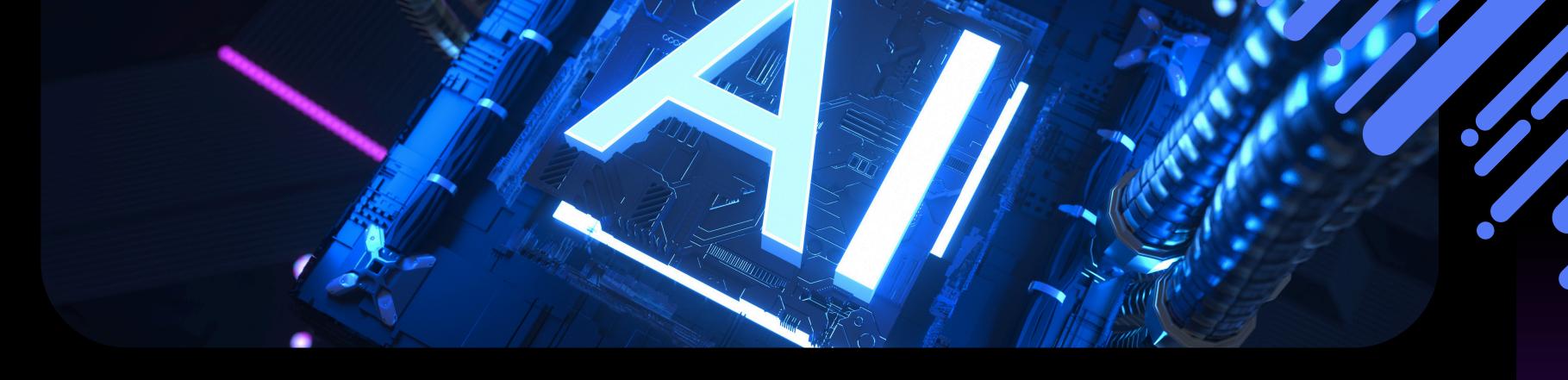


Al-Powered Document Processing with IBM Granite





Introduction

Problem Statement:

- Handling unstructured documents is time-consuming.
- Extracting meaningful insights from PDFs, images, and text requires automation.

Solution Overview:

- Al-powered document processing using IBM Granite models for text embedding & language understanding.
- Integrated with Gradio for an interactive and user-friendly interface.



Project Objectives







 Automate document processing with Alpowered text extraction.



Improve efficiency by using OCR and NLP models.

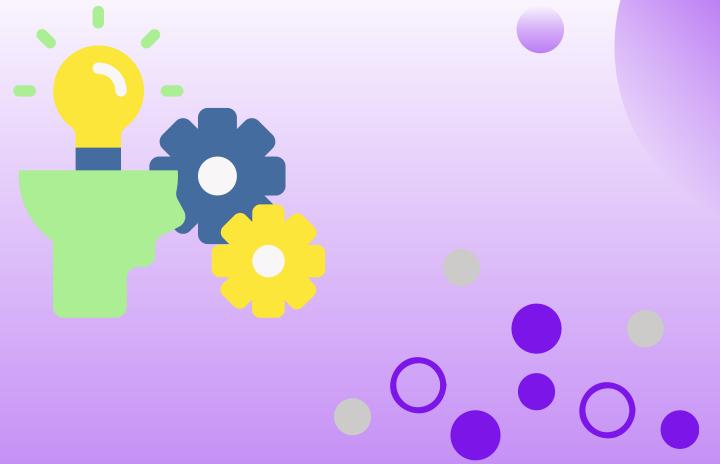


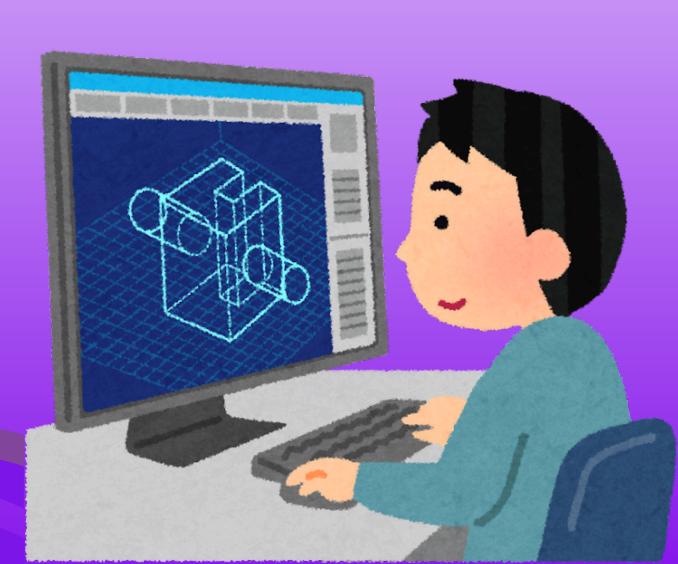
Enable interactive analysis through a webbased UI.

Technology Stack

- Programming Language: Python
- Libraries & Frameworks:
 - Gradio UI for document processing
 - Transformers IBM Granite models for NLP
 Embedding Model:granite-embedding-278m-multilingual

 Text Processing Model:granite-3.1-8b-instruct
 - Tesseract OCR Image-to-text conversion
 - pdfplumber PDF text extraction
 - SQLite Storing processing history
 - langchain: Various langchain tools are utilized for processing
- Hardware & Deployment:
 - Runs on a local machine or cloud.
 - Can be shared via a web interface.





Supports multiple file types – PDFs, images, and text files

OCR-enabled – Extracts text from scanned documents

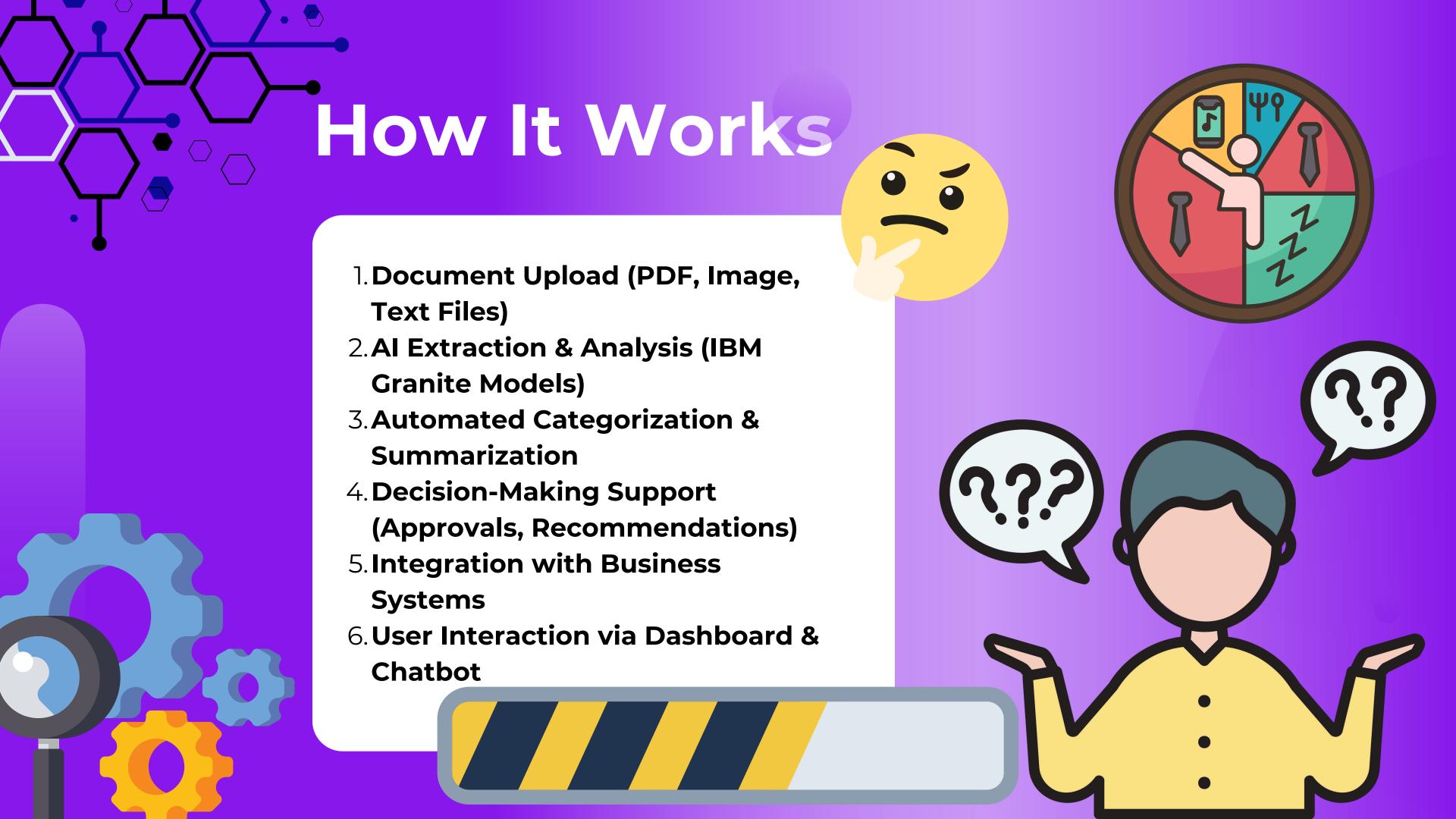






Al-driven text processing – Uses IBM Granite models

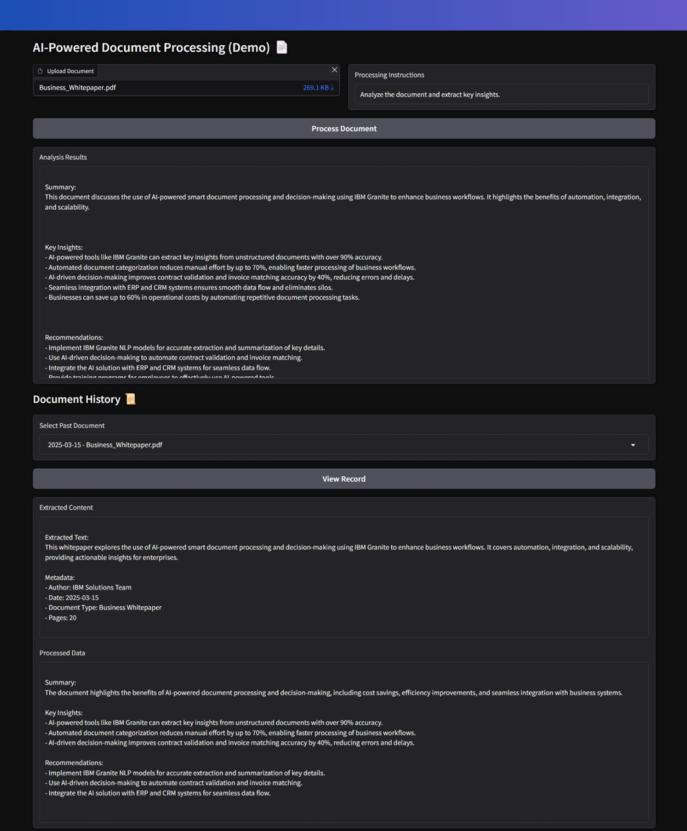
Interactive UI with Gradio – Easy-to-use interface



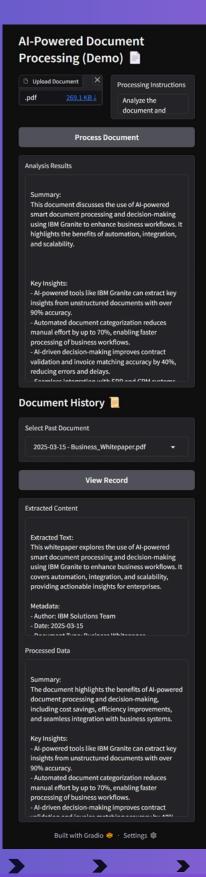




Result Expected



Use via API 🗸 · Built with Gradio 😂 · Settings 🕸



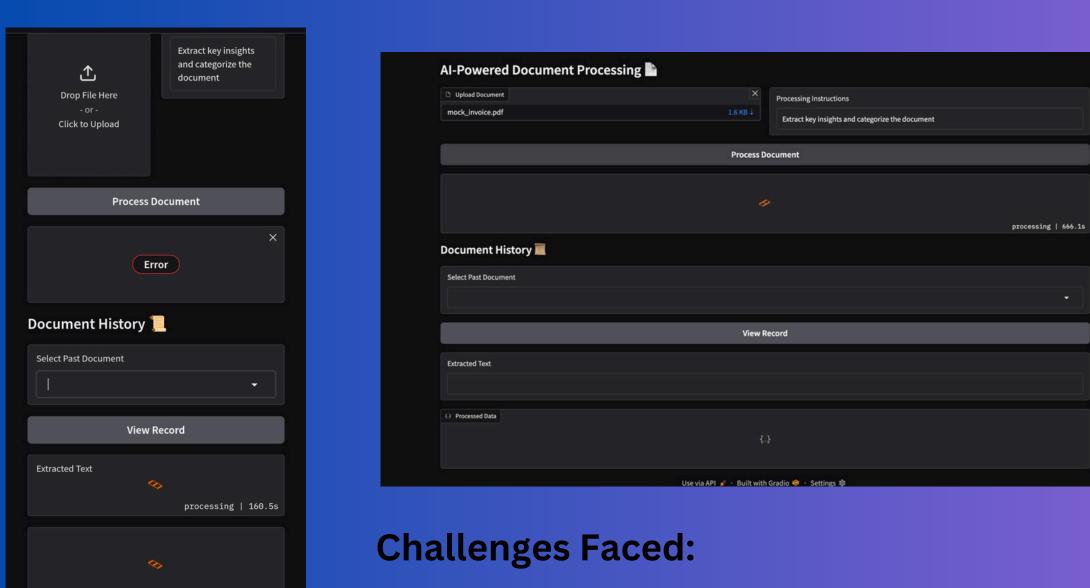


Ideal Situation (With better hardware)

- Faster processing of unstructured
- documents
 - Improved efficiency in document classification and extraction
- Automated decision-making reducing manual workload
- Seamless integration with business workflows
- Faster inference times for document processing
 - Higher throughput for handling multiple
- documents simultaneously
 - More complex and deeper NLP analysis
- with better results



Result Obtained





Real World Scenario (Inferior hardware)

- Due to hardware limitations, the processing time was significantly high.
- The system failed to deliver results within the stipulated time.
- Some large documents were not fully processed due to resource constraints.

- Running locally required 32GB RAM, which was unavailable.
- Attempts to deploy on IBM Cloud were hindered by strict access restrictions.
- Team members were not added to the same resource group, preventing collaboration.
- Low availability of computational resources impacted efficiency.
- Default agent in IBM WatsonX was not functioning correctly.
- Estimated processing time was excessively high, making real-time use impractical.
- Cloud execution costs were prohibitive, with \$68 per hour on IBM WatsonX, making long-running tasks unfeasible.





Limitations & Future Scope:

- Optimization required for handling large-scale document processing efficiently.
- Exploring cloud-based deployment for enhanced scalability.
- Potential integration with high-performance computing resources to overcome hardware bottlenecks.







Conclusion

Despite our best efforts, we were unable to achieve our goal due to significant hardware constraints. Running the AI models locally required more powerful machines, and cloud deployment presented challenges such as access restrictions and high operational costs. These limitations hindered real-time document processing, making the system impractical for large-scale use. Moving forward, optimizing resource allocation and exploring alternative cloud solutions with better cost efficiency and accessibility will be crucial to improving deployment and performance.





