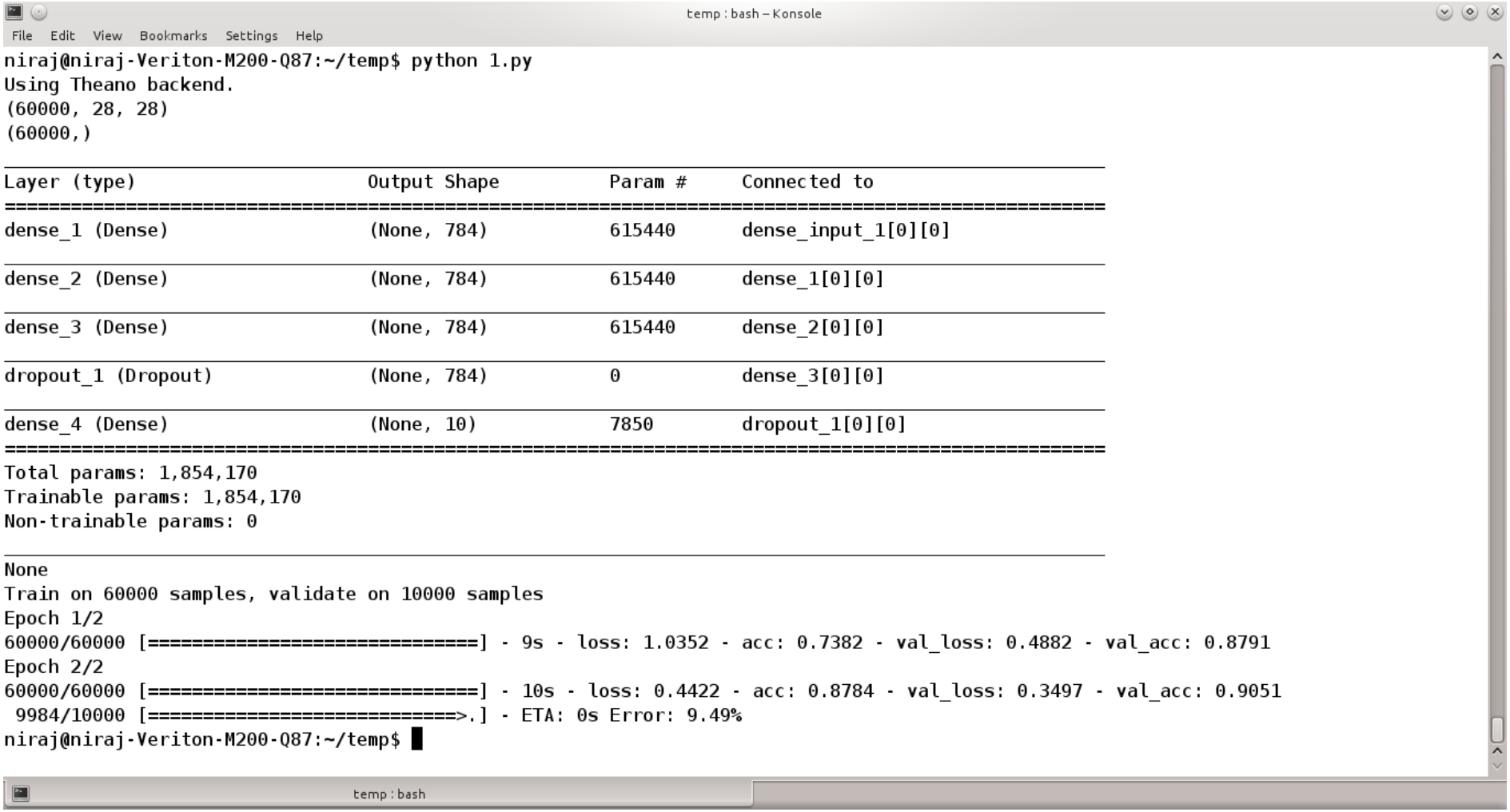

temp : bash

Improving Performance of Simple Network: using different optimizers



Improving Performance of Simple Network: training for more number of epochs

temp : bash – Konsole

File Edit View Bookmarks Settings Help

```
niraj@niraj-Veriton-M200-Q87:~/temp$ python 1.py
Using Theano backend.
(60000, 28, 28)
(60000,)

_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
dense_1 (Dense)              (None, 784)               615440
_________________________________________________________________
dense_2 (Dense)              (None, 784)               615440
_________________________________________________________________
dense_3 (Dense)              (None, 784)               615440
_________________________________________________________________
dropout_1 (Dropout)          (None, 784)               0
_________________________________________________________________
dense_4 (Dense)              (None, 10)                7850
=================================================================
Total params: 1,854,170
Trainable params: 1,854,170
Non-trainable params: 0
_________________________________________________________________
None
Train on 60000 samples, validate on 10000 samples
Epoch 1/20
60000/60000 [==============================] - 10s - loss: 1.0537 - acc: 0.7378 - val_loss: 0.4933 - val_acc: 0.8815
Epoch 2/20
60000/60000 [==============================] - 9s - loss: 0.4407 - acc: 0.8813 - val_loss: 0.3505 - val_acc: 0.9054
Epoch 3/20
60000/60000 [==============================] - 10s - loss: 0.3512 - acc: 0.9016 - val_loss: 0.3008 - val_acc: 0.9187
Epoch 4/20
```

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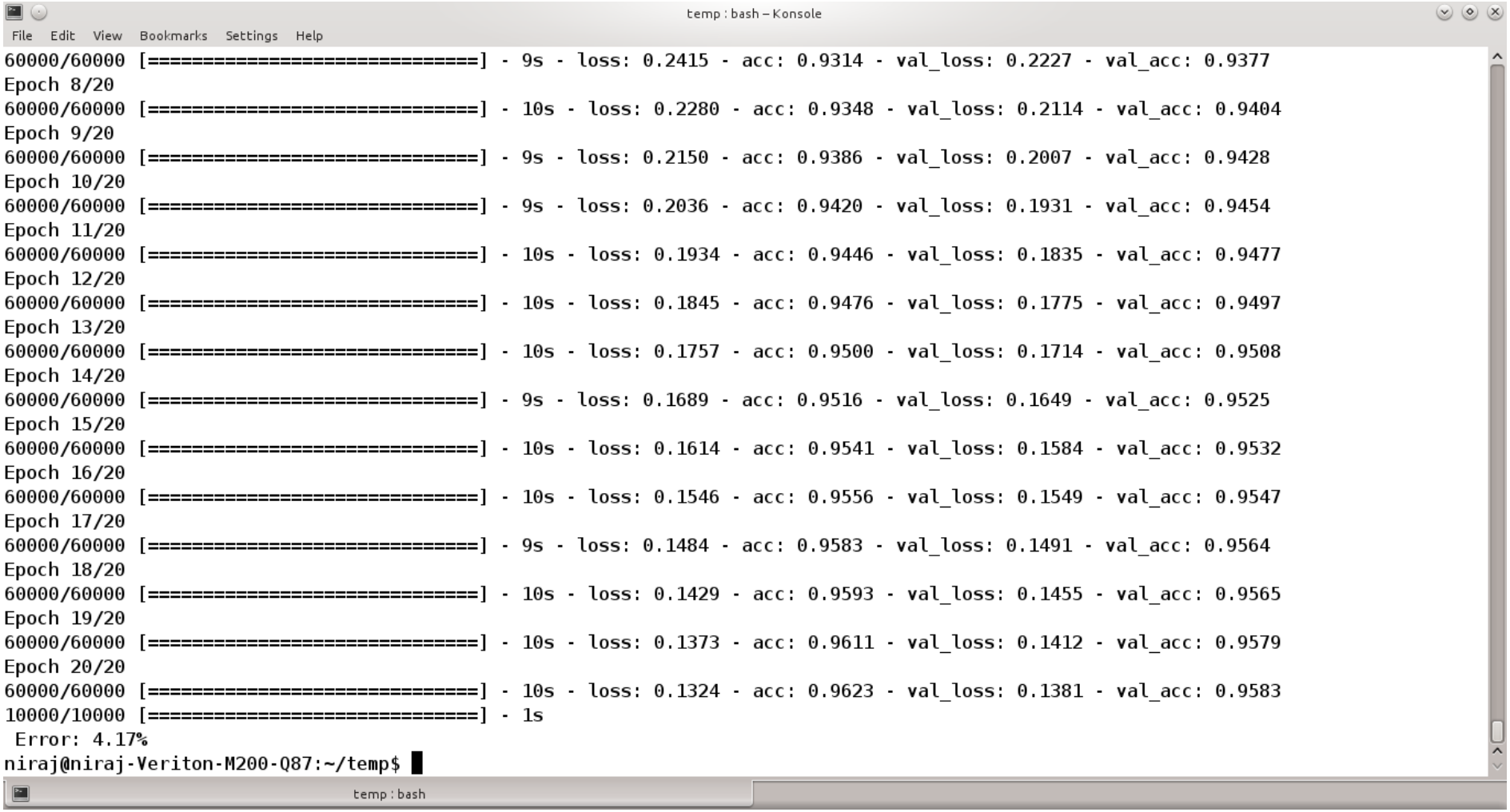
Improving Performance of Simple Network: training for more number of epochs



Improving Performance of Simple Network

other options to explore

- different learning rate for optimizer
- number of neurons in hidden layer
- batch size
- with additional hidden layers
- with dropout
- with different optimizers
- with more number of epochs
- Controlling the optimizer learning rate
- Increasing the number of internal hidden neurons
- Increasing the size of batch computation

Constructing the Right Network

steps to follow to make an efficient image classifier?

- lot of experimentation and testing to get the optimal structure and parameters

Important Links

Links

1. Keras Official Documentation Page
2. keras official github
3. Keras GitHub page
4. Another GitHub Page
5. GitHub Page MNIST example
6. Keras Tutorial
7. An Example
8. Another Example
9. Deep Learning with Keras (Book)

The End