```
#CNN (A)
from keras.datasets import imdb
from keras.preprocessing import sequence
max_features = 10000
maxlen = 100
print('Loading data...')
(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=max_features)
print(len(x_train), 'train sequences')
print(len(x_test), 'test sequences')
print('Pad sequences (samples x time)')
x_train = sequence.pad_sequences(x_train, maxlen=maxlen)
x test = sequence.pad sequences(x test, maxlen=maxlen)
print('x_train shape:', x_train.shape)
print('x_test shape:', x_test.shape)

    □ Loading data...

    Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz">https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz</a>
    25000 train sequences
    25000 test sequences
    Pad sequences (samples x time)
    x_train shape: (25000, 100)
    x_test shape: (25000, 100)
#CNN (B)
from keras.datasets import imdb
from keras.models import Sequential
from keras.layers import Embedding, Conv1D, GlobalMaxPooling1D, Dense
from keras.preprocessing import sequence
max features = 10000
maxlen = 100
print('Loading data...')
(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=max_features)
print(len(x_train), 'train sequences')
print(len(x_test), 'test sequences')
print('Pad sequences (samples x time)')
x_train = sequence.pad_sequences(x_train, maxlen=maxlen)
x_{test} = sequence.pad_sequences(x_{test}, maxlen=maxlen)
\label{eq:print('x_train shape:', x_train.shape)} \\
print('x_test shape:', x_test.shape)
model = Sequential()
model.add(Embedding(max_features, 128, input_length=maxlen))
model.add(Conv1D(64, 5, activation='relu'))
model.add(GlobalMaxPooling1D())
model.add(Dense(1, activation='sigmoid'))
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
print('Training model...')
model.fit(x_train, y_train, epochs=3, batch_size=32, validation_split=0.2)
print('Evaluating model...')
loss, accuracy = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', accuracy)
    Loading data...
    25000 train sequences
    25000 test sequences
    Pad sequences (samples x time)
    x_train shape: (25000, 100)
    x_test shape: (25000, 100)
    Training model...
    Epoch 1/3
    Enoch 2/3
    Epoch 3/3
    Evaluating model..
    Test loss: 0.3903481364250183
```

Test accuracy: 0.8460400104522705

```
# RNN sentiment analysis on movie reviews
from keras.datasets import imdb
from keras.preprocessing.text import Tokenizer
from keras.utils import pad_sequences
from keras import Sequential
from keras.layers import Dense, SimpleRNN, Embedding, Flatten
# Load IMDb dataset
(X_train, y_train), (X_test, y_test) = imdb.load_data(num_words=10000)
# Pad sequences
X_train = pad_sequences(X_train, padding='post', maxlen=50)
X_test = pad_sequences(X_test, padding='post', maxlen=50)
# Define model
model = Sequential()
# Embedding layer
model.add(Embedding(10000, 32, input_length=50))
# SimpleRNN layer
model.add(SimpleRNN(32))
# Dense layer
model.add(Dense(1, activation='sigmoid'))
# Compile model
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
# Print model summary
model.summary()
    Model: "sequential_1"
     Layer (type)
                               Output Shape
                                                       Param #
     embedding_1 (Embedding)
                               (None, 50, 32)
                                                       320000
     simple_rnn_1 (SimpleRNN)
                                                       2080
                               (None, 32)
     dense_1 (Dense)
                               (None, 1)
                                                       33
    _____
    Total params: 322113 (1.23 MB)
    Trainable params: 322113 (1.23 MB)
    Non-trainable params: 0 (0.00 Byte)
# ADD TWO NUMBERS
#Step-1: Import
import numpy as np
import tensorflow as tf
from random import randrange
#Step-2: Generate Training Data
trainingInput = [[i, i + randrange(5000)] for i in range(1, 5000)]
trainingOutput = [(input [0] + input [1]) for input in trainingInput ]
testInput = [[5, 5], [1, 9], [2, 5], [6, 3], [1, 4]]
testOutput = [10, 10, 7, 9, 5]
#Step-3: Build the model
model = tf.keras.Sequential()
model.add(tf.keras.layers.Flatten(input_shape=(2,)))
model.add(tf.keras.layers.Dense(64, activation=tf.nn.relu))
model.add(tf.keras.layers.Dense(64, activation=tf.nn.relu))
model.add(tf.keras.layers.Dense(1))
#Step-4: Compile the model
model.compile(optimizer='adam', loss=tf.keras.losses.mae, metrics=['mae'])
#Step-5: Training
model.fit(trainingInput, trainingOutput, batch_size=5, epochs=3)
#Step-6: Final Evaluation and Prediction
test_loss, test_acc = model.evaluate(testInput, testOutput)
print("Test Accuracy : ", test_acc)
a = np.array([[3, 3000], [4, 5], [1,10], [2,10],[5,9], [4,10], [1,15]])
print(model.predict(a))
    Enoch 1/3
    1000/1000 [
                   Epoch 2/3
```

```
Epoch 3/3
            1/1 [============] - 1s 581ms/step - loss: 0.2576 - mae: 0.2576
            Test Accuracy : 0.25757989287376404
            1/1 [=======] - 0s 175ms/step
            [[3010.5828]
                    9.251566]
              [ 11.226904]
                   12.218241]
              [ 14.256031]
             [ 14.250777]
[ 16.240612]]
#ANN
from keras.models import Sequential
from keras.layers import Dense, Activation
import numpy as np
import pandas as pd
from sklearn import datasets
iris =datasets.load_iris()
X, y = datasets.load_iris( return_X_y = True)
from sklearn.model_selection import train_test_split
X_{\text{train}}, X_{\text{test}}, y_{\text{train}}, y_{\text{test}} = train_test_split(X_{\text{test}}, Y_{\text{test}}) = Define the network model and its arguments.
# Set the number of neurons/nodes for each layer:
model = Sequential()
model.add(Dense(2,input_shape=(4,)))
model.add(Activation('sigmoid'))
model.add(Dense(1))
model.add(Activation('sigmoid'))
#sgd = SGD(lr=0.0001, decay=1e-6, momentum=0.9, nesterov=True)
#model.compile(loss='categorical_crossentropy', optimizer=sgd,
metrics=['accuracy']# Compile the model and calculate its accuracy:
\verb|model.compile(loss='mean_squared_error', optimizer='sgd', \verb|metrics=['accuracy']|| \\ \verb|model.fit(X_train, y_train, batch_size=32, epochs=3)| \\ \\ \verb|model.fit(X_train, y_train, batch_size=32, epochs=32, 
# Print a summary of the Keras model:
model.summary()
#model.fit(X_train, y_train)
#model.fit(X_train, y_train,batch_size=32, epochs=3)
model.fit(X_train, y_train, epochs=5)
score = model.evaluate(X_test, y_test)
print(score)
           Model: "sequential 2"
```

Layer (type)	Output Shape	Param #
dense_2 (Dense)	(None, 2)	10
activation (Activation)	(None, 2)	0
dense_3 (Dense)	(None, 1)	3
activation_1 (Activation)	(None, 1)	0
Total params: 13 (52.00 Byte Trainable params: 13 (52.00 Non-trainable params: 0 (0.0 Epoch 1/5	Byte) 00 Byte)	
3/3 [===================================	•	
3/3 [===================================	======] - 0s 4ms/st	ep - loss: 0.678
3/3 [===================================	======] - 0s 4ms/st	ep - loss: 0.677
3/3 [===================================	======] - 0s 4ms/st	ep - loss: 0.676
3/3 [===================================		

 $\hbox{\tt [0.7189559936523438, 0.33333333432674408]}$

```
# importing Python library
import numpy as np
# define Unit Step Function
def unitStep(v):
   if v >= 0:
        return 1
    else:
        return 0
# design Perceptron Model
def perceptronModel(x, w, b):
   v = np.dot(w, x) + b
    y = unitStep(v)
    return y
# NOT Logic Function
# wNOT = -1, bNOT = 0.5
def NOT_logicFunction(x):
    wNOT = -1
    bNOT = 0.5
    return perceptronModel(x, wNOT, bNOT)
# AND Logic Function
\# here w1 = wAND1 = 1,
# w2 = wAND2 = 1, bAND = -1.5
{\tt def\ AND\_logicFunction}(x)\colon
    w = np.array([1, 1])
    bAND = -1.5
    return perceptronModel(x, w, bAND)
# OR Logic Function
\# w1 = 1, w2 = 1, bOR = -0.5
def OR_logicFunction(x):
   w = np.array([1, 1])
    bOR = -0.5
    return perceptronModel(x, w, bOR)
# XOR Logic Function
# with AND, OR and NOT
# function calls in sequence
def XOR_logicFunction(x):
    y1 = AND_logicFunction(x)
    y2 = OR_logicFunction(x)
    y3 = NOT_logicFunction(y1)
    final_x = np.array([y2, y3])
    finalOutput = AND_logicFunction(final_x)
    return finalOutput
# testing the Perceptron Model
test1 = np.array([0, 1])
test2 = np.array([1, 1])
test3 = np.array([0, 0])
test4 = np.array([1, 0])
print("XOR({}), {}) = {}".format(0, 1, XOR_logicFunction(test1)))
print("XOR(\{\},\ \{\})\ =\ \{\}".format(1,\ 1,\ XOR\_logicFunction(test2)))
print("XOR({}, {}) = {}".format(0, 0, XOR_logicFunction(test3)))
print("XOR({}, {}) = {}".format(1, 0, XOR_logicFunction(test4)))
     XOR(0, 1) = 1
     XOR(1, 1) = 0
     XOR(0, 0) = 0
     XOR(1, 0) = 1
```