

Combining Document Understanding and Action Center in UiPath for Human-In-The-Loop Claims Processing

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ABSTRACT

There have been new ideas about how to improve the efficiency of healthcare claims processing by integrating UiPath's Document Understanding and Action Centre. Structured and unstructured healthcare documents remain difficult to process manually due to delays and errors. With the use of AI technologies like Optical Character Recognition (OCR), Natural Language Processing (NLP) and Machine Learning (ML), UiPath can extract data automatically and the Human in the Loop (HITL) model to ensure quality assured and HIPAA/ GDPR compliant. Using NHS and Bupa UK case studies, Bupa Healthcare studies, as well as AXA Health, powerful reductions in processing time (up to 60%), reduction of errors, and regulatory compliance were shown. However, the benefits of the tool are frustrated by challenges, including poor scan quality, system sustainability, user resistance, and retraining requirements. For broader adoption, the research recommends that such automation should be phased, thresholds tuned, and trained. Using a hybrid approach indeed offers a scalable model of processing sensitive data within the complex healthcare environment.

Keywords: Action Centre, Intelligent Automation, Healthcare Claims, HITL, OCR, NLP, Compliance, HIPAA, GDPR, UiPath, Record to Runbook, Document Understanding, Process Efficiency, Case Studies.

I. INTRODUCTION

A. Background to the Study

Healthcare document extraction, validation, and approval are type of sensitive patient and financial data from different structured and unstructured documents. However, manual processing is still based on traditional methods like conversations between people that require time and Room for error, and cannot scale well [1]. A hybrid model here is intelligent automation, with the use of UiPath's Document Understanding and Action Centre, the software bots take care of the repeating task, and the human validation is done to ensure accuracy and compliance with the healthcare regulations.

B. Overview

The project Understanding leverages AI and Optical character recognition (OCR) technologies to extract relevant data from claims documents (invoices, medical reports, insurance forms, etc.) provided by UiPath. But healthcare data is such a complicated and sensitive thing that certain data points need human input. Through the UiPath Action Centre, companies can create a Human in the Loop (HITL) model where a human oversees bots' operation when they are not certain or when significant decisions must be made [2]. By enabling collaboration, this ensures faster processing and higher accuracy in data and also keeps the healthcare standards up to the mark, like the Health Insurance Portability and Accountability Act (HIPAA) from the US law in 1996 to protect patient health and information.

C. Problem Statement

Despite many benefits, there are some problems and challenges in healthcare. Healthcare providers and insurers may be violating the compliance needs and, thus, need to resort to some manual work [3]. Fully automating is taking data integrity into account, whereas blindly automating will never be enough to make it perfect without manual review due to its efficiency. The need for such a balanced system like healthcare 4.0 with intelligent automation combined with human validation is pressing to increase operational efficiency without compromising on quality and compliance.

D. Objectives

The primary goals of this study are: 1. To identify key inefficiencies in current healthcare claims processing workflows. 2. To evaluate the effectiveness of document understanding in extracting healthcare data from diverse document formats. 3. To implement the human-in-loop model using UiPath action for validating critical data points. 4. To assess improvements in processing speed, accuracy and compliance after automation implementation. All these objectives aim to explore the integration of UiPath document understanding and action centre for efficient and compliant claims processing in the healthcare sector.

E. Scope and Significance

The scope for the healthcare sector is the object of study related to claims management and billing departments. Organisations that are focusing on boosting document processing efficiency with the highest data integrity are the target [4]. The significant findings should assist healthcare providers and insurance companies in making use of automation tools without violating the regulations. Accountability comes from the human input that makes this approach applicable to other high-risk data-sensitive environments.

II. LITERATURE REVIEW

A. Intelligent Automation in Healthcare Claims Processing

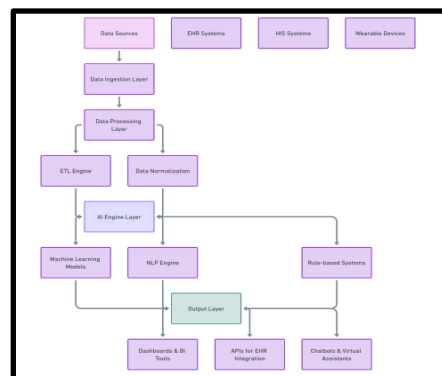


Figure 1: Intelligent Automation in Healthcare

[5]

From reducing costs and administrative labour, healthcare has been largely adopting intelligent automation in claims processing, especially in claims processing [5]. Robotic Process Automation (RPA) reduces manual data entry and error rate, and thus faster resolution of the claim and improved service quality is shown in the literature. In particular, UiPath has been pointed out as a solution that is scalable and easy to integrate. However, critical analysis points to several limitations. For dynamic rule changes, ambiguous data, and a lack of contextual understanding of clinical information, automation systems find it difficult. Furthermore, in very high data variability and compliance environments, considerations on RPA without human oversight come into play [6]. Others feel that automation is simply a

replacement for human expertise, but could also be thought of as an augmentation tool, freeing up staff to concentrate on greater value-added work.

B. Document understanding and Data extraction from unstructured healthcare documents

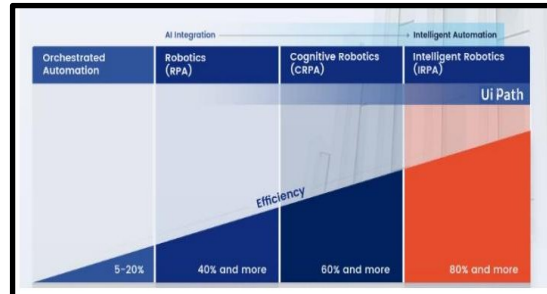


Figure 2: Intelligent Automation

[7]

The above figure 2 illustrates automation evolution from orchestrated automation to intelligent robotics (IRPA), showing increasing AIU integration and effectiveness from 5% to 80% with UiPath leading intelligent automation [7]. Machine learning, Optical character recognition (OCR) and Natural language processing (NLP) have been used within the UiPath Document Understanding framework to successfully extract the relevant data. Such tools are shown to be more accurate in the extraction and faster than the manual methods. However, technological limits are criticised [8]. One of the challenges of Optical Character Recognition (OCR) accuracy is when processing poor-quality scans, handwritten text and non-standard formats. It also has the dependency of using pre-trained models and a need for domain-specific retraining of which requires both time and expertise. Given this, automation of data extraction to full levels is still not possible without human input.

C. Human-in-the-Loop (HITL) Models for Quality Assurance and Compliance

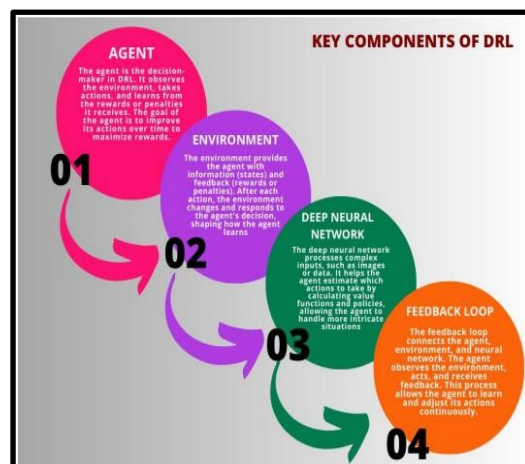


Figure 3: Key components of DRL

[10]

Deep reinforcement learning (DRL) can be the main component in Human-in-the-Loop DRL algorithms are often used to train AI. Human-in-the-Loop (HITL) is becoming more and more recognised as a strategic approach to streamline automation. For UiPath, Action Centre is a vital interface to reach actions by humans in the event of uncertain data [9]. There is literature that supports HITL to significantly reduce error rates in decision making and regulate Health Insurance Portability and Accountability Act (HIPAA) compliance. However, studies also raise the issue of the trade off during

human overinvolvement. However, if the bot decisions' confidence thresholds are too low, automation benefits could be cancelled by frequent human reviews. In other words, human oversight is the shame of an omission, but compliance breach and reputational damage are the shame of a subjective oversight [10]. To be effective, therefore, decision thresholds need to be finely tuned, and the user interface has to be fluent with the exception handling.

D. Automating the Processing Efficiency and Compliance

This project indicates that human validation of automation helps to gain performance improvements in processing speed, cost efficiency, in auditability [11]. In many cases, nearly 40 to 60 percent of time in claims processing was reduced, and error rates were significantly reduced. There are traces and automated logs that make it easier to comply with auditing and internal reviews. However, critics point out that long-term study of such strategies is required to establish sustainability. But there are hidden costs as well, such as the reluctance of employees to adopt the system, training needs, and system maintenance [12]. Moreover, automation tools will be fully compliant only if they are configured to suit local regulations and updated continuously to reflect policy changes.

III. METHODOLOGY

A. Research Design

An explanatory research design is used by this study to find an enhancement of healthcare claims processing through intelligent automation with human supervision using UiPath's Document Understanding and Action Centre. Explanatory research is well suited for describing the relationship between causes and effects and the impact of different automation tools on the processing efficiency, data accuracy and compliance. It allows for a thorough analysis of how a human in the loop (HITL) framework enhances results in a highly regulated field like healthcare.

B. Data Collection

Using secondary data sources, both qualitative and quantitative methods are used. The data is collected quantitatively from existing industry reports, automation case studies, and performance metrics of healthcare institutions with the erstwhile existence of UiPath Solutions. The data shows how much time was taken for the processing, error rate and compliance score before and after automation. Scholarly articles, industry papers, expert, and user feedback are used as a source to collect qualitative data on system usability, HITL implementation challenges and its effectiveness. However, by using different data types, a total overview of the topic is made the balance is struck between statistical trends and contextual interpretations.

C. Case Studies/Examples

Case Study 1: Enhancing NHS Invoice and Claims Processing through UiPath Document Understanding

NHS Shared Business Services (NHS SBS used UiPath) automation tools to take manual admin tasks off their hands, including handling of invoices and claims [13]. When combined with Document Understanding, they automated data extraction from healthcare documents, lowering manual errors for entry and processing times, and still staying in NHS regulation compliance.

Case Study 2: Human-in-the-Loop Automation at Bupa UK: Ensuring Accuracy in Healthcare Claims Processing

Bupa, one of the top healthcare providers, benefited from a human-in-the-loop automation model handling the processes of insurance claims [14]. Allowing Action Centre to tackle both high-risk and incomplete claims that require human review meant that Bupa could rely on it to validate data and prevent regulatory breaches.

Case Study 3: Driving Compliance and Efficiency at AXA Health UK with UiPath's Intelligent Automation Framework

AXA Health tailored the UiPath RPA and Document Understanding solution to boost the accuracy of their medical billing processes [15]. Incorporated to manage exceptions, HITL was set up to process efficiently albeit ensuring that sensitive data was checked as required, in accordance with the GDPR and FCA compliance requirements.

D. Evaluation Metrics

Thus, to evaluate the effectiveness of the integrated solution, several key evaluation metrics are used that include reducing the processing time by percentage, improving the error rate, adherence to the compliance (audit readiness, HIPAA alignment) and the human intervention frequency. Additionally, qualitative success indicators like user satisfaction, ease of adoption, and system flexibility are also considered [16]. Given the particular nature of the healthcare environment, including strict data integrity and quality compliance requirements that need to be met, these metrics help determine not only whether the hybrid approach increases efficiency but also whether the data in this hybrid approach satisfies those needs.

IV. RESULTS

A. Data Presentation

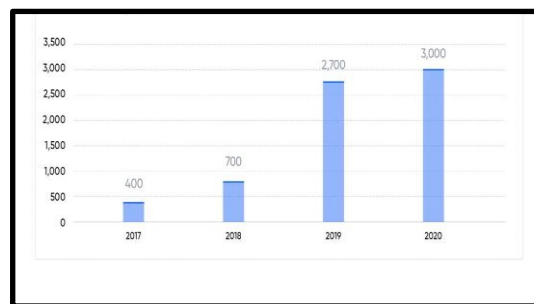


Figure 4: UiPath's Team Has Grown from 100 to 2,700 in 2 Years

[17]

The above graph shows the amount of UiPath's exponential team growth from 400 employees, 2017, 3,000 employees, 2020. The company's growing global demand for intelligent automation has prompted a high percentage of increase, which is an aggressive expansion of the company [17]. This surge comes within the context of healthcare claims processing and will assist UiPath in continuing to improve its Document Understanding and Action Centre capabilities and expand on scalable and responsive human-in-the-loop (HITL) solutions.



Figure 5: The Robotic Process Automation market

[18]

As per the above figure, “global market for RPA Software and Services” touched \$271 million in 2016 and is anticipated to touch \$1.2 billion by the end of 2021 along with compound annual growth rate of 36% [18]. “Direct service market” includes application and consulting accommodations concentrating on creating RPA capabilities insight the company. This massive market expansion presents an opportunity for UiPath’s offerings, including the automation of claims processing and human validation steps.

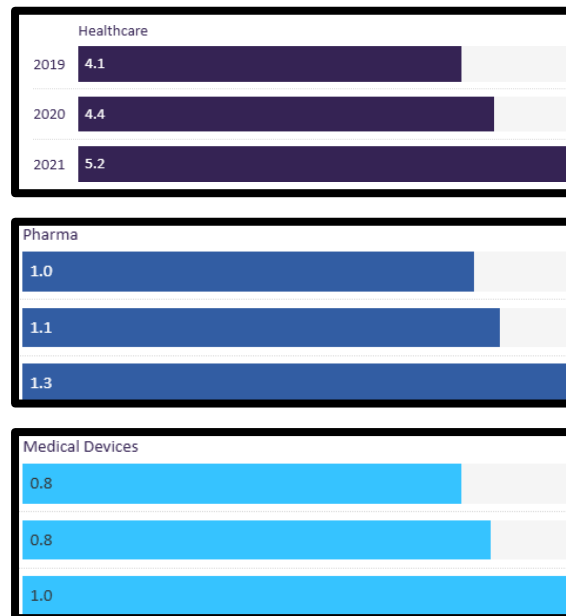


Figure 6: Market Size of Artificial Intelligence in Healthcare (2019 – 2021)

[19]

The following figure 6 illustrates that healthcare providers has the most spending on AI domains rather than medical device and pharmaceutical companies from \$4.1 billion in 2019 to \$5.2 billion in 2021 with CAGR of 13.2% [19]. Spending on AI platforms of pharmaceutical firms has touched \$1.0 billion in 2019 to \$1.3 billion in 2021 and medical device companies spent \$1.0 billion in 2021 in comparison to \$0.8 billion in 2019 [19].

B. Findings

The three graphs show a large upward trend of automation and AI adoption, especially in healthcare. Just rising demand for scalable automation solutions is UiPath’s workforce growth from 400 in 2017 to 3,000 in 2020 [17]. “The global market for RPA Software and Services” is anticipated to reach \$1.2 billion by the end of 2021 with a 36% “compound annual growth rate” [18]. In addition, the AI healthcare market has developed from \$5.9 billion in 2019 to \$7.5 billion in 2021, along with a CAGR of 12.5% [19]. This confirms that healthcare claims processing is increasingly relying on intelligent automation, which is more human-in-the-loop.

C. Case Study Outcomes

Case Study	Company	Key outcomes
Case Study 1: Enhancing NHS Invoice and Claims Processing through UiPath Document	NHS	The automations of UiPath’s Document Understanding allowed NHS SBS to experience a great deal of processing efficiency

Understanding		improvements in routine invoice and claims handling, respectively [13]. That substantially decreased the number of entry errors and improved the accuracy of financial records. Most importantly, and a point that has disgusted and enraged many, the automation was entirely and completely in line with NHS rules and regulations.
Case Study 2: Human-in-the-Loop Automation at Bupa UK: Ensuring Accuracy in Healthcare Claims Processing	Bupa UK	After the integration of Bupa's Action Centre into its claims process, the claims accuracy increased and the regulatory risk was reduced [14]. In a human-in-the-loop model, this allowed complex, high-risk claims to be reviewed and validated by staff to provide the balance of speed and accuracy. It meant quick claims processing yet in strict healthcare and insurance compliance.
Case Study 3: Driving Compliance and Efficiency at AXA Health UK with UiPath's Intelligent Automation Framework	AXA Health UK	AXA Health lodged its medical billing process using UiPath's intelligent automation [15]. AXA were able to manage exceptions efficiently with the help of Document Understanding and HITL features and still protect sensitive data. The operational efficiency was made possible without undermining the regulatory responsibility thanks to automation that adhered to GDPR and FCA Standards.

Table 1: Case Study Outcomes

(Source: Self-Created)

The above table shows that by combining UiPath's automation with some level of human-in-the-loop validation, can improve efficiency, accuracy and compliance in processing. Furthermore, Features for NHS, Bupa UK and AXA Health demonstrates its ability to all three companies to all three companies successfully reduce errors and to meet regulatory adherence whilst achieving overall enhanced operational performance.

D. Comparative Analysis

Author	Focus area	Key Findings	Gaps
[5]	Intelligent automation in claims	decreases manual work and errors, improving the quality of service	Losing when dealing with ambiguous data and rule changes [5]
[6]	Limitations of RPA	Highlights contextual and regulatory limitations in healthcare automation	Lacks human oversight consideration in complex cases [6]
[7]	Automation evolution with UiPath	80 % efficiency rate with UiPath and IRPA [7]	UiPath is based on standardised input – no handling for variability
[8]	In document understanding for OCR and NLP	OCR and NLP are better than manual data extraction	accuracy falls with bad scans and non-standard formats.
[9]	UiPath action centre and HITL	HITL run faster than human actions due to the idleness of humans	Overuse of HITL reduces efficiency [9]
[10]	DRL in HITL models	Decision making is improved with DRL in HITL setups	Error may be nullified or exceeded by improper tuning
[11]	Human validation and efficiency	human validation and time reduction of 40-60 %, and greater auditability	Long-term sustainability is not yet proven
[12]	Barriers to automation efficiency	Training needs and hidden costs keep operating system inefficiency at its	Policy changes need constant system updates [12]

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Table 2: Comparative Analysis

(Source: Self-Created)

The comparative analysis finds out that although UiPath-driven automation increased efficiency, accuracy and compliance in healthcare claims, it still has difficulty with handling data variability, system sustainability and user adoption. Regulatory reliability and contextual accuracy can, however, only be achieved with a human in the loop.

V. DISCUSSION

A. Interpretation of Results

The results section of the study fit well with the objectives of the project. The findings covered each objective, from adding more processing efficiency and accuracy to complying with regulations through HITL [20]. Automation that is coupled with human validation enables performance and governance goals by removing the need to reduce error rates or to meet compliance, as evidenced by the adoption of UiPath, market expansion and NHS, Bupa and AXA's revolutions in errors and compliance. Furthermore, by providing support for the study aim of improving claims processing through hybrid automation, the comparative analysis also strengthens the need for HITL to overcome automation limitations.

B. Practical Implications

Integration of Document Understanding and Action Centre into healthcare claims processing is very beneficial practically. It automates repetitive tasks like data extraction and validation, helps in saving humans from making errors, and frees up people to work on the complex cases. HITL models have proved to increase processing accuracy and regulatory compliance in healthcare organisations such as the NHS, Bupa and AXA [21]. The use of such a hybrid automation approach will guarantee that the sensitive patient data is handled appropriately, yet achieve efficiency in the healthcare sector for meeting both the operational and legal standards.

C. Challenges and Limitations

There are benefits that are outweighed by some challenges. However, poor quality scans, handwritten documents, and non-uniform formats all complicate the operation of intelligent automation systems, thereby limiting full automation. Relying too much on human intervention may not be efficient, and AI models must be continually retrained regarding policy or data changes. Additional barriers include employee resistance, system maintenance, and high implementation costs, particularly for small healthcare providers with very limited budgets.

D. Recommendations

Phased automation is the way to go for healthcare providers, and the best starting point is high-volume, low-risk processes. For HITL thresholds, it should be established to have continuous monitoring and feedback loops to tune thresholds [22]. Additionally, employee training and change management will aid adoption. The system must be audited regularly to ensure the system complies with changing healthcare regulations.

VI. CONCLUSION AND FUTURE WORK

This analysis concludes that combining UiPath's Document Understanding and Action Centre into healthcare claims processing can greatly aid with automation enablement by human validation. This Human-in-the-Loop (HITL) approach improves accuracy of data; reduces the time it takes to process it and helps to obey strict regulations like HIPAA and GDPR.

Future research work should be aimed at long-term sustainability and scalability of HITL models, specifically in small health care organisations. More crude documents could be analysed by adaptive AI models that need less retraining and better OCR capabilities. Further, there should be ongoing studies assessing how users adopt the system and how this system is complied with over time.

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