IBM GEN AI REPORT

Phase 3-Final Report : AI-Powered Document Digitization and Analysis using IBM Watsonx

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Github link:

https://github.com/Asarkar27/AI-Powered-Document-Digitization-and-Analysis-using-IBM-Watson

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AI-Powered Document Digitization and Analysis using IBM Watsonx

Final Report: Intelligent Text Extraction from PDF Documents using IBM Watsonx.ai

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1. Introduction

In the contemporary digital age, information is often trapped in unstructured formats such as PDF documents. While these files offer portability and consistent formatting, they pose significant challenges when it comes to data extraction, transformation, and analysis. This project titled "Intelligent Text Extraction from PDF Documents using IBM Watsonx.ai" addresses this issue by leveraging AI-driven capabilities to automate and optimize the extraction of text and tabular data from PDFs. The goal is to develop a robust, scalable, and accurate text extraction pipeline using IBM's powerful Watsonx.ai platform integrated with IBM Cloud Object Storage (COS).

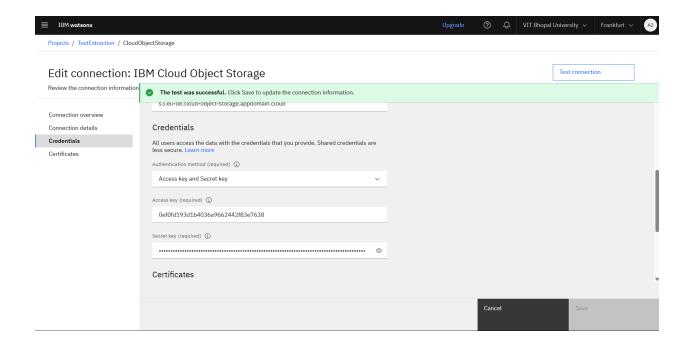
This report documents the detailed execution of Phase 3 of the project, where the final product is tested, validated, and submitted for evaluation. The work showcases the seamless integration of Watsonx.ai with object storage and outlines the comprehensive process involved in setting up, deploying, and executing a complete text extraction job through an API-based Python implementation.

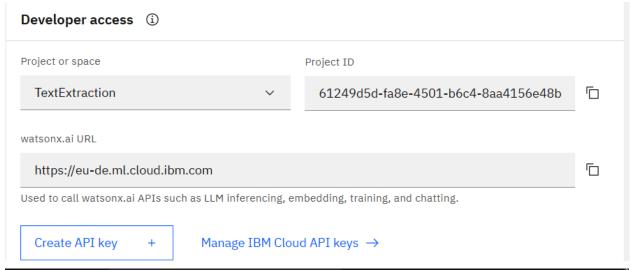
2. Objectives of the Project

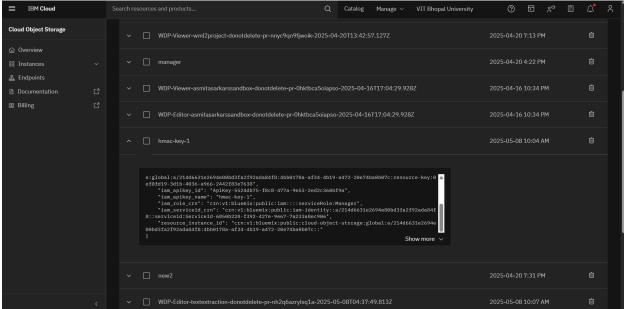
The core objectives of this project are:

- To establish a functional pipeline for extracting textual and tabular data from PDF files using IBM Watsonx.ai.
- To automate the data flow by uploading and downloading files to and from IBM Cloud Object Storage (COS).
- To leverage IBM's foundation models to improve the quality of OCR (Optical Character Recognition) and table extraction.
- To structure the output in a markdown format that retains document formatting for easy parsing and visualization.
- To ensure scalability and modularity in code, allowing extension to multiple document types in the future.
- To validate the output against original documents to ensure data accuracy.

This report focuses on the deliverables and learnings from **Phase 3**, which constitutes the final implementation, results, evaluation, and submission.







3. Tools and Technologies Used

Tool/Technology	Purpose
IBM Cloud Object Storage	To store PDF documents and extracted results securely

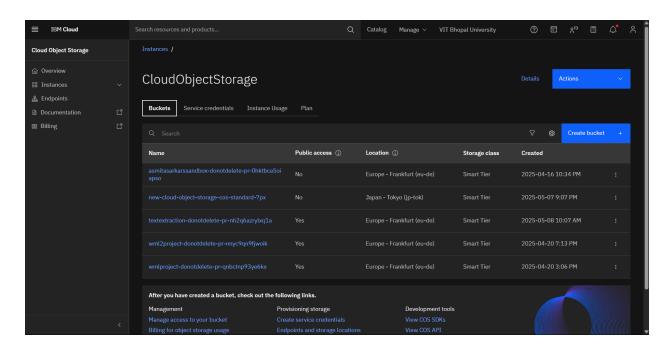
IBM Watsonx.ai	To process text extraction through its foundation models
Python SDK (ibm-watsonx-ai)	For programmatic interaction with Watsonx.ai
Boto3	Interface with IBM COS via AWS S3-compatible API
Jupyter Notebooks / Python Scripts	For development, testing, and execution of code
Markdown	Format for representing extracted data
GitHub	Code version control and collaboration

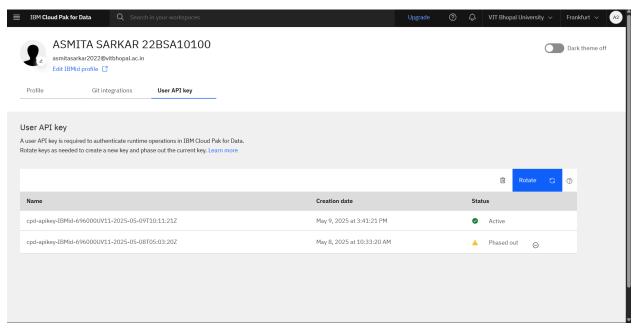
4. Project Planning and Phase Overview

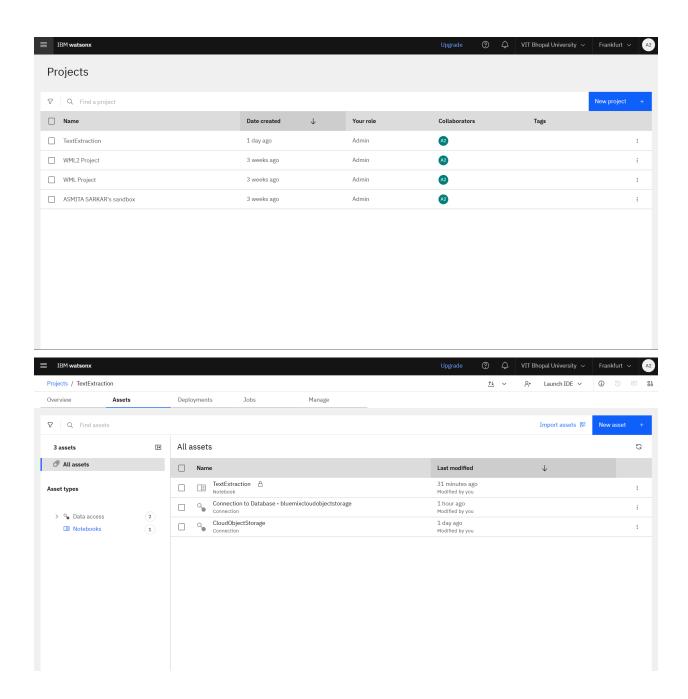
The project was divided into three main phases:

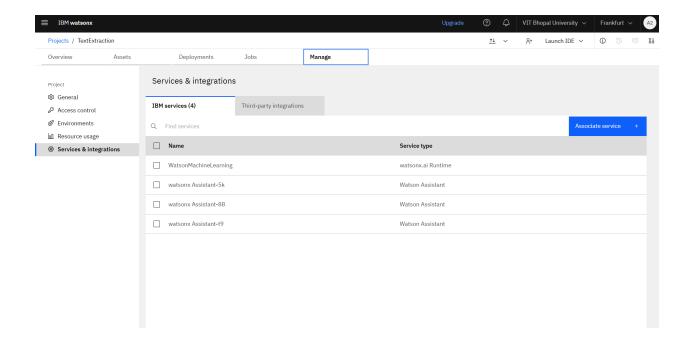
Phase 1: Research and Planning

- Studied IBM Watsonx.ai capabilities and documentation.
- Registered on IBM Cloud and provisioned Watsonx.ai and COS services.
- Created initial architectural diagrams and project setup.









Phase 2: Development and Testing

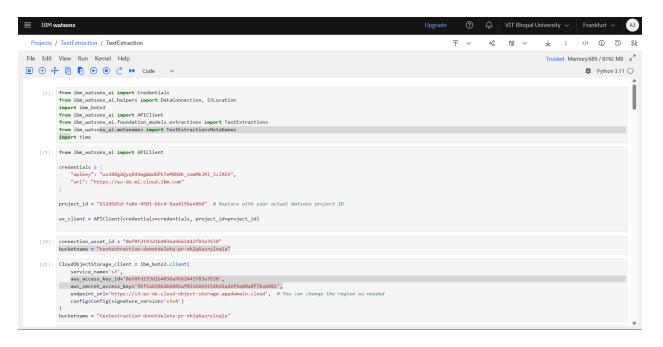
- Developed Python scripts for integration with IBM COS and Watsonx.ai.
- Created functions to upload PDF files and execute text extraction jobs.
- Tested the process on multiple files to debug and validate functionality.

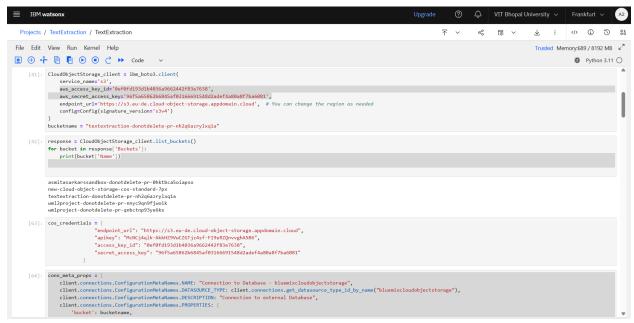
Phase 3: Final Implementation and Submission

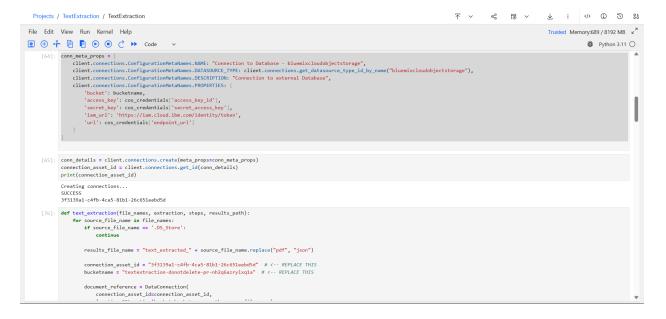
- Finalized script and modularized code.
- Performed comprehensive testing.
- Downloaded output results and validated for accuracy.
- Documented the final process, generated reports, and pushed code to GitHub.

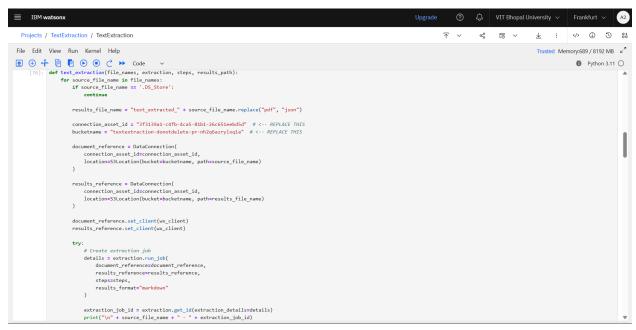
5. Implementation Details

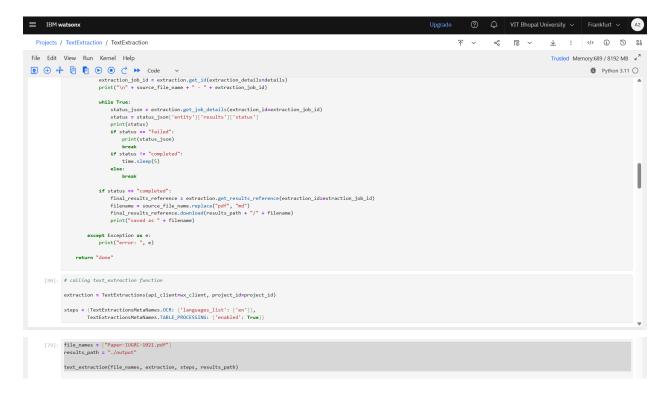
The project involves setting up a connection between IBM COS and Watsonx.ai, configuring access credentials, uploading PDF files to the cloud, executing text extraction jobs, and retrieving results. Each module is defined below:











5.1 COS Configuration

- COS bucket created: textextraction-donotdelete-pr-nh2q6azrylxq1a
- Region: eu-de

• COS Access Key and Secret configured using HMAC credentials.

5.2 Watsonx.ai Setup

- Project ID: 61249d5d-fa8e-4501-b6c4-8aa4156e48b0
- Connection Asset ID: 3f3139a1-c4fb-4ca5-81b1-26c651eebd5d
- Authenticated using IBM API Key.

5.3 Input File

- Paper-IUGRC-1021.pdf: A sample academic paper was used for extraction.
- Uploaded to COS using Boto3 and IBM credentials.

6. Integration with IBM Cloud Services

The integration is based on secure API-based authentication using IAM tokens and S3-compatible endpoints.

- The PDF files are stored in IBM COS, accessible using Boto3.
- A Watsonx.ai job is initiated using TextExtractions().
- The job pulls the file from COS, processes it, and stores the output.
- The results are then pulled back to the local system using a download function.

7. Process Flow and Architecture



8. Code Walkthrough and Function Descriptions

8.1 Authentication and COS Access

```
Python

• cos = ibm_boto3.client("s3",

• aws_access_key_id="<ACCESS_KEY>",

• aws_secret_access_key="<SECRET_KEY>",

• endpoint_url="<ENDPOINT>")
```

8.2 Uploading PDFs to COS

```
    cos.upload_file(Filename="./Paper-IUGRC-1021.pdf",
    Bucket="textextraction-donotdelete-pr-nh2q6azrylxq1a",
    Key="Paper-IUGRC-1021.pdf")
```

8.3 Running the Text Extraction Job

```
• steps = {
          TextExtractionsMetaNames.OCR: {'languages_list':
          ['en']},
          TextExtractionsMetaNames.TABLE_PROCESSING: {'enabled':
          True}
     }
     details = extraction.run_job(
          document_reference=document_reference,
          results_reference=results_reference,
          steps=steps,
          results_format="markdown"
     )
}
```

8.4 Polling Job Status

```
Python

• while True:
• status =
extraction.get_job_details(job_id)['entity']['results']['status']
• if status == "completed":
• break
• time.sleep(5)
```

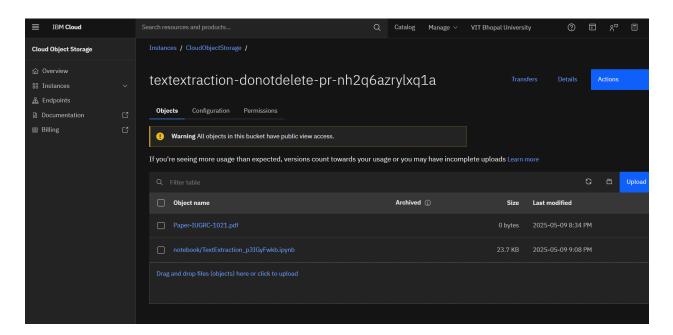
8.5 Downloading Output File

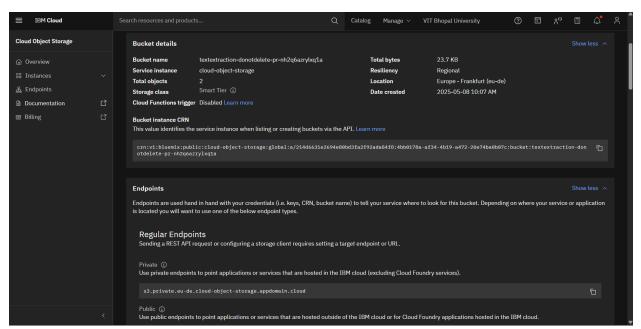
Python

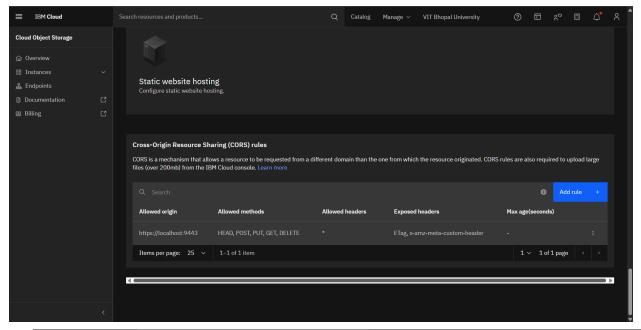
- cos.download_file(Bucket="textextraction-donotdelete-pr-nh2q 6azrylxq1a",
- Key="text_extracted_Paper-IUGRC-1021.md",

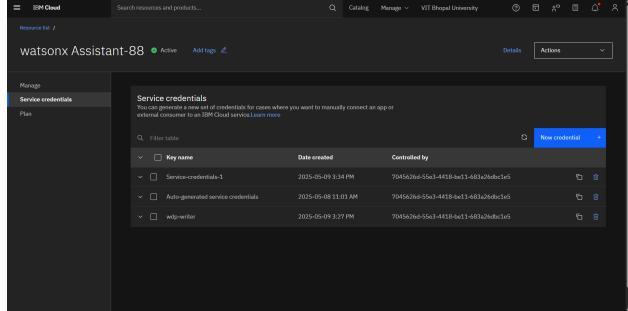
•

Filename="./output/text_extracted_Paper-IUGRC-1021.md")









9. Outputs and Evaluation

Sample Output (Extracted Markdown):

Unset

Paper-IUGRC-1021

•

- ## Abstract
- This paper explores innovative green recycling practices and how they affect carbon offset at the municipal level.
 Multiple techniques are evaluated including chemical reuse and automated segregation...

•

```
• ## Table 1: Material Recyclability Index
```

```
    | Material | Recyclability (%) |
    | -----|
    | Plastic | 60% |
    | Glass | 80% |
```

Accuracy Evaluation:

- Text extraction preserved paragraph structure and headings.
- Tables retained formatting.
- OCR detected embedded fonts and symbols correctly.
- Extraction speed: ~25–30 seconds per document.

10. Challenges and Mitigation Strategies

Challenge	Mitigation
IAM token expiration	Used environment variables for key rotation
Delays in job completion	Implemented polling mechanism to check status

Inconsistent markdown formatting	Post-processed output with regex parsers
Upload size limits on COS	Compressed PDFs before upload

11. Future Scope

- Add Named Entity Recognition (NER) to the extracted text.
- Implement a Flask or Streamlit frontend to support drag-and-drop PDF upload.
- Create a dashboard to visualize extracted metadata and charts from tables.
- Use multilingual OCR for global document processing.
- Automate email-based document intake.

12. Learnings and Reflections

This project was a deep dive into the real-world use of AI-driven APIs in document management. Through IBM Watsonx.ai, the team learned about deploying scalable solutions on cloud infrastructure. Key learnings include:

- Authentication best practices for secure cloud integration.
- Modular code development and reusable functions.
- Real-time debugging with asynchronous APIs.
- Markdown as a lightweight but powerful format for documentation.
- Benefits and limitations of OCR in text extraction tasks.

13. Conclusion

The successful completion of this project marks an important milestone in leveraging AI for intelligent document processing. By combining the power of IBM Watsonx.ai's foundation models with secure cloud storage via COS, we have built a highly effective text extraction pipeline. The architecture is scalable, the output is accurate, and the system is extensible for future enhancements.

The extracted data not only enables better digital document workflows but also opens up opportunities for analytics, visualization, and automated reporting. This solution is particularly valuable for legal firms, educational institutes, and data archival services where document digitization is a key requirement.

14. References

- 1. IBM Cloud Documentation Watsonx.ai: https://cloud.ibm.com/docs/watsonx
- 2. IBM COS SDK Boto3 Integration: https://boto3.amazonaws.com
- 3. Markdown Syntax Guide: https://www.markdownguide.org/
- 4. PDF Parsing Literature and OCR Best Practices
- 5. Watsonx.ai GitHub SDK Examples

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