

# Development and Seamless Integration of an On-premise AI Agent for Clinical Drafting: Insights from the Y-KNOT Project



Hanjae Kim, B.S.<sup>1</sup>; So-Yeon Lee, M.D., Ph.D.<sup>2,3</sup>; Seng Chan You, M.D., Ph.D.<sup>1,2,3</sup>

<sup>1</sup> Department of Biomedical Systems Informatics, Yonsei University College of Medicine, Seoul, South Korea

<sup>2</sup> PHI Digital Healthcare, Seoul, South Korea

<sup>3</sup> Institute for Innovation in Digital Healthcare, Yonsei University, Seoul, South Korea

## Brief Summary

Large Language Models (LLMs) have shown promise in reducing clinical documentation burden, yet their real-world implementation faces significant challenges, particularly in non-English speaking countries. Here we present Your-Knowledgeable Navigator of Treatment (Y-KNOT) project, the first successful implementation of an on-premise bilingual LLM-based artificial intelligence agent integrated with electronic health records for automated clinical drafting. We successfully deployed the service in a tertiary hospital in South Korea, which seamlessly aligns with existing clinical workflows. This demonstrates a practical framework for implementing LLM-based systems for other healthcare institutions, paving the way for broader adoption of LLM solutions.

## Introduction

### Background

- Clinical documentation significantly burdens healthcare providers,<sup>1,2</sup> and there is growing optimism about Large Language Models' (LLMs) potential to alleviate this burden.<sup>3</sup>
- However, implementing existing LLM solutions in South Korea presents unique challenges, such as data sovereignty issues and the complexity of bilingual documents.
- Moreover, due to their separate interfaces, manually retrieving information from Electronic Health Records (EHRs) and typing it into LLMs may ironically be time-consuming.

### Objectives

- To introduce Your-Knowledgeable Navigator of Treatment (Y-KNOT) project, aimed at developing a LLM-based artificial intelligence (AI) agent that seamlessly integrates a bilingual small LLM with EHR systems for automatic clinical drafting.

## Methods

From June through November 2024 at Severance Hospital (Seoul, South Korea), we carried out three parallel processes.

### Development of Medical Foundation LLM

Base architecture: Llama 3<sup>4</sup> (8B parameters)

First pre-training:

- general corpus (1.5 TB, Korean & English)

Second pre-training: (instruction pre-training<sup>5</sup>)

- medical corpus (90 GB, Korean & English)
- general corpus (9 GB, Korean & English)

### Clinical Co-development

- In collaboration with physicians, data scientists, software engineers, and medical record specialists, we conducted iterative cycles of aligning clinical requirements, hospital data availability, documentation standards, and technical feasibility.
- Through this process, the LLM was instruction-tuned for specific documentation tasks: emergency department (ED) discharge summaries and pre-anesthetic assessments.

### EHR Integration

We integrated the LLM into EHR by proceeding through three key components:

- Document standardization based on Fast Healthcare Interoperability Resource<sup>6</sup> (FHIR).
- Defining trigger points for the activation.
- Optimizing the user interaction framework for seamless clinical workflow incorporation.

## Results

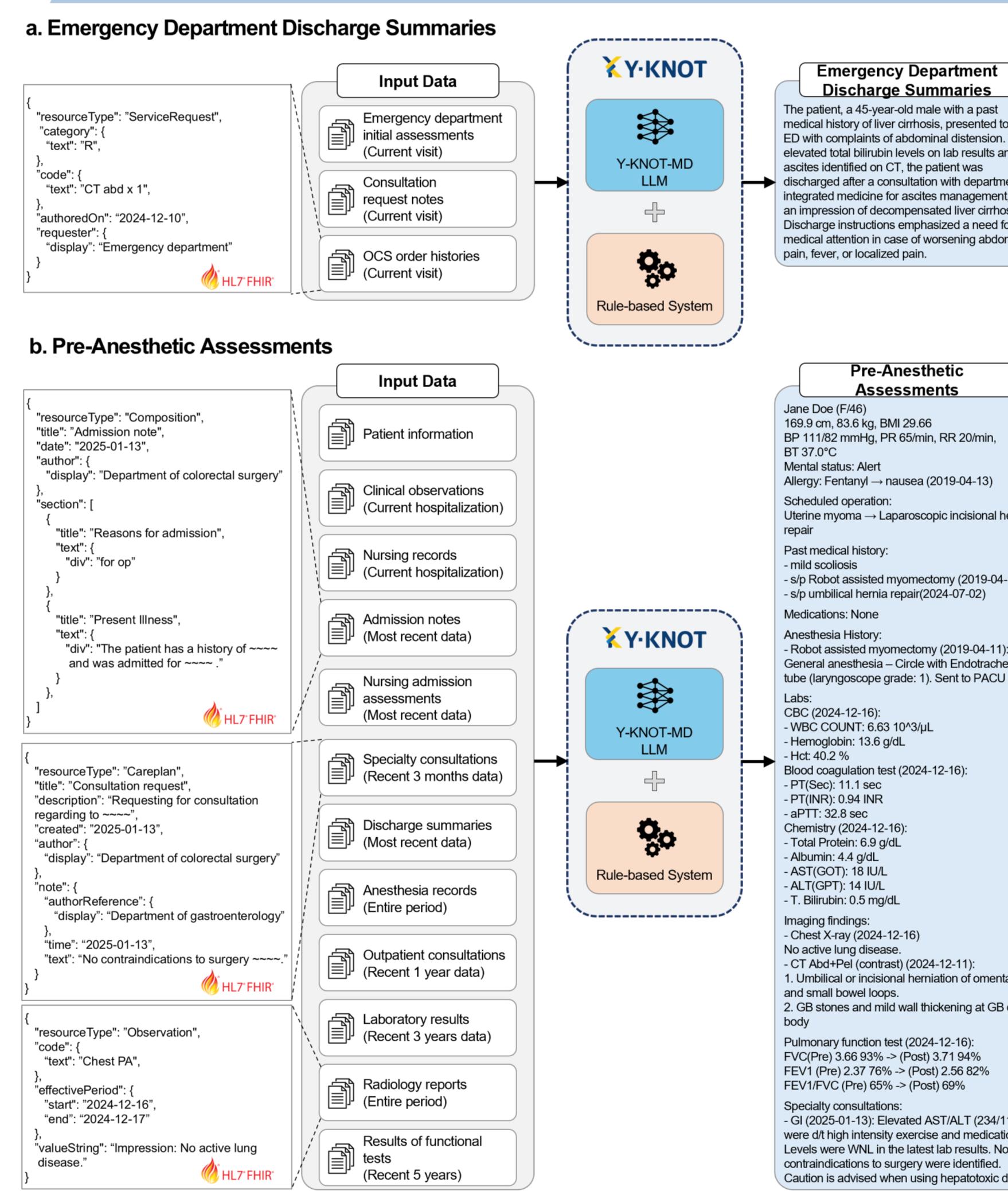


Figure 1. Drafting clinical notes with LLMs: from FHIR input to clinical output.

All medical records used as input data are converted into FHIR standards. Criteria for selecting input data are stated in parentheses. The examples provided in the figure are simplified versions of the actual data.

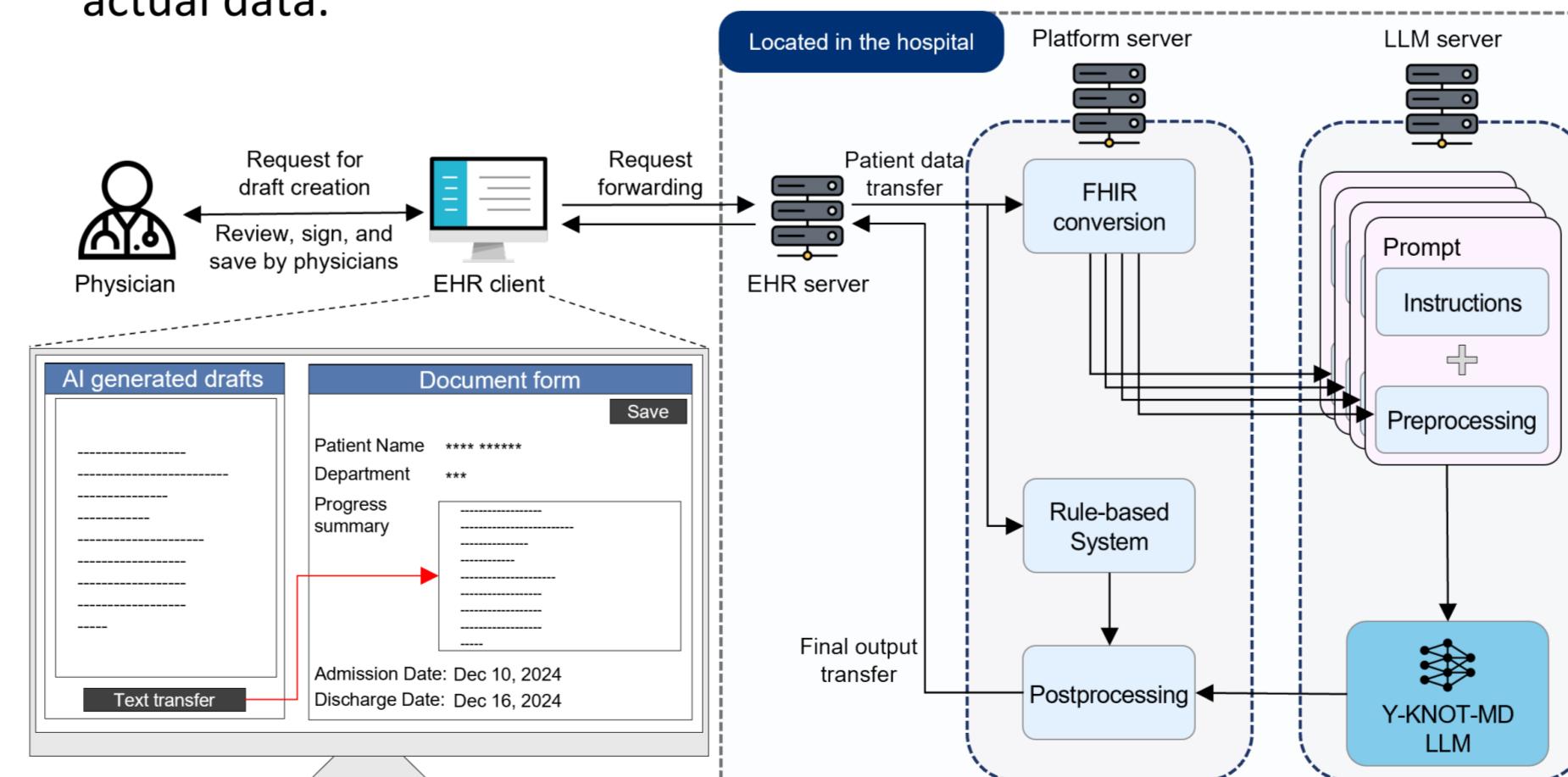


Figure 2. Overview of the automated drafting process with the AI agent in the EHR system.

When the drafting is initiated, patients' data is transferred from the EHR server to the Y-KNOT system. The data is converted into FHIR structures and processed in combination of LLM and rule-based approaches. Multiple prompts are created for the LLM, each designed to extract specific aspects of the document. Data exchange between system components operates through predefined APIs. All servers and databases are located within the hospital's secure on-premise environment.

## Contact information

oneash082498@yuhs.ac (Hanjae Kim) ; imipenem@phidigital.co.kr (So-Yeon Lee) ; chandryou@yuhs.ac (Seng Chan You)

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