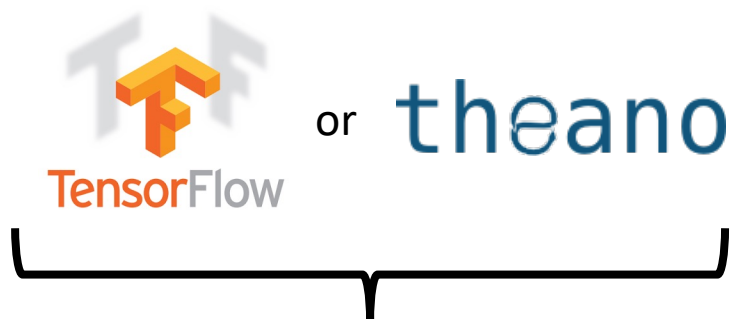


Keras



Keras



Very flexible
Need some
effort to learn

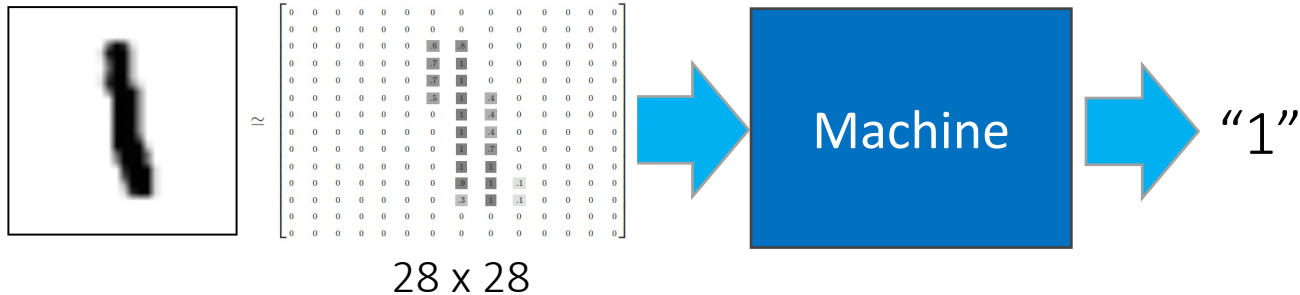
Interface of
TensorFlow
or Theano



Easy to learn and use
(still have some flexibility)
You can modify it if you can write
TensorFlow or Theano

“Hello World”

Handwriting Digit Recognition



MNIST Data: <http://yann.lecun.com/exdb/mnist/>

Keras provides data sets loading function: <http://keras.io/datasets/>

Keras: Part I

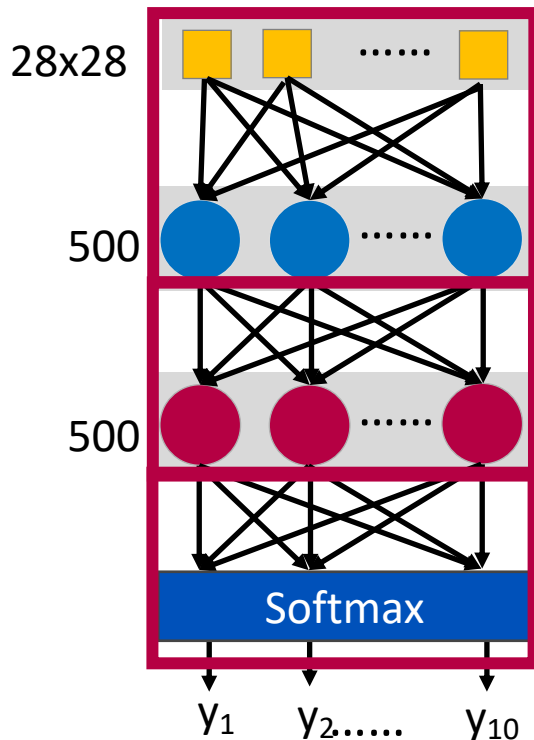
Step 1:
define a set
of function



Step 2:
goodness of
function



Step 3: pick
the best
function



```
model = Sequential()
```

```
model.add( Dense( input_dim=28*28,  
                  output_dim=500 ) )  
model.add( Activation('sigmoid') )
```

softplus, softsign, relu, tanh,
hard_sigmoid, linear

```
model.add( Dense( output_dim=500 ) )  
model.add( Activation('sigmoid') )
```

```
model.add( Dense( output_dim=10 ) )  
model.add( Activation('softmax') )
```

Keras: Part II

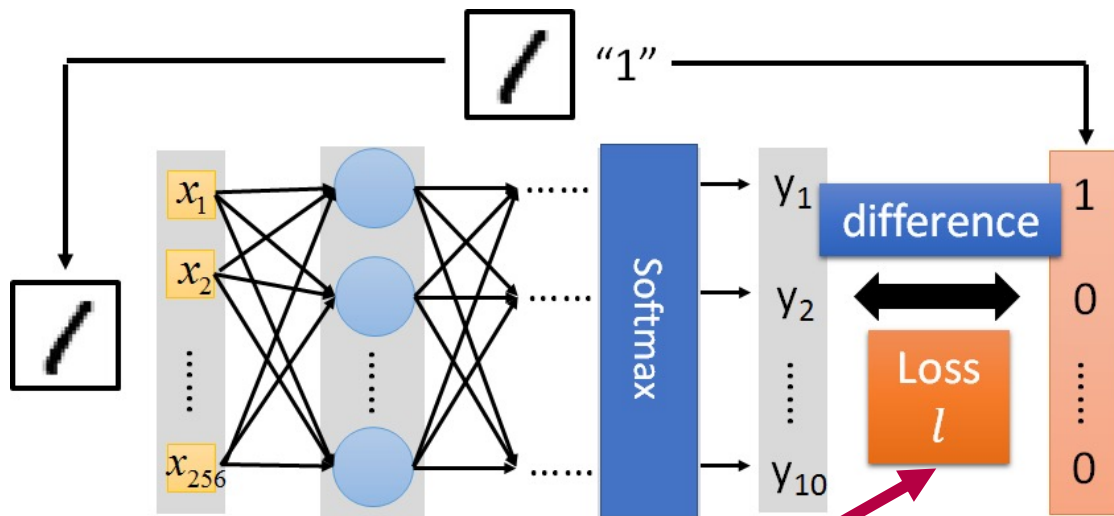
Step 1:
define a set
of function



Step 2:
goodness of
function



Step 3: pick
the best
function



```
model.compile(loss='categorical_crossentropy',  
              optimizer='adam',  
              metrics=['accuracy'])
```

Several alternatives: <https://keras.io/objectives/>



Keras: Part III



Step 3.1: Configuration

```
model.compile(loss='categorical_crossentropy',  
              optimizer='adam',  
              metrics=['accuracy'])
```

SGD, RMSprop, Adagrad, Adadelata, Adam, Adamax, Nadam

Step 3.2: Find the optimal network parameters

```
model.fit(x_train, y_train, batch_size=100, nb_epoch=20)
```

Training data
(Images)

Labels (digits)

In the following slides



Keras: Part IV

Step 1:
define a set
of function



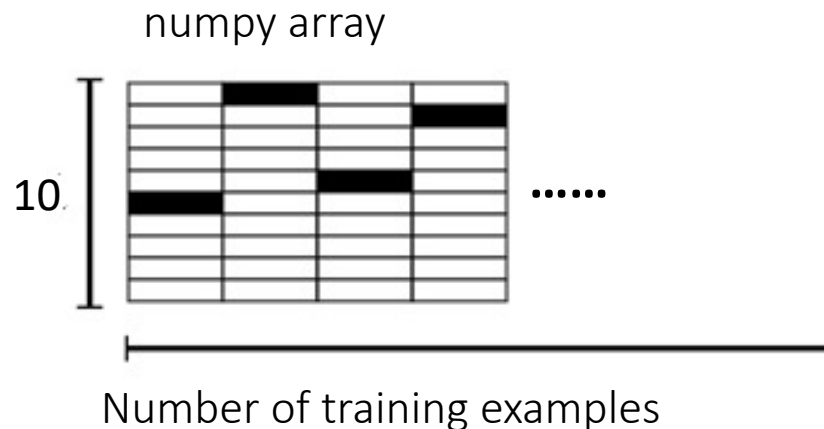
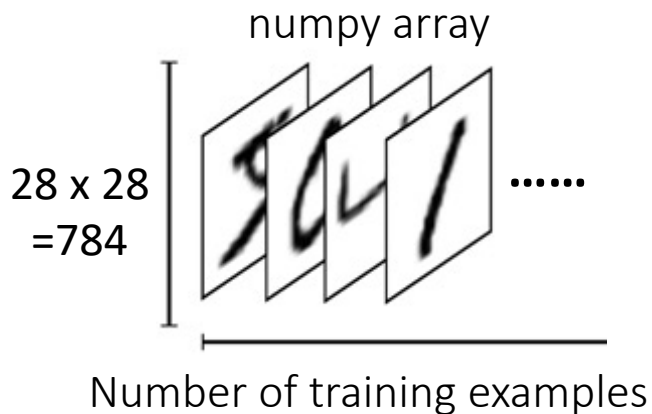
Step 2:
goodness of
function



Step 3: pick
the best
function

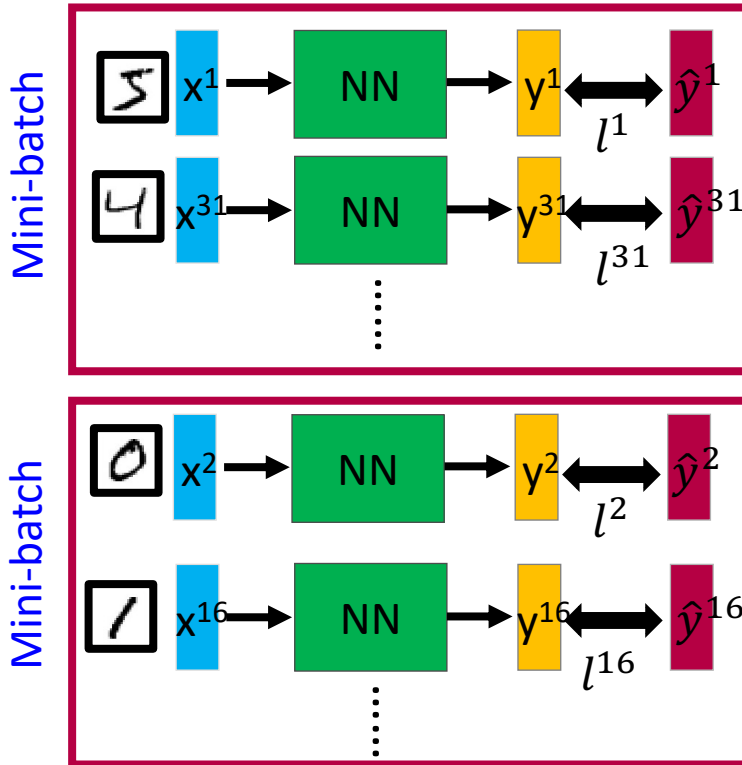
Step 3.2: Find the optimal network parameters

```
model.fit(x_train, y_train, batch_size=100, nb_epoch=20)
```



We do not really minimize total loss!

Mini-Batch



- Randomly initialize network parameters

- Pick the 1st batch
 $L' = l^1 + l^{31} + \dots$
Update parameters once
- Pick the 2nd batch
 $L'' = l^2 + l^{16} + \dots$
Update parameters once
 \vdots
- Until all mini-batches have been picked

one epoch

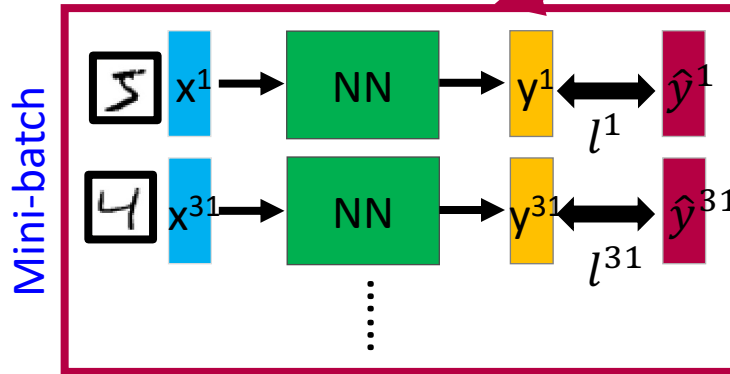
Repeat the above process



Mini-Batch, Continued

Batch size influences both *speed* and *performance*. You have to tune it.

```
model.fit(x_train, y_train, batch_size=100, nb_epoch=20)
```



100 examples in a mini-batch
Batch size = 1
Stochastic gradient descent

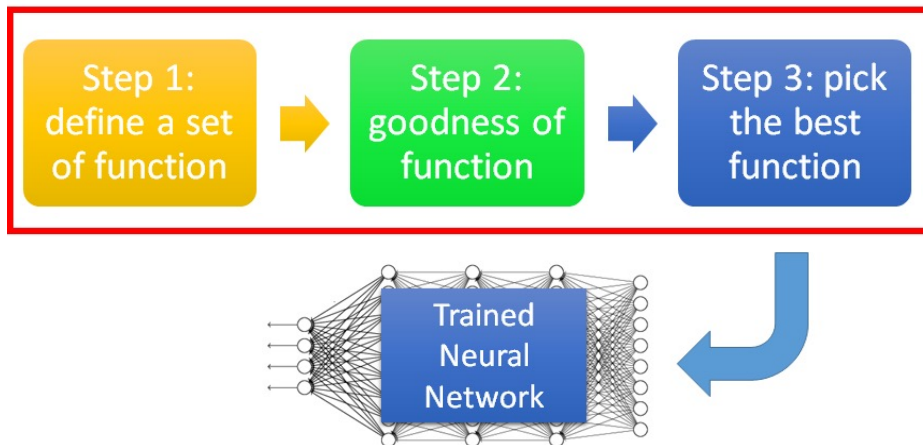
- Pick the 1st batch
 $L' = l^1 + l^{31} + \dots$
Update parameters once
- Pick the 2nd batch
 $L'' = l^2 + l^{16} + \dots$
Update parameters once
- ⋮
- Until all mini-batches have been picked

Repeat 20 times

one epoch



Keras: Part V



How to use the neural network (testing):

case 1:

```
score = model.evaluate(x_test,y_test)
print('Total loss on Testing Set:', score[0])
print('Accuracy of Testing Set:', score[1])
```

case 2:

```
result = model.predict(x_test)
```