## In [1]: pip install keras

Requirement already satisfied: keras in c:\users\91944\anaconda3\lib\site-packages (2.4. 3)

Requirement already satisfied: scipy>=0.14 in c:\users\91944\anaconda3\lib\site-packages (from keras) (1.5.2)

Requirement already satisfied: numpy>=1.9.1 in c:\users\91944\anaconda3\lib\site-package s (from keras) (1.19.2)

Requirement already satisfied: pyyaml in c:\users\91944\anaconda3\lib\site-packages (fro m keras) (5.3.1)

Requirement already satisfied: h5py in c:\users\91944\anaconda3\lib\site-packages (from keras) (2.10.0)

Requirement already satisfied: six in c:\users\91944\anaconda3\lib\site-packages (from h 5py->keras) (1.15.0)

Note: you may need to restart the kernel to use updated packages.

## In [2]: pip install tensorflow

Requirement already satisfied: tensorflow in c:\users\91944\anaconda3\lib\site-packages (2.4.1)

Requirement already satisfied: protobuf>=3.9.2 in c:\users\91944\anaconda3\lib\site-pack ages (from tensorflow) (3.15.7)

Requirement already satisfied: tensorflow-estimator<2.5.0,>=2.4.0 in c:\users\91944\anac onda3\lib\site-packages (from tensorflow) (2.4.0)

Requirement already satisfied: google-pasta~=0.2 in c:\users\91944\anaconda3\lib\site-packages (from tensorflow) (0.2.0)

Requirement already satisfied: grpcio~=1.32.0 in c:\users\91944\anaconda3\lib\site-packa ges (from tensorflow) (1.32.0)

Requirement already satisfied: opt-einsum~=3.3.0 in c:\users\91944\anaconda3\lib\site-packages (from tensorflow) (3.3.0)

Requirement already satisfied: keras-preprocessing~=1.1.2 in c:\users\91944\anaconda3\lib\site-packages (from tensorflow) (1.1.2)

Requirement already satisfied:  $six\sim=1.15.0$  in c:\users\91944\anaconda3\lib\site-packages (from tensorflow) (1.15.0)

Requirement already satisfied: tensorboard~=2.4 in c:\users\91944\anaconda3\lib\site-pac kages (from tensorflow) (2.4.1)

Requirement already satisfied: wrapt~=1.12.1 in c:\users\91944\anaconda3\lib\site-packag es (from tensorflow) (1.12.1)

Requirement already satisfied: astunparse~=1.6.3 in c:\users\91944\anaconda3\lib\site-pa ckages (from tensorflow) (1.6.3)

Requirement already satisfied: flatbuffers~=1.12.0 in c:\users\91944\anaconda3\lib\site-packages (from tensorflow) (1.12)

Requirement already satisfied: absl-py~=0.10 in c:\users\91944\anaconda3\lib\site-packag es (from tensorflow) (0.12.0)

Requirement already satisfied: termcolor~=1.1.0 in c:\users\91944\anaconda3\lib\site-pac kages (from tensorflow) (1.1.0)

Requirement already satisfied: typing-extensions~=3.7.4 in c:\users\91944\anaconda3\lib\site-packages (from tensorflow) (3.7.4.3)

Requirement already satisfied: gast==0.3.3 in c:\users\91944\anaconda3\lib\site-packages (from tensorflow) (0.3.3)

Requirement already satisfied: numpy~=1.19.2 in c:\users\91944\anaconda3\lib\site-packag es (from tensorflow) (1.19.2)

Requirement already satisfied: wheel~=0.35 in c:\users\91944\anaconda3\lib\site-packages (from tensorflow) (0.35.1)

Requirement already satisfied: h5py~=2.10.0 in c:\users\91944\anaconda3\lib\site-package s (from tensorflow) (2.10.0)

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Requirement already satisfied: markdown>=2.6.8 in c:\users\91944\anaconda3\lib\site-pack ages (from tensorboard~=2.4->tensorflow) (3.3.4)

Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in c:\users\91944\anaconda3 \lib\site-packages (from tensorboard~=2.4->tensorflow) (1.8.0)

Requirement already satisfied: setuptools>=41.0.0 in c:\users\91944\anaconda3\lib\site-p ackages (from tensorboard~=2.4->tensorflow) (50.3.1.post20201107)

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>tensorflow) (3.1.0)
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es (from rsa<5,>=3.1.4; python version >= "3.6"->google-auth<2,>=1.6.3->tensorboard~=2.4
->tensorflow) (0.4.8)
Note: you may need to restart the kernel to use updated packages.
```

```
import nltk
In [3]:
         from nltk.stem import WordNetLemmatizer
         lemmatizer = WordNetLemmatizer()
         import json
         import pickle
         import numpy as np
         from keras.models import Sequential
         from keras.layers import Dense, Activation, Dropout
         from keras.optimizers import SGD
         import random
         words=[]
         classes = []
         documents = []
         ignore_words = ['?', '!']
         data file = open('chatbot data.json').read()
         intents = json.loads(data_file)
```

```
if intent['tag'] not in classes:
    classes.append(intent['tag'])
```

## In [5]: | print(words)

['Hi', 'there', 'How', 'are', 'you', 'Is', 'anyone', 'there', '?', 'Hey', 'Hola', 'Hell o', 'Good', 'day', 'Bye', 'See', 'you', 'later', 'Goodbye', 'Nice', 'chatting', 'to', 'you', ',' bye', 'Till', 'next', 'time', 'Thanks', 'Thanks', 'Thanks', 'Yhanks', 'Yhanks', 'Thanks', 'Thanks', 'Yhanks', 'Yhanks', 'Yhanks', 'Yhanks', 'Yhanks', 'Yhanks', 'Yhanks', 'Yhanks', 'Yhank', 'you', 'want', 'to', 'exchange', 'Ga', 'I', 'replace', 'I', 'mant', 'to', 'exchange', 'Ga', 'I', 'replace', 'I', 'mant', 'to', 'exchange', 'Gan', 'I', 'replace', 'I', 'mant', 'to', 'replace', 'my', 'I', 'need', 'to', 'swap', 'my', 'Gan', 'Joneing', 'Out', 'my', 'for', 'another', 'one', '?', 'I', 'need', 'to', 'swap', 'my', 'Gan', 'Joneing', 'hours', 'Yi, 'another', 'one', '?', 'What', 'are', 'you', 'hours', '?', 'What', 'are', 'you', 'store', 'hours', 'I', 'want', 'to', 'a', 'houran', 'how', 'do', 'I', 'to', 'a', 'person', 'I', 'want', 'to', 'talk', 'to', 'a', 'human', how', 'do', 'I', 'talk', 'to', 'a', 'human', 'how', 'do', 'I', 'talk', 'to', 'a', 'human', 'kepresentative', 'Speak', 'to', 'a', 'Garvic', 'agent', 'Where', 'do', 'you', 'have', 'stores', 'Y', 'Are', 'there', 'stores', 'in', 'are', 'you', 'in', 'do', 'you', 'have', 'any', 'stores', 'what', 'are', 'you', 'have', 'stores', 'by', 'are', 'you', 'hy', 'Where', 'are', 'you', 'have', 'stores', 'by', 'are', 'you', 'hy', 'Where', 'are', 'you', 'have', 'stores', 'what', 'are', 'you', 'have', 'stores', 'hours', 'Brimington', 'to', 'elalyed', 'are', 'you', 'brimington', 'store', 'hours', 'Brimington', 'to', 'be', 'delayed', 'are', 'you', 'brimington', 'store', 'hours', 'Brimington', 'to', 'be', 'delayed', 'are', 'you', 'brimington', 'store', 'hours', 'Brimington', 'to', 'be', 'delayed', 's', 'what', 'are', 'you', 'brimington', 'to', 'be', 'delayed', 's', 'what', 'are', 'you', 'brex', 'so', 'yo', 'brey', 'delayed', 's', 'yo', 'ho', 's', 'yo', 'ho', 'time', 's', 'yo', 'ho', ' ['Hi', 'there', 'How', 'are', 'you', 'Is', 'anyone', 'there', '?', 'Hey', 'Hola', 'Hell e', 'are', 'my', 'items', 'Are', 'my', 'items', 'on', 'the', 'way', 'Why', 'do', 'I', 'h ave', 'to', 'pay', 'restocking', 'fees', 'Can', 'I', 'return', 'in', 'store', 'How', 'lo ng', 'do', 'I', 'have', 'to', 'return', 'an', 'return', 'mail', 'return', 'store']

```
# lemmatize, lower each word and remove duplicates
In [6]:
         words = [lemmatizer.lemmatize(w.lower()) for w in words if w not in ignore words]
         words = sorted(list(set(words)))
         # sort classes
         classes = sorted(list(set(classes)))
         # documents = combination between patterns and intents
```

```
print (len(documents), "documents")
# classes = intents
print (len(classes), "classes", classes)
# words = all words, vocabulary
print (len(words), "unique lemmatized words", words)

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

113 documents
15 classes ['birmingham', 'detroit', 'exchange', 'farmington', 'goodbye', 'greeting', 'h ours', 'location', 'order', 'representative', 'restocking', 'return', 'return\_mail', 're turn\_store', 'thanks']
120 unique lemmatized words ["'s", '(', ')', ',', 'a', 'address', 'agent', 'an', 'anothe r', 'any', 'anyone', 'are', 'awesome', 'be', 'birmingham', 'by', 'bye', 'call', 'can', 'chatting', 'close', 'customer', 'day', 'delayed', 'detroit', 'different', 'do', 'doe', 'exchange', 'farmington', 'fee', 'for', 'framington', 'going', 'good', 'goodbye', 'hav e', 'hello', 'helpful', 'helping', 'hey', 'hi', 'hola', 'hour', 'how', 'human', 'i', 'i n', 'is', 'item', 'label', 'larger', 'later', 'located', 'location', 'long', 'mail', 'm e', 'michigan', 'my', 'near', 'need', 'next', 'nice', 'not', 'number', 'on', 'one', 'ope n', 'opening', 'order', 'other', 'out', 'package', 'pay', 'person', 'pick', 'process', 'provide', 'real', 'refund', 'refunded', 'replace', 'representative', 'restocking', 'ret urn', 'schedule', 'see', 'send', 'service', 'shipping', 'size', 'smaller', 'speak', 'ste p', 'store', 'street', 'swap', 'take', 'talk', 'thank', 'thanks', 'that', 'the', 'ther e', 'till', 'time', 'to', 'track', 'tracking', 'up', 'want', 'way', 'what', 'when', 'whe re', 'why', 'will', 'you', 'your']

In [7]: from sklearn.model\_selection import train\_test\_split

```
# create our training data
In [8]:
         training = []
         # create an empty array for our output
         output empty = [0] * len(classes)
         # training set, bag of words for each sentence
         for doc in documents:
             # initialize our bag of words
             bag = []
             # list of tokenized words for the pattern
             pattern words = doc[0]
             # lemmatize each word - create base word, in attempt to represent related words
             pattern words = [lemmatizer.lemmatize(word.lower()) for word in pattern words]
             # create our bag of words array with 1, if word match found in current pattern
             for w in words:
                 bag.append(1) if w in pattern_words else bag.append(0)
             # output is a '0' for each tag and '1' for current tag (for each pattern)
             output row = list(output empty)
             output_row[classes.index(doc[1])] = 1
             training.append([bag, output row])
         # shuffle our features and turn into np.array
         random.shuffle(training)
         training = np.array(training)
         # create train and test lists. X - patterns, Y - intents
         X = list(training[:,0])
         y = list(training[:,1])
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.10, shuffle = Fal
         print("Training data created")
```

Training data created

<ipython-input-8-5eb5ef77b0b8>:24: VisibleDeprecationWarning: Creating an ndarray from r
agged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with dif

ferent lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtyp
e=object' when creating the ndarray
 training = np.array(training)

In [9]: | print(training)

```
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 0, 0, 0, 0, 1, 0])
 list([0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0])]]
In [10]:
 print(output_row)
 [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0]
 print(pattern_words)
In [11]:
 ['return', 'store']
 print(X)
In [12]:
 0, 0, 0, 0, 0, 0, 0, 0], [0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
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0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0], [0, 1, 1, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0. 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0. 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0], [0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0], [0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0. 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0], [0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0. 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 

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0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0], [0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0. 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0. 0, 0, 0, 1, 0], [0, 0, 0, 0, 0, 0, 0, 0. 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0], [0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0], [0, 0, 0, 0, 0, 0, 0, 1, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0. 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,  0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0], [0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0]

```
In [13]: print(y)
```

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

 $X_{train} = np.array(X_{train}) X_{test} = np.array(X_{test}) y_{train} = np.array(y_{train}) y_{test} = np.array(y_{test}) X = np.array(X) y = np.array(y)$ 

```
In [14]: # Create model - 3 layers. First layer 128 neurons, second layer 64 neurons and 3rd out
# equal to number of intents to predict output intent with softmax
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'))
# Compile model. Stochastic gradient descent with Nesterov accelerated gradient gives g
sgd = SGD(lr=0.01, decay=1e-6, momentum=0.9, nesterov=True)
model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])
#fitting and saving the model
hist = model.fit(np.array(X_train), np.array(y_train), epochs=200, batch_size=5, verbo
#ynew = model.predict classes(np.array(test x))
#model.save('chatbot_model.h5',hist)
print("model created")
Epoch 1/200
Epoch 3/200
Epoch 4/200
Epoch 5/200
Epoch 6/200
Epoch 7/200
Epoch 8/200
Epoch 9/200
Epoch 10/200
Epoch 11/200
Epoch 12/200
Epoch 13/200
Epoch 14/200
Epoch 15/200
Epoch 16/200
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Epoch 21/200
Epoch 22/200
Epoch 23/200
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Epoch 24/200
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Epoch 89/200
Epoch 90/200
Epoch 91/200
Epoch 92/200
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Epoch 94/200
Epoch 95/200
Epoch 96/200
Epoch 97/200
00us/step - loss: 0.1413 - accuracy: 0.9799
Epoch 98/200
Epoch 99/200
Epoch 100/200
Epoch 101/200
Epoch 102/200
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Epoch 185/200
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Chatbot with Evaluation
  Epoch 186/200
  Epoch 187/200
  Epoch 188/200
  Epoch 189/200
  Epoch 190/200
  Epoch 191/200
  Epoch 192/200
  Epoch 193/200
  Epoch 194/200
  Epoch 195/200
  Epoch 196/200
  Epoch 197/200
  Epoch 198/200
  Epoch 199/200
  Epoch 200/200
  model created
In [15]: __, accuracy = model.evaluate(X_test, y_test)
  print('Accuracy: %.2f' % (accuracy*100))
  Accuracy: 83.33
  Running the final model to save for GUI
In [16]:
  # equal to number of intents to predict output intent with softmax
  model = Sequential()
```

```
# Create model - 3 Layers. First Layer 128 neurons, second Layer 64 neurons and 3rd out
# equal to number of intents to predict output intent with softmax
model = Sequential()
model.add(Dense(128, input_shape=(len(X[0]),), activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(64, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(len(y[0]), activation='softmax'))

# Compile model. Stochastic gradient descent with Nesterov accelerated gradient gives g
sgd = SGD(lr=0.01, decay=1e-6, momentum=0.9, nesterov=True)
model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])

#fitting and saving the model
hist_1 = model.fit(np.array(X), np.array(y), epochs=1000, batch_size=5, verbose=1)

#ynew = model.predict_classes(np.array(test_x))
model.save('chatbot_model.h5',hist_1)
print("model created")
```

Epoch 1/1000

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Epoch 2/1000
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Epoch 4/1000
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s/step - loss: 0.1557 - accuracy: 0.9460
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s/step - loss: 0.0341 - accuracy: 0.9804
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s/step - loss: 0.0361 - accuracy: 1.0000
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s/step - loss: 0.0147 - accuracy: 1.0000
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s/step - loss: 0.0041 - accuracy: 1.0000
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s/step - loss: 0.0084 - accuracy: 1.0000
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s/step - loss: 0.0784 - accuracy: 0.9578
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s/step - loss: 0.0102 - accuracy: 1.0000
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s 1ms/step - loss: 0.0147 - accuracy: 0.9912
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s 1ms/step - loss: 0.0140 - accuracy: 1.0000
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s/step - loss: 0.0069 - accuracy: 1.0000
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s/step - loss: 0.0030 - accuracy: 1.0000
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Epoch 710/1000
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s 1ms/step - loss: 0.0057 - accuracy: 0.9967
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s 1ms/step - loss: 0.0118 - accuracy: 0.9901
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s 2ms/step - loss: 0.0023 - accuracy: 1.0000
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s/step - loss: 0.0486 - accuracy: 0.9879
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s/step - loss: 0.1769 - accuracy: 0.9685
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s 1ms/step - loss: 0.0128 - accuracy: 0.9875
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     s 1ms/step - loss: 6.8770e-04 - accuracy: 1.0000
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     model created
    saving the same model for chat bot GUI
In [17]:
     import nltk
     from nltk.stem import WordNetLemmatizer
     lemmatizer = WordNetLemmatizer()
     import pickle
     import numpy as np
     from keras.models import load_model
     model = load_model('chatbot model.h5')
     import json
     import random
     intents = json.loads(open('chatbot data.json').read())
     words = pickle.load(open('words.pkl','rb'))
     classes = pickle.load(open('classes.pkl','rb'))
In [18]:
     from nltk.stem import PorterStemmer
     from nltk.tokenize import word tokenize
     ps = PorterStemmer()
     #for w in words:
      # print(w, ": ", ps.stem(w))
     def clean up sentence(sentence):
In [19]:
        # tokenize the pattern - split words into array
        sentence words = nltk.word tokenize(sentence)
        #sentence_words = [ps.stem(word.lower()) for word in sentence_words]
        # stem each word - create short form for word
        sentence words = [lemmatizer.lemmatize(word.lower()) for word in sentence words]
        return sentence words
     # return bag of words array: 0 or 1 for each word in the bag that exists in the sentenc
     def bow(sentence, words, show_details=True):
```

```
# tokenize the pattern
             sentence words = clean up sentence(sentence)
             # bag of words - matrix of N words, vocabulary matrix
             bag = [0]*len(words)
             for s in sentence_words:
                 for i,w in enumerate(words):
                    if w == s:
                        # assign 1 if current word is in the vocabulary position
                        bag[i] = 1
                        if show_details:
                            print ("found in bag: %s" % w)
             return(np.array(bag))
         def predict_class(sentence, model):
             # filter out predictions below a threshold
             p = bow(sentence, words, show_details=False)
             res = model.predict(np.array([p]))[0]
             ERROR THRESHOLD = 0.70
             results = [[i,r] for i,r in enumerate(res) if r>ERROR THRESHOLD]
             # sort by strength of probability
             results.sort(key=lambda x: x[1], reverse=True)
             return list = []
             for r in results:
                 return_list.append({"intent": classes[r[0]], "probability": str(r[1])})
             return return list
         print(np.array(bag))
In [20]:
         0 0 0 0 0 0 0 0 0]
         def getResponse(ints, intents_json):
In [21]:
             tag = ints[0]['intent']
             list of intents = intents json['intents']
             for i in list_of_intents:
                 if(i['tag']== tag):
                    result = random.choice(i['responses'])
                    break
             return result
         def chatbot response(text):
             ints = predict class(text, model)
             res = getResponse(ints, intents)
             return res
        def getResponse(intents_list, intents_ison): tag = intents_list[0]['intent'] list_of_intents =
        intents_json['intents'] for i in list_of_intents: if(i['tag']== tag): result = random.choice(i['responses'])
        break return result
        while True: message = input("") ints=predict_class(message) res = getResponse(ints, intents) print
        (res)
         import tkinter
In [22]:
         from tkinter import *
In [23]:
```

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In [24]:
          def send():
              msg = EntryBox.get("1.0", 'end-1c').strip()
              EntryBox.delete("0.0",END)
              if msg != '':
                  ChatLog.config(state=NORMAL)
                  ChatLog.insert(END, "You: " + msg + '\n\n')
                  ChatLog.config(foreground="white", font=("Verdana", 15, 'bold'))
                  res = chatbot response(msg)
                  ChatLog.insert(END, "Bot: " + res + '\n\n')
                  ChatLog.config(state=DISABLED)
                  ChatLog.yview(END)
In [25]:
          base = Tk()
          base.title("Chatbot")
          base.geometry("400x500")
          base.resizable(width=TRUE, height=TRUE)
Out[25]:
          #Create Chat window
In [26]:
          ChatLog = Text(base, bd=0, bg="black", height="10", width="100", font="Verdana",)
          ChatLog.config(state=DISABLED)
          #Bind scrollbar to Chat window
In [27]:
          scrollbar = Scrollbar(base, command=ChatLog.yview, cursor="mouse")
          ChatLog['yscrollcommand'] = scrollbar.set
In [28]:
          #Create Button to send message
          SendButton = Button(base, font=("Verdana",12,'bold'), text="Send", width="12", height=5
                              bd=0, bg="#096484", activebackground="black",fg='white',
                              command= send )
          #Create the box to enter message
In [29]:
          EntryBox = Text(base, bd=0, bg="white",width="29", height="5", font="Verdana")
          EntryBox.bind("<Return>", send)
Out[29]: '1869754265536send'
          #Place all components on the screen
In [30]:
          scrollbar.place(x=376,y=6, height=386)
          ChatLog.place(x=6,y=6, height=386, width=370)
          EntryBox.place(x=128, y=401, height=90, width=265)
          SendButton.place(x=6, y=401, height=90)
 In [ ]:
          base.mainloop()
 In [ ]:
 In [ ]:
 In [ ]:
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