

▼ Write a Program tutorial on tensorflow (XOR, AND)

Lab_Internal_1 _VDSS09_Abhishek_122021601009

LAB INTERNAL 1

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MTECH DSS

NUMPY : NumPy is a Python library used for working with arrays. It has functions for working in domain of linear algebra, fourier transform, and matrices.

TENSORFLOW :

1. It's an open source artificial intelligence library, that uses the data flow graphs, to build models.
2. It helps to create large-scale neural networks with many layers.
3. It is mainly used for:
 - Classification,
 - Perception,
 - Understanding,
 - Discovering,
 - Prediction
 - Creation

KERAS:

1. It offers consistent & simple APIs designed for human beings, not machines.
2. It helps in reducing cognitive load.
3. It minimizes the number of user actions required for common use cases, and it provides clear and actionable feedback upon user error.

▼ AND

```
#we import libraries
import numpy as np
from tensorflow.keras.models import Sequential
```

```
from tensorflow.keras.layers import Dense
```

```
#Dense layer is the regular deeply connected neural network layer.
```

```
#It does the below operation on the input and return the output.
```

```
training_data = np.array([[0,0],[0,1],[1,0],[1,1]], "float32")
```

```
# numpy array for training data
```

```
target_data = np.array([[0],[0],[0],[1]], "float32")
```

```
# numpy array for target data
```

```
model = Sequential()
```

```
#A Sequential model is appropriate for a plain stack of layers where
```

```
#each layer has exactly one input tensor and one output tensor & is not
```

```
#appropriate when our model has multiple inputs or multiple outputs.
```

```
#Any of your layers has multiple inputs or multiple outputs
```

```
model.add(Dense(16, input_dim=2, activation='relu'))
```

```
# we use activation function RELU, and has 16 nodes, 2D input dimension
```

```
model.add(Dense(1, activation='sigmoid'))
```

```
#we add output layer with activation function of sigmoid and has 1 output node
```

```
model.compile(loss='mean_squared_error',optimizer='adam',metrics=['binary_accuracy'])
```

```
#then we compile the model, we use adam optimiser for
```

```
model.fit(training_data, target_data, epochs=1000)
```

```
#then we fit the model with 1000 epochs and we give the arguments as
```

```
# training and target data
```

```
scores = model.evaluate(training_data, target_data)
```

```
# then we evaluate the model and find the scores
```

```
print("\n%s: %.2f%%" % (model.metrics_names[1], scores[1]*100))
```

```
# then we find the binary accuracy which was our metrices
```

```
print (model.predict(training_data).round())
```

```
#AND GATE
```

```
1/1 [=====] - 0s 9ms/step - loss: 0.0108 - binary_accuracy: 0.5000
```

```
Epoch 976/1000
```

```
1/1 [=====] - 0s 8ms/step - loss: 0.0108 - binary_accuracy: 0.5000
```

```
Epoch 977/1000
```

```
1/1 [=====] - 0s 10ms/step - loss: 0.0108 - binary_accuracy: 0.5000
```

```
Epoch 978/1000
```

```
1/1 [=====] - 0s 4ms/step - loss: 0.0108 - binary_accuracy: 0.5000
```

```
Epoch 979/1000
```

```
1/1 [=====] - 0s 11ms/step - loss: 0.0107 - binary_accuracy: 0.5000
```

```
Epoch 980/1000
```

```
1/1 [=====] - 0s 8ms/step - loss: 0.0107 - binary_accuracy: 0.5000
```

```
Epoch 981/1000
```

```
1/1 [=====] - 0s 8ms/step - loss: 0.0107 - binary_accuracy: 0.5000
```

```
Epoch 982/1000
```

```
1/1 [=====] - 0s 7ms/step - loss: 0.0106 - binary_accuracy: 0.5000
```

```
Epoch 983/1000
```

```
1/1 [=====] - 0s 12ms/step - loss: 0.0106 - binary_accuracy: 0.5000
```

```

Epoch 984/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0106 - binary_accuracy: 1.0000
Epoch 985/1000
1/1 [=====] - 0s 8ms/step - loss: 0.0106 - binary_accuracy: 1.0000
Epoch 986/1000
1/1 [=====] - 0s 7ms/step - loss: 0.0105 - binary_accuracy: 1.0000
Epoch 987/1000
1/1 [=====] - 0s 5ms/step - loss: 0.0105 - binary_accuracy: 1.0000
Epoch 988/1000
1/1 [=====] - 0s 7ms/step - loss: 0.0105 - binary_accuracy: 1.0000
Epoch 989/1000
1/1 [=====] - 0s 10ms/step - loss: 0.0105 - binary_accuracy: 1.0000
Epoch 990/1000
1/1 [=====] - 0s 16ms/step - loss: 0.0104 - binary_accuracy: 1.0000
Epoch 991/1000
1/1 [=====] - 0s 8ms/step - loss: 0.0104 - binary_accuracy: 1.0000
Epoch 992/1000
1/1 [=====] - 0s 5ms/step - loss: 0.0104 - binary_accuracy: 1.0000
Epoch 993/1000
1/1 [=====] - 0s 10ms/step - loss: 0.0103 - binary_accuracy: 1.0000
Epoch 994/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0103 - binary_accuracy: 1.0000
Epoch 995/1000
1/1 [=====] - 0s 5ms/step - loss: 0.0103 - binary_accuracy: 1.0000
Epoch 996/1000
1/1 [=====] - 0s 4ms/step - loss: 0.0103 - binary_accuracy: 1.0000
Epoch 997/1000
1/1 [=====] - 0s 7ms/step - loss: 0.0102 - binary_accuracy: 1.0000
Epoch 998/1000
1/1 [=====] - 0s 7ms/step - loss: 0.0102 - binary_accuracy: 1.0000
Epoch 999/1000
1/1 [=====] - 0s 9ms/step - loss: 0.0102 - binary_accuracy: 1.0000
Epoch 1000/1000
1/1 [=====] - 0s 7ms/step - loss: 0.0102 - binary_accuracy: 1.0000
1/1 [=====] - 0s 159ms/step - loss: 0.0101 - binary_accuracy: 1.0000

binary_accuracy: 100.00%
[[0.]
 [0.]
 [0.]
 [1.]]

```

▼ XOR Gate

```

#importing the libraries
import numpy as np
from keras.models import Sequential
from keras.layers.core import Dense

#Dense layer is the regular deeply connected neural network layer.
#It does the below operation on the input and return the output.

training_data = np.array([[0,0],[0,1],[1,0],[1,1]], "float32")
# numpy array for training data

```

```

target_data = np.array([[0],[1],[1],[0]], "float32")
# numpy array for target data

model = Sequential()
#A Sequential model is appropriate for a plain stack of layers where
#each layer has exactly one input tensor and one output tensor & is not
#appropriate when our model has multiple inputs or multiple outputs.
#Any of your layers has multiple inputs or multiple outputs

model.add(Dense(16, input_dim=2, activation='relu'))
# we use activation function RELU, and has 16 nodes, 2D input dimension

model.add(Dense(1, activation='sigmoid'))
#we add output layer with activation function of sigmoid and has 1 output node

model.compile(loss='mean_squared_error', optimizer='adam', metrics=['binary_accuracy'])
#then we compile the model, we use adam optimiser for

model.fit(training_data, target_data, epochs=1000)
#then we fit the model with 1000 epochs and we give the arguments
#as training and target data

scores = model.evaluate(training_data, target_data)
# then we evaluate the model and find the scores

print("\n%s: %.2f%%" % (model.metrics_names[1], scores[1]*100))
# then we find the binary accuracy which was our metrices

print (model.predict(training_data).round())
# then we print the model predictions

```

```

1/1 [=====] - 0s 7ms/step - loss: 0.0263 - binary_accuracy: 0.5000
Epoch 976/1000
1/1 [=====] - 0s 4ms/step - loss: 0.0263 - binary_accuracy: 0.5000
Epoch 977/1000
1/1 [=====] - 0s 8ms/step - loss: 0.0262 - binary_accuracy: 0.5000
Epoch 978/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0262 - binary_accuracy: 0.5000
Epoch 979/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0261 - binary_accuracy: 0.5000
Epoch 980/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0261 - binary_accuracy: 0.5000
Epoch 981/1000
1/1 [=====] - 0s 5ms/step - loss: 0.0260 - binary_accuracy: 0.5000
Epoch 982/1000
1/1 [=====] - 0s 14ms/step - loss: 0.0260 - binary_accuracy: 0.5000
Epoch 983/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0259 - binary_accuracy: 0.5000
Epoch 984/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0259 - binary_accuracy: 0.5000
Epoch 985/1000
1/1 [=====] - 0s 4ms/step - loss: 0.0258 - binary_accuracy: 0.5000
Epoch 986/1000
1/1 [=====] - 0s 5ms/step - loss: 0.0258 - binary_accuracy: 0.5000
Epoch 987/1000
1/1 [=====] - 0s 7ms/step - loss: 0.0257 - binary_accuracy: 0.5000
Epoch 988/1000

```

```
Epoch 988/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0257 - binary_accuracy: 1.0000
Epoch 989/1000
1/1 [=====] - 0s 9ms/step - loss: 0.0256 - binary_accuracy: 1.0000
Epoch 990/1000
1/1 [=====] - 0s 5ms/step - loss: 0.0256 - binary_accuracy: 1.0000
Epoch 991/1000
1/1 [=====] - 0s 7ms/step - loss: 0.0255 - binary_accuracy: 1.0000
Epoch 992/1000
1/1 [=====] - 0s 4ms/step - loss: 0.0255 - binary_accuracy: 1.0000
Epoch 993/1000
1/1 [=====] - 0s 7ms/step - loss: 0.0254 - binary_accuracy: 1.0000
Epoch 994/1000
1/1 [=====] - 0s 4ms/step - loss: 0.0254 - binary_accuracy: 1.0000
Epoch 995/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0253 - binary_accuracy: 1.0000
Epoch 996/1000
1/1 [=====] - 0s 9ms/step - loss: 0.0253 - binary_accuracy: 1.0000
Epoch 997/1000
1/1 [=====] - 0s 6ms/step - loss: 0.0252 - binary_accuracy: 1.0000
Epoch 998/1000
1/1 [=====] - 0s 8ms/step - loss: 0.0252 - binary_accuracy: 1.0000
Epoch 999/1000
1/1 [=====] - 0s 9ms/step - loss: 0.0251 - binary_accuracy: 1.0000
Epoch 1000/1000
1/1 [=====] - 0s 8ms/step - loss: 0.0251 - binary_accuracy: 1.0000
1/1 [=====] - 0s 274ms/step - loss: 0.0250 - binary_accuracy: 1.0000

binary_accuracy: 100.00%
[[0.]
 [1.]
 [1.]
 [0.]]
```