

Report:

Overview of the Implementation

Our implementation is a Gradio-based chatbot designed to recognize intents related to emergencies and respond appropriately based on the recognized intent. We utilized Hugging Face Transformers for the intent recognition model and Gradio to create a user-friendly web interface. Below is an evaluation of each component:

Model Integration

- **Model Loading:**
We used the BERT-base-uncased model as the backbone for sequence classification. While this is a valid Hugging Face model, it is a generic pre-trained BERT model that we have not fine-tuned for the specific emergency intents defined in our `intent_action_mapping`.
 - **Improvement:** We should fine-tune this model on a labeled dataset of emergency-related utterances to enhance the accuracy of intent recognition.
 - **Tokenizer:**
We correctly utilized the tokenizer to preprocess user input into the format required by the model.
-

2. Intent-to-Action Mapping

- We created a dictionary to map the intents recognized by the model to predefined actions. This worked well for our proof of concept.
 - **Potential Issue:**
 - The number of intents in the mapping must align with the number of labels used to train our classification model. If there's a mismatch, the system may fail or misclassify intents.
 - **Improvement:** We could dynamically match labels from the model's configuration or training dataset to ensure consistency.
-

3. Prediction and Classification

- **Prediction Function:**
We used the model's logits to predict the intent and mapped it to the corresponding action.

- **Improvement:** We should add error handling for unexpected inputs or predictions that fall outside the predefined intent labels.
-

4. Chatbot Interaction

- **Gradio Interface:**

We designed a clean and user-friendly web interface where users can type emergency requests and receive chatbot responses.

- **Customization:**

- We could enhance the chatbot by adding features such as:
 - **Voice Input/Output** for accessibility.
 - **Context Awareness** to handle multi-turn conversations.
 - Additionally, we could display more detailed responses or integrate APIs for real-time responses (e.g., Twilio for SMS or emergency services).
-

5. Missing Components

- **Fine-tuned Model:** We are currently using the BERT model without training it specifically for emergency-related intents, which limits its effectiveness for this task.
 - **Error Handling:**
 - We need to address how the chatbot handles input that doesn't map to a defined intent to avoid incorrect responses.
 - **Deployment:**
 - Our chatbot is designed for local use with Gradio. For production, we should:
 - Deploy it on a web server (e.g., AWS, Heroku, or Hugging Face Spaces).
 - Secure endpoints to ensure data privacy, especially when dealing with sensitive emergency data.
-

Recommendations

1. **Dataset Preparation:**

We should create or use a labeled dataset of emergency-related utterances and fine-tune bert-base-uncased on this dataset.

2. **Model Evaluation:**

We need to evaluate the model's performance on a test dataset (e.g., accuracy, precision, recall). Additionally, confidence thresholds for predictions can help us avoid misclassification.

3. **Advanced Features:**

We could integrate APIs for real-time emergency responses and add multi-turn conversational capabilities using a dialogue management system.

4. **Testing:**

We must test the chatbot with diverse emergency requests to identify and address edge cases.

5. **Accessibility:**

Adding features like voice commands, multilingual support, and TTS (text-to-speech) will make the chatbot more accessible to elderly users.

Strengths

- **Ease of Use:** Gradio made it simple for us to create a user-friendly interface for non-technical users.
- **Modular Design:** Our mapping and prediction functions are well-structured, making them easy to extend.
- **Customizable:** We can quickly update intents and actions based on new requirements.

Conclusion

With this chatbot, we created a strong prototype for handling emergency intents using NLP. By fine-tuning the model, integrating emergency APIs, and improving robustness, we can transform it into a highly effective tool for assisting elderly individuals during emergencies.

Here is the implementation

<https://colab.research.google.com/drive/1kEcXtqLcPXvhVCt7ZP259qVJfBpfL7r1?usp=sharing>

here is the output

Emergency Assistant for Elderly

A voice-activated assistant to understand and respond to emergency requests from elderly individuals.

Enter your emergency request

my heart beat is not normal , i need a help

Clear

Submit

output

****Intent Recognized:**** Medical Emergency

****Action Taken:**** Notifying an ambulance and contacting your caregiver.

Flag