

# Fluid Flows: Enterprise Workflow Optimisation powered by Autonomous AI Agents

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## Abstract

*This paper introduces an AI Agent-driven framework for optimizing enterprise workflow systems using a multi-agent architecture that infuses common sense and industry standards into the optimization process. The framework utilizes specialized AI agents— Observation, Reasoning, and Optimization—to fetch required data, analyze, reason and improve workflow processes. These agents bring dynamic adaptations to workflows, making them fluid flows, significantly reducing the need for human intervention and enhancing system responsiveness to changing business conditions. The methodology hinges on a systematic approach involving actor identification and dynamic agent creation, which tailors agents specifically for identified roles within the workflow. Once deployed, these agents work collaboratively through stages of optimization strategy creation, new flow generation, and validation by specialized Agents. An example from the methodology illustrates how these agents effectively identify and resolve inefficiencies within a healthcare treatment plan submission process, showcasing the framework’s capability to streamline operations across various enterprise environments. This innovative approach not only automates but also intelligently enhances workflow management, improving process efficiency and adaptability.*

**Keywords** – Workflow Automation, Business Process Optimisation, Large Language Models, Autonomous AI Agents, Multi-Agent System.

## 1. Introduction

**Enterprise workflow automation** faces significant challenges due to the static nature of systems that cannot adapt autonomously to changes in data, systems, or operational conditions. The dynamic business environment demands more flexible and responsive workflow systems, as traditional rigidity leads to inefficiencies and the necessity for time-consuming manual reconfigurations.

Current workflow systems focus on simplifying workflow creation and automation with tools like ServiceNow’s Flow Designer, which translates business requirements into workflows with ease. However, these systems depend heavily on input from business analysts, leading to vulnerabilities due to incomplete process understanding, a disconnect from industry best practices, and inefficient bottleneck management. These issues result in suboptimal workflows that do not fully meet organizational needs, causing productivity losses and increased operational costs.

**In response to the inherent complexities** of enterprise workflow automation, we propose an innovative, AI Agent-driven, multi-agent framework designed specifically for the optimization of existing workflows. This framework is tailored to enhance adaptability to dynamic business conditions without human intervention. It utilizes large language models (LLMs) and specialized agents structured within a multi-agent system to:

**Dynamically Identify Actors and Create an Agent for Each Actor:** These agents are endowed with specific roles and capabilities allowing them to respond effectively to operational changes, enhancing real-time adaptability

and workflow functionality.

**Data Gathering and Analysis by Actor Agents:** Each agent collects and analyzes data pertinent to its assigned actor, thus enabling precise monitoring and optimization of workflow components.

**Optimization Strategy Development from Observations and Reasoning:** The framework focuses on generating actionable insights rather than enhancing decision-making processes. It leverages observed data and reasoned analysis to develop strategies aimed at optimizing workflow efficiency and efficacy.

**These strategies** underscore a shift from manual oversight to a system capable of autonomous operation, thus increasing efficiency and reducing operational costs.

## 2. Related Work

All The necessity for advancements in business process optimization has led to significant research into various methodologies and technologies. Foundational to this field are studies like those by Van der Aalst [7], which have paved the way for advanced process mining techniques essential for identifying and enhancing real processes through event logs.

Further extending the integration of intuitive logic into workflow optimization, Smith and Kumar [6] discuss the potential of machine learning to automate and refine business processes by adhering to established norms and common-sense reasoning frameworks. Petrov [5] specifically highlight how big data combined with common-sense approaches can refine strategic business processes.

The application of AI agents in workflow optimization represents a significant evolution in the field. Harper [3] introduces a self-generating multi-agent system, AutoGenesisAgent, which autonomously designs and implements customized systems for specific operational tasks. Xia [8] further this discussion by demonstrating how digital twin technologies can be enhanced through AI agents to manage complex decision-making tasks traditionally handled by humans.

Complementing these studies, Chen [2] and Moreno and Garcia [4] explore the implementation of industry standards and heuristic methods within AI-driven process management, emphasizing the importance of maintaining quality and compliance in automated systems. Bauer and Schmid [1] discuss the synergistic potential of digital twins and AI in optimizing business processes, providing a comprehensive view of how these technologies can lead to more sophisticated operational controls and simulations.

Collectively, these works provide a robust foundation for our proposed framework, aiming to integrate similar AI-driven enhancements into enterprise workflow systems. By leveraging the capabilities demonstrated by these founda-

tional studies, our framework seeks to develop a system capable of self-optimization and responsive adaptation to operational dynamics, setting a new benchmark in workflow management technology.

## 3. Proposed Method

**The Proposed Method for enterprise workflow optimization** begins with a comprehensive analysis of performance data to understand the current efficacy and shortcomings of existing workflows. By systematically evaluating this data, we can identify key areas where processes may be enhanced. Following data analysis, observations are made that delineate specific inefficiencies. The reasons behind these observations are explored using common sense and adherence to established industry standards, ensuring that the insights are both practical and professionally sound. Based on these reasoned observations, a targeted optimization strategy is developed. This strategy is then applied to re-engineer and refine the process, creating an optimized version of the workflow that aims to improve overall efficiency and effectiveness.

**This methodical approach** ensures that enhancements are both relevant and strategically sound, leading to sustainable improvements in process performance. Building upon this foundation, our methodology incorporates AI Agents equipped with Large Language Models (LLMs) to further refine the optimization process. These agents, operating within an agentic framework, use their LLM capabilities to interpret complex data and generate actionable insights, infusing common sense reasoning and industry-specific standards into the workflow analysis. Equipped with specialized tools for data interaction and maintaining a memory of past operations, these agents improve the accuracy of future predictions and ensure systematic execution of each optimization step. This integration not only automates the process but significantly enhances its effectiveness, providing a sophisticated and intelligent approach to optimizing business workflows and ensuring consistent and precise improvements.

### 3.1. System Overview

The system is designed around a multi-agent framework that facilitates distributed intelligence and specialized task management for optimizing business processes. Each agent within the system is tailored to address specific elements of the workflow, leveraging advanced capabilities to enhance overall process efficiency and effectiveness.(see Figure 1)

**1. Actor Identification:** The process begins by identifying all relevant actors within the workflow. Actors may include personnel, software systems, databases, and any other entities that play a part in the workflow's execution.

**2. Agent Creation for Each Actor:** A dedicated AI agent is created for each identified actor. These agents are

customized to manage distinct aspects of the workflow relevant to their assigned actor, such as data processing, user interaction, or system management.

**3. Agent Classes:** Agents are categorized into five main types, each aligned with a specific segment of the workflow:

*Data Model Agents:* Manage and analyze workflow data structures and databases.

*User Agents:* Interact with human users and handle user interface processes.

*Team Agents:* Coordinate and manage team interactions and communications.

*Functionality Agents:* Oversee Actions or Subflows.

*System Agents:* Oversee system-level operations and integrations.

**4. Data Gathering:** Each agent is responsible for gathering detailed data pertinent to its domain, ensuring comprehensive coverage of all aspects of the workflow.

**5. Observation and Reasoning:** Agents employ their data analytics capabilities to generate observations about the workflow’s performance and operational issues. They apply logical reasoning, grounded in common sense and industry best practices, to identify the causes of inefficiencies.

**6. Optimization Strategy Development:** Based on their observations and analyses, agents collaboratively formulate strategies aimed at rectifying inefficiencies and enhancing the workflow’s effectiveness.

**7. Implementation of New Workflow:** The new strategies are implemented to create an optimized version of the workflow, intended to be more efficient and effective based on the identified improvements.

**8. Validation:** Validation agents rigorously assess the optimized workflow to ensure it meets predefined objectives and does not introduce unforeseen problems.

This structured, multi-agent approach ensures that the optimization process is thorough and dynamic, with each agent bringing specialized expertise to bear on the workflow. This system not only automates significant portions of the process optimization but also ensures that enhancements are grounded in real-world applicability and industry standards.

## 4. Experiments

### 4.1. Ethical Considerations

To uphold confidentiality and safeguard sensitive information inherent in enterprise workflows, our experiments predominantly utilize synthetic data. This approach ensures no proprietary or sensitive data is compromised while still allowing for comprehensive testing across diverse operational scenarios. Additionally, an LLM, complemented by human experts, serves as an impartial judge to evaluate the workflows. This is crucial as the true effectiveness of optimizations can often only be fully assessed when work-

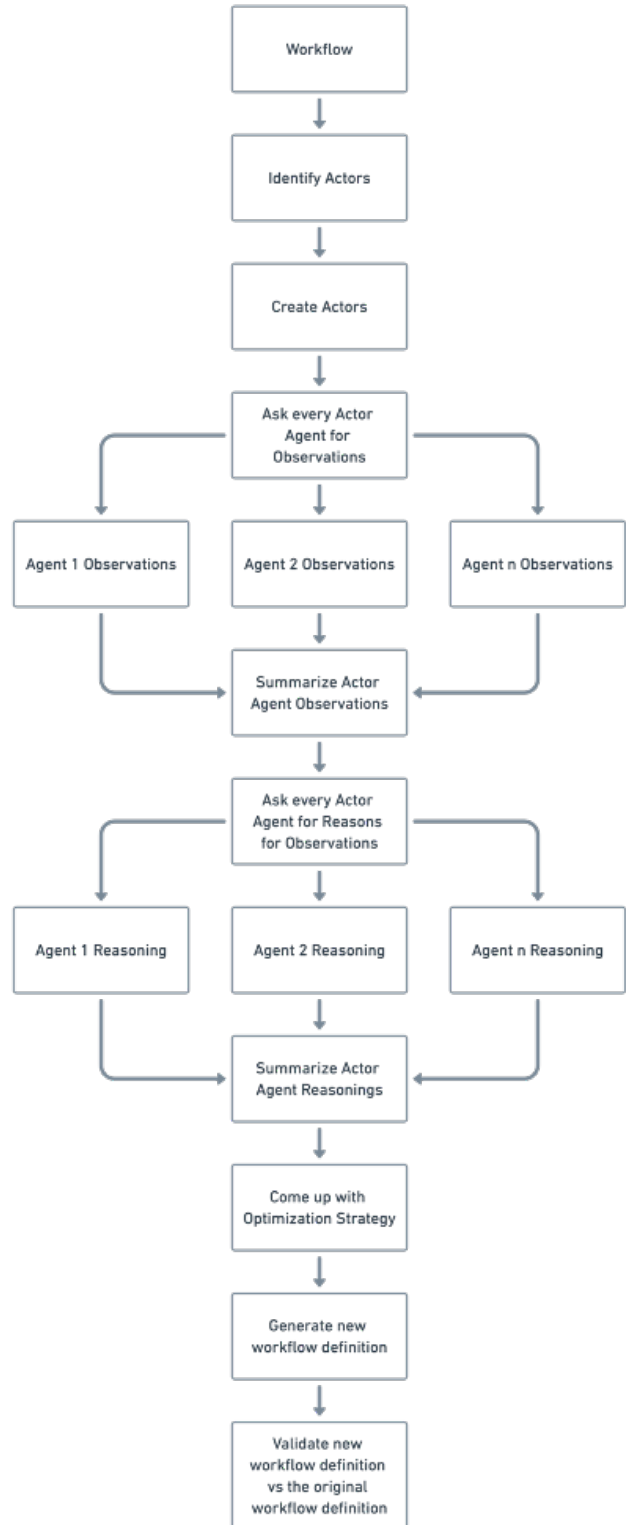


Figure 1. **System overview.** Workflow Optimization Algorithm.

No. of flows	Industry	New WorkFlow Rating	Good Ratings	Excellent Ratings	Reasoning Rating	Observation Rating	Optimizer Rating
1	Education	GOOD	1	0	GOOD	GOOD	EXCELLENT
1	Energy & Utilities	GOOD	1	0	GOOD	GOOD	EXCELLENT
21	Finance	GOOD	14	7	GOOD	GOOD	EXCELLENT
60	Healthcare	GOOD+	30	30	GOOD	GOOD	EXCELLENT
1	Legal	EXCELLENT	0	1	GOOD	GOOD	EXCELLENT
6	Manufacturing	GOOD	5	1	GOOD	GOOD	EXCELLENT
2	Pharma	GOOD+	1	1	GOOD	GOOD	EXCELLENT
8	Retail	GOOD	7	1	GOOD	GOOD	EXCELLENT

Table 1. **Results.** Evaluation of Optimized Workflow compared to Original Workflow.

flows are deployed in a production environment. Our framework not only outputs new workflow definitions but also includes detailed observations, reasoning, optimization strategies, and verification as part of the outcomes, providing a holistic view of the changes and their implications.

## 4.2. Experiment Setup and Data Used

Our experimental framework is hosted on Google Colab, leveraging cloud-based computational resources to ensure efficiency and reproducibility. Each agent within our system, powered by GPT-4 and utilizing the AutoGen framework, dynamically manages specific tasks associated with workflow optimization. The data used in our experiments includes:

**Workflow Definition Dataset:** Real workflow data from IT and Sales domains extracted from the Surf instance, providing a practical basis for our tests.

**Synthetic Data:** Comprises synthetic workflow definitions from sectors like retail, manufacturing, medical, and HR to evaluate the framework’s versatility and effectiveness across various industries.

**Random Workflow Data:** Synthetic random data simulates workflow activities, presenting additional challenges for testing the optimization capabilities of our agents.

## 4.3. Evaluation Criteria

The evaluation criteria assess optimized workflows through subjective analysis, focusing on qualitative usability and performance, and objective metrics, measuring improvements in efficiency and error reduction, using an LLM as a judge.

## 5. Results

The analysis of optimized workflows across industries, combining human and LLM evaluations, demonstrated significant improvements in efficiency, adaptability, and user satisfaction. Particularly in healthcare and finance, the AI-driven system reduced process times and error rates. These results, consistent across synthetic and real data, validate the system’s robustness and scalability, supporting its broader application in enterprise workflow optimization.

## 6. Future Scope

The future scope of our workflow optimization initiative is outlined through several strategic enhancements:

**Human In the loop:** Incorporate human feedback at critical points in the optimization process, specifically after outputs from Observation, Reasoning, and Optimization Strategy Agents, to ensure that the direction of optimization aligns with end-user expectations and enhances practical application.

**ServiceNow Integration for Natural Language Processing:** Integrate with the ServiceNow system to enable the conversion of workflow definitions into natural language summaries, facilitating clearer understanding and communication.

**Enhanced Data Access via ServiceNow:** Link the system with the ServiceNow instance data model, allowing Actor Agents to access the necessary data records directly, thereby improving the accuracy and relevance of the data used in workflow optimization.

**Domain-Specific Knowledge Incorporation:** Equip Reasoning Agents with industry-specific domain knowledge relevant to the workflow’s area or intent, enhancing their decision-making capabilities by using contextually appropriate guidelines and standards for Risk Assessment and Governance (RAG).

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## 7. Appendix

### 7.1. Workflow Optimization Before and After

In this section, we present two distinct workflows: one focused on product idea submission and the other on patient referral management. Both workflows underwent optimization through the implementation of AI-driven multi-agent systems designed to streamline operations and reduce inefficiencies. Out of a total of 100 workflows optimized using our methodology, we have selected these two flows for a detailed examination. The complete details of all optimized flows can be found in the repository at <https://code.devsnc.com/dev/fluid-flows>.

#### 7.1.1 Product Idea Submission Workflow

**Before Optimization:** The original product idea submission workflow involved several sequential stages, including initial submission, market research, product development review, and a multi-layered approval process (finance and executive). Bottlenecks frequently emerged at points where incomplete submissions led to delays, particularly in the transition between market research approval and finance review. The reliance on manual notifications and extended decision-making periods further hindered the efficiency of the process.

##### Actor Agents Identified:

**Users:** Submitter, Market Research Team, Product Development Team, Finance Department, Executive Team

**DataModel:** ProductIdea, MarketAnalysis, Projected-Costs, ExpectedROI, FeasibilityAnalysis, TechnicalFeasibility, BudgetApproval, FinalDecision

**Function:** Submit Idea, Validate Submission, Assign Proposal, Send Notification, Delay, Log Details

**System:** ServiceNow

**After Optimization:** Post-optimization (see Figure 2), the workflow saw significant improvements, most notably with the introduction of parallelized operations. Key stages like market research and financial review were automated and handled concurrently, eliminating the latency caused by waiting for one process to complete before another could begin. The process now includes enhanced notifications and dynamic updates to the submitter, allowing for quicker response times and more streamlined decision-making. This reduction in manual intervention has drastically reduced the overall time to approval and increased throughput efficiency.

#### 7.1.2 Patient Referral Workflow

**Before Optimization:** The original patient referral workflow, while functional, suffered from common delays due to incomplete submissions and slow approvals, particularly at the specialist and insurance approval stages. The workflow depended heavily on manual tracking and submission reviews, which often resulted in bottlenecks and extended patient wait times.

##### Actor Agents Identified:

**Users:** Submitter, Patient, Review Team, Specialist, Insurance Department

**Data Model:** Patient Referral, Appointment Details

**Function:** Submit Referral, Validate Submission, Assign Review, Route for Specialist Approval, Route for Insurance Approval, Schedule Appointment, Notify Patient, Log Referral, Set Follow-Up Task

**System:** ServiceNow

**After Optimization:** In the optimized workflow (see Figure 3), the referral process was enhanced through automation of the review stages. Specialist and insurance approvals are now automated, eliminating unnecessary delays. Additionally, an optimized appointment scheduling system has been introduced, allowing patients to select preferred time slots via an online portal, with automatic notifications sent upon confirmation. The entire referral process is dynamically tracked, and automated prompts ensure that incomplete submissions are flagged and resolved in real time. This new system has resulted in a marked improvement in both process speed and patient satisfaction.

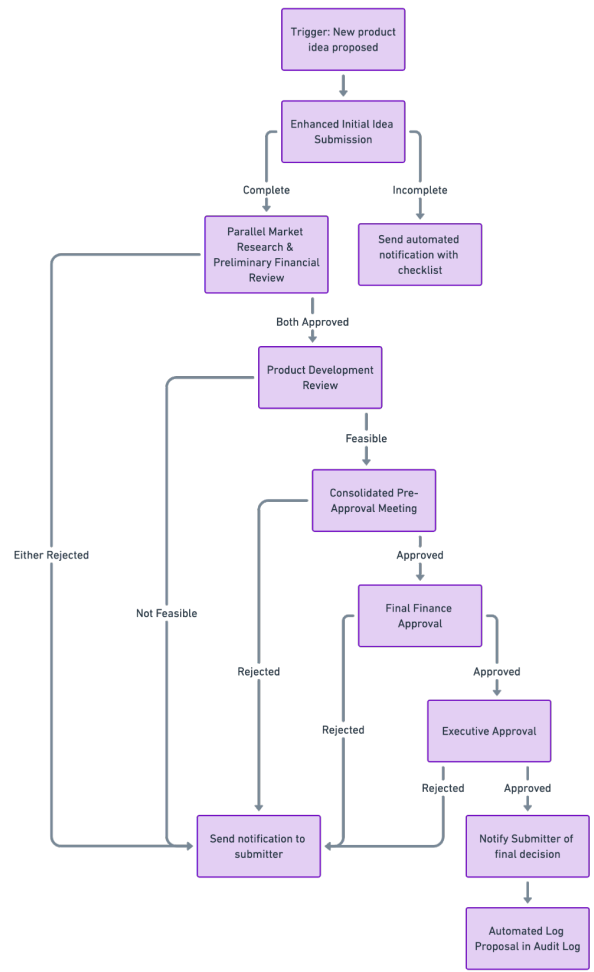
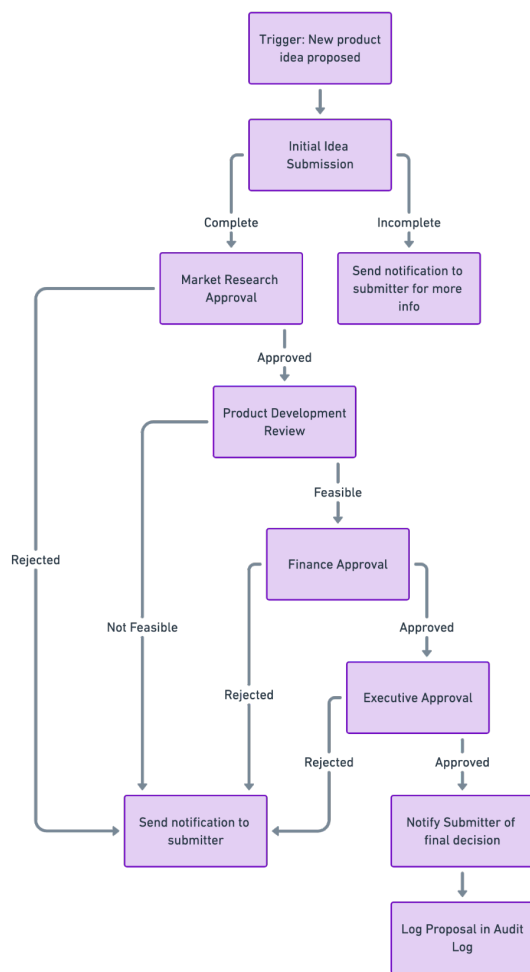
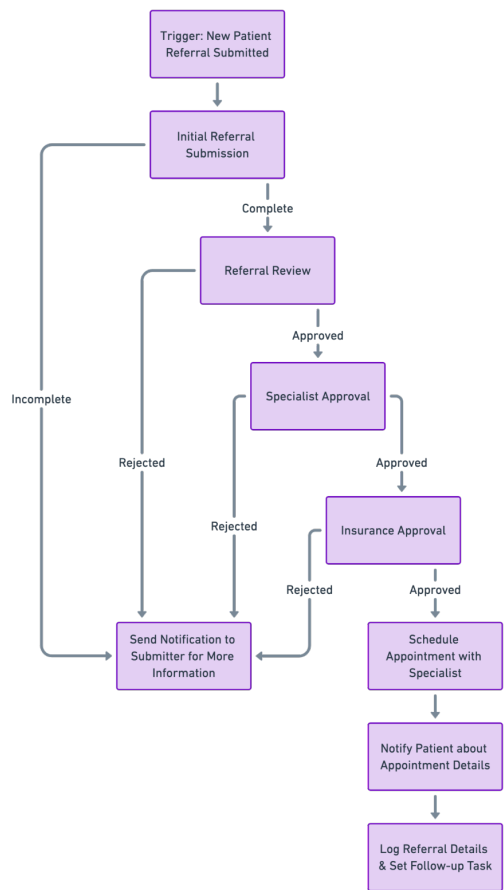
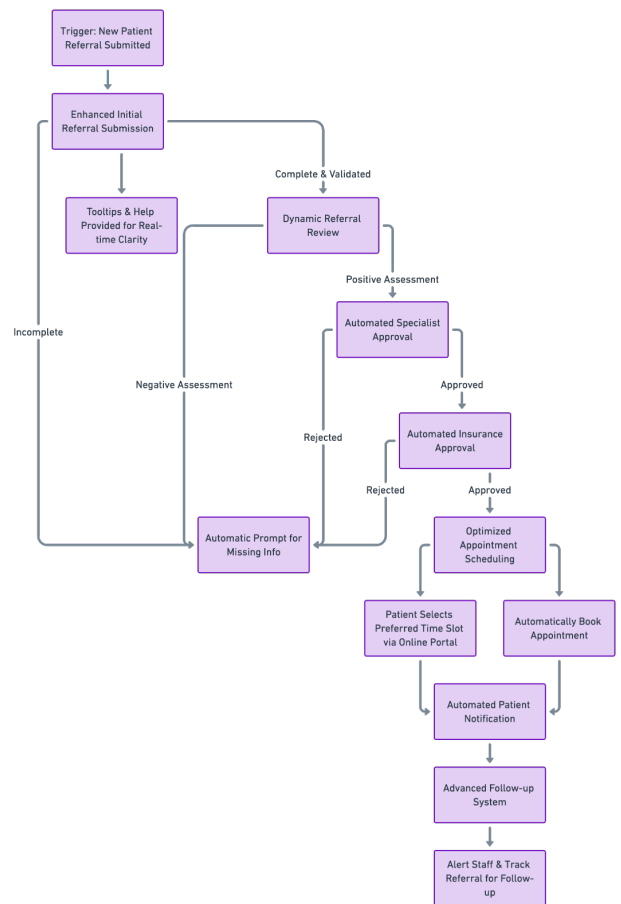


Figure 2. **Original Flow versus Optimized Flow - 1.** Product Idea Submission Workflow before and after optimization.

These visualizations illustrate the transformation of enterprise workflows through fluid flow optimization, showcasing the adaptability and efficiency brought about by intelligent, AI-based workflow agents.



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Figure 3. **Original Flow versus Optimized Flow - 2** Patient Referral Workflow before and after optimization.