

# AI for Good Workshop

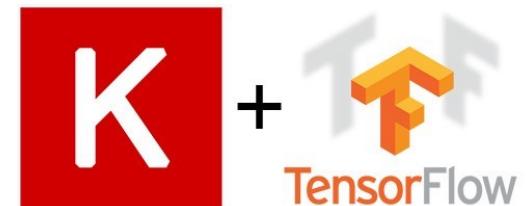
March 2019

Session 4

# Artificial Intelligence for everyone

Build your 1<sup>st</sup> solution on  
Deep Learning

<https://sites.google.com/view/AlforEveryone>



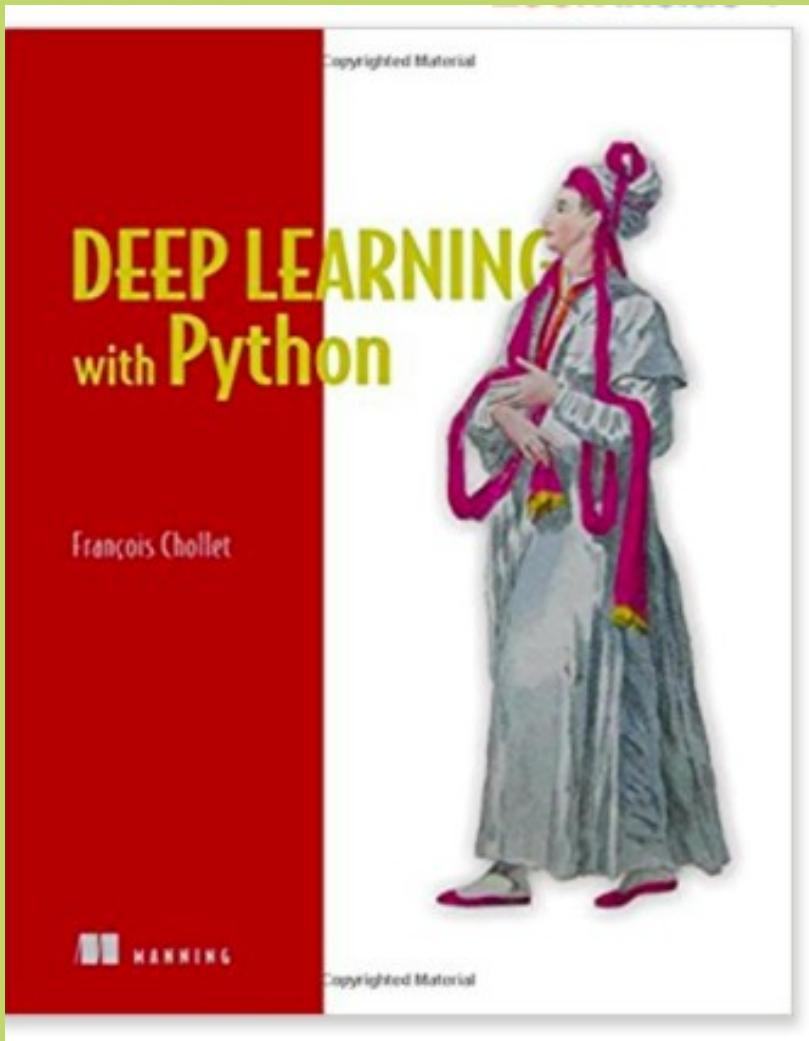
# Topics in this video

Jump start your AI journey !

1. Define a purpose
2. Apply the Strategy to jump start
3. Hands-on : Create your 1<sup>st</sup> Deep Learning solution



# The inspiration



Francois Chollet

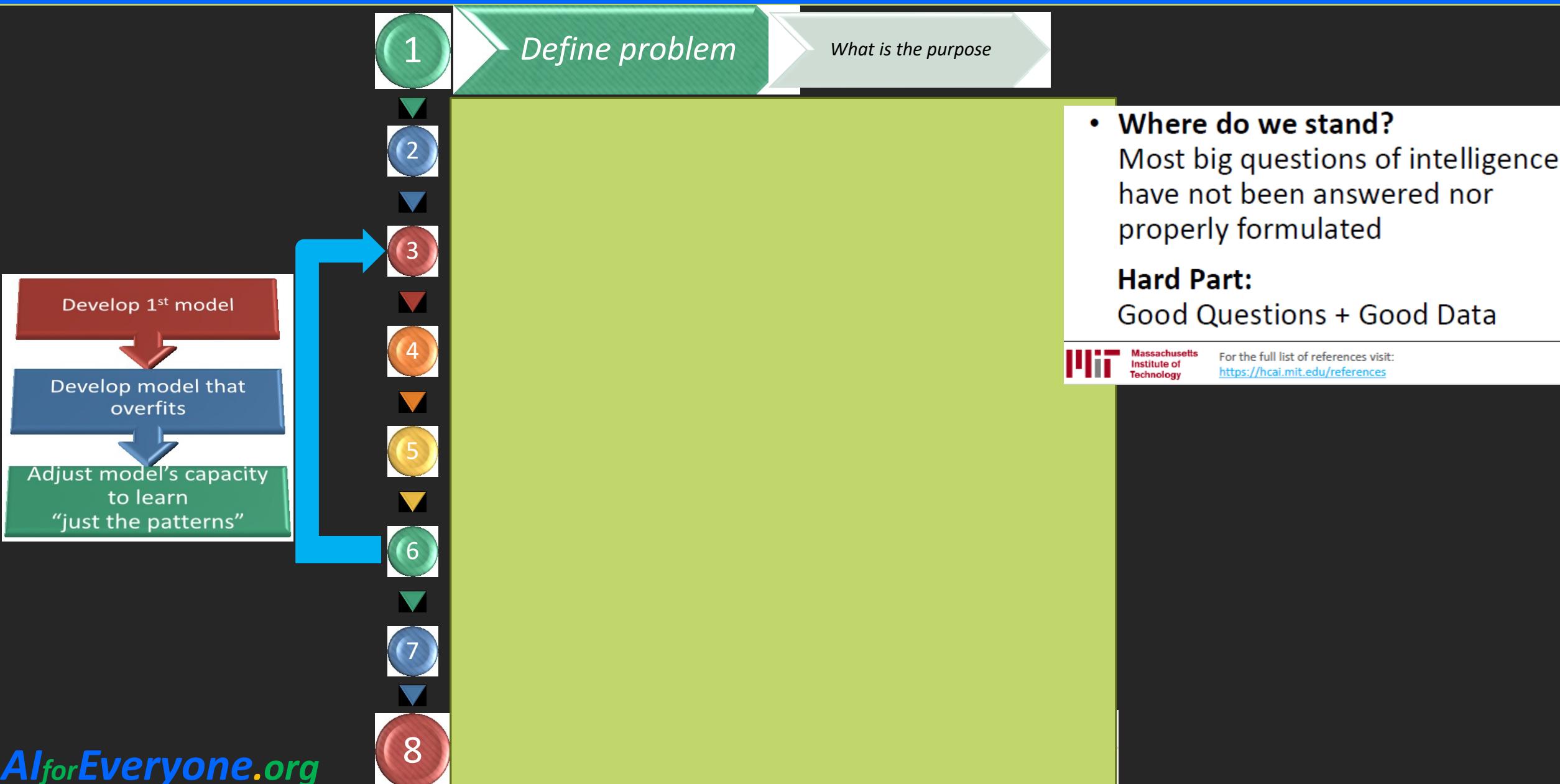
Artificial Intelligence Researcher,  
**Google**

**Deep Learning with Python**

<https://www.manning.com/books/deep-learning-with-python>

**ISBN-13:** 978-1617294433

# Design of Deep Learning experiment



# Define the idea

- App idea: Recognize your hand gesture to unlock your smartphone (Secure unlock)
- Problem: Recognize hand draw gestures to unlock your phone



# Regression vs Classification



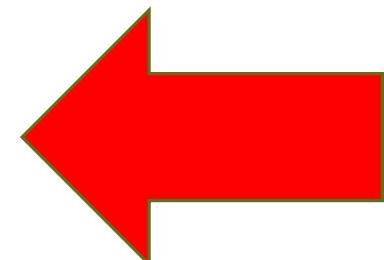
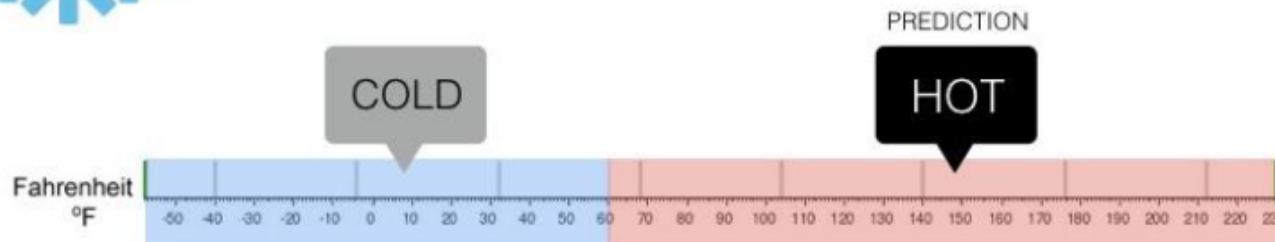
## Regression

What is the temperature going to be tomorrow?

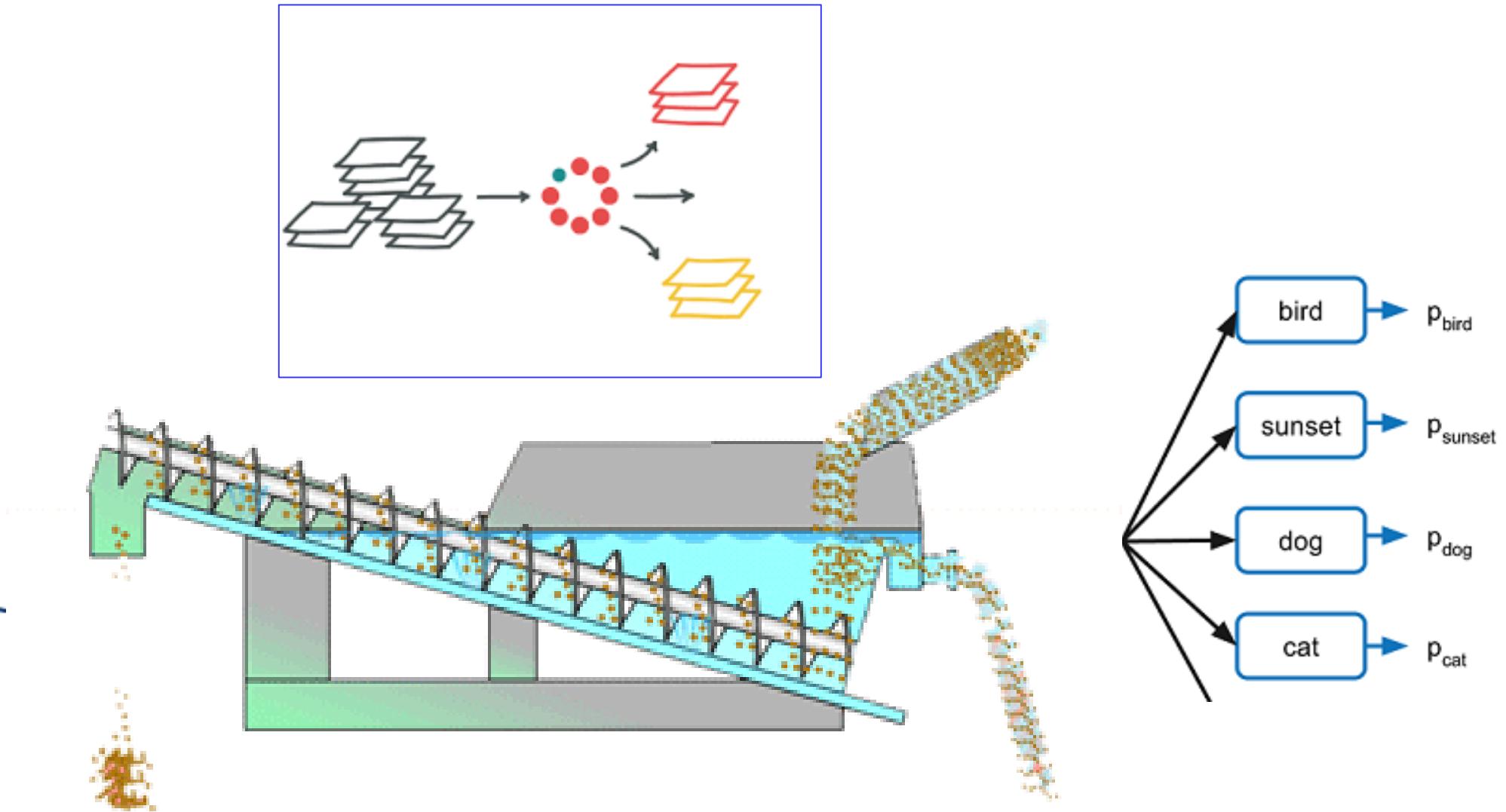


## Classification

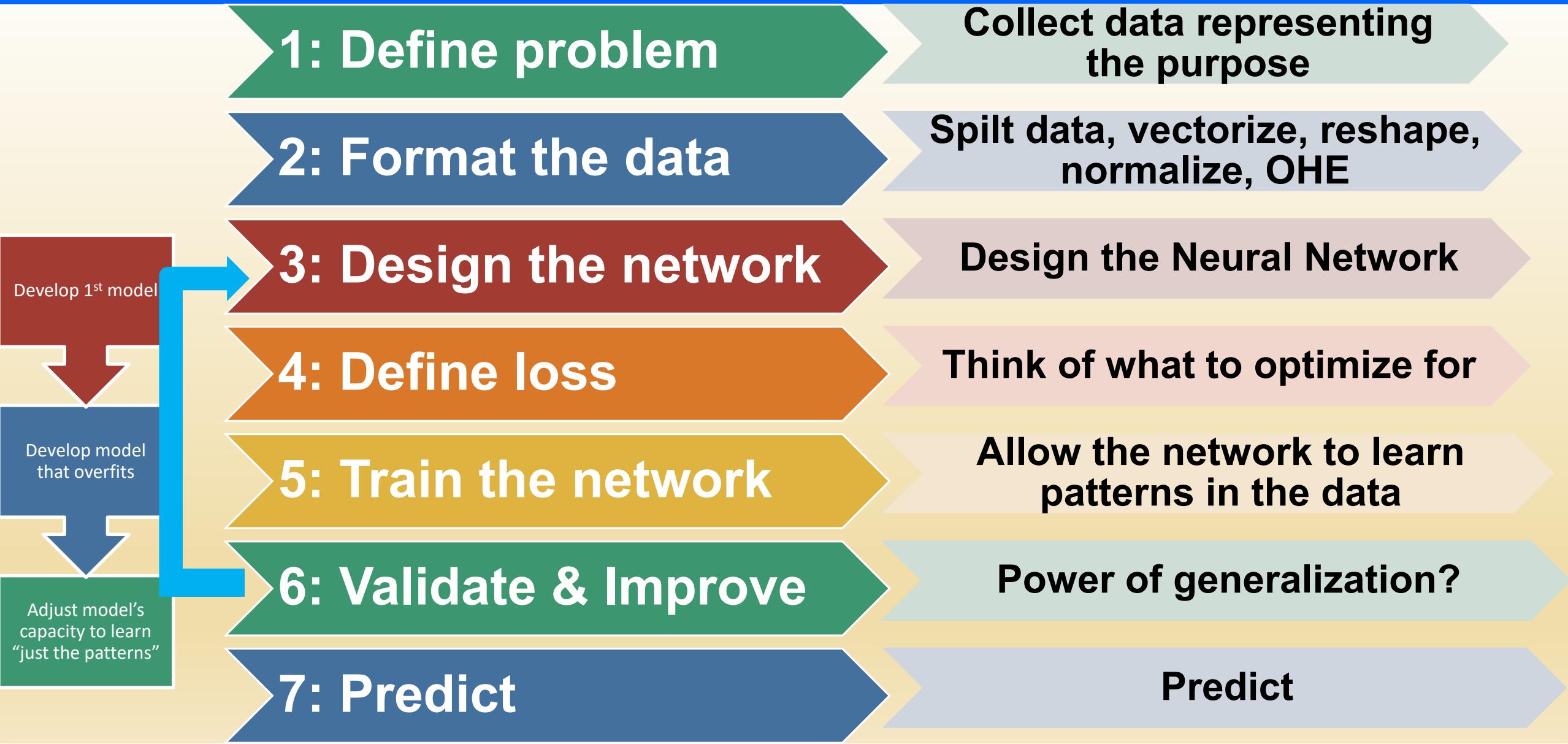
Will it be Cold or Hot tomorrow?



# visual recognition – image classification



# Method: 7 steps to develop your Deep Learning model



A → B



$X \rightarrow Y$

$X = \text{image}$

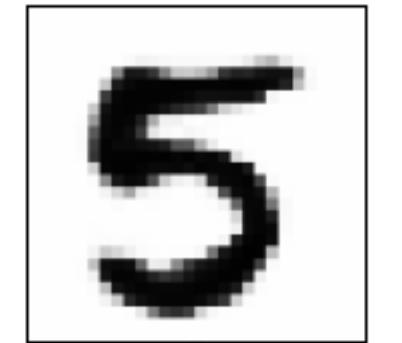


Input Image:

$Y = \text{Category of image}$

TensorFlow  
Model:

Output:



Neural  
Network

5

# Simplify your goal

Problem: Users draws a hand drawn gestures, your app identifies it.

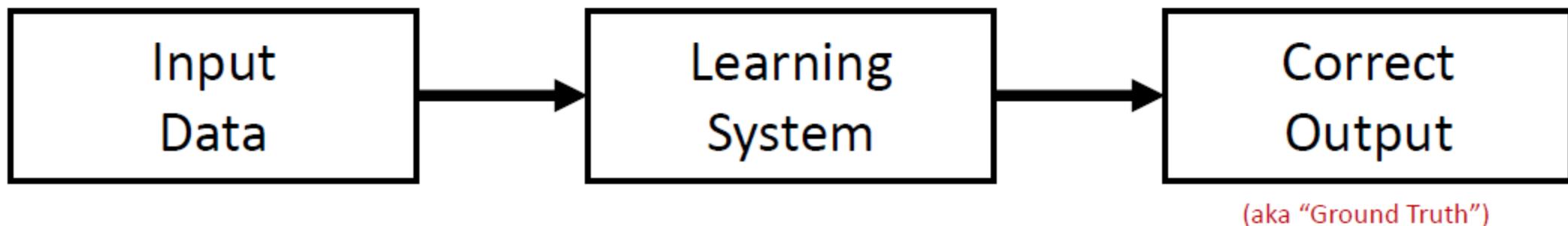
Simplified Problem: User can draw any number between 0 to 9



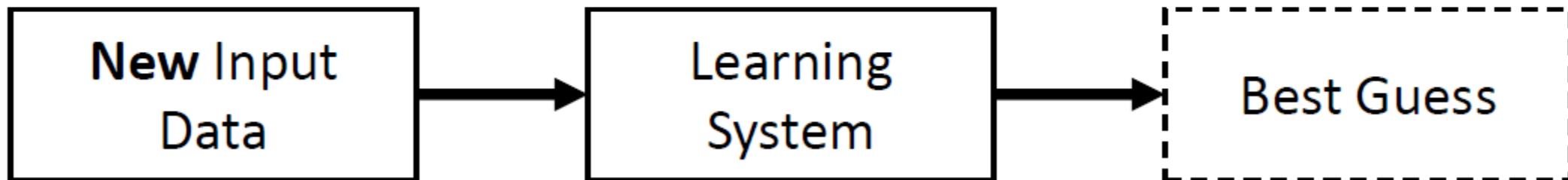
label = 5	label = 0	label = 4	label = 1	label = 9
5	0	4	1	9
label = 2	label = 1	label = 3	label = 1	label = 4
2	1	3	1	4
label = 3	label = 5	label = 3	label = 6	label = 1
3	5	3	6	1
label = 7	label = 2	label = 8	label = 6	label = 9
7	2	8	6	9

# Deep Learning: Training and Testing

## Training Stage:



## Testing Stage:



# The Challenge of Deep Learning: Efficient Teaching + Efficient Learning

- Humans can learn from very few examples
- Machines (in most cases) need thousands/millions of examples



# 1. Define problem

Collect the data representing the world you want to predict

Load the data

Visualize data

label = 5



label = 0



label = 4



label = 1



label = 9



label = 2



label = 1



label = 3



label = 1



label = 4



label = 3



label = 5



label = 3



label = 6



label = 1



label = 7



label = 2



label = 8



label = 6



label = 9



## Labeled dataset

airplane



automobile



bird



cat



deer



dog



frog



horse



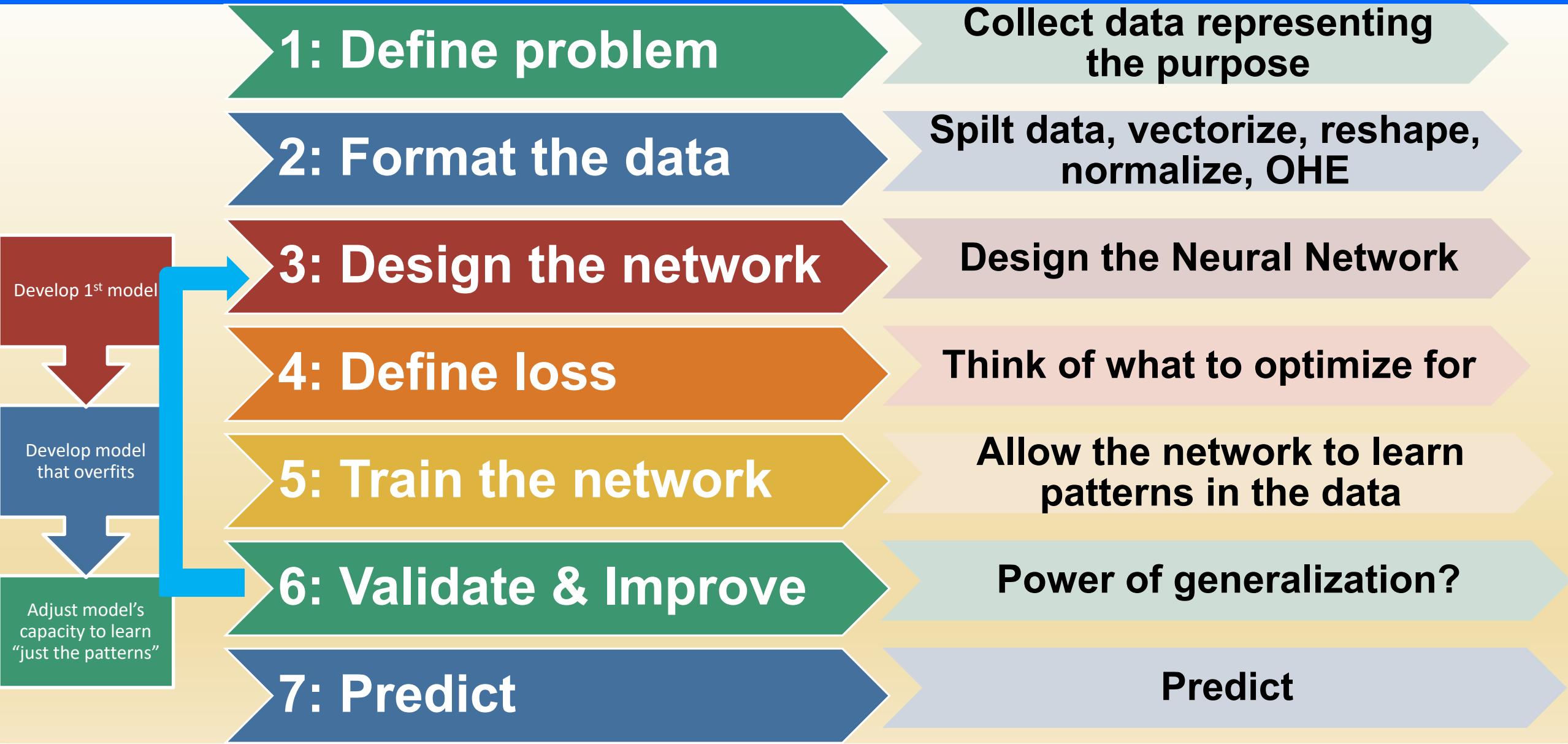
ship



truck



# Method: 7 steps to develop your Deep Learning model



# 1. Define problem

Collect the data representing the world you want to predict

Load the data

Visualize data

## `keras.datasets.mnist.load_data()`

```
from keras.datasets import mnist
```

```
(trainX, trainY), (testX, testY) = mnist.load_data()
```

```
len(trainX)
```

**type ( myObject )**

# 1. Define problem

Collect the data representing the world you want to predict

Load the data

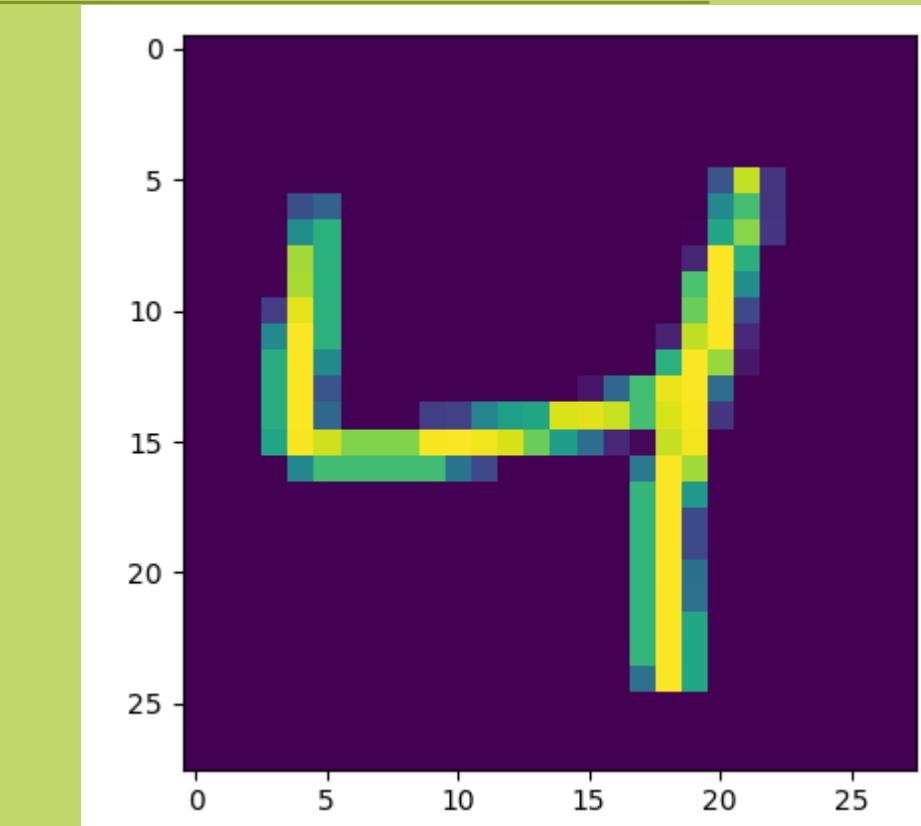
Visualize data

## matplotlib.pyplot.imshow( myArrayImage )

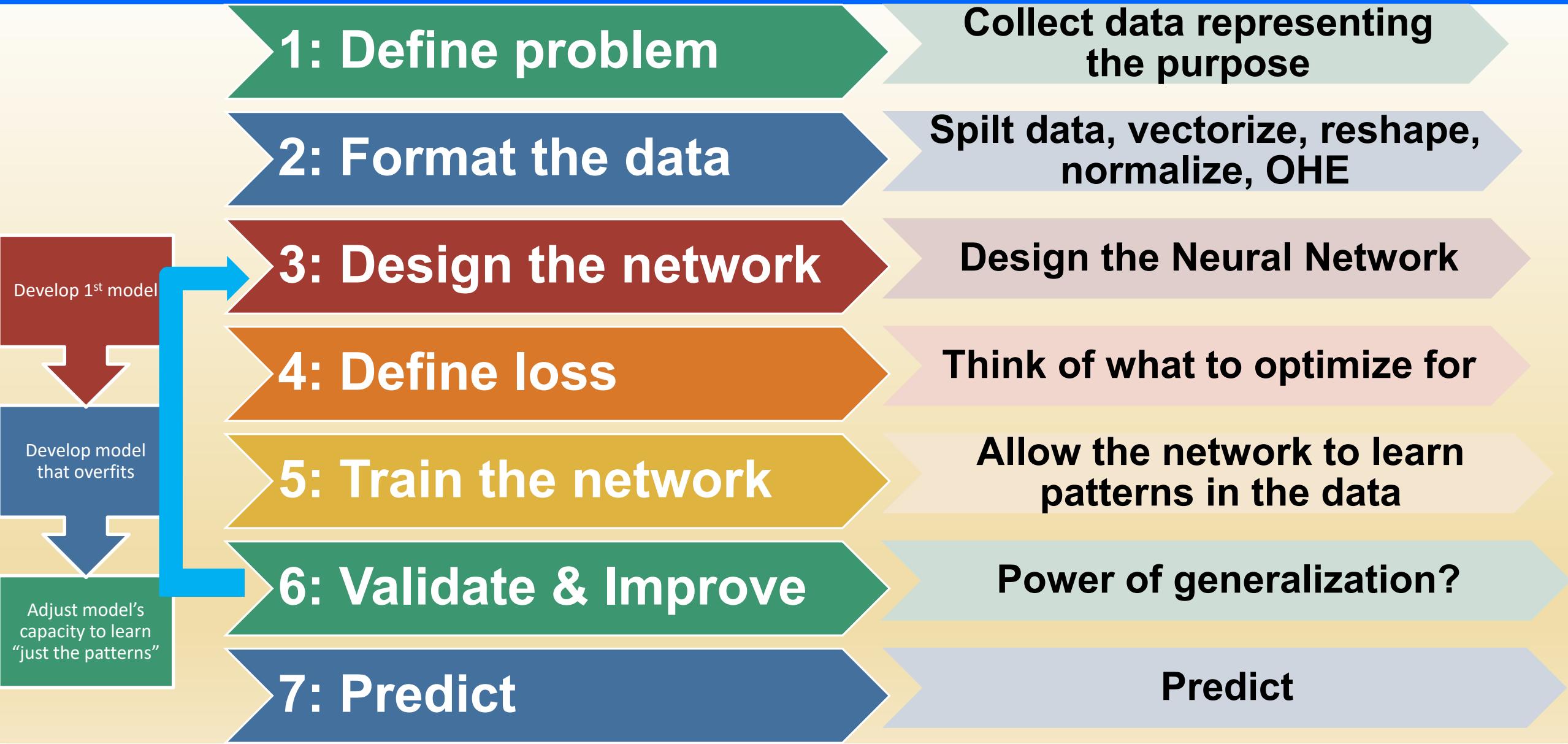
```
from matplotlib import pyplot
```

```
exampleImage = trainY[2]
```

```
pyplot.imshow(exampleImage)  
pyplot.show()
```



# Method: 7 steps to develop your Deep Learning model



## 2. Prepare data

2. Prepare  
the data

Spilt data in  
training &  
testing

Vectortize  
as tensors

Reshape

Normalize

One Hot  
encode

## 2. Prepare the data

Spilt data in  
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60000

No of training records = 60000

No of test records = 10000

Shape of training image = (60000, 28, 28)

## 2. Prepare the data

Spilt data in training & testing

Vectortize as tensors

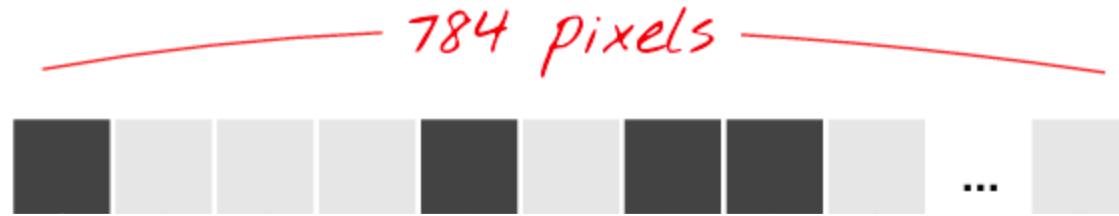
Reshape

Normalize

One Hot encode



28x28 pixels



```
newArray = myArray . reshape (28 * 28)
```

```
trainXready = trainX. reshape (800, 28*28)
```

```
trainX.shape
```

Shape of trainX: (800, 28, 28)

Shape of trainXready: (800, 784)

# Multi dimensional Array == Tensor

What\_is\_size\_of\_array = myArray.shape

1D array

7	2	9	10
---	---	---	----

axis 0 →

shape: (4,)

2D array

The diagram illustrates the transformation of a 1D array into a 2D array. On the left, a 1D array with four elements (7, 2, 9, 10) is shown. An arrow labeled "axis 0" points down to a 2D array on the right. This 2D array has two rows and three columns, with elements 5.2, 3.0, 4.5 in the top row and 9.1, 0.1, 0.3 in the bottom row.

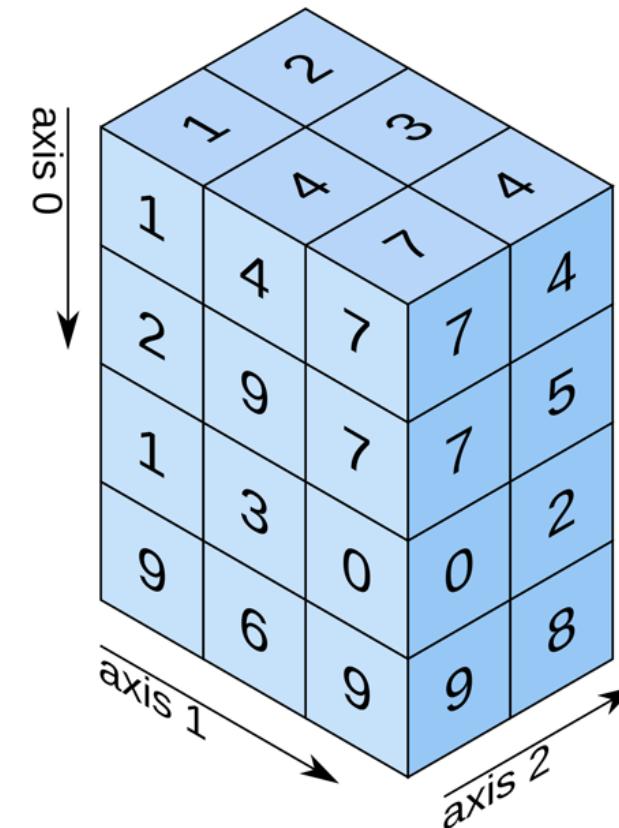
5.2	3.0	4.5
9.1	0.1	0.3

axis 0 ↓

axis 1 →

shape: (2, 3)

3D array



shape: (4, 3, 2)

# Reshape() an array

`reshape()` takes a NumPy array of one shape ...

And reconfigures it into an array with a new shape

1	2	3	4	5	6
7	8	9	10	11	12



1	2
3	4
5	6
7	8
9	10
11	12

## 2. Prepare the data

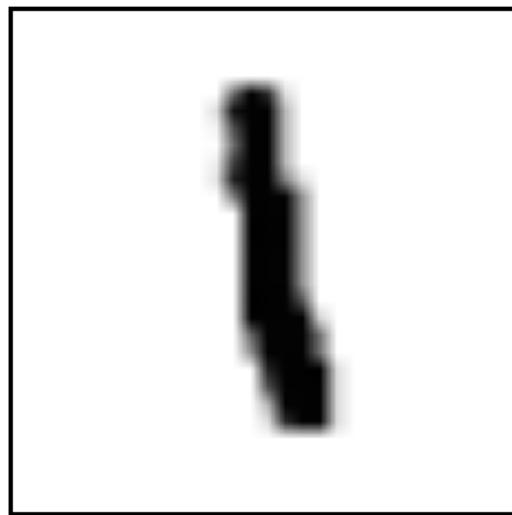
Spilt data in  
training &  
testing

Vectortize  
as tensors

Reshape

Normalize

One Hot  
encode



$$\approx \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & .6 & .8 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & .7 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & .7 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & .5 & 1 & .4 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & .4 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & .7 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & .1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & .3 & 1 & .1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

### Data Types

Boolean

Integer

Unsigned Integer

Float

Complex

String

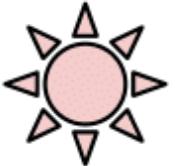
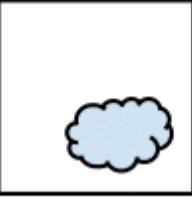
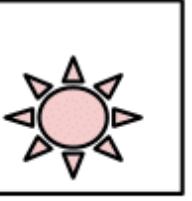
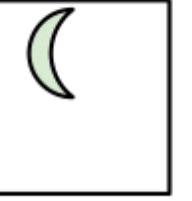
`afloat = ainteger.astype('float32')`

`newArray = myArray / 255`

`trainXready = temptrainX.astype('float32') / 255`

# One Hot Encoding

Multi-Class

$C = 3$	Samples
	
	
	

Labels ( $t$ )

[0 0 1] [1 0 0] [0 1 0]

`yNew = to_categorical (y)`

Data:  
[2]

Data after **One Hot Encoding**:  
[[ 0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]]



2. Prepare  
the data

Spilt data in  
training &  
testing

Vectortize  
as tensors

Reshape

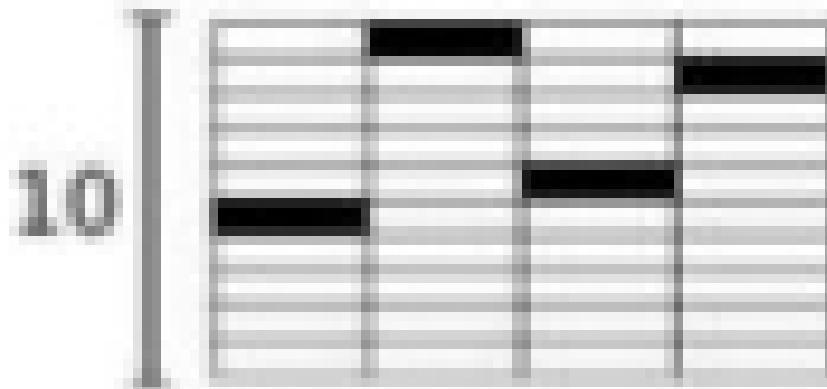
Normalize

One Hot  
encode

**myArray = keras.utils. to\_categorical ( alnteger)**

Data: [2]

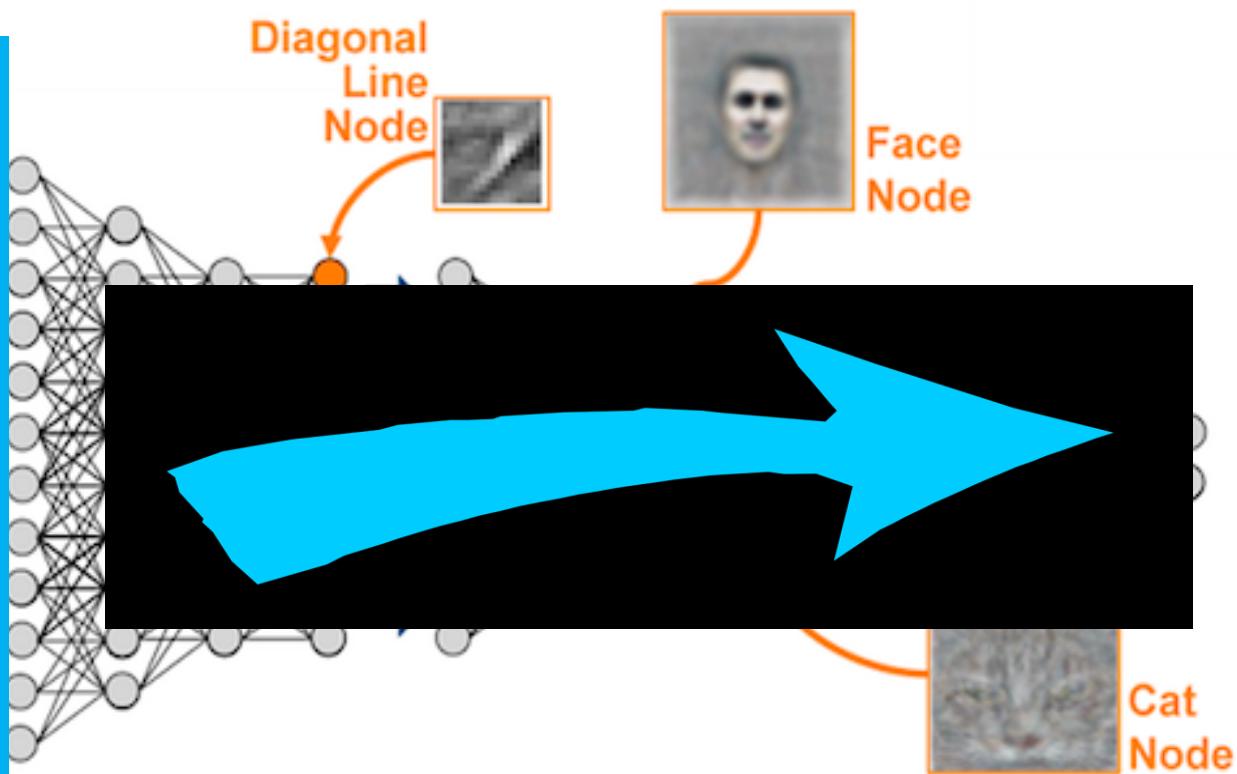
Data after OHE : [[ 0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]]



40000

# 3. Design the network

Data



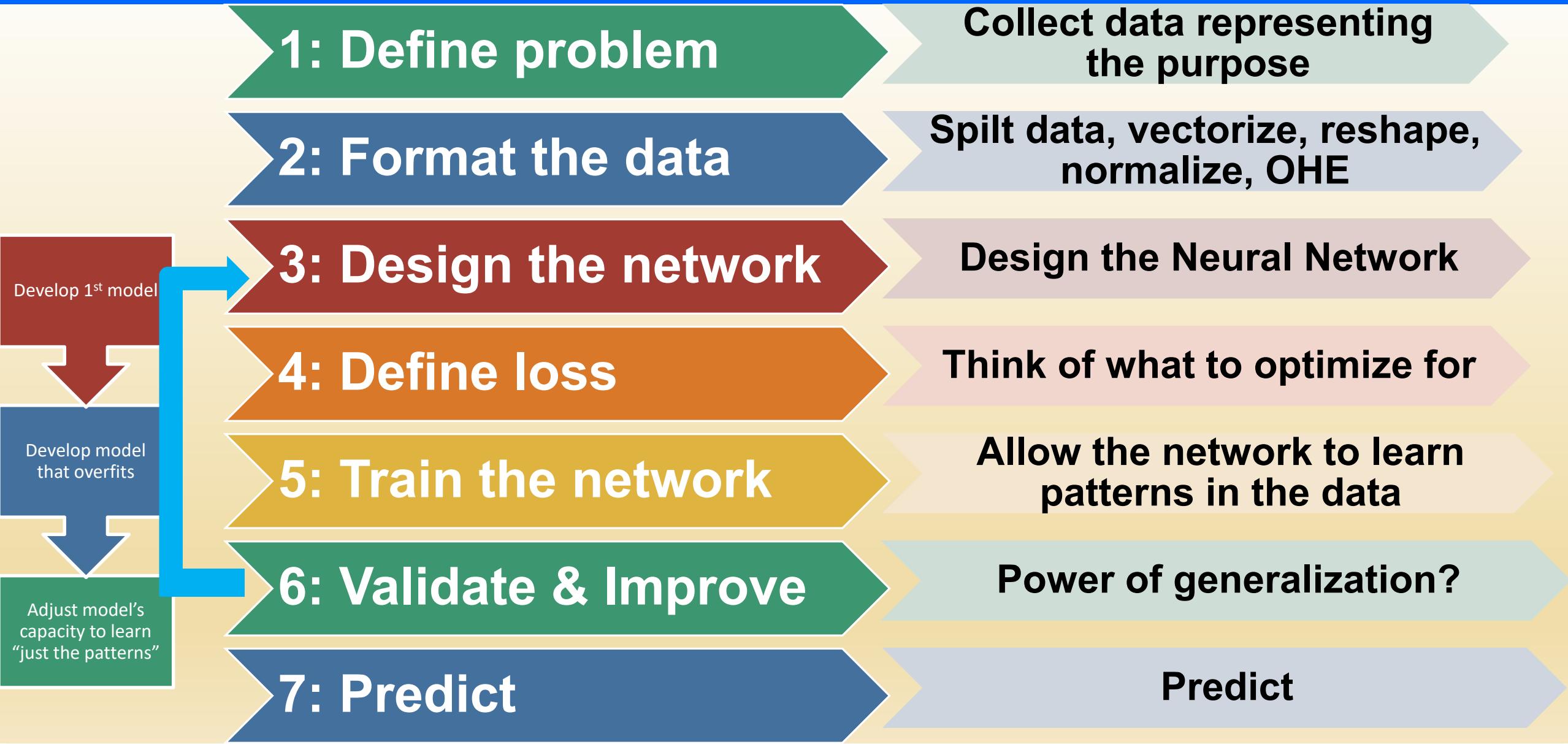
Representation of Data

Develop 1<sup>st</sup> model

Develop model that overfits

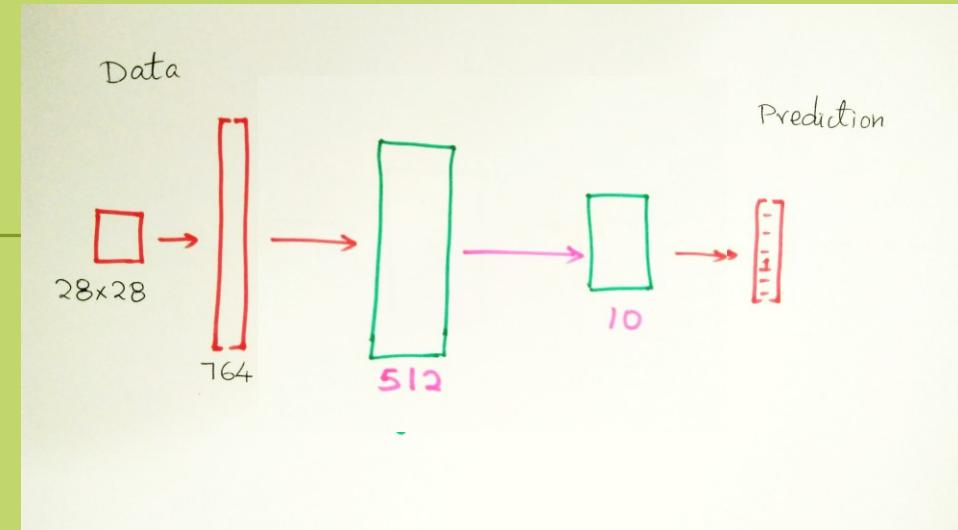
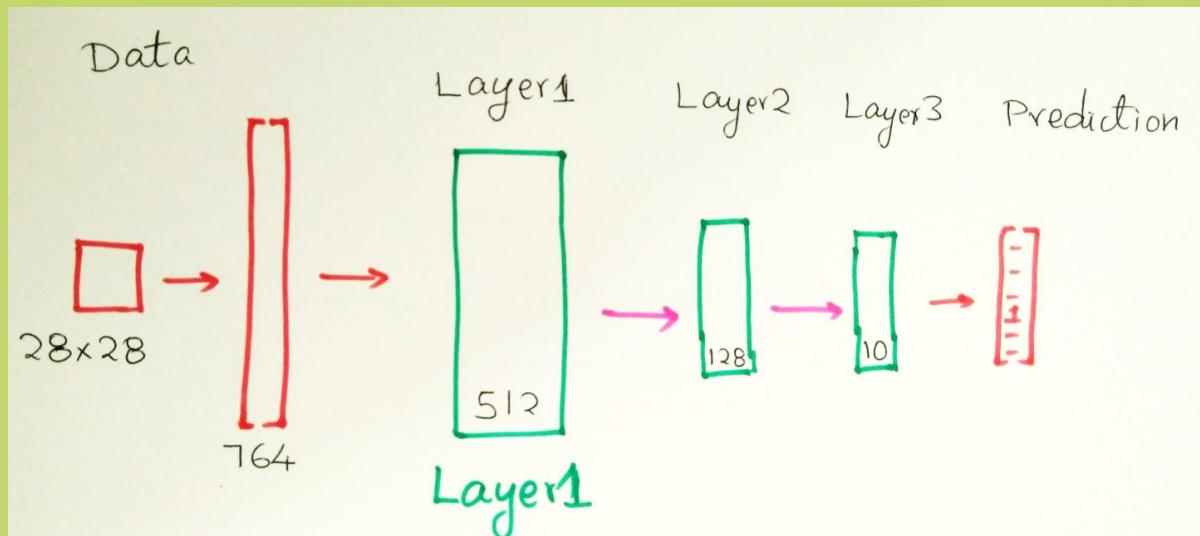
Adjust model's capacity to learn  
“just the patterns”

# Method: 7 steps to develop your Deep Learning model





# 3. Design the network



#Step 3: Define the network architecture

```
network = models.Sequential()  
layer1 = layers.Dense(512, input_shape=(28*28,), activation='relu')  
network.add(layer1)  
layer2 = layers.Dense(128, activation='relu')  
network.add(layer2)  
layer3 = layers.Dense(10, activation='softmax')  
network.add(layer3)
```



# 3. Design the network

#Step 3: Define the network architecture

```
network = models.Sequential()
```

```
layer1 = layers.Dense(512, input_shape=(28*28,))
```

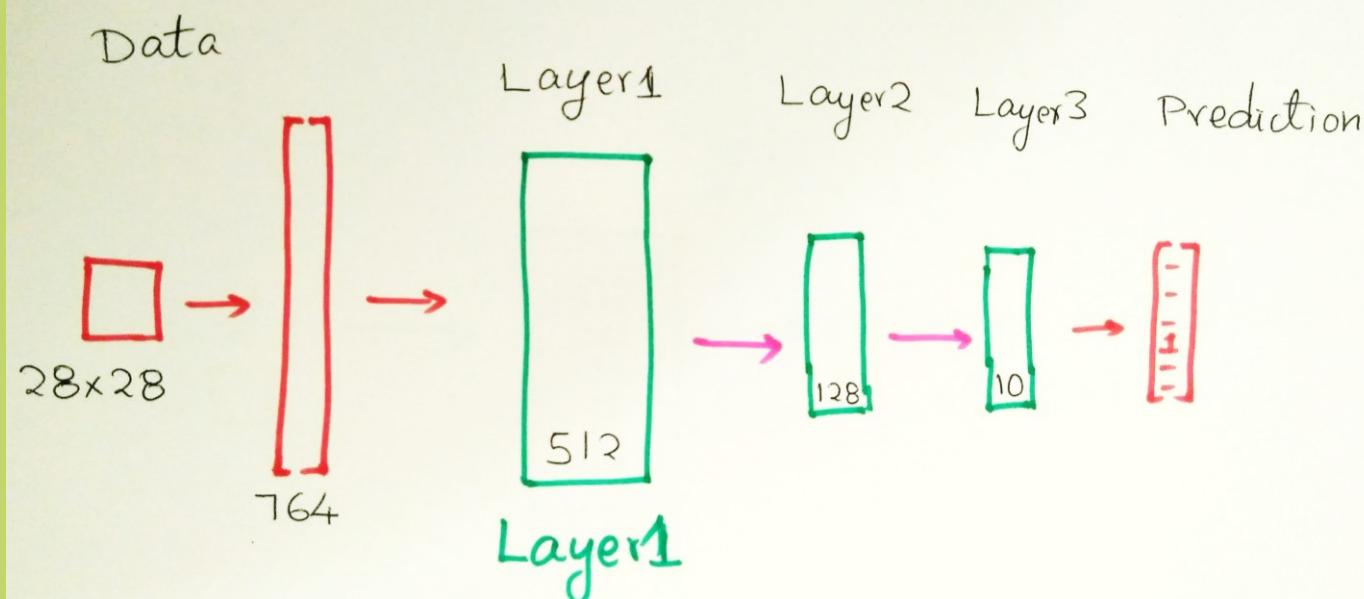
```
network.add(layer1)
```

```
layer2 = layers.Dense(128)
```

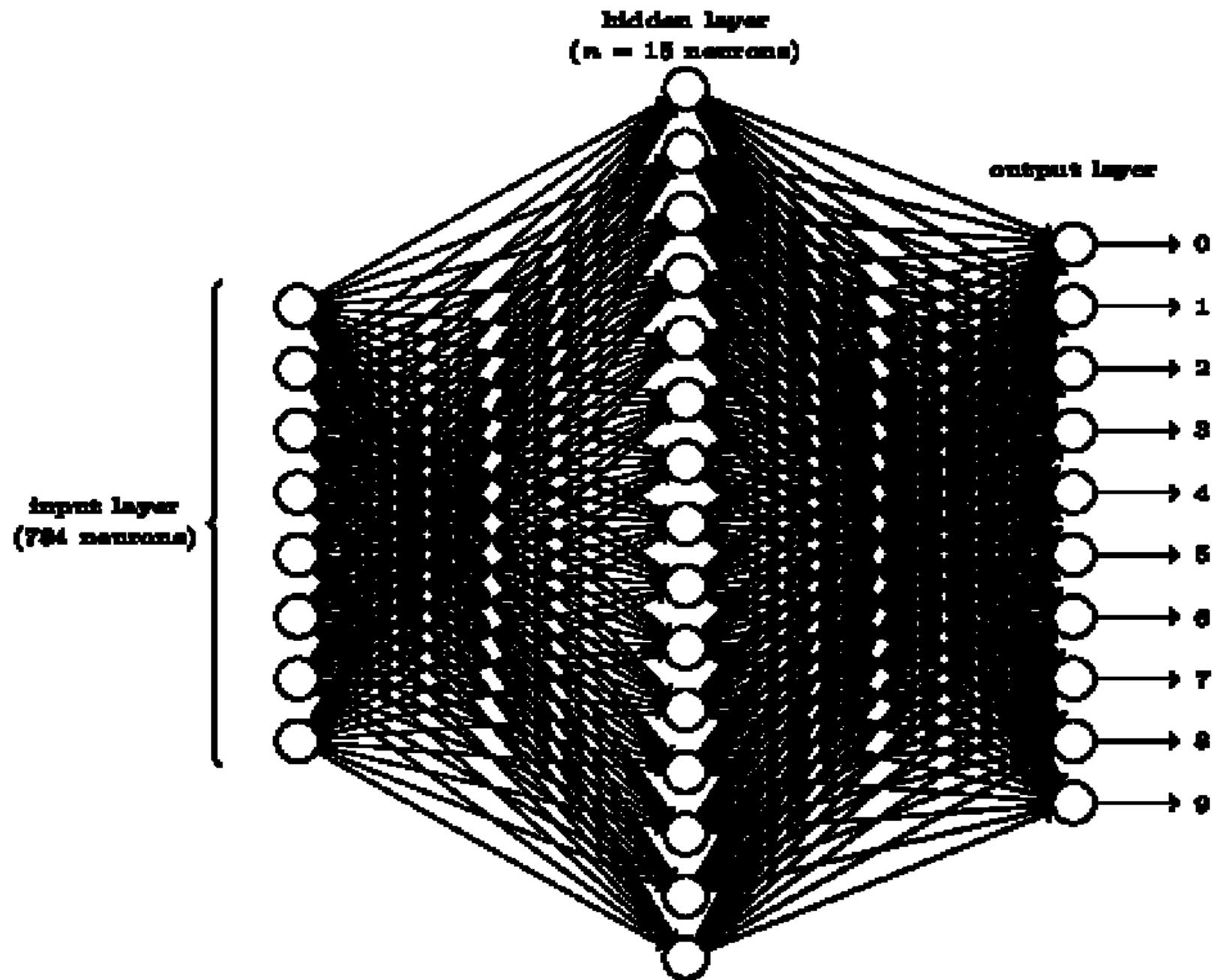
```
network.add(layer2)
```

```
layer3 = layers.Dense(10)
```

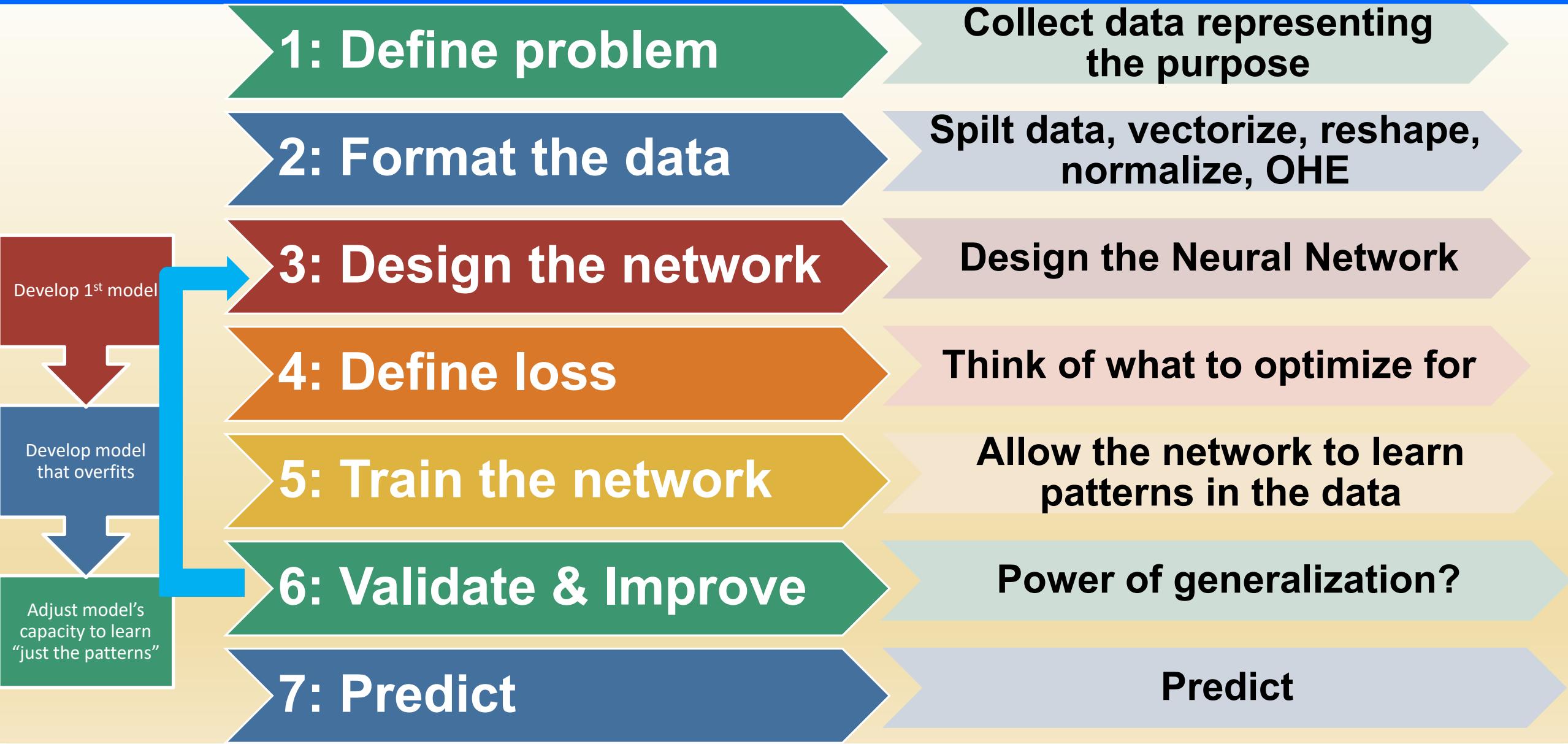
```
network.add(layer3)
```



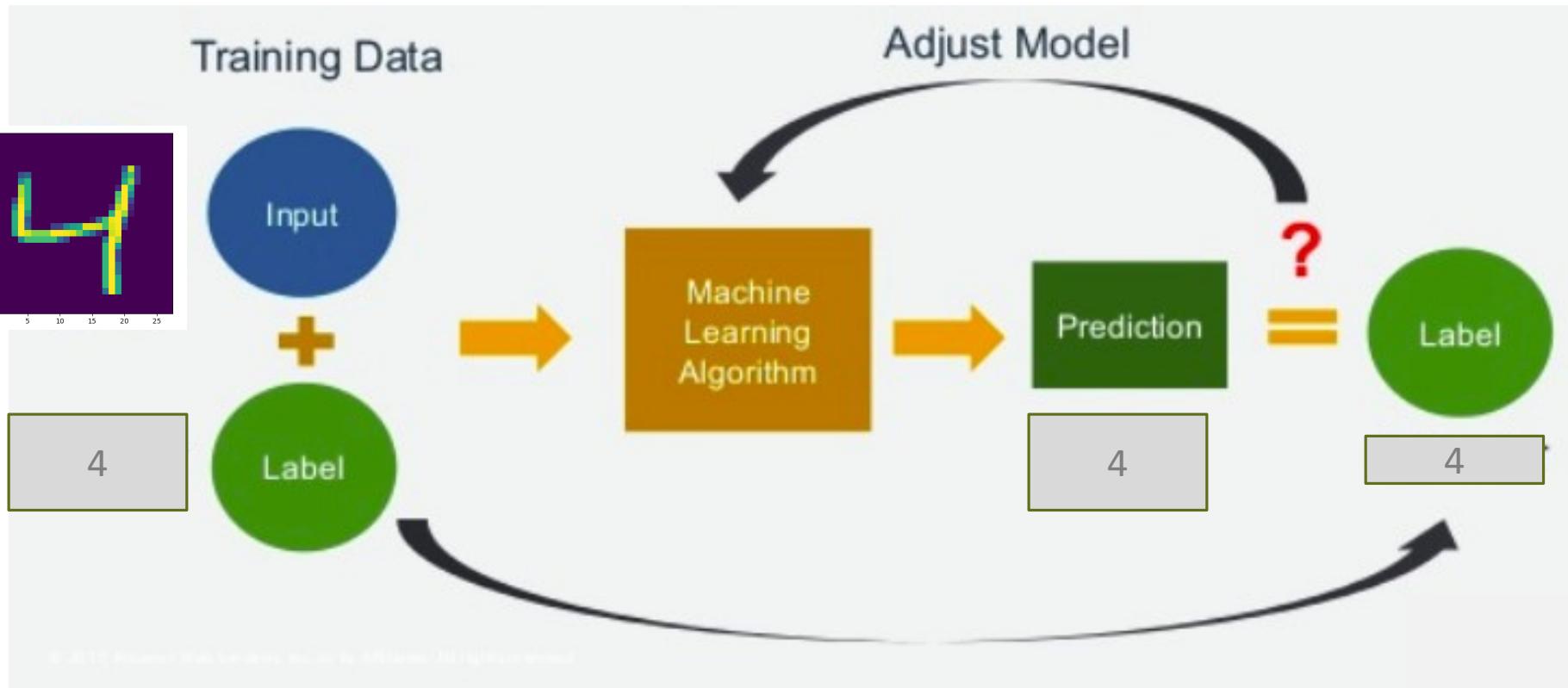
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3
4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4
5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5
6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6
7	7	7	7	7	7	7	7	7	7	7	7	7	7	7	7	7
8	8	8	8	8	8	8	8	8	8	8	8	8	8	8	8	8
9	9	9	9	9	9	9	9	9	9	9	9	9	9	9	9	9



# Method: 7 steps to develop your Deep Learning model



# 5. Train



# myModel.fit(X,Y)

```
import keras  
from keras import layers  
  
model = keras.Sequential()  
model.add(layers.Dense(20, activation='relu', input_shape=(10,)))  
model.add(layers.Dense(20, activation='relu'))  
model.add(layers.Dense(10, activation='softmax'))  
  
model.fit(x, y, epochs=10, batch_size=32)
```

$X \rightarrow Y$

$X = \text{image}$

Sample image

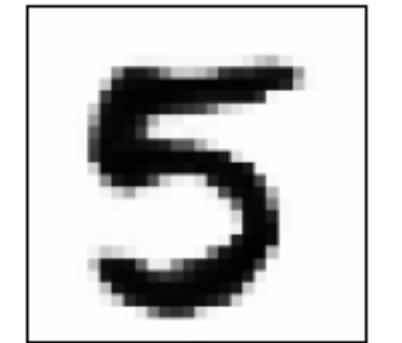
$Y = 5$

Label of the image

Input Image:



TensorFlow  
Model:



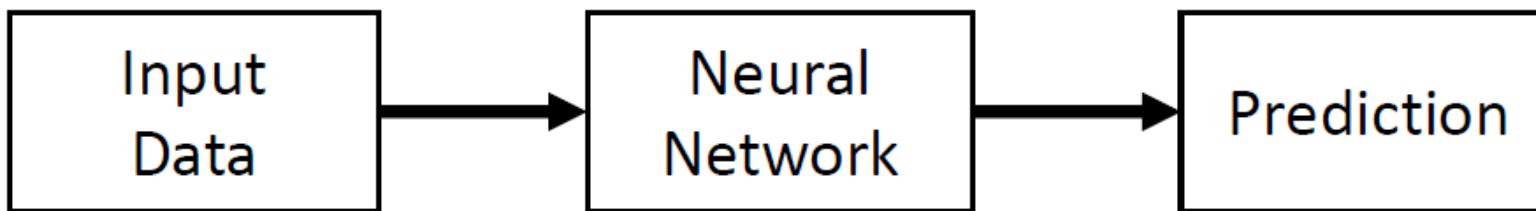
Neural  
Network

Output:

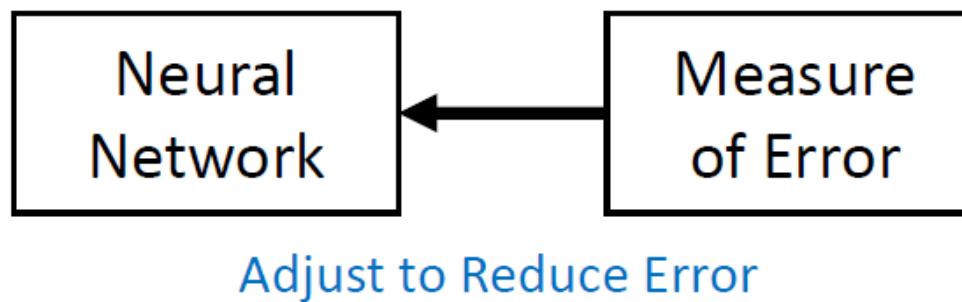
5

# How Neural Networks Learn: Backpropagation

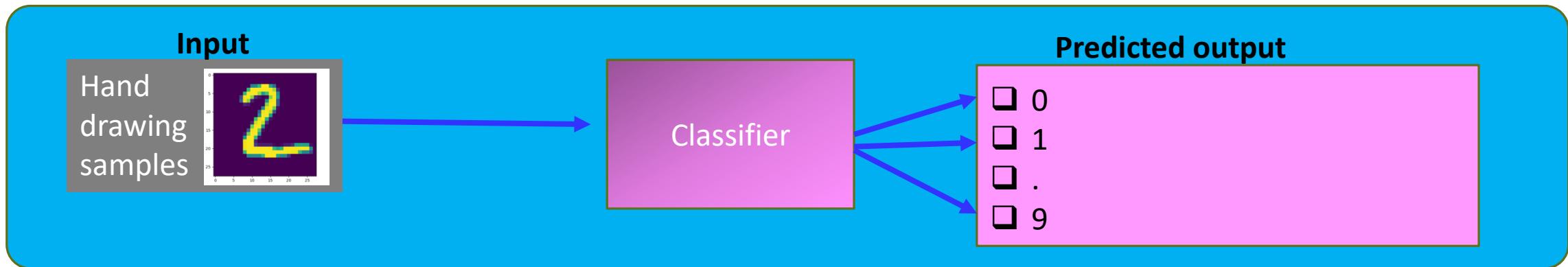
Forward Pass:



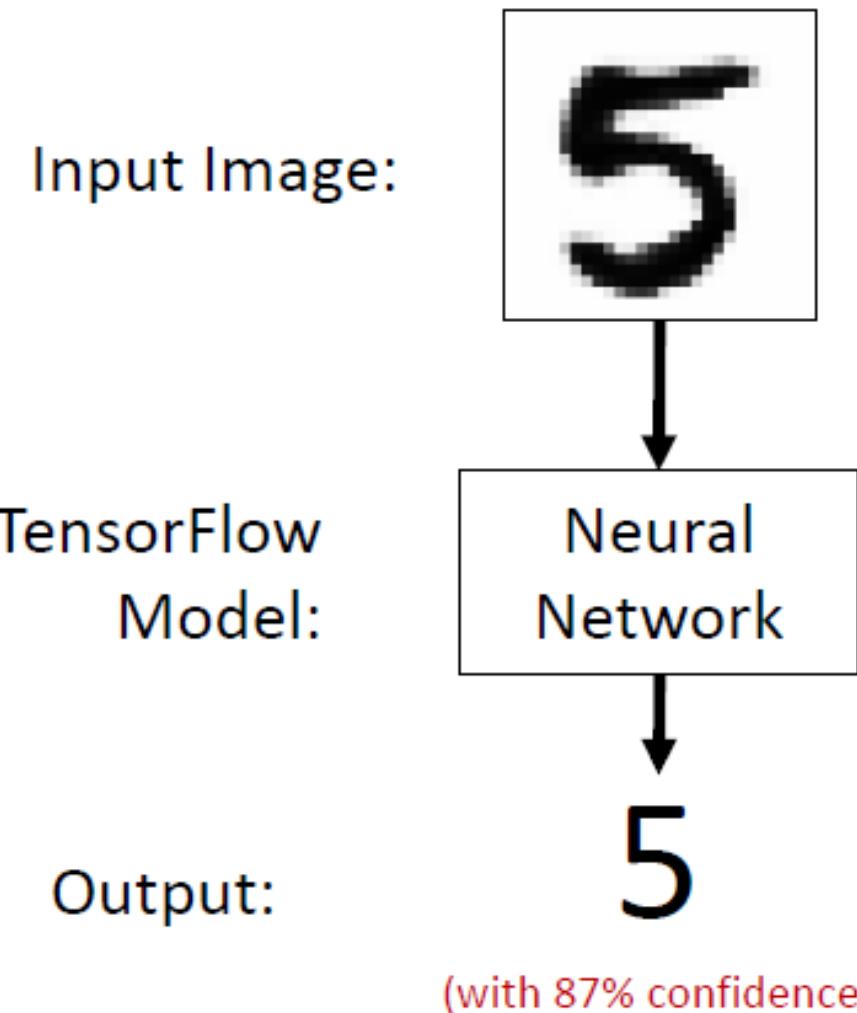
Backward Pass (aka Backpropagation):



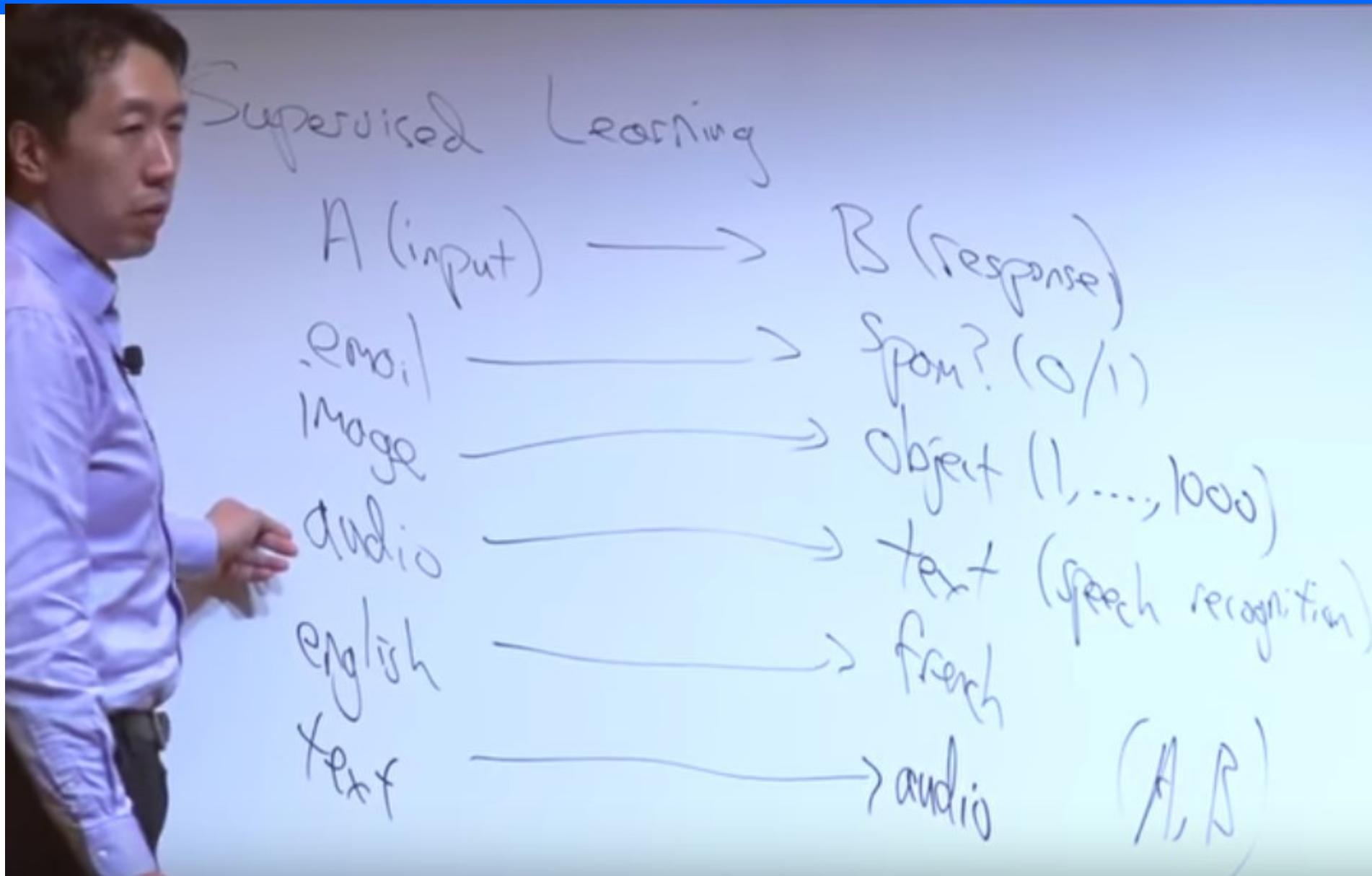
# 7. Predict



# $y = \text{myModel} . \text{Predict}(X)$



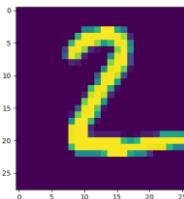
# $X \rightarrow Y$



# 7. Predict

Problem: Users draws a hand drawn gestures , your app predicts it.

```
predictedResults = network.predict(testXready[1:3])
```



**Predicted Results:**

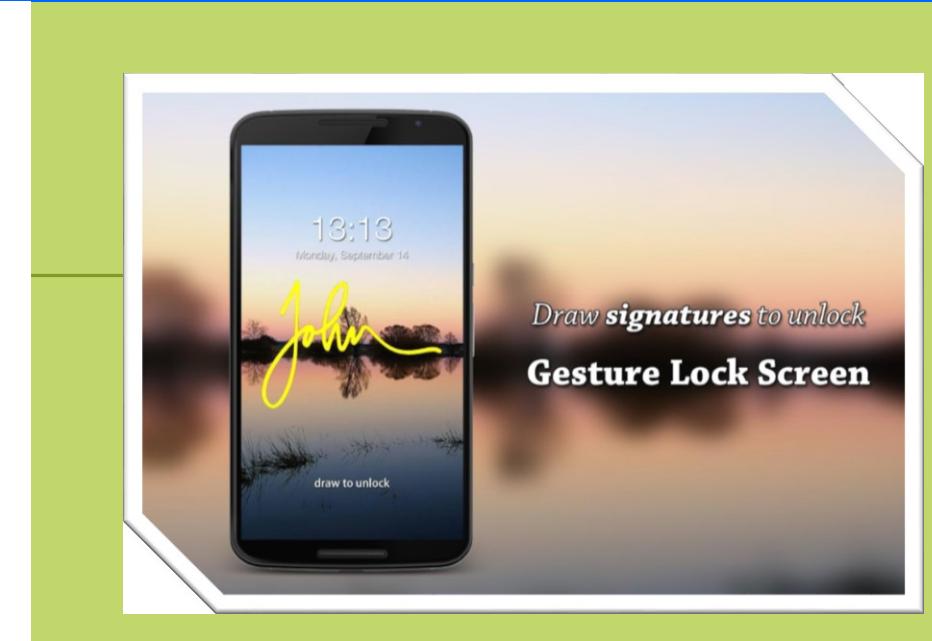
```
[ 1.35398892e-08  3.91871922e-08  9.99876499e-01  1.22501457e-04  
  1.77109036e-10  1.30937892e-07  2.99046292e-07  2.58880729e-11  
  4.58700157e-07  2.09212196e-11]
```

**Ground Truth :** [2]

**Ground Truth OHE :** [[ 0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]]

# Let's it together

- App idea: Recognize your hand gesture to unlock your smartphone (Secure unlock)
- Biz outcome: Recognize hand draw gestures to unlock your phone



#Step 6: Evaluate the network

```
metrics_test_loss, metrics_test_accuracy = network.evaluate(testXready, testYready)
```

Test accuracy is: 0.9157



Type this URL

**github.com/  
rajagopalmotivate1/  
AILab/**

# Design of Deep Learning experiment

