

## Project 3: Create a chatbot in Python

### Phase 3:Development Part 1

#### Introduction:

To build a chatbot using GPT-3 (or any other version of GPT), you'll need to follow a few steps. I'll outline the process for you, including setting up your environment, installing required libraries, and implementing basic user interactions. In this example, we'll use the Hugging Face Transformers library for GPT-3 integration and Flask for web app development.

#### Necessary step to follow:

##### 1. Environment Setup:

First, you'll need to set up your development environment. You can use a virtual environment to keep your project dependencies isolated. You'll need Python installed on your system.

#### CODE:

```
# Create a virtual environment
```

```
python -m venv chatbot-env
```

```
# Activate the virtual environment (on Windows)
```

```
chatbot-env\Scripts\activate
```

##### 2. Install Required Libraries:

Need to install the necessary libraries for chatbot:

Hugging Face Transformers: This library provides pre-trained models, including GPT-3.

Flask: A lightweight web framework for building the chatbot's user interface.

#### CODE:

```
pip install transformers
```

pip install flask

Importance of loading and processing dataset:

- Loading and preprocessing the dataset is an important first step in building any machine learning model. However, it is especially important for house price prediction models, as house price datasets are often complex and noisy.
- By loading and preprocessing the dataset, we can ensure that the machine learning algorithm is able to learn from the data effectively and accurately.

Challenges involved in loading and preprocessing a house price Dataset:

There are a number of challenges involved in loading and preprocessing a house price dataset, including:

- Handling missing values:  
House price datasets often contain missing values, which can be due to a variety of factors, such as human error or incomplete data collection. Common methods for handling missing values include dropping the rows with missing values, imputing the missing values with the mean or median of the feature, or using a more sophisticated method such as multiple imputation.
- Encoding categorical variables:  
House price datasets often contain categorical features, such as the type of house, the neighborhood, and the school district. These features need to be encoded before they can be used by machine learning models.
- One common way to encode categorical variables is to use one-hot encoding.

Loading the dataset:

- Loading the dataset using machine learning is the process of bringing the data into the machine learning environment so that it can be used to train and evaluate a model.
- The specific steps involved in loading the dataset will vary depending on the machine learning library or framework that is being used.

However, there are some general steps that are common to most machine learning frameworks:

a. Identify the dataset:

The first step is to identify the dataset that you want to load. This dataset may be stored in a local file, in a database, or in a cloud storage service.

b. Load the dataset:

Once you have identified the dataset, you need to load it into the machine learning environment. This may involve using a built-in function in the machine learning library, or it may involve writing your own code.

c. Preprocess the dataset:

Once the dataset is loaded into the machine learning environment, you may need to preprocess it before you can start training and evaluating your model. This may involve cleaning the data, transforming the data into a suitable format, and splitting the data into training and test sets.

Preprocessing the dataset:

- Data preprocessing is the process of cleaning, transforming, and integrating data in order to make it ready for analysis.
- This may involve removing errors and inconsistencies, handling missing values, transforming the data into a consistent format, and scaling the data to a suitable range.

### 3. Get GPT-3 Model:

You can choose a GPT-3 model from Hugging Face's model hub (or any other available model). For example, you can use "gpt2" or "gpt2-medium." Download the model and save it to your project directory.

Program:

```
from transformers import AutoModelForCausalLM, AutoTokenizer
```

```
# Load the GPT-3 model
```

```
model_name = "gpt2" # or any other model you prefer
```

```
model = AutoModelForCausalLM.from_pretrained(model_name)
```

```
tokenizer = AutoTokenizer.from_pretrained(model_name)
```

### 4. Create a Flask Web Application:

need to create a web interface for user interactions. Here's a basic Flask app structure

Program:

```
from flask import Flask, request, render_template
from transformers import AutoModelForCausalLM, AutoTokenizer

app = Flask(__name__)

model_name = "gpt2" # or any other model you prefer
model = AutoModelForCausalLM.from_pretrained(model_name)
tokenizer = AutoTokenizer.from_pretrained(model_name)

@app.route('/')
def chat():
    return render_template('chat.html')

@app.route('/ask', methods=['POST'])
def ask():
    user_input = request.form['user_input']
    # Process user input and get the model's response
    # You'll need to implement this part using the Transformers library
    model_response = generate_response(user_input)
    return model_response

if __name__ == '__main__':
```

```
app.run(debug=True)
```

## 5. Create the HTML Template:

Create an HTML file (e.g., chat.html) for the chat interface where users can input questions and receive responses. You can use basic HTML and JavaScript for this part.

## 6. Implement the generate\_response Function:

In the ask route of your Flask app, you need to implement the generate\_response function to interact with the GPT-3 model and get its response.

## 7. Run the Chatbot:

Run your Flask app

Program:

```
python your_app.py
```

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