

# Get to Know Tensorflow

October 2, 2020

## 1 WELCOME to this guided project “Get to Know TensorFlow” on Coursera Labs!

This project course is part of “Tensorflow for AI” Series of project courses on Coursera. We will go through 4 parts to implement our project: **Task 1:** Introduction and overview of the whole project (non-technical task). **Task 2:** Install and Import TensorFlow and Practice Simple Examples. **Task 3:** Create TensorFlow Pipeline. **Task 4:** Define, Compile and Train a Neural Network. **Task 5:** Solve Simple Exercises.

```
[1]: print("Let's learn \n Tensorflow! ")
```

```
Let's learn
Tensorflow!
```

### 1.0.1 Task 1: Introduction and overview of the whole project (non-technical task)

In this project, you will get a soft introduction to what Machine Learning and Deep Learning are, and how they offer you a new programming paradigm, giving you a new set of tools to open new scenarios. All you need to know is some basic programming skills, and you’ll pick the rest up as you go along...

**Definition of Tensorflow** TensorFlow is the second machine learning framework that Google created and used to design, build, and train deep learning models. You can use the TensorFlow library to do numerical computations, which in itself doesn’t seem all too special, but these computations are done with data flow graphs. In these graphs, nodes represent mathematical operations, while the edges represent the data, which usually are multidimensional data arrays or tensors, that are communicated between these edges. Tensorflow is an open-source machine learning library developed by Google. One of its applications is to develop deep neural networks.

Currently, the most famous deep learning library in the world is Google’s TensorFlow. Google product uses machine learning in all of its products to improve the search engine, translation, image captioning or recommendations. TensorFlow was first made public in late 2015, while the first stable version appeared in 2017. It is open source under Apache Open Source license. You can use it, modify it and redistribute the modified version for a fee without paying anything to Google.

It is called Tensorflow because it takes input as a multi-dimensional array, also known as tensors. You can construct a sort of flowchart of operations (called a Graph) that you want to perform on that input. The input goes in at one end, and then it flows through this system of multiple

operations and comes out the other end as output. This is why it is called TensorFlow because the tensor goes in it flows through a list of operations, and then it comes out the other side. |

## 1.0.2 Task 2: Install and Import TensorFlow and Practice Simple Examples

### Let's first install TensorFlow

```
[5]: # !pip install tensorflow==2.0.0-alpha0
```

And now let's start with our imports to use TensorFlow easily:

1. **numpy** helps us to represent our data as lists easily and quickly.
2. **keras** is the framework for defining a neural network as a set of Sequential layers.

```
[6]: import tensorflow as tf
import numpy as np
from tensorflow import keras
```

### Start with simple TensorFlow Examples

**Example 1** Let's now practice the elementary workflow of Tensorflow with a simple example. Let's create a computational graph that multiplies two numbers together.

During the example, we will multiply  $X_1$  and  $X_2$  together. Tensorflow will create a node to connect the operation. In our example, it is called multiply. When the graph is determined, Tensorflow computational engines will multiply together  $X_1$  and  $X_2$ .

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Finally, we will run a TensorFlow session that will run the computational graph with the values of  $X_1$  and  $X_2$  and print the result of the multiplication.

Let's define the  $X_1$  and  $X_2$  input nodes. When we create a node in Tensorflow, we have to choose what kind of node to create. The  $X_1$  and  $X_2$  nodes will be a placeholder node. The placeholder assigns a new value each time we make a calculation. We will create them as a TF dot placeholder node.

```
[7]: ### Step 1: Define the Variable
X_1 = tf.placeholder(tf.float32, name = "X_1")
X_2 = tf.placeholder(tf.float32, name = "X_2")
```

When we create a placeholder node, we have to pass in the data type will be adding numbers here so we can use a floating-point data type, let's use `tf.float32`. We also need to give this node a name. This name will show up when we look at the graphical visualizations of our model. Let's name this node  $X_1$  by passing in a parameter called `name` with a value of  $X_1$  and now let's define  $X_2$  the same way.  $X_2$ .

```
[ ]: ### Step 2: Define the Computation
multiply = tf.multiply(X_1, X_2, name = "multiplication")
```

Now we can define the node that does the multiplication operation. In Tensorflow we can do that by creating a `tf.multiply` node.

We will pass in the X\_1 and X\_2 nodes to the multiplication node. It tells tensorflow to link those nodes in the computational graph, so we are asking it to pull the values from x and y and multiply the result. Let's also give the multiplication node the name multiply. It is the entire definition for our simple computational graph.

```
[ ]: ### Step 3: Execute the operation
X_1 = tf.placeholder(tf.float32, name = "X_1")
X_2 = tf.placeholder(tf.float32, name = "X_2")

multiply = tf.multiply(X_1, X_2, name = "multiplication")

with tf.Session() as session:
    result = session.run(multiply, feed_dict={X_1:[1,2,3], X_2:[4,5,6]})
    print(result)
```

To execute operations in the graph, we have to create a session. In Tensorflow, it is done by tf.Session(). Now that we have a session we can ask the session to run operations on our computational graph by calling session.run(). To run the computation, we need to use run.

When the addition operation runs, it is going to see that it needs to grab the values of the X\_1 and X\_2 nodes, so we also need to feed in values for X\_1 and X\_2. We can do that by supplying a parameter called feed\_dict. We pass the value 1,2,3 for X\_1 and 4,5,6 for X\_2.

We print the results with print(result). We should see 4, 10 and 18 for 1x4, 2x5 and 3x6

**Example 2** Let's initialize two variables that are actually constants. Pass an array of four numbers to the constant() function.

Note that you could potentially also pass in an integer, but that more often than not, you'll find yourself working with arrays. As you saw in the introduction, tensors are all about arrays! So make sure that you pass in an array :) Next, you can use multiply() to multiply your two variables. Store the result in the result variable. Lastly, print out the result with the help of the print() function.

```
[9]: # Initialize two constants
x1 = tf.constant([1,2,3,4])
x2 = tf.constant([5,6,7,8])

# Multiply
result = tf.multiply(x1, x2, name="multiplication_2")

# Print the result
print(result)
```

```
tf.Tensor([ 5 12 21 32], shape=(4,), dtype=int32)
```

### Example 3

**In this exercise, we will build a basic neural network with TensorFlow that predicts the price of a house using a simple formula.** Now imagine if house pricing was as easy as a house costs 30k + 30k per swimming pool, so that a house with one swimming pool in costs 60k, a house with 2 swimming pools costs 90k etc.

You will create a neural network that learns this pattern or relationship so that it would predict a villa with 7 swimming pools as costing close to 240k etc.

Hint: Your neural network may work better if you scale the house price down. You don't have to give the answer 250...it might be better to create something that predicts the number 2.4, and then your answer is in the 'hundreds of thousands' etc.

```
[10]: from tensorflow import keras

model = tf.keras.Sequential([keras.layers.Dense(units=1, input_shape=[1])])
model.compile(optimizer='sgd', loss='mean_squared_error')

xs = np.array([1.0, 2.0, 3.0, 4.0, 5.0, 6.0], dtype=float)
ys = np.array([0.6, 0.9, 1.2, 1.5, 1.8, 2.1], dtype=float)

model.fit(xs, ys, epochs=1000)

print(model.predict([7.0]))
```

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6/6 [=====] - 0s 9ms/sample - loss: 21.0308
Epoch 2/1000
6/6 [=====] - 0s 166us/sample - loss: 19.7114
Epoch 3/1000
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Epoch 4/1000
6/6 [=====] - 0s 166us/sample - loss: 17.3165
Epoch 5/1000
6/6 [=====] - 0s 166us/sample - loss: 16.2307
Epoch 6/1000
6/6 [=====] - 0s 166us/sample - loss: 15.2133
Epoch 7/1000
6/6 [=====] - 0s 0s/sample - loss: 14.2599
Epoch 8/1000
6/6 [=====] - 0s 166us/sample - loss: 13.3665
Epoch 9/1000
6/6 [=====] - 0s 166us/sample - loss: 12.5293
Epoch 10/1000
6/6 [=====] - 0s 166us/sample - loss: 11.7448
Epoch 11/1000
6/6 [=====] - 0s 332us/sample - loss: 11.0096
Epoch 12/1000
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Epoch 13/1000
6/6 [=====] - 0s 166us/sample - loss: 9.6751
Epoch 14/1000
6/6 [=====] - 0s 166us/sample - loss: 9.0701
Epoch 15/1000
6/6 [=====] - 0s 166us/sample - loss: 8.5032
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Epoch 16/1000  
6/6 [=====] - 0s 166us/sample - loss: 7.9719  
Epoch 17/1000  
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Epoch 22/1000  
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6/6 [=====] - 0s 166us/sample - loss: 0.0526  
Epoch 194/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0526  
Epoch 195/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0525  
Epoch 196/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0525  
Epoch 197/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0525  
Epoch 198/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0524  
Epoch 199/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0524  
Epoch 200/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0523  
Epoch 201/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0523  
Epoch 202/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0523  
Epoch 203/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0522  
Epoch 204/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0522  
Epoch 205/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0521  
Epoch 206/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0521  
Epoch 207/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0521

Epoch 208/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0520  
Epoch 209/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0520  
Epoch 210/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0519  
Epoch 211/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0519  
Epoch 212/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0519  
Epoch 213/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0518  
Epoch 214/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0518  
Epoch 215/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0517  
Epoch 216/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0517  
Epoch 217/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0517  
Epoch 218/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0516  
Epoch 219/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0516  
Epoch 220/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0515  
Epoch 221/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0515  
Epoch 222/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0515  
Epoch 223/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0514  
Epoch 224/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0514  
Epoch 225/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0513  
Epoch 226/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0513  
Epoch 227/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0513  
Epoch 228/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0512  
Epoch 229/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0512  
Epoch 230/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0512  
Epoch 231/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0511

Epoch 232/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0511  
Epoch 233/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0510  
Epoch 234/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0510  
Epoch 235/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0510  
Epoch 236/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0509  
Epoch 237/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0509  
Epoch 238/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0509  
Epoch 239/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0508  
Epoch 240/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0508  
Epoch 241/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0507  
Epoch 242/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0507  
Epoch 243/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0507  
Epoch 244/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0506  
Epoch 245/1000  
6/6 [=====] - 0s 216us/sample - loss: 0.0506  
Epoch 246/1000  
6/6 [=====] - 0s 116us/sample - loss: 0.0506  
Epoch 247/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0505  
Epoch 248/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0505  
Epoch 249/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0504  
Epoch 250/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0504  
Epoch 251/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0504  
Epoch 252/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0503  
Epoch 253/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0503  
Epoch 254/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0503  
Epoch 255/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0502

Epoch 256/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0502  
Epoch 257/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0502  
Epoch 258/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0501  
Epoch 259/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0501  
Epoch 260/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0500  
Epoch 261/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0500  
Epoch 262/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0500  
Epoch 263/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0499  
Epoch 264/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0499  
Epoch 265/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0499  
Epoch 266/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0498  
Epoch 267/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0498  
Epoch 268/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0498  
Epoch 269/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0497  
Epoch 270/1000  
6/6 [=====] - 0s 499us/sample - loss: 0.0497  
Epoch 271/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0496  
Epoch 272/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0496  
Epoch 273/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0496  
Epoch 274/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0495  
Epoch 275/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0495  
Epoch 276/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0495  
Epoch 277/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0494  
Epoch 278/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0494  
Epoch 279/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0494

Epoch 280/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0493  
Epoch 281/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0493  
Epoch 282/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0492  
Epoch 283/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0492  
Epoch 284/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0492  
Epoch 285/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0491  
Epoch 286/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0491  
Epoch 287/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0491  
Epoch 288/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0490  
Epoch 289/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0490  
Epoch 290/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0490  
Epoch 291/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0489  
Epoch 292/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0489  
Epoch 293/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0489  
Epoch 294/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0488  
Epoch 295/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0488  
Epoch 296/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0487  
Epoch 297/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0487  
Epoch 298/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0487  
Epoch 299/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0486  
Epoch 300/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0486  
Epoch 301/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0486  
Epoch 302/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0485  
Epoch 303/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0485



Epoch 304/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0485  
Epoch 305/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0484  
Epoch 306/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0484  
Epoch 307/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0484  
Epoch 308/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0483  
Epoch 309/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0483  
Epoch 310/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0483  
Epoch 311/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0482  
Epoch 312/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0482  
Epoch 313/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0481  
Epoch 314/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0481  
Epoch 315/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0481  
Epoch 316/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0480  
Epoch 317/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0480  
Epoch 318/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0480  
Epoch 319/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0479  
Epoch 320/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0479  
Epoch 321/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0479  
Epoch 322/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0478  
Epoch 323/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0478  
Epoch 324/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0478  
Epoch 325/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0477  
Epoch 326/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0477  
Epoch 327/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0477

Epoch 328/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0476  
Epoch 329/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0476  
Epoch 330/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0476  
Epoch 331/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0475  
Epoch 332/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0475  
Epoch 333/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0474  
Epoch 334/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0474  
Epoch 335/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0474  
Epoch 336/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0473  
Epoch 337/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0473  
Epoch 338/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0473  
Epoch 339/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0472  
Epoch 340/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0472  
Epoch 341/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0472  
Epoch 342/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0471  
Epoch 343/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0471  
Epoch 344/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0471  
Epoch 345/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0470  
Epoch 346/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0470  
Epoch 347/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0470  
Epoch 348/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0469  
Epoch 349/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0469  
Epoch 350/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0469  
Epoch 351/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0468

Epoch 352/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0468  
Epoch 353/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0468  
Epoch 354/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0467  
Epoch 355/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0467  
Epoch 356/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0467  
Epoch 357/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0466  
Epoch 358/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0466  
Epoch 359/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0466  
Epoch 360/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0465  
Epoch 361/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0465  
Epoch 362/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0465  
Epoch 363/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0464  
Epoch 364/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0464  
Epoch 365/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0464  
Epoch 366/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0463  
Epoch 367/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0463  
Epoch 368/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0462  
Epoch 369/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0462  
Epoch 370/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0462  
Epoch 371/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0461  
Epoch 372/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0461  
Epoch 373/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0461  
Epoch 374/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0460  
Epoch 375/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0460

Epoch 376/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0460  
Epoch 377/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0459  
Epoch 378/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0459  
Epoch 379/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0459  
Epoch 380/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0458  
Epoch 381/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0458  
Epoch 382/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0458  
Epoch 383/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0457  
Epoch 384/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0457  
Epoch 385/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0457  
Epoch 386/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0456  
Epoch 387/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0456  
Epoch 388/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0456  
Epoch 389/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0455  
Epoch 390/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0455  
Epoch 391/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0455  
Epoch 392/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0454  
Epoch 393/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0454  
Epoch 394/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0454  
Epoch 395/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0453  
Epoch 396/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0453  
Epoch 397/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0453  
Epoch 398/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0452  
Epoch 399/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0452

Epoch 400/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0452  
Epoch 401/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0451  
Epoch 402/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0451  
Epoch 403/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0451  
Epoch 404/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0451  
Epoch 405/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0450  
Epoch 406/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0450  
Epoch 407/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0450  
Epoch 408/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0449  
Epoch 409/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0449  
Epoch 410/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0449  
Epoch 411/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0448  
Epoch 412/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0448  
Epoch 413/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0448  
Epoch 414/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0447  
Epoch 415/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0447  
Epoch 416/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0447  
Epoch 417/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0446  
Epoch 418/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0446  
Epoch 419/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0446  
Epoch 420/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0445  
Epoch 421/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0445  
Epoch 422/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0445  
Epoch 423/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0444

Epoch 424/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0444  
Epoch 425/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0444  
Epoch 426/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0443  
Epoch 427/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0443  
Epoch 428/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0443  
Epoch 429/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0442  
Epoch 430/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0442  
Epoch 431/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0442  
Epoch 432/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0441  
Epoch 433/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0441  
Epoch 434/1000  
6/6 [=====] - 0s 162us/sample - loss: 0.0441  
Epoch 435/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0440  
Epoch 436/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0440  
Epoch 437/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0440  
Epoch 438/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0439  
Epoch 439/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0439  
Epoch 440/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0439  
Epoch 441/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0438  
Epoch 442/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0438  
Epoch 443/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0438  
Epoch 444/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0438  
Epoch 445/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0437  
Epoch 446/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0437  
Epoch 447/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0437

Epoch 448/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0436  
Epoch 449/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0436  
Epoch 450/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0436  
Epoch 451/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0435  
Epoch 452/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0435  
Epoch 453/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0435  
Epoch 454/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0434  
Epoch 455/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0434  
Epoch 456/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0434  
Epoch 457/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0433  
Epoch 458/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0433  
Epoch 459/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0433  
Epoch 460/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0432  
Epoch 461/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0432  
Epoch 462/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0432  
Epoch 463/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0432  
Epoch 464/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0431  
Epoch 465/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0431  
Epoch 466/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0431  
Epoch 467/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0430  
Epoch 468/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0430  
Epoch 469/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0430  
Epoch 470/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0429  
Epoch 471/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0429

Epoch 472/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0429  
Epoch 473/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0428  
Epoch 474/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0428  
Epoch 475/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0428  
Epoch 476/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0427  
Epoch 477/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0427  
Epoch 478/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0427  
Epoch 479/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0426  
Epoch 480/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0426  
Epoch 481/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0426  
Epoch 482/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0426  
Epoch 483/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0425  
Epoch 484/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0425  
Epoch 485/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0425  
Epoch 486/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0424  
Epoch 487/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0424  
Epoch 488/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0424  
Epoch 489/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0423  
Epoch 490/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0423  
Epoch 491/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0423  
Epoch 492/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0422  
Epoch 493/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0422  
Epoch 494/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0422  
Epoch 495/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0422



Epoch 496/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0421  
Epoch 497/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0421  
Epoch 498/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0421  
Epoch 499/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0420  
Epoch 500/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0420  
Epoch 501/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0420  
Epoch 502/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0419  
Epoch 503/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0419  
Epoch 504/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0419  
Epoch 505/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0418  
Epoch 506/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0418  
Epoch 507/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0418  
Epoch 508/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0418  
Epoch 509/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0417  
Epoch 510/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0417  
Epoch 511/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0417  
Epoch 512/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0416  
Epoch 513/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0416  
Epoch 514/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0416  
Epoch 515/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0415  
Epoch 516/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0415  
Epoch 517/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0415  
Epoch 518/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0415  
Epoch 519/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0414

Epoch 520/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0414  
Epoch 521/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0414  
Epoch 522/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0413  
Epoch 523/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0413  
Epoch 524/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0413  
Epoch 525/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0412  
Epoch 526/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0412  
Epoch 527/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0412  
Epoch 528/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0412  
Epoch 529/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0411  
Epoch 530/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0411  
Epoch 531/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0411  
Epoch 532/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0410  
Epoch 533/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0410  
Epoch 534/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0410  
Epoch 535/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0409  
Epoch 536/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0409  
Epoch 537/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0409  
Epoch 538/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0409  
Epoch 539/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0408  
Epoch 540/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0408  
Epoch 541/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0408  
Epoch 542/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0407  
Epoch 543/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0407

Epoch 544/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0407  
Epoch 545/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0406  
Epoch 546/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0406  
Epoch 547/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0406  
Epoch 548/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0406  
Epoch 549/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0405  
Epoch 550/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0405  
Epoch 551/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0405  
Epoch 552/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0404  
Epoch 553/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0404  
Epoch 554/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0404  
Epoch 555/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0403  
Epoch 556/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0403  
Epoch 557/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0403  
Epoch 558/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0403  
Epoch 559/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0402  
Epoch 560/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0402  
Epoch 561/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0402  
Epoch 562/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0401  
Epoch 563/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0401  
Epoch 564/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0401  
Epoch 565/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0401  
Epoch 566/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0400  
Epoch 567/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0400

Epoch 568/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0400  
Epoch 569/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0399  
Epoch 570/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0399  
Epoch 571/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0399  
Epoch 572/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0399  
Epoch 573/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0398  
Epoch 574/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0398  
Epoch 575/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0398  
Epoch 576/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0397  
Epoch 577/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0397  
Epoch 578/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0397  
Epoch 579/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0396  
Epoch 580/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0396  
Epoch 581/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0396  
Epoch 582/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0396  
Epoch 583/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0395  
Epoch 584/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0395  
Epoch 585/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0395  
Epoch 586/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0394  
Epoch 587/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0394  
Epoch 588/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0394  
Epoch 589/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0394  
Epoch 590/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0393  
Epoch 591/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0393

Epoch 592/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0393  
Epoch 593/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0392  
Epoch 594/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0392  
Epoch 595/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0392  
Epoch 596/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0392  
Epoch 597/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0391  
Epoch 598/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0391  
Epoch 599/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0391  
Epoch 600/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0390  
Epoch 601/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0390  
Epoch 602/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0390  
Epoch 603/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0390  
Epoch 604/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0389  
Epoch 605/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0389  
Epoch 606/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0389  
Epoch 607/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0388  
Epoch 608/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0388  
Epoch 609/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0388  
Epoch 610/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0388  
Epoch 611/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0387  
Epoch 612/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0387  
Epoch 613/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0387  
Epoch 614/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0386  
Epoch 615/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0386

Epoch 616/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0386  
Epoch 617/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0386  
Epoch 618/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0385  
Epoch 619/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0385  
Epoch 620/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0385  
Epoch 621/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0385  
Epoch 622/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0384  
Epoch 623/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0384  
Epoch 624/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0384  
Epoch 625/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0383  
Epoch 626/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0383  
Epoch 627/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0383  
Epoch 628/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0383  
Epoch 629/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0382  
Epoch 630/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0382  
Epoch 631/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0382  
Epoch 632/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0381  
Epoch 633/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0381  
Epoch 634/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0381  
Epoch 635/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0381  
Epoch 636/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0380  
Epoch 637/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0380  
Epoch 638/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0380  
Epoch 639/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0379

Epoch 640/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0379  
Epoch 641/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0379  
Epoch 642/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0379  
Epoch 643/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0378  
Epoch 644/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0378  
Epoch 645/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0378  
Epoch 646/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0378  
Epoch 647/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0377  
Epoch 648/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0377  
Epoch 649/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0377  
Epoch 650/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0376  
Epoch 651/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0376  
Epoch 652/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0376  
Epoch 653/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0376  
Epoch 654/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0375  
Epoch 655/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0375  
Epoch 656/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0375  
Epoch 657/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0375  
Epoch 658/1000  
6/6 [=====] - 0s 253us/sample - loss: 0.0374  
Epoch 659/1000  
6/6 [=====] - 0s 164us/sample - loss: 0.0374  
Epoch 660/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0374  
Epoch 661/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0373  
Epoch 662/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0373  
Epoch 663/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0373

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Epoch 664/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0373
Epoch 665/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0372
Epoch 666/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0372
Epoch 667/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0372
Epoch 668/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0372
Epoch 669/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0371
Epoch 670/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0371
Epoch 671/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0371
Epoch 672/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0370
Epoch 673/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0370
Epoch 674/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0370
Epoch 675/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0370
Epoch 676/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0369
Epoch 677/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0369
Epoch 678/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0369
Epoch 679/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0369
Epoch 680/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0368
Epoch 681/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0368
Epoch 682/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0368
Epoch 683/1000
6/6 [=====] - 0s 333us/sample - loss: 0.0367
Epoch 684/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0367
Epoch 685/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0367
Epoch 686/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0367
Epoch 687/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0366

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Epoch 688/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0366  
Epoch 689/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0366  
Epoch 690/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0366  
Epoch 691/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0365  
Epoch 692/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0365  
Epoch 693/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0365  
Epoch 694/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0365  
Epoch 695/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0364  
Epoch 696/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0364  
Epoch 697/1000  
6/6 [=====] - 0s 499us/sample - loss: 0.0364  
Epoch 698/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0363  
Epoch 699/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0363  
Epoch 700/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0363  
Epoch 701/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0363  
Epoch 702/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0362  
Epoch 703/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0362  
Epoch 704/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0362  
Epoch 705/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0362  
Epoch 706/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0361  
Epoch 707/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0361  
Epoch 708/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0361  
Epoch 709/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0361  
Epoch 710/1000  
6/6 [=====] - 0s 333us/sample - loss: 0.0360  
Epoch 711/1000  
6/6 [=====] - 0s 333us/sample - loss: 0.0360

Epoch 712/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0360  
Epoch 713/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0360  
Epoch 714/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0359  
Epoch 715/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0359  
Epoch 716/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0359  
Epoch 717/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0358  
Epoch 718/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0358  
Epoch 719/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0358  
Epoch 720/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0358  
Epoch 721/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0357  
Epoch 722/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0357  
Epoch 723/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0357  
Epoch 724/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0357  
Epoch 725/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0356  
Epoch 726/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0356  
Epoch 727/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0356  
Epoch 728/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0356  
Epoch 729/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0355  
Epoch 730/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0355  
Epoch 731/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0355  
Epoch 732/1000  
6/6 [=====] - 0s 499us/sample - loss: 0.0355  
Epoch 733/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0354  
Epoch 734/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0354  
Epoch 735/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0354

Epoch 736/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0354  
Epoch 737/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0353  
Epoch 738/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0353  
Epoch 739/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0353  
Epoch 740/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0353  
Epoch 741/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0352  
Epoch 742/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0352  
Epoch 743/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0352  
Epoch 744/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0351  
Epoch 745/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0351  
Epoch 746/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0351  
Epoch 747/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0351  
Epoch 748/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0350  
Epoch 749/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0350  
Epoch 750/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0350  
Epoch 751/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0350  
Epoch 752/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0349  
Epoch 753/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0349  
Epoch 754/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0349  
Epoch 755/1000  
6/6 [=====] - 0s 333us/sample - loss: 0.0349  
Epoch 756/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0348  
Epoch 757/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0348  
Epoch 758/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0348  
Epoch 759/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0348

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Epoch 760/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0347
Epoch 761/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0347
Epoch 762/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0347
Epoch 763/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0347
Epoch 764/1000
6/6 [=====] - 0s 190us/sample - loss: 0.0346
Epoch 765/1000
6/6 [=====] - 0s 142us/sample - loss: 0.0346
Epoch 766/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0346
Epoch 767/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0346
Epoch 768/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0345
Epoch 769/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0345
Epoch 770/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0345
Epoch 771/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0345
Epoch 772/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0344
Epoch 773/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0344
Epoch 774/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0344
Epoch 775/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0344
Epoch 776/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0343
Epoch 777/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0343
Epoch 778/1000
6/6 [=====] - 0s 332us/sample - loss: 0.0343
Epoch 779/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0343
Epoch 780/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0342
Epoch 781/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0342
Epoch 782/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0342
Epoch 783/1000
6/6 [=====] - 0s 0s/sample - loss: 0.0342

```

Epoch 784/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0341  
Epoch 785/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0341  
Epoch 786/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0341  
Epoch 787/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0341  
Epoch 788/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0340  
Epoch 789/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0340  
Epoch 790/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0340  
Epoch 791/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0340  
Epoch 792/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0339  
Epoch 793/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0339  
Epoch 794/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0339  
Epoch 795/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0339  
Epoch 796/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0338  
Epoch 797/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0338  
Epoch 798/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0338  
Epoch 799/1000  
6/6 [=====] - 0s 101us/sample - loss: 0.0338  
Epoch 800/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0337  
Epoch 801/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0337  
Epoch 802/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0337  
Epoch 803/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0337  
Epoch 804/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0336  
Epoch 805/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0336  
Epoch 806/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0336  
Epoch 807/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0336

Epoch 808/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0335  
Epoch 809/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0335  
Epoch 810/1000  
6/6 [=====] - 0s 184us/sample - loss: 0.0335  
Epoch 811/1000  
6/6 [=====] - 0s 148us/sample - loss: 0.0335  
Epoch 812/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0334  
Epoch 813/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0334  
Epoch 814/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0334  
Epoch 815/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0334  
Epoch 816/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0333  
Epoch 817/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0333  
Epoch 818/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0333  
Epoch 819/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0333  
Epoch 820/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0333  
Epoch 821/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0332  
Epoch 822/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0332  
Epoch 823/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0332  
Epoch 824/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0332  
Epoch 825/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0331  
Epoch 826/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0331  
Epoch 827/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0331  
Epoch 828/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0331  
Epoch 829/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0330  
Epoch 830/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0330  
Epoch 831/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0330

Epoch 832/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0330  
Epoch 833/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0329  
Epoch 834/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0329  
Epoch 835/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0329  
Epoch 836/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0329  
Epoch 837/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0328  
Epoch 838/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0328  
Epoch 839/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0328  
Epoch 840/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0328  
Epoch 841/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0327  
Epoch 842/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0327  
Epoch 843/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0327  
Epoch 844/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0327  
Epoch 845/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0326  
Epoch 846/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0326  
Epoch 847/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0326  
Epoch 848/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0326  
Epoch 849/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0326  
Epoch 850/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0325  
Epoch 851/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0325  
Epoch 852/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0325  
Epoch 853/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0325  
Epoch 854/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0324  
Epoch 855/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0324

Epoch 856/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0324  
Epoch 857/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0324  
Epoch 858/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0323  
Epoch 859/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0323  
Epoch 860/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0323  
Epoch 861/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0323  
Epoch 862/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0322  
Epoch 863/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0322  
Epoch 864/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0322  
Epoch 865/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0322  
Epoch 866/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0322  
Epoch 867/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0321  
Epoch 868/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0321  
Epoch 869/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0321  
Epoch 870/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0321  
Epoch 871/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0320  
Epoch 872/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0320  
Epoch 873/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0320  
Epoch 874/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0320  
Epoch 875/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0319  
Epoch 876/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0319  
Epoch 877/1000  
6/6 [=====] - 0s 332us/sample - loss: 0.0319  
Epoch 878/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0319  
Epoch 879/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0318



Epoch 880/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0318  
Epoch 881/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0318  
Epoch 882/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0318  
Epoch 883/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0318  
Epoch 884/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0317  
Epoch 885/1000  
6/6 [=====] - 0s 164us/sample - loss: 0.0317  
Epoch 886/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0317  
Epoch 887/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0317  
Epoch 888/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0316  
Epoch 889/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0316  
Epoch 890/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0316  
Epoch 891/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0316  
Epoch 892/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0315  
Epoch 893/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0315  
Epoch 894/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0315  
Epoch 895/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0315  
Epoch 896/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0315  
Epoch 897/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0314  
Epoch 898/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0314  
Epoch 899/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0314  
Epoch 900/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0314  
Epoch 901/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0313  
Epoch 902/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0313  
Epoch 903/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0313

Epoch 904/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0313  
Epoch 905/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0313  
Epoch 906/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0312  
Epoch 907/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0312  
Epoch 908/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0312  
Epoch 909/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0312  
Epoch 910/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0311  
Epoch 911/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0311  
Epoch 912/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0311  
Epoch 913/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0311  
Epoch 914/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0310  
Epoch 915/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0310  
Epoch 916/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0310  
Epoch 917/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0310  
Epoch 918/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0310  
Epoch 919/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0309  
Epoch 920/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0309  
Epoch 921/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0309  
Epoch 922/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0309  
Epoch 923/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0308  
Epoch 924/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0308  
Epoch 925/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0308  
Epoch 926/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0308  
Epoch 927/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0308

Epoch 928/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0307  
Epoch 929/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0307  
Epoch 930/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0307  
Epoch 931/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0307  
Epoch 932/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0306  
Epoch 933/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0306  
Epoch 934/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0306  
Epoch 935/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0306  
Epoch 936/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0306  
Epoch 937/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0305  
Epoch 938/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0305  
Epoch 939/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0305  
Epoch 940/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0305  
Epoch 941/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0304  
Epoch 942/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0304  
Epoch 943/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0304  
Epoch 944/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0304  
Epoch 945/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0304  
Epoch 946/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0303  
Epoch 947/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0303  
Epoch 948/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0303  
Epoch 949/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0303  
Epoch 950/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0302  
Epoch 951/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0302

Epoch 952/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0302  
Epoch 953/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0302  
Epoch 954/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0302  
Epoch 955/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0301  
Epoch 956/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0301  
Epoch 957/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0301  
Epoch 958/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0301  
Epoch 959/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0300  
Epoch 960/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0300  
Epoch 961/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0300  
Epoch 962/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0300  
Epoch 963/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0300  
Epoch 964/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0299  
Epoch 965/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0299  
Epoch 966/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0299  
Epoch 967/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0299  
Epoch 968/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0298  
Epoch 969/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0298  
Epoch 970/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0298  
Epoch 971/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0298  
Epoch 972/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0298  
Epoch 973/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0297  
Epoch 974/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0297  
Epoch 975/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0297

Epoch 976/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0297  
Epoch 977/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0297  
Epoch 978/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0296  
Epoch 979/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0296  
Epoch 980/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0296  
Epoch 981/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0296  
Epoch 982/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0295  
Epoch 983/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0295  
Epoch 984/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0295  
Epoch 985/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0295  
Epoch 986/1000  
6/6 [=====] - 0s 0s/sample - loss: 0.0295  
Epoch 987/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0294  
Epoch 988/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0294  
Epoch 989/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0294  
Epoch 990/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0294  
Epoch 991/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0293  
Epoch 992/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0293  
Epoch 993/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0293  
Epoch 994/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0293  
Epoch 995/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0293  
Epoch 996/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0292  
Epoch 997/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0292  
Epoch 998/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0292  
Epoch 999/1000  
6/6 [=====] - 0s 166us/sample - loss: 0.0292

```
Epoch 1000/1000
6/6 [=====] - 0s 166us/sample - loss: 0.0292
[[2.6471088]]
```

### 1.0.3 Task 3: Create TensorFlow Pipeline

Let's Create a TensorFlow Pipeline in 5 Simple Steps

**Step 1: Create the data** First of all, let's use numpy library to generate two random values.

```
[11]: import numpy as np
      x_input = np.random.sample((1,2))
      print(x_input)
```

```
[[0.14723435 0.61966029]]
```

**Step 2: Create the placeholder** Like in the previous example, we create a placeholder with the name X. We need to specify the shape of the tensor explicitly. In case, we will load an array with only two values. We can write the shape as shape=[1,2]

```
[ ]: # using a placeholder
      x = tf.placeholder(tf.float32, shape=[1,2], name = "x")
```

**Step 3: Define the dataset method** next, we need to define the Dataset where we can populate the value of the placeholder x. We need to use the method tf.data.Dataset.from\_tensor\_slices

```
[ ]: dataset = tf.data.Dataset.from_tensor_slices(x)
```

**Step 4: Create the pipeline** In step four, we need to initialize the pipeline where the data will flow. We need to create an iterator with make\_initializable\_iterator. We name it iterator. Then we need to call this iterator to feed the next batch of data, get\_next. We name this step get\_next. Note that in our example, there is only one batch of data with only two values.

```
[ ]: iterator = dataset.make_initializable_iterator()
      get_next = iterator.get_next()
```

**Step 5: Execute the operation** The last step is similar to the previous example. We initiate a session, and we run the operation iterator. We feed the feed\_dict with the value generated by numpy. These two value will populate the placeholder x. Then we run get\_next to print the result.

```
[ ]: with tf.Session() as sess:
      # feeding the placeholder with data
      sess.run(iterator.initializer, feed_dict={ x: x_input })
      print(sess.run(get_next)) # output [ 0.52374458 0.71968478]
```

### 1.0.4 Task 4: Define, Compile and Train a Neural Network

**Let's create the simplest neural network.** It has 1 layer, and that layer has 1 neuron, and the input shape to it is just 1 value.

```
[ ]: model = tf.keras.Sequential([keras.layers.Dense(units=1, input_shape=[1])])
```

Now we compile our Neural Network, so we have to specify 2 functions, a loss and an optimizer.

1. The LOSS function measures the guessed answers against the known correct answers and measures how well or how badly it did.
2. The OPTIMIZER function makes another guess. Based on how the loss function went, it will try to minimize the loss.

It will repeat this for the number of EPOCHS which you will see shortly. But first, here's how we tell it to use 'MEAN SQUARED ERROR' for the loss and 'STOCHASTIC GRADIENT DESCENT' for the optimizer.

```
[ ]: model.compile(optimizer='sgd', loss='mean_squared_error')
```

**Providing the Data** Next, we are taking some data : 6 xs and 6 ys. You can see that the relationship between these is that  $y=4x+1$ , so where  $x = -1$ ,  $y=-3$  etc. etc.

We are using **Numpy** library by specifying the values as an np.array[]

```
[ ]: xs = np.array([-2.0, -1.0, 0.0, 1.0, 2.0, 3.0], dtype=float)
     ys = np.array([-2.0, 0.0, 6.0, 10.0, 14.0, 18.0], dtype=float)
```

**Training the Neural Network** The neural network learns the relationship between the Xs and Ys in the model.fit call. Making a guess, measuring how good or bad it is (aka the loss), using the optimizer to make another guess etc. It will do it for the number of epochs you specify. You'll see the loss on the right hand side, after running the code.

```
[ ]: model.fit(xs, ys, epochs=500)
```

You can use the **model.predict** method to have it figure out the Y for a previously unknown X. So, for example, if  $X = 50$  :

```
[ ]: print(model.predict([50.0]))
```

### 1.0.5 Task 4: Solve Simple Exercises

**Exercise 1** Which library is the framework for defining a neural network as a set of Sequential layers?

1. Numpy
2. Keras
3. Matplotlib

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**Answer 1** The correct answer is (2)

Keras is the framework for defining a neural network as a set of Sequential layers. It is a neural network library while TensorFlow is the open-source library for a number of various tasks in machine learning

**Exercise 2** Why we specify a LOSS and an OPTIMIZER functions?

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**Answer 2** To compile our Neural Network, we have to specify 2 functions, a loss and an optimizer. The LOSS function measures the guessed answers against the known correct answers and measures how well or how badly it did. The OPTIMIZER function makes another guess. Based on how the loss function went, it will try to minimize the loss.

There is not a single loss function that works for all kind of data. For example: Mean Square Error, Mean Absolute Error, Huber Loss, Log-Cosh Loss, Quantile Loss

**Exercise 3** What changes in the code below you should perform to increase the number of neurons in this simple Neural Network?

```
[ ]: model = tf.keras.Sequential([keras.layers.Dense(units=1, input_shape=[1])])
```

1. Increasing the input\_shape.
2. Creating another Dense Layer.
3. Increasing the units.

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**Answer 3** The correct answer is (3)

Each unit goes for one single neuron in the neural network.

## 1.1 Bonus: Extra Exercise!

In this tutorial, we'll create a simple neural network classifier in TensorFlow. The key advantage of this model over the Linear Classifier trained in the previous tutorial is that it can separate data



which is NOT linearly separable. We will implement this model for classifying images of hand-written digits from the so-called MNIST data-set.

We assume that you have the basic knowledge over the concept and you are just interested in the Tensorflow implementation of the Neural Nets.

The structure of the neural network that we're going to implement is as follows. We're using images of handw-ritten digits of the MNIST data which has 10 classes (i.e. digits from 0 to 9). The implemented network has 2 hidden layers: the first one with 200 hidden units (neurons) and the second one (also known as classifier layer) with 10 (number of classes) neurons.

nn89.png

```
[ ]: # imports
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt

img_h = img_w = 28          # MNIST images are 28x28
img_size_flat = img_h * img_w # 28x28=784, the total number of pixels
n_classes = 10              # Number of classes, one class per digit

def load_data(mode='train'):
    """
    Function to (download and) load the MNIST data
    :param mode: train or test
    :return: images and the corresponding labels
    """
    from tensorflow.examples.tutorials.mnist import input_data
    mnist = input_data.read_data_sets("MNIST_data/", one_hot=True)
    if mode == 'train':
        x_train, y_train, x_valid, y_valid = mnist.train.images, mnist.train.
→labels, \
                                                mnist.validation.images, mnist.
→validation.labels
        return x_train, y_train, x_valid, y_valid
    elif mode == 'test':
        x_test, y_test = mnist.test.images, mnist.test.labels
    return x_test, y_test

def randomize(x, y):
    """ Randomizes the order of data samples and their corresponding labels"""
    permutation = np.random.permutation(y.shape[0])
    shuffled_x = x[permutation, :]
    shuffled_y = y[permutation]
    return shuffled_x, shuffled_y

def get_next_batch(x, y, start, end):
```

```

x_batch = x[start:end]
y_batch = y[start:end]
return x_batch, y_batch

```

```

[:]: # Load MNIST data
x_train, y_train, x_valid, y_valid = load_data(mode='train')
print("Size of:")
print("- Training-set:\t\t{}".format(len(y_train)))
print("- Validation-set:\t{}".format(len(y_valid)))

```

```

[:]: # To get a better sense of the data, let's checkout the shapes of the loaded
      ↪arrays.
print('x_train:\t{}'.format(x_train.shape))
print('y_train:\t{}'.format(y_train.shape))
print('x_train:\t{}'.format(x_valid.shape))
print('y_valid:\t{}'.format(y_valid.shape))

```

```

[:]: y_valid[:5, :]

```

```

[:]: # Hyper-parameters
epochs = 10           # Total number of training epochs
batch_size = 100      # Training batch size
display_freq = 100    # Frequency of displaying the training results
learning_rate = 0.001 # The optimization initial learning rate

h1 = 200              # number of nodes in the 1st hidden layer

```

```

[:]: # weight and bias wrappers
def weight_variable(name, shape):
    """
    Create a weight variable with appropriate initialization
    :param name: weight name
    :param shape: weight shape
    :return: initialized weight variable
    """
    initer = tf.truncated_normal_initializer(stddev=0.01)
    return tf.get_variable('W_' + name,
                           dtype=tf.float32,
                           shape=shape,
                           initializer=initer)

def bias_variable(name, shape):
    """
    Create a bias variable with appropriate initialization
    :param name: bias variable name
    :param shape: bias variable shape
    :return: initialized bias variable
    """

```

```

initial = tf.constant(0., shape=shape, dtype=tf.float32)
return tf.get_variable('b_' + name,
                       dtype=tf.float32,
                       initializer=initial)

```

```

[: def fc_layer(x, num_units, name, use_relu=True):
    """
    Create a fully-connected layer
    :param x: input from previous layer
    :param num_units: number of hidden units in the fully-connected layer
    :param name: layer name
    :param use_relu: boolean to add ReLU non-linearity (or not)
    :return: The output array
    """
    in_dim = x.get_shape()[1]
    W = weight_variable(name, shape=[in_dim, num_units])
    b = bias_variable(name, [num_units])
    layer = tf.matmul(x, W)
    layer += b
    if use_relu:
        layer = tf.nn.relu(layer)
    return layer

```

```

[: # Create the graph for the linear model
# Placeholders for inputs (x) and outputs(y)
x = tf.placeholder(tf.float32, shape=[None, img_size_flat], name='X')
y = tf.placeholder(tf.float32, shape=[None, n_classes], name='Y')

```

```

[: # Create a fully-connected layer with h1 nodes as hidden layer
fc1 = fc_layer(x, h1, 'FC1', use_relu=True)
# Create a fully-connected layer with n_classes nodes as output layer
output_logits = fc_layer(fc1, n_classes, 'OUT', use_relu=False)

```

```

[: # Define the loss function, optimizer, and accuracy
loss = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits(labels=y,
    ↳ logits=output_logits), name='loss')
optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate, name='Adam-op').
    ↳ minimize(loss)
correct_prediction = tf.equal(tf.argmax(output_logits, 1), tf.argmax(y, 1),
    ↳ name='correct_pred')
accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32),
    ↳ name='accuracy')

# Network predictions
cls_prediction = tf.argmax(output_logits, axis=1, name='predictions')

```

```

[: # Create the op for initializing all variables
init = tf.global_variables_initializer()

```

```

[:]: # Create an interactive session (to keep the session in the other cells)
sess = tf.InteractiveSession()
# Initialize all variables
sess.run(init)
# Number of training iterations in each epoch
num_tr_iter = int(len(y_train) / batch_size)
for epoch in range(epochs):
    print('Training epoch: {}'.format(epoch + 1))
    # Randomly shuffle the training data at the beginning of each epoch
    x_train, y_train = randomize(x_train, y_train)
    for iteration in range(num_tr_iter):
        start = iteration * batch_size
        end = (iteration + 1) * batch_size
        x_batch, y_batch = get_next_batch(x_train, y_train, start, end)

        # Run optimization op (backprop)
        feed_dict_batch = {x: x_batch, y: y_batch}
        sess.run(optimizer, feed_dict=feed_dict_batch)

        if iteration % display_freq == 0:
            # Calculate and display the batch loss and accuracy
            loss_batch, acc_batch = sess.run([loss, accuracy],
                                             feed_dict=feed_dict_batch)

            print("iter {0:3d}: \t Loss={1:.2f}, \t Training Accuracy={2:.01%}".
                  format(iteration, loss_batch, acc_batch))

    # Run validation after every epoch
    feed_dict_valid = {x: x_valid[:1000], y: y_valid[:1000]}
    loss_valid, acc_valid = sess.run([loss, accuracy],
                                     feed_dict=feed_dict_valid)
    print('-----')
    print("Epoch: {0}, validation loss: {1:.2f}, validation accuracy: {2:.01%}".
          format(epoch + 1, loss_valid, acc_valid))
    print('-----')

[:]: # Test the network after training
# Accuracy
x_test, y_test = load_data(mode='test')
feed_dict_test = {x: x_test[:1000], y: y_test[:1000]}
loss_test, acc_test = sess.run([loss, accuracy], feed_dict=feed_dict_test)
print('-----')
print("Test loss: {0:.2f}, test accuracy: {1:.01%}".format(loss_test, acc_test))
print('-----')

[:]: def plot_images(images, cls_true, cls_pred=None, title=None):
    """
    Create figure with 3x3 sub-plots.

```

```

:param images: array of images to be plotted, (9, img_h*img_w)
:param cls_true: corresponding true labels (9,)
:param cls_pred: corresponding true labels (9,)
"""
fig, axes = plt.subplots(3, 3, figsize=(9, 9))
fig.subplots_adjust(hspace=0.3, wspace=0.3)
for i, ax in enumerate(axes.flat):
    # Plot image.
    ax.imshow(images[i].reshape(28, 28), cmap='binary')

    # Show true and predicted classes.
    if cls_pred is None:
        ax_title = "True: {0}".format(cls_true[i])
    else:
        ax_title = "True: {0}, Pred: {1}".format(cls_true[i], cls_pred[i])

    ax.set_title(ax_title)

    # Remove ticks from the plot.
    ax.set_xticks([])
    ax.set_yticks([])

if title:
    plt.suptitle(title, size=20)
plt.show(block=False)

def plot_example_errors(images, cls_true, cls_pred, title=None):
    """
    Function for plotting examples of images that have been mis-classified
    :param images: array of all images, (#imgs, img_h*img_w)
    :param cls_true: corresponding true labels, (#imgs,)
    :param cls_pred: corresponding predicted labels, (#imgs,)
    """
    # Negate the boolean array.
    incorrect = np.logical_not(np.equal(cls_pred, cls_true))

    # Get the images from the test-set that have been
    # incorrectly classified.
    incorrect_images = images[incorrect]

    # Get the true and predicted classes for those images.
    cls_pred = cls_pred[incorrect]
    cls_true = cls_true[incorrect]

    # Plot the first 9 images.
    plot_images(images=incorrect_images[0:9],
                cls_true=cls_true[0:9],

```

```
cls_pred=cls_pred[0:9],  
title=title)
```

```
[ ]: # Plot some of the correct and misclassified examples  
cls_pred = sess.run(cls_prediction, feed_dict=feed_dict_test)  
cls_true = np.argmax(y_test[:1000], axis=1)  
plot_images(x_test, cls_true, cls_pred, title='Correct Examples')  
plot_example_errors(x_test[:1000], cls_true, cls_pred, title='Misclassified_  
→Examples')  
plt.show()
```

## 1.2 Congratulations!

[Back To Tasks Page.](#)