CREATE A CHATBOT IN PYTHON

TEAM MEMBER

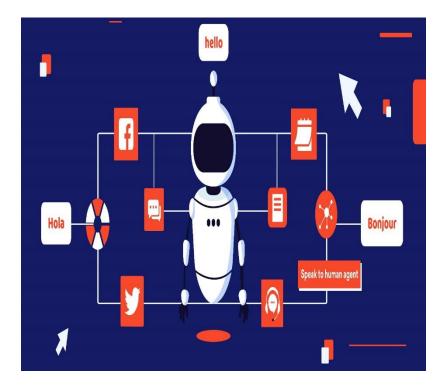
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PHASE-2: Innovation

SYNOPSIS:

- Overview
- Innovation techniques
 - 1. NLTK
 - 2. Ensemble learning
 - 3. Deep learning
- Visualizating Accuracy
- Pre-Trained Language Models (Gpt-3)
- Conclusion





OVERVIEW:

Chatbots are developed using innovative techniques like ensemble learning, deep learning, and natural language processing (NLP). Ensemble learning enhances accuracy, deep learning enables complex user queries, and NLP enhances human language understanding, making interactions more engaging.

INNOVATION TECHNIQUES:

Achieving robustness and accuracy in AI systems is a complex and ongoing challenge. There are several innovative techniques exploring to enhance the robustness and accuracy of AI systems. Here are some of them: Deep Learning, Ensemble Learning, Reinforcement Learning, and NLTK (Natural Language Toolkit)

1. NLTK:

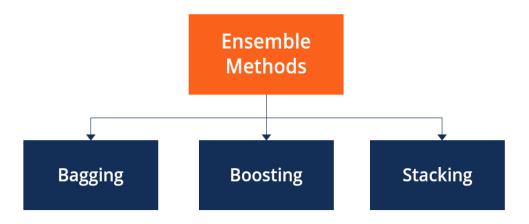
NLTK provides a wide range of tools, resources, and libraries for various NLP tasks, such as text processing, tokenization, stemming, tagging, parsing, and sentiment analysis. NLTK provides a set of tools and resources for text processing and analysis, making it for creating chatbots and improving the accuracy of prediction systems.

IMPLEMENTATION:

```
import nltk
import random
from sklearn.feature_extraction.text import TfidfVectorizer
nltk.download("punkt")
def tokenizer(text):
    text = text.lower()
    tokens = word_tokenize(text)
    tokens = [lemmatizer.lemmatize(token) for token in tokens]
    tokens = [token for token in tokens if token not in stop_words]
    return token
```

2. ENSEMBLE LEARNING:

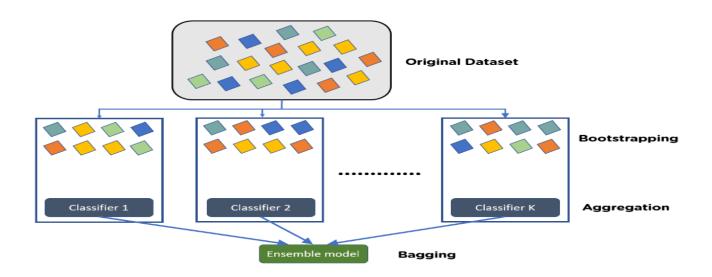
Ensemble learning is a technique in machine learning where multiple models are combined to improve predictive performance.



In this project, I used bagging method to achieve prediction system accuracy and robustness.

BAGGING:

Bagging (Bootstrap Aggregating) is an ensemble machine learning technique designed to improve the performance and robustness of predictive models by creating multiple subsets of training data, training individual models, and combining their predictions.



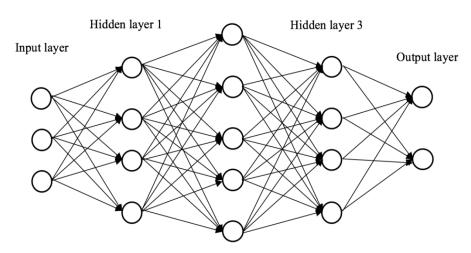
IMPLEMENTATION:

```
def bow(sentence, words, show_details=True):
    sentence_words = clean_up_sentence(sentence)
    bag = [0]*len(words)
    for s in sentence_words:
        for i,w in enumerate(words):
        if w == s:
            bag[i] = 1
        if show_details:
            print ("found in bag: %s" % w)
    return(np.array(bag))
```

3.DEEP LEARNING:

Deep Learning is a subfield of machine learning and artificial intelligence (AI) that focuses on the development of neural networks, particularly deep neural networks, inspired by the structure and function of the human brain.

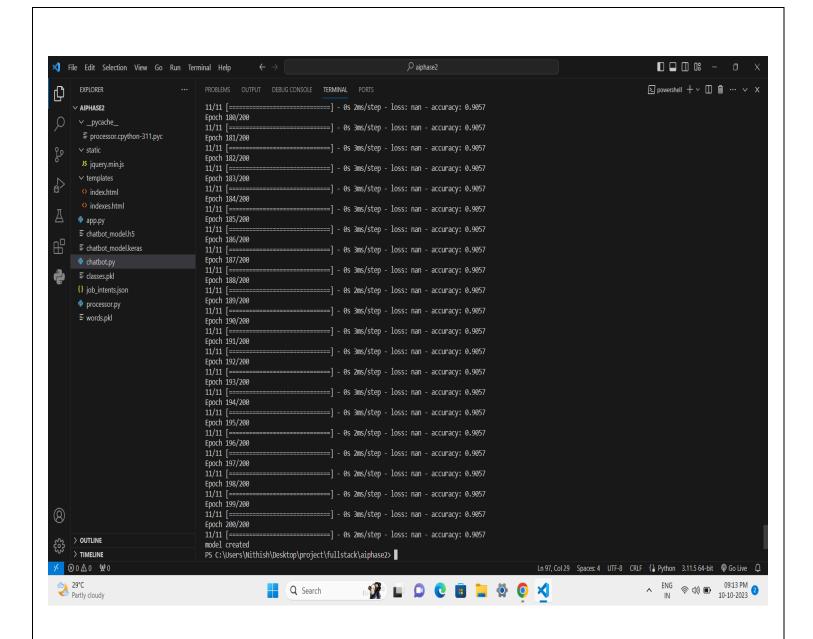
Hidden layer 2



IMPLEMENTATION:

```
import numpy as np
    from keras.models import Sequential
    from keras.layers import Dense, Activation, Dropout
    from keras.optimizers import SGD
    import random
    import tensorflow as tf
    words=[]
    classes = []
   documents = []
   ignore_words = ['?', '!']
   data_file = open('job_intents.json', encoding='utf-8').read()
   intents = ison.loads(data file)
   for intent in intents['intents']:
       for pattern in intent['patterns']:
          w = nltk.word_tokenize(pattern)
          words.extend(w)
          documents.append((w, intent['tag']))
         if intent['tag'] not in classes:
            classes.append(intent['tag'])
            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)))
            print (len(documents), "documents")
            print (len(classes), "classes", classes)
            print (len(words), "unique lemmatized words", words)
          pickle.dump(words,open('words.pkl','wb'))
```

```
pickle.dump(classes,open('classes.pkl','wb'))
          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:
               bag.append(1) if w in pattern_words else bag.append(0)
               output_row = list(output_empty)
               output row[classes.index(doc[1])] = 1
               training.append([bag, output_row])
               random.shuffle(training)
               max_{length} = max(len(item[0])) for item in training)
               training padded = np.array([item[0] + [0] * (max length - len(item[0])) +
item[1] for item in training])training = np.array(training_padded)
             train x = list(training[:, :-1])
             train_y = list(training[:, -1:])
             print("Training data created")
             model = Sequential()
             model.add(Dense(128, input_shape=(len(train_x[0]),), activation='relu'))
             model.add(Dropout(0.5))
             model.add(Dense(64, activation='relu'))
             model.add(Dropout(0.5))
             model.add(Dense(len(train_y[0]), activation='softmax'))
             sgd = tf.keras.optimizers.legacy.SGD(lr=0.01, decay=1e-6, momentum=0.9,
nesterov=True)
             model.compile(loss='categorical crossentropy', optimizer=sgd,
metrics=['accuracy'])
             hist = model.fit(np.array(train x), np.array(train y), epochs=200, batch size=5,
verbose=1)
             model.save('chatbot_model.h5', hist)
print("model created")
```



VISUALIZATING ACCURACY:

```
fig,ax=plt.subplots(nrows=1,ncols=2,figsize=(20,5))

ax[0].plot(history.history['loss'],label='loss',c='red')

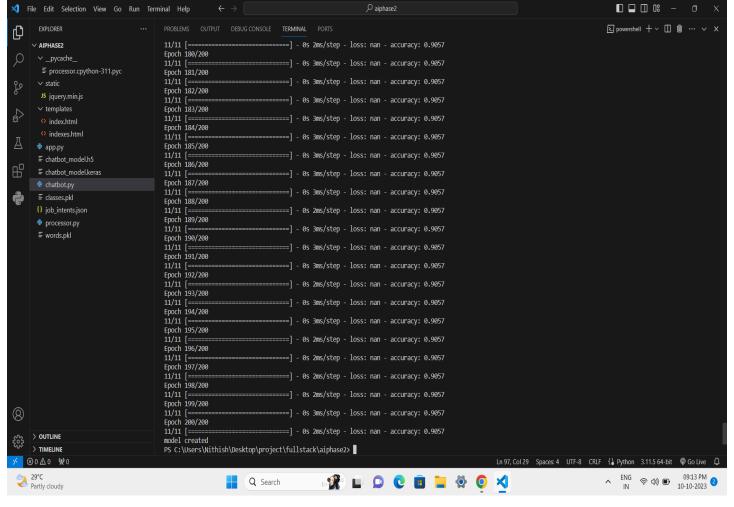
ax[0].plot(history.history['val_loss'],label='val_loss',c = 'blue')

ax[0].set_xlabel('Epochs')

ax[1].set_xlabel('Epochs')

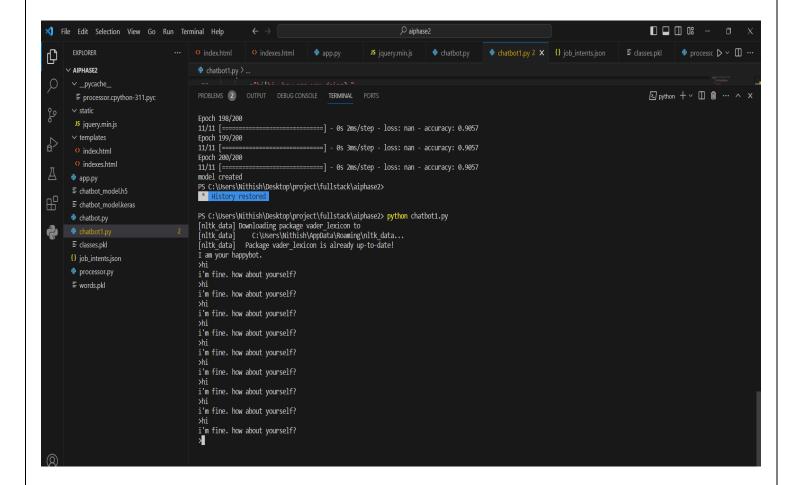
ax[0].set_ylabel('Loss')
```

```
ax[1].set_ylabel('Accuracy')
ax[0].set_title('Loss Metrics')
ax[1].set_title('Accuracy Metrics')
ax[1].plot(history.history['accuracy'],label='accuracy')
ax[1].plot(history.history['val_accuracy'],label='val_accuracy')
ax[0].legend()
ax[1].legend()
plt.show()
```



PRE-TRAINED LANGUAGE MODELS (GPT-3)

GPT-3 can be effectively utilized to enhance the quality of responses in a chatbot. By integrating GPT-3 into a chatbot system, you can take advantage of its natural language understanding and generation capabilities.



CONCLUSION:

The integration of NLP, Deep Learning, and Ensemble Learning has significantly improved chatbot development, enhancing their comprehension, response accuracy, and performance, leading to potential applications in various fields.

