# Develop a Chatbot Using Python, NLTK, and TensorFlow

```
In [1]: # Run this cell so you do not see GPU availability errors from tensorflow
import os
os.environ['TF_CPP_MIN_LOG_LEVEL'] = '3'
```

### Task 1: Import the Required Libraries

```
import json
import pickle
import random
import nltk
from nltk.stem import WordNetLemmatizer
import numpy as np
from tensorflow.keras.models import Sequential
from tensorflow.keras.optimizers import SGD
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Activation
from tensorflow.keras.layers import Dropout
import matplotlib.pyplot as plt
from matplotlib import rcParams
```

#### Task 2: Load the Data

```
In [5]: #Load the data
    nltk.download('omw-1.4')
    with open(r'intents.json') as data:
        intents = json.loads(data.read())

[nltk_data] Downloading package omw-1.4 to /root/nltk_data...
[nltk_data] Package omw-1.4 is already up-to-date!
```

#### **Task 3: Tokenization**

```
In [6]: words = []
    classes = []
    documents = []

for intent in intents['intents']:
        for pattern in intent['patterns']:
            w = nltk.word_tokenize(pattern)
            words.extend(w)
            classes.append(intent['tag'])
            documents.append((w, intent['tag']))
```

#### **Task 4: Lemmatization**

```
In [8]: lemmatizer = WordNetLemmatizer()
   ignore_words = ['?', '!']
   words = [lemmatizer.lemmatize(w.lower())for w in words if w not in ignore_words]
   words = sorted(list(set(words)))
   classes = sorted(list(set(classes)))

pickle.dump(words, open('words.pkl', 'wb'))
   pickle.dump(classes, open('classes.pkl', 'wb'))
```

## Task 5: Create Data for Training

```
In [9]: training = []
        output_empty = [0]*len(classes)
        for doc in documents:
            bag = []
            pattern_words = doc[0]
            pattern_words = [lemmatizer.lemmatize(word.lower())for word in pattern_words]
            for w in words:
                 if w in pattern_words:
                    bag.append(1)
                 else: bag.append(0)
            output_row = list(output_empty)
            output_row[classes.index(doc[1])] = 1
            training.append([bag, output_row])
        random.shuffle(training)
        a = int(0.7*len(training))
        training = np.array(training, dtype = 'object')
        X_train = list(training[:a, 0])
        y_train = list(training[:a, 1])
        X_val = list(training[a:, 0])
        y val = list(training[a:, 1])
```

# Task 6: Design the Model

```
In [10]: model = Sequential()
  model.add(Dense(128, input_shape=(len(X_train[0]),), activation='relu'))
  model.add(Dropout(0.5))
  model.add(Dense(64, activation='relu'))
  model.add(Dropout(0.5))
  model.add(Dense(len(y_train[0]), activation='softmax'))
```

#### Task 7: Train and Save the Model

```
Epoch 1/200
625 - val_loss: 2.2055 - val_accuracy: 0.0000e+00
562 - val_loss: 2.2325 - val_accuracy: 0.0000e+00
Epoch 3/200
7/7 [============] - 0s 12ms/step - loss: 2.2104 - accuracy: 0.0
938 - val_loss: 2.2512 - val_accuracy: 0.0000e+00
Epoch 4/200
188 - val_loss: 2.2456 - val_accuracy: 0.0000e+00
Epoch 5/200
125 - val_loss: 2.2549 - val_accuracy: 0.0000e+00
Epoch 6/200
750 - val_loss: 2.2266 - val_accuracy: 0.0667
Epoch 7/200
7/7 [================== ] - 0s 13ms/step - loss: 1.7730 - accuracy: 0.4
062 - val_loss: 2.2071 - val_accuracy: 0.0667
Epoch 8/200
438 - val_loss: 2.1752 - val_accuracy: 0.0667
Epoch 9/200
7/7 [==========] - 0s 14ms/step - loss: 1.5751 - accuracy: 0.4
688 - val_loss: 2.1223 - val_accuracy: 0.2000
Epoch 10/200
625 - val_loss: 2.0427 - val_accuracy: 0.2667
Epoch 11/200
7/7 [==========] - 0s 13ms/step - loss: 1.4701 - accuracy: 0.5
312 - val_loss: 1.9338 - val_accuracy: 0.2667
Epoch 12/200
500 - val_loss: 1.8588 - val_accuracy: 0.2667
875 - val_loss: 1.7940 - val_accuracy: 0.2667
Epoch 14/200
125 - val_loss: 1.7093 - val_accuracy: 0.3333
Epoch 15/200
7/7 [================== ] - 0s 12ms/step - loss: 1.1757 - accuracy: 0.6
250 - val_loss: 1.6763 - val_accuracy: 0.4000
Epoch 16/200
812 - val_loss: 1.6333 - val_accuracy: 0.4000
Epoch 17/200
562 - val_loss: 1.5206 - val_accuracy: 0.4667
Epoch 18/200
438 - val_loss: 1.4496 - val_accuracy: 0.5333
```

```
750 - val_loss: 1.4279 - val_accuracy: 0.4667
Epoch 20/200
438 - val_loss: 1.3996 - val_accuracy: 0.4667
Epoch 21/200
7/7 [========== ] - 0s 12ms/step - loss: 0.6128 - accuracy: 0.8
438 - val_loss: 1.3407 - val_accuracy: 0.6000
Epoch 22/200
438 - val_loss: 1.2556 - val_accuracy: 0.6667
Epoch 23/200
375 - val_loss: 1.2088 - val_accuracy: 0.6000
7/7 [=============== ] - 0s 13ms/step - loss: 0.4832 - accuracy: 0.8
438 - val_loss: 1.1384 - val_accuracy: 0.6667
Epoch 25/200
7/7 [===========] - 0s 13ms/step - loss: 0.4309 - accuracy: 0.9
062 - val_loss: 1.0939 - val_accuracy: 0.6667
Epoch 26/200
7/7 [===========] - 0s 13ms/step - loss: 0.4147 - accuracy: 0.9
062 - val_loss: 1.0767 - val_accuracy: 0.6667
Epoch 27/200
062 - val_loss: 1.0560 - val_accuracy: 0.6000
Epoch 28/200
7/7 [==========] - 0s 12ms/step - loss: 0.4768 - accuracy: 0.9
062 - val_loss: 1.0434 - val_accuracy: 0.6667
Epoch 29/200
750 - val loss: 1.0315 - val accuracy: 0.6667
7/7 [==========] - 0s 12ms/step - loss: 0.4184 - accuracy: 0.9
375 - val_loss: 1.0580 - val_accuracy: 0.6667
Epoch 31/200
375 - val_loss: 1.1151 - val_accuracy: 0.6667
Epoch 32/200
000 - val_loss: 1.1221 - val_accuracy: 0.6667
Epoch 33/200
375 - val_loss: 1.0824 - val_accuracy: 0.6667
Epoch 34/200
438 - val_loss: 1.0636 - val_accuracy: 0.6667
Epoch 35/200
750 - val_loss: 1.1538 - val_accuracy: 0.6667
Epoch 36/200
000 - val_loss: 1.1656 - val_accuracy: 0.6667
Epoch 37/200
000 - val_loss: 1.1479 - val_accuracy: 0.6667
Epoch 38/200
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000 - val_loss: 1.0596 - val_accuracy: 0.6667
Epoch 39/200
375 - val_loss: 1.0283 - val_accuracy: 0.6667
Epoch 40/200
750 - val_loss: 0.9859 - val_accuracy: 0.6667
Epoch 41/200
7/7 [=================] - 0s 12ms/step - loss: 0.1960 - accuracy: 1.0
000 - val_loss: 0.9695 - val_accuracy: 0.6667
Epoch 42/200
375 - val_loss: 0.9957 - val_accuracy: 0.6667
Epoch 43/200
062 - val_loss: 1.0210 - val_accuracy: 0.6667
Epoch 44/200
000 - val_loss: 1.0439 - val_accuracy: 0.6667
Epoch 45/200
7/7 [==========] - 0s 12ms/step - loss: 0.1400 - accuracy: 1.0
000 - val_loss: 1.0371 - val_accuracy: 0.6667
Epoch 46/200
688 - val_loss: 0.9631 - val_accuracy: 0.6667
Epoch 47/200
7/7 [==========] - 0s 13ms/step - loss: 0.2358 - accuracy: 0.9
688 - val_loss: 0.9645 - val_accuracy: 0.6667
Epoch 48/200
688 - val_loss: 0.9953 - val_accuracy: 0.6667
Epoch 49/200
7/7 [==========] - 0s 12ms/step - loss: 0.1232 - accuracy: 1.0
000 - val_loss: 1.0202 - val_accuracy: 0.6667
Epoch 50/200
7/7 [==========] - 0s 12ms/step - loss: 0.1437 - accuracy: 0.9
688 - val_loss: 1.0312 - val_accuracy: 0.6667
Epoch 51/200
000 - val_loss: 1.0649 - val_accuracy: 0.6667
Epoch 52/200
375 - val_loss: 1.1078 - val_accuracy: 0.6667
Epoch 53/200
000 - val_loss: 1.1394 - val_accuracy: 0.6667
Epoch 54/200
7/7 [==========] - 0s 13ms/step - loss: 0.0939 - accuracy: 1.0
000 - val loss: 1.1118 - val accuracy: 0.6667
Epoch 55/200
688 - val_loss: 1.0632 - val_accuracy: 0.6667
Epoch 56/200
000 - val_loss: 1.0198 - val_accuracy: 0.6667
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Epoch 57/200
7/7 [================== ] - 0s 13ms/step - loss: 0.1117 - accuracy: 1.0
000 - val_loss: 1.0248 - val_accuracy: 0.7333
Epoch 58/200
7/7 [===================] - 0s 12ms/step - loss: 0.0847 - accuracy: 0.9
688 - val_loss: 1.0436 - val_accuracy: 0.6667
Epoch 59/200
7/7 [=============] - 0s 13ms/step - loss: 0.0715 - accuracy: 1.0
000 - val loss: 1.0720 - val accuracy: 0.6667
Epoch 60/200
000 - val_loss: 1.0802 - val_accuracy: 0.6667
Epoch 61/200
000 - val_loss: 1.0460 - val_accuracy: 0.6667
Epoch 62/200
375 - val_loss: 1.0484 - val_accuracy: 0.6667
Epoch 63/200
7/7 [=================== ] - 0s 12ms/step - loss: 0.1170 - accuracy: 1.0
000 - val_loss: 1.0840 - val_accuracy: 0.7333
Epoch 64/200
000 - val_loss: 1.1159 - val_accuracy: 0.6667
Epoch 65/200
7/7 [==========] - 0s 12ms/step - loss: 0.0687 - accuracy: 1.0
000 - val_loss: 1.1345 - val_accuracy: 0.6667
Epoch 66/200
688 - val_loss: 1.1365 - val_accuracy: 0.6667
Epoch 67/200
7/7 [==========] - 0s 13ms/step - loss: 0.0745 - accuracy: 1.0
000 - val_loss: 1.1371 - val_accuracy: 0.6667
Epoch 68/200
000 - val_loss: 1.1289 - val_accuracy: 0.6667
000 - val_loss: 1.1221 - val_accuracy: 0.6667
Epoch 70/200
000 - val_loss: 1.1158 - val_accuracy: 0.6667
Epoch 71/200
7/7 [====================] - 0s 12ms/step - loss: 0.0769 - accuracy: 1.0
000 - val_loss: 1.1059 - val_accuracy: 0.6667
Epoch 72/200
000 - val_loss: 1.1015 - val_accuracy: 0.6667
Epoch 73/200
000 - val_loss: 1.0887 - val_accuracy: 0.6667
Epoch 74/200
688 - val_loss: 1.0694 - val_accuracy: 0.6667
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000 - val_loss: 1.0639 - val_accuracy: 0.6667
Epoch 76/200
000 - val_loss: 1.0649 - val_accuracy: 0.6667
Epoch 77/200
7/7 [=========== ] - 0s 12ms/step - loss: 0.0173 - accuracy: 1.0
000 - val_loss: 1.0722 - val_accuracy: 0.6667
Epoch 78/200
375 - val_loss: 1.0749 - val_accuracy: 0.6667
Epoch 79/200
000 - val_loss: 1.0506 - val_accuracy: 0.6667
7/7 [================ ] - 0s 13ms/step - loss: 0.0616 - accuracy: 1.0
000 - val_loss: 1.0815 - val_accuracy: 0.6667
Epoch 81/200
7/7 [===========] - 0s 12ms/step - loss: 0.0901 - accuracy: 0.9
688 - val_loss: 1.0586 - val_accuracy: 0.6667
Epoch 82/200
000 - val_loss: 1.0493 - val_accuracy: 0.6667
Epoch 83/200
000 - val_loss: 1.0689 - val_accuracy: 0.6667
Epoch 84/200
7/7 [==========] - 0s 12ms/step - loss: 0.1262 - accuracy: 0.9
688 - val_loss: 1.0842 - val_accuracy: 0.6667
Epoch 85/200
000 - val loss: 1.0930 - val accuracy: 0.6667
000 - val_loss: 1.1047 - val_accuracy: 0.6667
Epoch 87/200
000 - val_loss: 1.1062 - val_accuracy: 0.6667
Epoch 88/200
688 - val_loss: 1.0812 - val_accuracy: 0.6667
Epoch 89/200
000 - val_loss: 1.0862 - val_accuracy: 0.6667
Epoch 90/200
375 - val_loss: 1.0964 - val_accuracy: 0.6667
Epoch 91/200
688 - val_loss: 1.1316 - val_accuracy: 0.7333
Epoch 92/200
7/7 [==========] - 0s 12ms/step - loss: 0.0796 - accuracy: 0.9
688 - val_loss: 1.2488 - val_accuracy: 0.7333
Epoch 93/200
000 - val_loss: 1.2984 - val_accuracy: 0.7333
Epoch 94/200
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000 - val_loss: 1.2628 - val_accuracy: 0.6667
Epoch 95/200
000 - val_loss: 1.2647 - val_accuracy: 0.6667
Epoch 96/200
7/7 [================== ] - 0s 13ms/step - loss: 0.0161 - accuracy: 1.0
000 - val_loss: 1.2657 - val_accuracy: 0.6667
7/7 [====================] - 0s 13ms/step - loss: 0.0506 - accuracy: 1.0
000 - val_loss: 1.2252 - val_accuracy: 0.6667
Epoch 98/200
000 - val_loss: 1.2033 - val_accuracy: 0.6667
Epoch 99/200
000 - val_loss: 1.1942 - val_accuracy: 0.7333
Epoch 100/200
688 - val_loss: 1.1899 - val_accuracy: 0.6667
Epoch 101/200
7/7 [============] - 0s 12ms/step - loss: 0.0920 - accuracy: 0.9
688 - val_loss: 1.1606 - val_accuracy: 0.7333
Epoch 102/200
688 - val_loss: 1.1770 - val_accuracy: 0.7333
Epoch 103/200
7/7 [==========] - 0s 13ms/step - loss: 0.0521 - accuracy: 0.9
688 - val_loss: 1.2038 - val_accuracy: 0.6000
Epoch 104/200
000 - val_loss: 1.2060 - val_accuracy: 0.6000
Epoch 105/200
7/7 [==========] - 0s 13ms/step - loss: 0.0531 - accuracy: 0.9
688 - val_loss: 1.1534 - val_accuracy: 0.7333
Epoch 106/200
7/7 [==========] - 0s 12ms/step - loss: 0.1528 - accuracy: 0.9
688 - val_loss: 1.1300 - val_accuracy: 0.7333
Epoch 107/200
000 - val_loss: 1.1246 - val_accuracy: 0.7333
Epoch 108/200
000 - val_loss: 1.1287 - val_accuracy: 0.7333
Epoch 109/200
000 - val_loss: 1.1311 - val_accuracy: 0.7333
Epoch 110/200
7/7 [==========] - 0s 13ms/step - loss: 0.0279 - accuracy: 1.0
000 - val loss: 1.1332 - val accuracy: 0.7333
Epoch 111/200
688 - val_loss: 1.1470 - val_accuracy: 0.7333
Epoch 112/200
000 - val_loss: 1.1595 - val_accuracy: 0.7333
```

```
Epoch 113/200
7/7 [=================== ] - 0s 12ms/step - loss: 0.0697 - accuracy: 0.9
688 - val_loss: 1.1601 - val_accuracy: 0.7333
Epoch 114/200
7/7 [===================] - 0s 13ms/step - loss: 0.0377 - accuracy: 1.0
000 - val_loss: 1.1555 - val_accuracy: 0.7333
Epoch 115/200
7/7 [============] - 0s 13ms/step - loss: 0.0274 - accuracy: 1.0
000 - val_loss: 1.1457 - val_accuracy: 0.7333
Epoch 116/200
000 - val_loss: 1.1534 - val_accuracy: 0.6667
Epoch 117/200
000 - val_loss: 1.1671 - val_accuracy: 0.6667
Epoch 118/200
000 - val_loss: 1.1835 - val_accuracy: 0.6667
Epoch 119/200
7/7 [================== ] - 0s 13ms/step - loss: 0.0151 - accuracy: 1.0
000 - val_loss: 1.2058 - val_accuracy: 0.6667
Epoch 120/200
000 - val_loss: 1.2130 - val_accuracy: 0.6667
Epoch 121/200
000 - val_loss: 1.2168 - val_accuracy: 0.6667
Epoch 122/200
375 - val_loss: 1.2615 - val_accuracy: 0.6667
Epoch 123/200
7/7 [==========] - 0s 13ms/step - loss: 0.0150 - accuracy: 1.0
000 - val_loss: 1.3089 - val_accuracy: 0.6667
Epoch 124/200
688 - val_loss: 1.3417 - val_accuracy: 0.6667
7/7 [============= ] - 0s 13ms/step - loss: 0.0071 - accuracy: 1.0
000 - val_loss: 1.3625 - val_accuracy: 0.6667
Epoch 126/200
000 - val_loss: 1.3523 - val_accuracy: 0.6667
Epoch 127/200
7/7 [=================== ] - 0s 15ms/step - loss: 0.0283 - accuracy: 1.0
000 - val_loss: 1.3583 - val_accuracy: 0.6667
Epoch 128/200
000 - val_loss: 1.3704 - val_accuracy: 0.6667
Epoch 129/200
000 - val_loss: 1.3721 - val_accuracy: 0.6667
Epoch 130/200
000 - val_loss: 1.4315 - val_accuracy: 0.6667
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000 - val_loss: 1.4543 - val_accuracy: 0.6667
Epoch 132/200
000 - val_loss: 1.4592 - val_accuracy: 0.6667
Epoch 133/200
000 - val_loss: 1.4694 - val_accuracy: 0.6667
Epoch 134/200
000 - val_loss: 1.4783 - val_accuracy: 0.6667
Epoch 135/200
688 - val_loss: 1.4822 - val_accuracy: 0.6667
7/7 [=============== ] - 0s 13ms/step - loss: 0.0189 - accuracy: 1.0
000 - val_loss: 1.4774 - val_accuracy: 0.6667
Epoch 137/200
000 - val_loss: 1.4243 - val_accuracy: 0.6667
Epoch 138/200
000 - val_loss: 1.3780 - val_accuracy: 0.6667
Epoch 139/200
000 - val_loss: 1.3501 - val_accuracy: 0.6667
Epoch 140/200
7/7 [==========] - 0s 12ms/step - loss: 0.1253 - accuracy: 0.9
688 - val_loss: 1.3472 - val_accuracy: 0.6667
Epoch 141/200
000 - val loss: 1.3479 - val accuracy: 0.7333
000 - val_loss: 1.3665 - val_accuracy: 0.7333
Epoch 143/200
000 - val_loss: 1.3972 - val_accuracy: 0.7333
Epoch 144/200
000 - val_loss: 1.3821 - val_accuracy: 0.7333
Epoch 145/200
000 - val_loss: 1.3338 - val_accuracy: 0.7333
Epoch 146/200
688 - val_loss: 1.3184 - val_accuracy: 0.7333
Epoch 147/200
000 - val_loss: 1.3190 - val_accuracy: 0.7333
Epoch 148/200
000 - val_loss: 1.3083 - val_accuracy: 0.7333
Epoch 149/200
000 - val_loss: 1.2948 - val_accuracy: 0.7333
Epoch 150/200
```

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000 - val_loss: 1.2777 - val_accuracy: 0.6667
Epoch 151/200
000 - val_loss: 1.2774 - val_accuracy: 0.6667
Epoch 152/200
7/7 [=====================] - 0s 12ms/step - loss: 0.0038 - accuracy: 1.0
000 - val_loss: 1.2809 - val_accuracy: 0.6667
Epoch 153/200
7/7 [==================] - 0s 12ms/step - loss: 0.0819 - accuracy: 0.9
688 - val_loss: 1.2869 - val_accuracy: 0.6667
Epoch 154/200
000 - val_loss: 1.2930 - val_accuracy: 0.6667
Epoch 155/200
000 - val_loss: 1.3002 - val_accuracy: 0.6667
Epoch 156/200
000 - val_loss: 1.3148 - val_accuracy: 0.6667
Epoch 157/200
7/7 [==========] - 0s 12ms/step - loss: 0.0158 - accuracy: 1.0
000 - val_loss: 1.3186 - val_accuracy: 0.6667
Epoch 158/200
688 - val_loss: 1.3894 - val_accuracy: 0.6667
Epoch 159/200
000 - val_loss: 1.6595 - val_accuracy: 0.6667
Epoch 160/200
000 - val_loss: 1.7891 - val_accuracy: 0.6667
Epoch 161/200
7/7 [==========] - 0s 13ms/step - loss: 0.0390 - accuracy: 1.0
000 - val_loss: 1.8364 - val_accuracy: 0.6667
Epoch 162/200
7/7 [==========] - 0s 13ms/step - loss: 0.0553 - accuracy: 1.0
000 - val_loss: 1.8188 - val_accuracy: 0.6667
Epoch 163/200
000 - val_loss: 1.8457 - val_accuracy: 0.6667
Epoch 164/200
688 - val_loss: 1.7515 - val_accuracy: 0.6667
Epoch 165/200
000 - val_loss: 1.6808 - val_accuracy: 0.6667
Epoch 166/200
7/7 [==========] - 0s 13ms/step - loss: 0.0068 - accuracy: 1.0
000 - val loss: 1.6467 - val accuracy: 0.6667
Epoch 167/200
688 - val_loss: 1.6412 - val_accuracy: 0.6667
Epoch 168/200
688 - val_loss: 1.5411 - val_accuracy: 0.7333
```

```
Epoch 169/200
7/7 [=================== ] - 0s 13ms/step - loss: 0.0138 - accuracy: 1.0
000 - val_loss: 1.4700 - val_accuracy: 0.7333
Epoch 170/200
7/7 [====================] - 0s 13ms/step - loss: 0.0138 - accuracy: 1.0
000 - val_loss: 1.4348 - val_accuracy: 0.7333
Epoch 171/200
7/7 [============] - 0s 13ms/step - loss: 0.0042 - accuracy: 1.0
000 - val loss: 1.4233 - val accuracy: 0.7333
Epoch 172/200
688 - val_loss: 1.3906 - val_accuracy: 0.7333
Epoch 173/200
000 - val_loss: 1.3634 - val_accuracy: 0.7333
Epoch 174/200
000 - val_loss: 1.3575 - val_accuracy: 0.7333
Epoch 175/200
7/7 [=================== ] - 0s 13ms/step - loss: 0.0401 - accuracy: 1.0
000 - val_loss: 1.3522 - val_accuracy: 0.7333
Epoch 176/200
000 - val_loss: 1.3477 - val_accuracy: 0.6667
Epoch 177/200
000 - val_loss: 1.3502 - val_accuracy: 0.6667
Epoch 178/200
000 - val_loss: 1.3493 - val_accuracy: 0.6667
Epoch 179/200
7/7 [==========] - 0s 13ms/step - loss: 0.0656 - accuracy: 0.9
688 - val_loss: 1.3379 - val_accuracy: 0.6667
Epoch 180/200
688 - val_loss: 1.3714 - val_accuracy: 0.6667
Epoch 181/200
7/7 [============== ] - 0s 12ms/step - loss: 0.0176 - accuracy: 1.0
000 - val_loss: 1.4049 - val_accuracy: 0.6667
Epoch 182/200
000 - val_loss: 1.4250 - val_accuracy: 0.6667
Epoch 183/200
7/7 [================== ] - 0s 13ms/step - loss: 0.0114 - accuracy: 1.0
000 - val_loss: 1.4356 - val_accuracy: 0.6667
Epoch 184/200
688 - val_loss: 1.4570 - val_accuracy: 0.6667
Epoch 185/200
000 - val_loss: 1.5621 - val_accuracy: 0.6667
Epoch 186/200
000 - val_loss: 1.6246 - val_accuracy: 0.6667
```

```
000 - val_loss: 1.6463 - val_accuracy: 0.6667
Epoch 188/200
000 - val_loss: 1.6548 - val_accuracy: 0.6667
Epoch 189/200
000 - val_loss: 1.6476 - val_accuracy: 0.6667
Epoch 190/200
000 - val_loss: 1.6451 - val_accuracy: 0.6667
Epoch 191/200
7/7 [==========] - 0s 12ms/step - loss: 0.0042 - accuracy: 1.0
000 - val_loss: 1.6447 - val_accuracy: 0.6667
000 - val_loss: 1.6422 - val_accuracy: 0.6667
Epoch 193/200
000 - val_loss: 1.6428 - val_accuracy: 0.6667
Epoch 194/200
7/7 [=================== ] - 0s 12ms/step - loss: 0.0107 - accuracy: 1.0
000 - val_loss: 1.6432 - val_accuracy: 0.6667
Epoch 195/200
7/7 [==========] - 0s 13ms/step - loss: 0.0348 - accuracy: 1.0
000 - val_loss: 1.6430 - val_accuracy: 0.6667
Epoch 196/200
7/7 [==========] - 0s 13ms/step - loss: 0.0232 - accuracy: 1.0
000 - val_loss: 1.5998 - val_accuracy: 0.6667
Epoch 197/200
000 - val_loss: 1.5053 - val_accuracy: 0.6667
000 - val_loss: 1.4753 - val_accuracy: 0.6667
Epoch 199/200
000 - val_loss: 1.4856 - val_accuracy: 0.6667
Epoch 200/200
000 - val_loss: 1.4908 - val_accuracy: 0.6667
```

# Task 8: Print the training curves

```
In [13]: plt.rcParams["figure.figsize"]=(12,8)
N = np.arange(0, 200)
plt.style.use("ggplot")
plt.figure()
plt.plot(N, hist.history["loss"], label = 'train_loss')
plt.plot(N, hist.history["val_loss"], label = 'val_loss')
plt.plot(N, hist.history['accuracy'], label = 'accuracy')
plt.plot(N, hist.history["val_accuracy"], label = 'val_accuracy')
plt.title("Training Loss and Accuracy")
plt.xlabel("Epoch #")
```

```
plt.ylabel("Loss/Accuracy")
plt.legend()
```

Out[13]: <matplotlib.legend.Legend at 0x787766a52b20>



In []: