## **API DOCUMENTATION**

## **API ENDPOINT REPORT**

## Introduction

This report provides a detailed analysis of the API endpoints implemented in the FastAPI-based application. The application processes user queries based on extracted text from resumes available in CSV and PDF formats. It then uses embeddings and a pre-trained language model to generate a comprehensive answer to the user's question.

## **Endpoint Overview**

The application consists of a single API endpoint that enables users to interact with the system:

ATTRIBUTE	DESCRIPTION
HTTP METHOD	POST
ENDPOINT PATH	/query/
PURPOSE	Processes user queries and provides detailed answers based on information extracted from resumes.
INPUT FORMAT	JSON
OUTPUT FORMAT	JSON

## **Request Details**

## **Request Payload**

The API expects a JSON object with the following structure:

FIELD	ТҮРЕ	DESCRIPTION
question	string	Represents the question that the user wants to be answered based on the resumes' content.

#### **Response Details**

The API returns a JSON object containing the generated answer. The structure is as follows:

FIELD	ТҮРЕ	DESCRIPTION
response	string	The generated answer based on relevant resume content.

## **Functionality**

#### **How the Endpoint Works**

#### 1. User Input Handling:

 The user sends a POST request to the /query/ endpoint with a JSON payload containing their question.

#### 2. Processing the Question:

- The handle\_query function receives the request and extracts the question.
- The extracted question is passed to the handle\_user\_input function, which utilizes various techniques to find relevant information from the resumes, including:
  - **FAISS Index:** Uses the pre-trained HuggingFaceEmbeddings model to search for the most similar texts within the extracted resume content.
  - BM25 Algorithm: Applies a BM25 ranking mechanism to retrieve the most relevant resume texts based on the user's query.

#### 3. Generating the Response:

 The generate\_response\_from\_inference\_api function integrates with the Hugging Face Inference API to generate a comprehensive answer based on the relevant resume content and user query.

#### **Integration with External API**

- The API utilizes Hugging Face's Inference API (mistralai/Mistral-Nemo-Instruct-2407 model) to generate responses based on the preprocessed data from resumes.
- An environment variable (HUGGINGFACEHUB\_API\_TOKEN) is used to manage authentication with the Hugging Face Inference API.

## **Error Handling and Edge Cases**

• **Validation:** The endpoint relies on FastAPI's built-in validation to ensure that the question field is correctly provided as a string.

- **Missing Data:** If no relevant text is found for a given question, the API may still attempt to generate a response, but it might be less accurate.
- External API Issues: If there's a problem with the Hugging Face Inference API (e.g., invalid API token or connectivity issues), a generic error message will be returned, indicating that the response generation failed.

#### Conclusion

The /query/ API endpoint is a robust and versatile feature of this application, capable of processing complex user questions and generating detailed answers based on extracted resume data. By integrating advanced text retrieval methods and leveraging a powerful language model, this endpoint serves as an efficient tool for querying extensive resume databases.

#### **Summary of the Endpoint:**

Endpoint: /query/

Method: POST

Input: JSON with question field.

• Output: JSON with response field containing the generated answer.

• Technology Stack: FastAPI, FAISS, BM25, Hugging Face Inference API

## **SAMPLE QUERIES**

Google Colab link for the queries:

https://colab.research.google.com/drive/14Hrh-RyMpIN8 IcYfJMDvjvpzdScgW6v?usp=sharing

Query 1: Summarise the first HR Resume

#### Response:

```
***Nork History:**

- ***Professional Summary:** Passionate HR specialist with over 10 years of extensive experience in completing reference checks, preliminary phone interviews, job a standard interviews in the specialist Horizontal Engineer**

- ***Professional Summary:** Passionate HR specialist with over 10 years of extensive experience in completing reference checks, preliminary phone interviews, job a standard interviews in the standard interviews in the standard interviews in the standard interviews in the standard interview in the standard interview
```

## Query 2: What is the Technical Stack of a random candidate?

#### Response:

```
You: what are the techinal stack of a random candidates
Bot: Based on the provided resumes, here is a random candidate's technical stack:

**Java Developer** (from resume 1)

- Java, JEE, Web Services, JSP, Servlet, Struts, JDBC, JMS, MDB, JMX, Spring (IoC, Spring MVC, Spring Security, Spring Remoting, Spring AMQP), Hibernate/MyBatis, M
```

# **Query 3:** Which programming languages are commonly known by candidates for software engineering roles?

#### Response:

#### Query 4: Which data visualization tools are candidates familiar with?

#### Response:

```
You: Which data visualization tools are candidates familiar with?

Bot: Based on the provided resumes, here are the data visualization tools that candidates are familiar with:

1. **Microsoft Excel**: This is the most commonly mentioned tool, with 6 out of 8 candidates listing it. Excel is widely used for creating charts, graphs, and o 2. **Power BI**: 3 out of 8 candidates mentioned Power BI, which is a powerful data visualization tool by Microsoft that provides interactive visualizations and 3. **Tableau**: 2 out of 8 candidates mentioned Tableau, a popular data visualization tool that helps users to explore, analyze, and present data with interactic 4. **Qlikview**: 1 out of 8 candidates mentioned Qlikview, a data visualization tool that enables users to create interactive dashboards and reports to explore 5. **Minitab**: 1 out of 8 candidates mentioned Minitab, a statistical software suite that includes data visualization capabilities for creating charts, graphs, 6. **PMatplotlib/Seaborn**: 1 out of 8 candidates mentioned Matplotlib or Seaborn, which are popular data visualization libraries in Python for creating static, 7. **D3.js**: 1 out of 8 candidates mentioned D3.js, a JavaScript library for data visualization in the web browser. D3.js helps users create dynamic and intera 8. **Power Query**: 1 out of 8 candidates mentioned Power Query, a powerful data connectivity and transformation add-in for Excel that enables users to create interactive pivot tables an 10. **Google Data Studio**: 1 out of 8 candidates mentioned Power Pivot, a data analysis and visualization tool that enables users to create interactive das 10. **Google Data Studio**: 1 out of 8 candidates mentioned Power Pivot, a data on visualization tool that enables users to create interactive das 10. **Google Data Studio**: 1 out of 8 candidates mentioned Power Pivot, a data on visualization tool that enables users to create interactive das 10. **Google Data Studio**: 1 out of 8 candidates mentioned Power Pivot, a data consciled that enables u
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#### Query 5: How many years of experience do candidates usually have in machine learning?

#### Response: