

Cognitive Automation in Procure-To-Pay: Re-Engineering Workflows with Rule-Aware Robotics

Jagadeesh Vasanthada
Independent Researcher
Alpharetta, USA
Jagadeesh.Vasanthada@gmail.com

Abstract— Whether manual processes, inefficiencies, or operational risks, the procurement-to-pay (P2P) process is at the heart of every enterprise. The traditional P2P system is high-costing, has a long cycle time, and is highly error-prone. You could re-engineer workflows using cognitive automation and rule-aware robots. Financial control and relationship strain with suppliers limit the finance and procurement teams' ability to operate efficiently. We investigate how applying cognitive RPA's optical character recognition and rules can create a more agile, accurate, and cost-effective P2P ecosystem. We don't just automate tasks; we rethink your entire workflow: automated invoice processing, intelligent three-way matching, and streamlined exception handling. Providing P2P solutions across the globe. Organizations are looking to use rule-aware bots within existing platforms like SAP Ariba for straight-through processing, compliance, and to empower the finance team to move from transactional execution to financial management. The article presents a framework for this re-engineering process and shares the key operational and strategic benefits of a cognitive automation-first approach to the Procure-to-Pay cycle.

Keywords- Procure-to-Pay, P2P, Automation, RPA, SAP Ariba, Workflow Design

I. INTRODUCTION (HEADING 1)

Nowadays, efficiency and accuracy of the Procure-to-Pay (P2P) process are essential in modern business management. The entire process—from purchase requisition to supplier payment—is used by the procurement and finance team to perform end-to-end functions. Performance is of significant importance to an organization's liquidity, operational costs, compliance, and supplier relationships. The peer-to-peer cycle is notoriously heavy on manual intervention in many organizations, despite its mission-critical status. This opens it up to operational bottlenecks, errors, and rising costs.

It is well-known that traditional P2P workflows face challenges. Manual data entry from invoices is tedious and error-prone. Three-way matching takes up a lot of time. Plus, this matching process produces a high exception volume. Investigating takes a lot of manual effort. Without being in the same organization's chain, the payment process can be lengthy and complicated. Therefore, an early discount opportunity may be lost. Besides, the vendor relationship may also get strained. Moreover, the lack of visibility into the real-time process hampers cash flow and planning for sourcing strategies.

Robotic Process Automation (RPA) can be a great solution to these issues that we have faced in recent years. Robotic Process Automation (RPA) uses computerized bots to mimic human actions and perform rule-based tasks. To realize RPA's full transformative power, we need to go beyond RPA and augment it with cognitive capabilities. Bots that can work with semi-structured data and execute simple, logic-based decision-making are known as cognitive automation. This development brings intelligence to process automation.

The study focuses on re-engineering the P2P workflows using cognitive automation and rule-aware robotics. It is the study of how knowledge and data affect workflows. Intelligent bots can automate complex processes, including invoice data extraction and three-way matching, exception management with business rules, and integration with existing ERP and

procurement systems, such as SAP Ariba. The objective is to build and execute a rapid, controlled, scalable P2P feature that can generate essential financial and strategic value.

II. THE PROCURE-TO-PAY LANDSCAPE

A. Core Components of the P2P Cycle (Heading 2)

The P2P process is a series of interconnected processes that manage a company's purchases and payments. Although there may be variations, the basic stages are the same:

1. **Requisitioning:** The cycle begins with the internal department raising a purchase requisition when a need for goods or services is identified. The document describes the what, why, and when of the purchase that is needed [1].
2. **Purchasing:** From now on, the requisition will be converted to a Purchase Order (PO) upon approval. A Purchase Order (PO) is a legally binding document sent by a buyer to a supplier to authorize a purchase under the terms specified, such as price, quantity, delivery date, etc. [2].
3. **Receiving:** Once the goods or services are delivered, the organization records them in the form of a Goods Receipt (GR) or a Service Entry Sheet. The supplier has fulfilled their obligation under the contract as per the PO at this step.
4. **Invoice Processing:** A payment request is made when the supplier sends an invoice. Afterwards, the accounts payable (AP) department must capture invoice data and perform three-way match validation on the PO, the GR, and the invoice.
5. **Payment:** After an invoice is validated and approved, it will be scheduled for payment. The final stage is to make the payment to the supplier as per the contract and post it in the books [3].

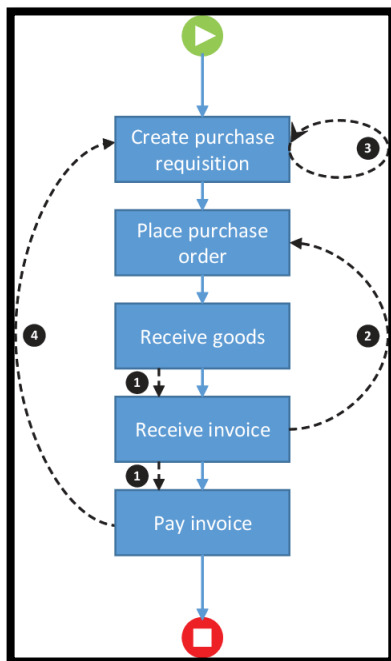


Figure 1: P2P-process-with-four-execution-variants

B. Inherent Inefficiencies in Traditional Workflows

In a non-automated or partially-automated environment, each of these stages presents opportunities for inefficiency and error.

Manual Data Entry and Validation: It is painfully slow and a rich source of errors [2] to manually key in data from hundreds or thousands of invoices received in multiple formats, such as paper, PDF, and e-mail. Just one mistyped digit can cause overpayments, underpayments, or payment rejections that require expensive rework.

Labor-Intensive Matching Process: Invoice Validation Through Three-Way Match. This step is the most labor-intensive and depends on considerable manual input. However, it is rather labor-intensive if done manually. AP clerks or Accounts Payable clerks are responsible for obtaining three sets of documents from various sources and verifying them [4].

High Cost of Exception Handling: When exceptions occur during matching, the processing cost increases. Those previous calculations are still being justified throughout the checks done one after the other. Talking to procurement staff, warehouse workers, suppliers, and others led to a delay in the earlier payment cycle, warranting a detailed probe.

Limited Visibility and Control: When you manually track invoices, it limits your visibility into invoice status and overall accounts payable (AP) performance. Hence, it becomes challenging to predict cash outflows, such as bills and invoices. Further, it hampers the ability to take any discounts for paying early. This causes slowdowns when processes often bottleneck and agree on a supplier. Audits/Compliance can only be possible with Transparency.

III. COGNITIVE AUTOMATION AND RULE-AWARE ROBOTICS

Cognitive automation is superior to traditional process automation by a long shot. It combines the operational advantages of RPA with an additional layer of digital intelligence, enabling the automation of more complex, judgment-based tasks that were previously achievable only by humans.

A. Defining Robotic Process Automation (RPA)

The essence of Robotic Process Automation is the configuration of software that can execute a business process. According to [5], these bots make use of application GUIs to interface with them in the same way a human would. They can access system login, navigate menus, copy and paste data, fill out forms, and transfer data between applications. One significant advantage of RPA is its non-invasiveness [12]. It does not require changes to existing IT infrastructure or application code. This ability makes RPA very affordable and fast to implement.

B. The Advent of “Cognitive” Capabilities

By using technologies that enable bots to learn to handle variability and make simple decisions, RPA gains’ cognitive’ abilities. In the context of P2P, the most relevant of these are:

Optical Character Recognition (OCR) and Intelligent Character Recognition (ICR): In simple terms, these bots can read PDFs, scanned documents, and images to extract valuable data. OCR enables the recognition of characters in an image to convert them into machine-readable codes. While that is true, ICR adds intelligence by leveraging context and validation rules to improve extraction accuracy from semi-structured formats [6].

Rule-Based Engines: This is the core of rule-based robots. A rule engine is a program component enabling the definition of business logic and policies using if-then rules [11]. The RPA bot can input data into this engine, which gives either a decision or an outcome. For example, you can say, if the invoice amount is equal to the P.O. amount and the invoiced quantity is equal to the goods receipt quantity, approve the invoice for payment.”

Natural Language Processing (NLP): This ability is more advanced. Basic NLP enables the bot to understand unstructured text and extract some information from it. For example, a body of email containing an invoice or a payment query from a supplier [7].

TABLE I. KEY COGNITIVE CAPABILITIES FOR RPA IN PROCURE-TO-PAY

Cognitive Capability	Core Function	Application in Procure-to-Pay (P2P)	Key Benefit
Optical/Intelligent Character Recognition (OCR/ICR)	Enables software bots to “read” and extract structured data from digital images and semi-structured documents (e.g., PDFs).	Automatically capturing data fields such as invoice number, date, line items, and total amount directly from supplier invoices.	Eliminates manual data entry, reduces human error, and accelerates the initial stage of invoice processing.

Rule-Based Engines	Executes predefined business logic and policies through a series of “if-then” conditions to make automated decisions.	Performing the three-way match by comparing invoice data against purchase order and goods receipt information to validate transactions.	Enables “straight-through processing” for compliant invoices and automates initial exception handling based on defined business rules.
Natural Language Processing (NLP)	Allows bots to interpret and extract relevant information from unstructured human language (text).	Understanding the content of supplier emails to identify attached invoices or to categorize and route payment status queries.	Expands automation to cover communication channels, reducing the need for manual monitoring of inboxes and supplier correspondence.

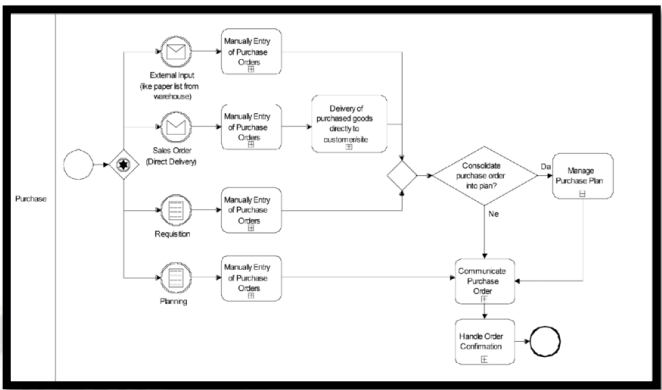


Figure 2: Automated Requisition and PO Management

B. Intelligent Invoice Processing

Cognitive automation would add the most value here. It can re-engineer the whole invoice processing system for STP. The change begins with allowing the software bot to monitor only for the arrival of supply invoices. The bot can be configured to issue commands and monitor dedicated intake channels, such as an email inbox or a network folder. This bot employs OCR technology and ICR techniques to read in arriving invoice data. The software can read through the layout and the format of the document. These software programs will be able to extract essential data points, such as invoice number, date, supplier name, PO number, and line-item details, in a single successful extraction. Capturing information digitally from the beginning eliminates the chance of making errors.

Once we digitize the data, the workflow automatically proceeds to the automated three-way match, which is the bot’s main validation logic [10]. The bot now performs this tedious manual job with speed and accuracy no human can match. The invoice data extracted is consciously interfaced with the ERP or P2P system to fetch the relevant purchase order and retrieve the associated goods receipt (GR). The bot’s rule engine checks the three documents by comparing them against the defined business rules and tolerance levels. It validates price, quantities, etc. “When the invoice, PO, and GR match in many cases, straight-through processing happens in the system.” The bot ends the workflow by posting the invoice for payment in the financial system with no human touch required. This ability speeds up invoice cycle times for a large portion of AP volume, reducing them from days to minutes, improving efficiency, and helping the organization take advantage of early payment discounts.

C. Rule-Based Exception Handling

The real smarts of the re-engineered workflow are its handling of exceptions. When the bot encounters an issue, it doesn’t just error out and leave the job in a stack for a human to deal with. Instead, it utilizes its rule engine to do initial triage and resolution:

Categorization and Routing: The bot can identify a mismatched error type. If there is a price discrepancy, the rule could be to route the exception to the appropriate category manager in the procurement team. The warehouse or receiving department may

C. The Synergy with P2P Platforms (Heading 2)

The aim of cognitive automation is not to replace strong P2P platforms like SAP Ariba but to supplement them. These platforms provide a structured environment for purchasing, billing, and payments. It usually takes manual effort to get information into these systems and to move information between modules or connected software programs [8]. Bots that are aware of rules are the digital glue that can do this automation. For instance, a bot can pick up invoice data from an email, log in to SAP Ariba, enter it, trigger the matching process, and, based on the outcome, route the invoice for payment or escalate it as an exception, all without human intervention.

IV. RE-ENGINEERING P2P WORKFLOWS WITH RULE-AWARE ROBOTICS

Using rule-aware robotics, the P2P workflow can be redesigned from a manual, sequential workflow to an automated, parallel, and exception-based workflow.

A. Automated Requisition and PO Management

Although less complex, the front end of the P2P cycle would also benefit from automation. A bot can monitor multiple request queues across different departments. It can check the requisition against the budget code and the procurement policy to determine whether it is valid. It can also check if a preferred supplier agreement is in place [9]. After a requisition is approved, a bot can create a standard PO from the template, fill in the details, and send an email to the supplier. That speeds up the procurement initiation phase.

handle a quantity discrepancy due to a potential measurement mode. This guarantees that the exception reaches the individual best able to fix it without delay [10].

Automated Remediation: For low-risk and standard exceptions, the bot can be programmed to take action autonomously. For example, if the bot recognizes an invoice as a duplicate of an existing one in the system, it can reject it and inform the supplier. If the term "PO Number" is missing in the invoice, the bot can send an automated email back to the supplier [9].

D. *Streamlining Supplier Onboarding and Management*

Supplier-related processes can also be automated. If a new supplier needs to be onboarded, a bot can kickstart the process. It can send out information forms, validate submitted data (such as tax ID and bank details) against external databases, and create vendor master records in the ERP. Bots can monitor supplier performance by automatically tracking on-time delivery and invoice accuracy [8].

V. WORKFLOW DESIGN CONSIDERATIONS FOR AUTOMATION

A. *Process Discovery and Standardization*

Understanding and documenting the current P2P processes is the first step. Many people try to automate a process that is already broken or overly complicated. Increase the productivity of administrative functions through workflow automation [7]. It means assessing to remove steps, roles, and responsibilities, and to align procedures within the organization.

B. *Designing the Human-in-the-Loop Workflow*

The purpose of automation is not to cut out human involvement, but to boost it. The workflow design needs to help create a human-bot collaborative model. Bots need to take on high-volume, repetitive rule-based tasks so that humans can engage in value-added activities [6]. This includes complex and sensitive exception management, negotiating with suppliers, conducting spend analysis and developing strategic insights, and managing time-sensitive and critical inquiries that require effort, initiative, and judgment. The process must have clear escalation paths from bots to humans.

C. *Building a Robust Rule Engine*

A rule-aware bot will be effective or not depending on how good the rule engine is. The business must establish the reasoning that will guide the bot's actions [5]. This includes setting tolerance thresholds for mismatches (e.g., a percentage threshold for price or quantity variance that triggers an exception), establishing how exceptions are routed, and defining the automated notifications sent to the supplier and the internal team.

D. *Governance and Control*

A strong governance framework requires urgency as organizations deploy a "digital workforce" of bots. To achieve this, one must establish protocols such as performance monitoring, data security manager policies, access control, and an access manager, and analyze a bot's actions [2]. It is essential

to be compliant and fully transparent on the automated activities taking place [13].

VI. CONCLUSION

By shifting from manual to intent-driven automated work with cognitive automation, Finance and Procurement will undergo a complete paradigm shift. When companies change the workflow rather than optimize it, they can achieve step-change improvements in efficiency, accuracy & control. The chargeable process is more of a chargeable function than a clean finance enabler, as it moves fast. Cognitive technology in automation processes reduces costs while increasing quality and speed (20 Words). You can quickly and efficiently process your invoices while lowering costs, improving data accuracy, and strengthening internal controls. If there is an error in issuing compliance orders, a need for human intelligence recruitment will arise, say experts. P2P specialists are increasingly becoming data problem-solvers and business partners. The current business model to develop has no set calendar. So, the period will be delayed initially. The company's wheels are about to be set in motion. The real launch date will be 6 December 2019, to coincide with the quarterly close; that date will again be fixed for Wheels on 6 December 2019. Likewise, official wheels are an agile, data-driven, and scalable financial operation that helps the organization better manage complexity and modern network supplier relationships. The research paper illustrates the technologies that businesses can leverage to enhance operational excellence.

REFERENCES

- [1] M. Brezzi, S. González, D. Nguyen, and M. Prats, "An updated OECD framework on drivers of trust in public institutions to meet current and future challenges," *www.oecd-ilibrary.org*, Dec. 2021, doi: <https://doi.org/10.1787/b6c5478c-en>.
- [2] O. Kolade and A. Owoseni, "Employment 5.0: The Work of the Future and the Future of Work," *Technology in Society*, vol. 71, no. 102086, p. 102086, Aug. 2022, doi: <https://doi.org/10.1016/j.techsoc.2022.102086>.
- [3] L. Maier-Hein *et al.*, "Surgical data science – from concepts toward clinical translation," *Medical Image Analysis*, vol. 76, p. 102306, Feb. 2022, doi: <https://doi.org/10.1016/j.media.2021.102306>.
- [4] K. K. Arun, S. K. Mydhili, S. Baskar, and P. M. Shakeel, "Fuzzy rule-based environment-aware autonomous mobile robots for actuated touring," *Intelligent Service Robotics*, Apr. 2020, doi: <https://doi.org/10.1007/s11370-020-00320-z>.
- [5] C. Jansen and E. Sklar, "Exploring Co-creative Drawing Workflows," *Frontiers in Robotics and AI*, vol. 8, May 2021, doi: <https://doi.org/10.3389/frobt.2021.577770>.
- [6] M. F. Zia, M. Benbouzid, E. Elbouchikhi, S. M. Mueen, K. Techato, and J. M. Guerrero, "Microgrid Transactive Energy: Review, Architectures, Distributed Ledger Technologies, and Market Analysis," *IEEE Access*, vol. 8, pp. 19410–19432, 2020, doi: <https://doi.org/10.1109/access.2020.2968402>.
- [7] B. K. Wiederhold, "Connecting Through Technology During the Coronavirus Disease 2019 Pandemic: Avoiding

- ‘Zoom Fatigue,’” *Cyberpsychology, Behavior, and Social Networking*, vol. 23, no. 7, Jun. 2020, doi: <https://doi.org/10.1089/cyber.2020.29188.bkw>.
- [8] E. Křenková, K. Rieser, and A. Sato, “How software robots can facilitate the procurement process? A case study of Siemens in the Czech Republic,” *Entrepreneurial Business and Economics Review*, vol. 9, no. 3, pp. 191–203, 2021, doi: <https://doi.org/10.15678/eber.2021.090312>.
- [9] H. Abou-Ibrahim and F. Hamzeh, “Understanding Stakeholders’ impact on design workflow dynamics using agent-based modeling,” *Automation in Construction*, vol. 138, p. 104254, Jun. 2022, doi: <https://doi.org/10.1016/j.autcon.2022.104254>.
- [10] J. Corral-Acero *et al.*, “The ‘Digital Twin’ to enable the vision of precision cardiology,” *European Heart Journal*, vol. 41, no. 48, Mar. 2020, doi: <https://doi.org/10.1093/eurheartj/ehaa159>.
- [11] International Journal on Recent and Innovation Trends in Computing and Communication
<https://ijritcc.org/index.php/ijritcc>
- [12] D. Linke and S. Strahringer, “Integration einer Blockchain in ein ERP-System für den Procure-to-Pay-Prozess: Prototypische Realisierung mit SAP S/4HANA und Hyperledger Fabric am Beispiel der Daimler AG,” *HMD Praxis der Wirtschaftsinformatik*, vol. 55, no. 6, pp. 1341–1359, Nov. 2018, doi: <https://doi.org/10.1365/s40702-018-00472-8>.
- [13] Y. Li and C. M. Burns, “Modeling Automation With Cognitive Work Analysis to Support Human-Automation Coordination,” *Journal of Cognitive Engineering and Decision Making*, vol. 11, no. 4, pp. 299–322, May 2017, doi: <https://doi.org/10.1177/1555343417709669>.