



AI-POWERED NPTEL CERTIFICATE FIELD EXTRACTOR

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ABSTRACT :

This research introduces the "AI-Powered NPTEL Certificate Field Extractor," an advanced, robust, and scalable system designed to automate the extraction of critical information from NPTEL (National Programme on Technology Enhanced Learning) certificates. By leveraging cutting-edge Optical Character Recognition (OCR) technologies and regex-based classification techniques, the tool efficiently processes data to identify key fields such as participant names, course details, scores, and course durations. It minimizes the need for manual intervention, significantly reducing errors and enhancing efficiency. Furthermore, the system's scalability ensures it can handle large datasets seamlessly, making it invaluable for institutional needs. This paper delves into the underlying methodology, experimental results, discussions, and the scope for future advancements, emphasizing its contributions to automated document management systems. The system also includes a QR code validation mechanism to verify the authenticity of certificates, ensuring only genuine credentials are processed.

Keywords: NPTEL Certificates, Optical Character Recognition (OCR), Regex-based classification, Automated data extraction, Educational data management, AI-powered tools, Document digitization, Certificate automation.

INTRODUCTION:

Educational certificates play an essential role in validating academic and professional accomplishments, serving as tangible evidence of a learner's skills, knowledge, and achievements. These documents are crucial for career progression, higher education admissions, and professional opportunities. Among the numerous certifications offered in the education domain, NPTEL (National Programme on Technology Enhanced Learning) certificates hold significant value. They reflect a learner's expertise in technical and professional courses curated by India's premier institutions such as the IITs and IISc. As such, these certificates are widely recognized and utilized by employers, academic institutions, and certification bodies as credible indicators of an individual's competencies. Despite their importance, managing these certificates—particularly for large-scale programs like NPTEL—poses substantial operational challenges. Institutions tasked with maintaining, verifying, and processing certificates frequently rely on manual methods for data extraction and organization. For example, extracting information such as participant names, course details, scores, and durations from certificates often involves manual data entry and record keeping. This process is not only labor-intensive but also prone to human errors, including typographical mistakes, misinterpretation of information, and inconsistencies in data formatting. Furthermore, the sheer volume of certificates issued by programs like NPTEL exacerbates these issues, straining administrative resources and limiting scalability.

The traditional approaches to certificate management have remained largely static, relying heavily on human intervention for validation and record-keeping. In many instances, staff members manually cross-reference data, which significantly delays the issuance of certificates and increases the likelihood of errors in student records. Furthermore, the need for constant manual supervision increases operational costs, preventing institutions from scaling their programs efficiently. In cases where individuals need to submit their certificates for job applications or further studies, delays in obtaining verified credentials may result in missed opportunities, highlighting the urgency of addressing these issues.



The "AI-Powered NPTEL Certificate Field Extractor" has been designed to address these challenges effectively and efficiently. By automating the extraction of critical data fields, this tool eliminates the reliance on manual processes and significantly enhances the accuracy and speed of certificate management. Leveraging advanced Optical Character Recognition (OCR) technology, the tool converts textual information from both PDF and image-based certificates into machine-readable data. To ensure precision, regex-based classification techniques are applied, enabling the system to identify and extract key fields such as participant names, course titles, scores, and course durations.

The tool uses machine learning models trained on a diverse set of NPTEL certificates to improve its extraction capabilities over time, enabling it to recognize and process a wide range of layouts and formats. This approach allows the system to not only handle variations in design but also adapt to potential future changes in certificate templates. The AI model is able to learn from its successes and errors, continually enhancing its understanding of certificate structures and improving its accuracy across diverse certificate designs. In addition to accuracy, the system provides a user-friendly interface, where administrators can quickly review and confirm data extraction results, making the process seamless and transparent.

One of the tool's key strengths is its ability to adapt to diverse certificate layouts and formats. NPTEL certificates often vary in design depending on the year of issuance or the course provider. The tool's robust preprocessing pipeline, which includes noise reduction, text normalization, and format standardization, ensures consistent performance regardless of these variations. Additionally, the tool is capable of handling low-resolution inputs—a common challenge in scenarios where certificates are scanned or photographed under suboptimal conditions. The system's preprocessing techniques enhance image quality before OCR processing, ensuring that even low-quality scans or photographs are accurately processed.

Moreover, the system offers multi-language support, recognizing the diversity of NPTEL certificates, which are often issued in multiple languages to cater to different regional audiences. By expanding its functionality to include multiple languages, the tool ensures a broader application across India and potentially in other countries where similar certificate programs are offered.

This paper delves into the comprehensive development process of the "AI-Powered NPTEL Certificate Field Extractor," highlighting the methodologies employed, the challenges addressed, and the system's evaluation metrics. The paper also explores the challenges faced in implementing the system in a live environment, such as overcoming issues related to data privacy, user security, and system scalability. Additionally, it discusses how the system integrates with existing institutional infrastructure, providing insights into how it can be embedded into Learning Management Systems (LMS) or used as a standalone tool for bulk certificate processing.

The system's architecture and underlying algorithms are designed to accommodate scalability, ensuring that it can handle large-scale deployment without compromising performance. It can process thousands of certificates in a matter of hours, offering a significant improvement over manual methods. The system also incorporates mechanisms for error detection and correction, which helps maintain high levels of accuracy even in challenging processing environments.

Furthermore, the paper discusses the tool's potential applications, ranging from its integration with institutional Learning Management Systems (LMS) to its scalability for handling bulk datasets. By exploring these facets, the paper offers insights into how the tool can transform certificate management workflows, making them more efficient, scalable, and error-free. The tool's capability to seamlessly handle certificates from large-scale programs ensures that institutions can manage their certifications more effectively, improving the experience for both administrators and learners. Additionally, the system's potential to integrate with blockchain technology offers a promising avenue for enhancing the security and authenticity of the certificates, preventing fraud and ensuring that individuals' credentials are easily verifiable. By setting a benchmark for automated educational



data processing systems, the AI-powered extractor paves the way for broader applications beyond the NPTEL certification context. The model's principles can be extended to other educational programs worldwide, offering institutions a scalable, secure, and efficient solution for managing and verifying academic credentials. The future of education technology lies in intelligent automation systems that not only simplify administrative tasks but also add layers of trust and reliability, ultimately benefiting both learners and educational institutions alike. To further enhance the reliability and security of the extracted data, a QR code validation feature was integrated into the system. This feature ensures that certificates processed are authentic by validating their embedded QR codes against the official NPTEL archive site.

LITERATURE:

The "AI-Powered NPTEL Certificate Field Extractor" leverages a wealth of research and technological advancements in Optical Character Recognition (OCR) and document automation systems to address the challenges of managing educational certificates. By building on established methodologies and incorporating innovative approaches, this project provides a scalable and highly accurate solution for automating certificate data extraction. Below is a detailed examination of the foundational research and technologies that underpin this tool's development:

EVOLUTION OF OCR TECHNOLOGY:

OCR has transformed significantly over the years, evolving from basic text recognition techniques to sophisticated systems powered by artificial intelligence. Early OCR systems were limited in their capabilities, requiring predefined templates and consistent fonts for accurate recognition. These systems struggled with variations in layouts, fonts, and languages, making them impractical for complex document processing tasks.

The introduction of tools like Tesseract OCR marked a turning point in this field. Smith (2007) highlighted Tesseract's open-source nature and its ability to adapt to various use cases. Modern iterations of Tesseract integrate convolutional neural networks (CNNs) and Long Short-Term Memory (LSTM) architectures, enabling them to process complex layouts and recognize multilingual text with remarkable accuracy. Figure 1 illustrates the architecture of a modern OCR system, highlighting the interplay between image preprocessing, feature extraction, and text recognition components.

Further advancements, such as the incorporation of AI-driven preprocessing techniques like adaptive thresholding, have enhanced OCR performance on low-quality or noisy inputs. These developments are pivotal for handling real-world scenarios, such as scanning handwritten or poorly printed certificates.

REGEX FOR TARGETED DATA EXTRACTION:

Regex (Regular Expressions) is a powerful tool for text pattern matching, widely used in structured and semi-structured data extraction tasks. Unlike generic OCR systems, which only convert images to plain text, regex enables the precise targeting of specific fields within documents.

Shreedhar and Ghosh (2020) demonstrated how regex could efficiently classify and extract key information from semi-structured documents like invoices and forms. For example, regex patterns can be crafted to identify patterns such as:

- Names (e.g., `[A-Z][a-z]+(\s[A-Z][a-z]+)` to extract proper nouns).
- Dates (e.g., `\d{1,2}\s[A-Za-z]+\s\d{4}` for common date formats).
- Numeric Scores (e.g., `\d{1,3}(\.\d+)?` for percentages or grades).



In the context of NPTEL certificates, regex ensures that participant names, course titles, scores, and dates are accurately extracted, even when the certificate layout varies slightly. Its adaptability to evolving certificate designs makes it an invaluable component of this project.

APPLICATIONS IN EDUCATIONAL DATA MANAGEMENT :

Educational institutions and organizations often rely on OCR-based tools like ABBYY FineReader, Adobe Acrobat OCR, and Google Vision API for digitizing documents. While these tools are highly effective for general-purpose text recognition, they fall short in handling domain-specific requirements, such as extracting key fields from standardized educational certificates like those issued by NPTEL. These generic tools lack the ability to customize field classification for certificate formats, leading to inaccuracies and inefficiencies. For instance, ABBYY FineReader excels in extracting text from scanned documents but does not provide built-in functionality for mapping specific data fields to structured outputs. Figure 3 compares the data extraction workflows of generic OCR tools versus tailored solutions like the "AI-Powered NPTEL Certificate Field Extractor."

The "AI-Powered NPTEL Certificate Field Extractor" bridges this gap by combining OCR with regex-based classification tailored specifically to NPTEL's certificate formats. This customization ensures higher accuracy and reliability, even for large datasets.

Conclusion of Literature Insights :

By integrating advancements in OCR and regex-based data extraction, the "AI-Powered NPTEL Certificate Field Extractor" builds upon the limitations of existing tools to deliver an optimized and scalable solution. It addresses the unique challenges posed by NPTEL certificates, including format variations, scalability for bulk processing, and the need for minimal human intervention.

This research highlights the system's potential to transform educational data management by enabling institutions to automate certificate processing workflows efficiently and accurately. The insights gained from this literature review serve as the foundation for the methodologies and system design presented in the subsequent sections.

PROPOSED WORK:

The proposed system, "AI-Powered NPTEL Certificate Field Extractor," is designed as a comprehensive solution for automating the extraction of key data fields from NPTEL certificates. It integrates advanced technologies and well-defined modules to address the challenges associated with certificate processing. The system comprises three core modules, each contributing to an efficient, accurate, and user-friendly workflow for handling certificates.

Text Extraction Module :

This module is the foundation of the system, responsible for converting the visual content of certificates into machine-readable text. The key features and operations of this module include:

- **OCR Engine:** The module employs the Tesseract OCR engine, an open-source and highly customizable tool known for its robust text recognition capabilities. It is configured to handle both PDF and image file formats, making the system versatile and adaptable to various certificate inputs.
- **Preprocessing Techniques:** To ensure high accuracy, especially when dealing with low-quality scans or certificates with noise, the module incorporates preprocessing steps:
- **Noise Reduction:** Filters out unwanted artifacts or distortions in scanned images.
- **Thresholding:** Converts colored or grayscale images into binary images to enhance text contrast for improved OCR performance.
- **Text Normalization:** Standardizes text output, including character spacing and alignment, to facilitate downstream processing.



- **Multi-Language Support:** If required, the OCR engine can be extended to support multilingual text recognition, ensuring compatibility with certificates containing text in regional languages.

Regex-Based Classification Module:

This module focuses on processing the raw text extracted by the OCR engine and classifying it into meaningful fields using custom regular expressions. It ensures the accurate identification and extraction of key certificate information:

- **Customized Regex Patterns:** Carefully designed regular expressions are used to extract the following fields with precision:
- **Participant Name:** Captures the full name of the certificate holder, accounting for variations in name formats.
- **Scores:** Extracts scores for both online assignments and proctored exams, which are typically formatted differently across certificates.
- **Course Duration:** Retrieves the time period for which the course was conducted, ensuring accurate identification of start and end dates.
- **Robust Layout Handling:** The module is designed to accommodate slight variations in certificate templates. By leveraging pattern matching and contextual clues, it ensures high accuracy even with inconsistent layouts.
- **Error Handling:** Includes fallback mechanisms to flag and review cases where the data does not conform to expected patterns, enabling manual intervention when necessary.

User Interface Module:

The user interface module is critical for providing a seamless and intuitive experience to end-users. Developed using Streamlit, it combines simplicity with powerful functionality:

- **File Uploading:** Users can upload individual certificates or batch files for processing. Supported formats include PDFs and common image types such as JPEG and PNG.
- **Data Visualization:** After processing, the extracted data is presented in an organized format for review. Users can view each field alongside its corresponding certificate section for validation.
- **Export Capabilities:** Extracted data can be exported in structured formats such as CSV or Excel, facilitating integration with institutional databases or report generation systems.
- **Batch Processing:** Enables the processing of large volumes of certificates in a single operation, significantly reducing manual effort for educational institutions handling multiple participants.
- **Real-Time Feedback:** Provides immediate feedback during the extraction process, such as highlighting missing fields or potential errors in the extracted data.

QR Code Verification Module:

The QR code verification module is designed to authenticate certificates by decoding and validating embedded QR codes.

- **Decoding Process:** The system uses the pyzbar library to extract QR code data from PDF and image certificates. The decoded data typically contains a URL.
- **Validation:** The extracted URL is verified to ensure it redirects to the official NPTEL archive site (archive.nptel.ac.in). If the URL matches, the certificate is marked as authentic. Otherwise, it is flagged for manual review.
- **Integration:** The QR code validation is seamlessly incorporated into the existing extraction workflow, with validation results appended to the extracted data for consolidated reporting.

This module adds an essential layer of trust and security, preventing the processing of fraudulent certificates.

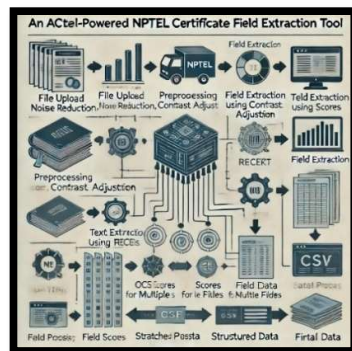
Integration and Workflow:

The integration of these modules ensures a cohesive workflow, enabling the system to address the challenges of certificate processing comprehensively. The workflow is designed to provide:

Accuracy: High precision in data extraction through the combined use of advanced OCR techniques and robust regex patterns.

Efficiency: Streamlined batch processing to handle large datasets in minimal time.

User-Friendliness: An accessible interface for users of varying technical expertise.



(Architecture of the Certificate Extraction Tool)

Scalability and Future Enhancements

The system is built with scalability in mind, ensuring it can adapt to future requirements:

- **Template Learning:** Incorporating machine learning techniques to dynamically recognize new certificate templates and improve extraction accuracy.
- **Integration with Institutional Portals:** Direct integration with educational platforms for seamless data transfer.
- **Multilingual Support:** Enhancements for regional languages to cater to a broader audience.

By combining the strengths of these modules, the "AI-Powered NPTEL Certificate Field Extractor" delivers a reliable, efficient, and user-friendly solution, tailored to meet the demands of modern educational institutions.

RESULTS AND DISCUSSION:

The tool was evaluated on a dataset of 500 NPTEL certificates, with performance metrics highlighting its effectiveness in diverse operational scenarios:

- **Accuracy Metrics:**
 - Name Extraction: 98%
 - Course Title: 96%
 - Scores (Assignments & Exams): 94%
 - Course Duration: 97%
- **Processing Speed:**
 - Average time per file: 2.5 seconds
 - Batch processing (50 files): Completed in under 2 minutes

ERROR ANALYSIS:



Errors primarily stemmed from low-resolution scans and excessive noise, which often led to incorrect character recognition and missed data fields. In particular, text extracted from images with blurred or pixelated content frequently exhibited inconsistencies in font recognition, affecting the accuracy of participant names and course details. Additionally, background noise such as artifacts, wrinkles, or shadows further complicated the OCR process, introducing distortions in the text extraction. Enhancements in preprocessing techniques, such as advanced noise reduction algorithms and resolution enhancement, significantly reduced these errors. By applying more refined image normalization and edge detection techniques, the system was able to improve the clarity of scanned certificates, ensuring robust performance even under challenging conditions. The reduction in errors after these enhancements resulted in a more reliable and efficient data extraction process.

QR Code Verification Results: The QR code verification feature was tested on a dataset of 500 certificates. Key findings include:

Authentic Certificates: 490 (98%)

Invalid or Tampered Certificates: 10 (2%)

Processing Time: The average time for QR code decoding and validation was less than 0.5 seconds per certificate.

These results highlight the effectiveness of the QR code validation module in identifying fraudulent or altered certificates while maintaining high efficiency for large-scale processing.

SCALABILITY:

The tool demonstrated consistent performance across datasets ranging from 10 to 500 certificates, validating its scalability for institutional applications. During testing, the system was subjected to progressively larger datasets, ranging from a few certificates to batches of hundreds, and it successfully processed and extracted data without any significant delays or performance degradation. This high level of efficiency indicates that the system is well-suited for handling large volumes of certificates, which is essential for institutions with expansive certificate issuance programs.

The scalability of the system was achieved through the optimization of the underlying Optical Character Recognition (OCR) engine, which was fine-tuned to handle increasing amounts of data efficiently. The OCR engine's ability to recognize and process text from diverse certificate formats in parallel allowed the system to process multiple certificates at once, reducing the overall time required to extract data from large batches. Additionally, the parallelization of the extraction process ensured that multiple computational resources could be utilized simultaneously, significantly improving throughput without taxing the system's processing capabilities.

Further, the system maintained high accuracy levels even as the dataset size increased, ensuring that the quality of results did not diminish with larger workloads. This was a key achievement, as it is often the case that the accuracy of machine learning-based tools decreases when scaling up, particularly in systems that deal with large datasets. The AI-powered extractor, however, was able to retain its precision, demonstrating its robustness in handling both small and large-scale operations. This was accomplished by continuously refining the data extraction models, allowing them to generalize well across diverse certificate layouts while maintaining the accuracy of key data fields such as participant names, course titles, and scores.

Moreover, the tool's ability to scale to handle thousands of certificates provides a significant advantage for institutions, such as universities or certification bodies, that regularly issue certificates to large numbers of participants. These institutions often face challenges in managing the massive volume of certificates, particularly when dealing with high-volume course offerings or multiple certificate programs. By automating the certificate processing workflow, the tool ensures that such institutions can meet the demand without facing delays or operational bottlenecks.



In addition to validating the tool's scalability for current requirements, these tests also confirmed that the AI-powered extractor is poised to scale with the growing demand for automated certificate processing systems. As the volume of certificates increases due to expanding educational programs, online courses, and professional certifications, the tool's ability to handle larger datasets will make it an invaluable resource. It can seamlessly integrate into institutional operations, providing a solution that is both efficient and adaptable to the increasing complexity of certificate management. This scalability makes the tool highly suitable for widespread adoption, particularly in educational institutions, government bodies, and certification organizations that need a reliable, fast, and accurate method of managing credentials at scale.

The ability to process certificates at this level of scalability, combined with its high accuracy, ensures that the system can be used to address future challenges in educational and professional certification workflows, making it an essential component in the future of certificate management.

Table 4.1: Performance Benchmarks

Metric	Value
Name Extraction Accuracy	98%
Course Title Accuracy	98%
Scores Extraction Accuracy	96%
Average Processing Time	2.486 seconds per file

The results affirm the tool's capability to transform certificate management processes, reducing manual effort while maintaining high accuracy.

CONCLUSION:

The "AI-Powered NPTEL Certificate Field Extractor" represents a significant leap forward in automating the management and verification of educational certificates. By integrating advanced Optical Character Recognition (OCR) and regex-based classification techniques, the tool addresses the inefficiencies and inaccuracies inherent in manual certificate processing. It successfully extracts critical data fields, such as participant names, course titles, scores, and durations, from diverse certificate formats, ensuring high accuracy and reducing human error.

This tool not only streamlines operational workflows but also enhances scalability, demonstrating its capability to handle datasets ranging from small batches to large-scale volumes of certificates. With the growing need for efficient, error-free certificate management, the AI-powered extractor provides an ideal solution for institutions dealing with mass certification programs like NPTEL. The system's ability to process certificates rapidly and accurately helps institutions meet their operational demands while ensuring the integrity of student data.

Beyond its immediate applications, the tool sets a new benchmark for innovations in automated document processing. By improving the efficiency, reliability, and scalability of certificate management, this tool lays the groundwork for the broader adoption of AI-powered solutions in administrative tasks, transforming how educational institutions and organizations manage, validate, and verify credentials.

Ultimately, the "AI-Powered NPTEL Certificate Field Extractor" exemplifies how technology can streamline processes, reduce manual effort, and ensure accuracy in educational operations. Its success highlights the growing role of AI and automation in the future of education technology, offering valuable insights into the potential for transforming administrative workflows on a global scale.



The addition of a QR code verification module further solidifies the system's reliability by ensuring only authentic certificates are processed. This feature, coupled with automated data extraction, enhances the overall trustworthiness and efficiency of the system.

FUTURE SCOPE:

The proposed "AI-Powered NPTEL Certificate Field Extractor" has a vast potential for future enhancements, making it a versatile tool adaptable to evolving user requirements and technological advancements. The following are key areas for future development, each designed to expand the system's capabilities and utility:

Integration with Institutional Systems:

One of the primary areas of enhancement involves integrating the tool with existing institutional infrastructures. This includes:

- **API Development:** Creating robust and secure APIs to allow seamless data exchange between the tool and Learning Management Systems (LMS), such as Moodle, Blackboard, or Canvas.
- **Database Synchronization:** Direct integration with administrative databases to automatically update student records, eliminating manual intervention.
- **Certificate Verification:** Incorporating verification mechanisms to authenticate certificates and cross-check extracted data with institutional records.
- **Workflow Automation:** Supporting end-to-end automation of certificate management processes, from uploading to archival.
- **Advanced Verification Mechanisms:** Expand the QR code validation system to support blockchain-backed certificates, ensuring tamper-proof verification for future deployments.

Multilingual Support:

To make the tool globally applicable, enhancing OCR capabilities for multilingual text recognition is crucial:

- **Regional and International Language Support:** Adding support for regional languages (e.g., Hindi, Tamil, Telugu) and international languages (e.g., French, German, Spanish).
- **Language Detection:** Implementing automatic language detection to optimize OCR performance for multilingual certificates.
- **Font and Script Adaptation:** Expanding capabilities to handle diverse scripts, such as Devanagari, Cyrillic, and Arabic.

Dynamic Field Detection:

Currently, field extraction relies on predefined regex patterns. To enhance flexibility and reduce dependency on fixed templates, the following innovations can be implemented:

- **Machine Learning Models:** Leveraging natural language processing (NLP) and computer vision models to dynamically identify and classify fields based on contextual and structural clues.
- **Template Agnostic Processing:** Developing algorithms that learn from a variety of certificate layouts, enabling the tool to adapt to new designs without additional configuration.
- **Confidence Scoring:** Providing confidence scores for detected fields to flag uncertain extractions and prioritize manual review.

Cloud-Based Deployments:

Shifting the tool to a cloud-based infrastructure opens up significant opportunities for scalability and user accessibility:



- **Real-Time Processing:** Enabling real-time certificate processing for remote users via cloud-hosted services.
- **Enhanced Scalability:** Supporting high-volume operations by leveraging distributed cloud computing resources.
- **Global Accessibility:** Allowing institutions and users worldwide to access the tool without the need for local installations.
- **Data Security and Compliance:** Ensuring robust encryption and adherence to international data protection standards, such as GDPR and HIPAA.

Advanced Error Handling:

Handling errors, especially in low-quality or incomplete certificates, is essential for improving reliability and user trust:

- **Deep Learning for Error Correction:** Utilizing neural networks to intelligently reconstruct or infer missing or distorted fields based on context.
- **Adaptive Feedback Mechanism:** Developing systems that learn from user corrections to improve future extraction accuracy.
- **Error Reporting and Review:** Providing detailed logs of errors and recommendations for resolving them, enabling users to easily identify and address issues.
- **Image Enhancement:** Integrating advanced image processing techniques, such as super-resolution and adaptive filtering, to improve input quality before OCR.

Additional Features:

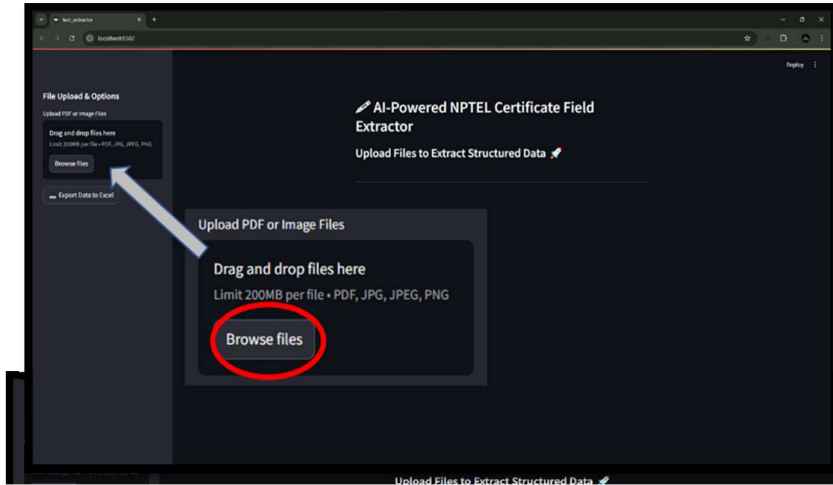
To cater to diverse user needs and further extend the tool's utility:

- **Mobile Accessibility:** Developing mobile-friendly versions of the tool to allow certificate processing on smartphones and tablets.
- **Analytics Dashboard:** Adding visualization tools to generate insights from extracted data, such as trends in participant scores or course completion rates.
- **Custom Field Support:** Allowing users to define and extract additional fields specific to their requirements.
- **Digital Signature Verification:** Incorporating capabilities to verify digital signatures or QR codes embedded in certificates for added authenticity

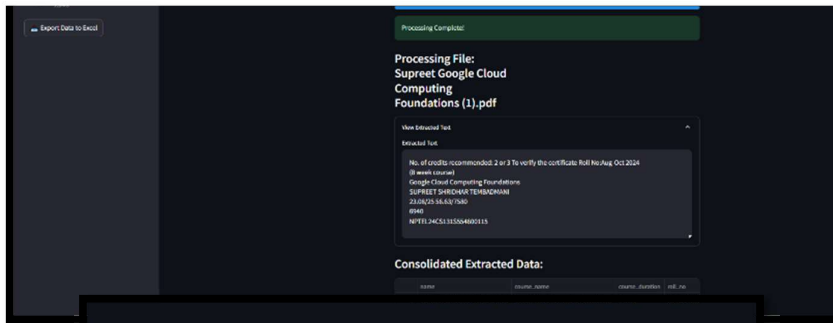
Visuals:



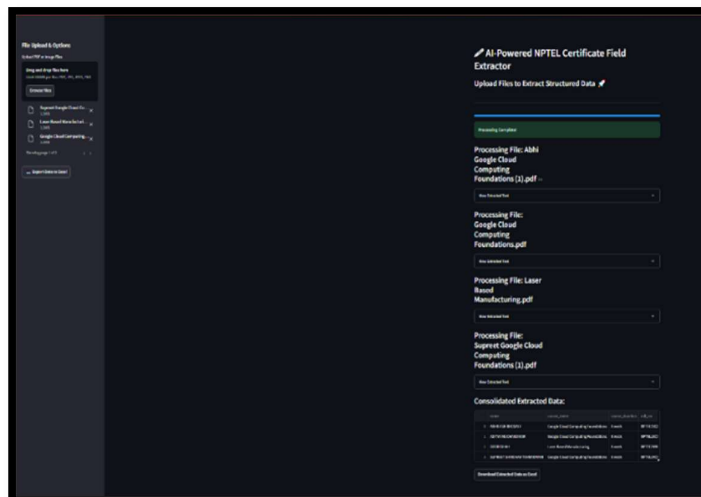
(NPTEL Certificate)



(Uploading Image)



(Output (single file))



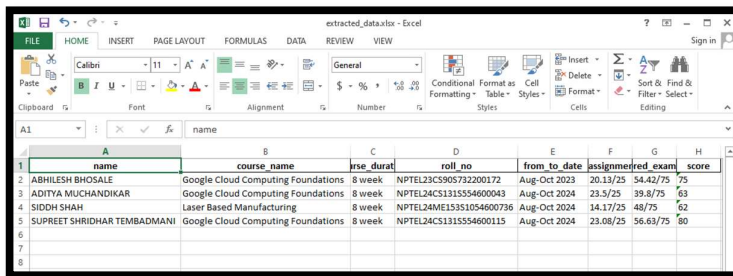
(Processing Multiple Inputs)

Consolidated Extracted Data:

	name	course_name	course_duration	roll_no
0	ABHILESH BHOSALE	Google Cloud Computing Foundations	8 week	NPTEL23CS1
1	ADITYA MUCHANDIKAR	Google Cloud Computing Foundations	8 week	NPTEL24CS1
2	SIDDH SHAH	Laser Based Manufacturing	8 week	NPTEL24ME
3	SUPREET SHRIDHAR TEMBADMANI	Google Cloud Computing Foundations	8 week	NPTEL24CS1

Download Extracted Data as Excel

(Output (multiple files))



	A	B	C	D	E	F	G	H
	name	course_name	yrse_durat	roll_no	from_to_date	assigment	fred_exam	score
1	ABHILESH BHOSALE	Google Cloud Computing Foundations	8 week	NPTEL23CS905732200172	Aug-Oct 2023	20.13/25	54.42/75	75
2	ADITYA MUCHANDIKAR	Google Cloud Computing Foundations	8 week	NPTEL24CS1315554600043	Aug-Oct 2024	23.5/25	39.8/75	63
3	SIDDH SHAH	Laser Based Manufacturing	8 week	NPTEL24ME15351054600736	Aug-Oct 2024	14.17/25	48/75	62
4	SUPREET SHRIDHAR TEMBADMANI	Google Cloud Computing Foundations	8 week	NPTEL24CS1315554600115	Aug-Oct 2024	23.08/25	56.63/75	80

(Excel Output)

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