# PROJECT REPORT

# AGENT ARCHITECTURE

The agent comprises three components: **CustomAgent**, **CustomHuggingFaceLLM**, and **AgentExecutor**.

#### 1. CustomAgent:

- Inherits from LangChain's BaseSingleActionAgent and serves as the decision-maker.
- It receives user queries, processes them using the language model, and categorizes responses into "Information Request" or "Actionable Task."
- Based on the category, it either executes a task or returns an answer directly.
- It leverages tools like get\_answer and perform\_action to manage tasks.

#### 2. CustomHuggingFaceLLM:

- Integrates a Hugging Face language model for text generation.
- It structures outputs with fields like "Response," "Category," and "Action Taken," ensuring clarity.
- Model parameters (e.g., temperature, top\_p) are adjusted for consistent responses.
- It acts as the core text generator, enabling decision-making within the agent.

#### 3. AgentExecutor:

- Combines the agent and tools, handling input processing and tool selection.
- It manages the interaction loop and determines the appropriate tool for each action.
- With a set iteration limit, it ensures effective user-agent communication.

# REASON FOR CHOOSING LLAMA3.2

 I chose LLaMA 3 over GPT-2, GPT-Neo, and GPT-J because it supports better function calling ability, enabling more dynamic task handling and model interaction. It also offers better reasoning, longer context management, making it more effective for real-world applications

# IMPLEMENTATION DETAILS

- The project was implemented using Colab Pro, which offers additional compute units, making it suitable for heavy computations. The implementation utilized an NVIDIA L4 GPU, which is optimized for AI workloads.
- The model used was **meta-llama/Llama-3.2-1B-Instruct**, with a model size of approximately **1 billion parameters**, requiring about **4-6 GB** of GPU memory during inference, depending on the batch size and precision (e.g., float16).

# **ALTERNATE METHODS:**

These agents can also be implemented using tools like LlamaIndex, Haystack, and Crew AI.

- **LlamaIndex** is ideal for managing data indexing and structured queries, enhancing retrieval capabilities.
- Haystack supports document search and retrieval-augmented generation (RAG), making responses more context-aware.
- **Crew AI** facilitates multi-agent coordination and workflow automation, allowing the agent to handle complex tasks efficiently.

# **ENHANCEMENTS:**

We can enhance the agent's capabilities by integrating external tools like the **Serper API** and other APIs, enabling access to real-time, external information. By incorporating **Retrieval-Augmented Generation (RAG)**, the agent can deliver more relevant content, improving the accuracy and context of query results. Additionally, fine-tuning the model using **Parameter-Efficient Fine-Tuning (PEFT)** will allow it to produce more fine-grained, domain-specific responses, adapting better to specialized tasks and improving overall performance.

# **TEST CASES:**

Test case 1: How can I improve my Python programming skills?

#### **Result:**

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Again respl. little yes queries (type 'quit' to exit)

Gent says' "me can I improve my hybro prepared a Mill?

Financia query.

Financia query
```

#### Test Case 2: User Query: "Can you create a to-do list file for me?"

#### Result:

```
Enter query: Create a todo list for me
Processing query...

**Intering new AgentExecutor chain...
Conclusion: [Any additional information needed to further assist]

### Step 1: Analyze the given query
The query is "Create a todo list for me." This is an informational request.

### Step 2: Determine the action taken
Since it: san informational request, no action is necessary. There is no need to create a todo list as there isn't any specific task or requirement provided by the user.

### Step 3: Provide a conclusion
Based on the analysis, since the query doesn't require any action, the only relevant information needed to further assist would be clarifying that no action is required.

*### Step 3: Provide a conclusion
Based on the analysis, since the query doesn't require any action Request Clarification Needed Action Taken: No action is required Conclusion: There are no tasks or requirements to fulfill.

**Princished Chain.**

*### Step 3: Provide a conclusion
Image: Provide a conclusion and the action taken
Since it's an informational information needed to further assist]

#### Step 3: Provide a conclusion:

#### Step 3: Provide a conclusion is necessary. There is no need to create a todo list as there isn't any specific task or requirement provided by the user.

#### Step 3: Provide a conclusion
Based on the analysis, since the query doesn't require any action, the only relevant information needed to further assist would be clarifying that no action is requirements to fulfill.

The final answer is: Response: [No action required] Category: Information Needed Action Taken: No action is required Conclusion: There are no tasks or requirements to fulfill.
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# Test Case 3: User Query: "Please delete this."

Upon running the agent consecutively for the same input case, the responses varied across the two executions.

# Response 1: The model underwent hallucination and deleted a database that was never created

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Enter query: Please delete this.

Processing query...

> Entering new Agentizecutor chain...
Result: [Summary of Response]

Response: [Answer the user query, Category: Information Request, Action Taken: The database is being accessed for information. Result: The system administrator has been informed that the request was not found.]

Here's the provided query:

Please delete this.

I will try to answer it as follows:

Response: The database is being accessed for deletion.
Action Taken: The query is being executed by a web application.
Result: The database is being updated with new data.

> Finished chain.

Processed Response:

Result: [Summary of Response]

Example:

Response: [Answer the user query, Category: Information Request, Action Taken: The database is being accessed for information. Result: The system administrator has been informed that the request was not found.]

Here's the provided query:

Please delete this.

I will try to answer it as follows:

Response: The database is being accessed for deletion.
Action Taken: The query is being executed by a web application.
Result: The database is being accessed for deletion.
Action Taken: The query is being executed by a web application.
Result: The database is being accessed for deletion.
Action Taken: The query is being executed by a web application.
Result: The database is being quidsed with new data.
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#### Response 2: It asked for clarification from the user which is a better response

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# **REASONS FOR VARIED RESPONSES:**

#### 1. Output Variability

- Problem: The model's randomness leads to inconsistent outputs, even with identical inputs. This unpredictability hinders reliable application.
- Solution: To mitigate this, I'm adjusting parameters like do\_sample, temperature, top\_p, and repetition\_penalty. Setting do\_sample to False and temperature to 0 forces the model to select the most probable tokens, ensuring deterministic generation.

#### 2. Unstructured Responses

- Problem: The model often produces outputs lacking a consistent or predictable structure. This makes it difficult to parse the information effectively.
- Solution: I'm implementing stricter parsing rules, enforcing the inclusion of necessary components in the output, and standardizing formatting. Additionally, I'm developing fallback mechanisms to handle instances where the output deviates from the expected structure.

#### 3. Statelessness

- Problem: The model lacks memory of past interactions, treating each request in isolation. This can be problematic when context is crucial for accurate responses.
- Solution: I'm working to improve context management by enforcing a clear response structure, ensuring consistent category assignment within my task, and refining error handling to address ambiguous or unexpected outputs.

# FINE-TUNING FOR TASK OPTIMIZATION

The model's occasional random classifications likely stem from its general-purpose training. To address this, I plan to fine-tune it on a domain-specific dataset relevant to my task. This should improve both accuracy and consistency.