# Converting Natural Language From Complex Procedural Documents To Structured Diagrams

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Introduction •0000

- Introduction



#### Motivation

Introduction

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- The volume of procedural documents is rapidly increasing across technical, legal, and organizational domains.
- Manually interpreting such documents is time-consuming, labor-intensive, and often requires domain expertise.
- Visual representations such as flowcharts significantly reduce cognitive load and aid procedural comprehension.
- Advances in large language models and retrieval techniques offer new potential for intelligent document understanding and visualization.
- There is currently no standard metric for evaluating the transformation from textual procedures to structured flowcharts.



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# **Objectives**

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- Develop a method to transform procedural documents into interactive flowcharts with support for clarification queries.
- Implement an evaluation framework and experiment to measure the effectiveness of AI in converting documents into flowcharts
- Deliver a system that is user-friendly and supports expert-guided refinement through a human-in-the-loop approach.

#### Natural Language Understanding

LLMs apply transformer-based attention to understand context, semantics, and procedural flow, allowing them to identify steps, conditions, and relationships in unstructured text.



### Actors & Features

#### Actors:

Introduction

- User (Document Provider): technical staff, analyst, or researcher who uploads documents and explores flowcharts.
- Al Engine (IUFlowGen): automated system that uses LLMs and retrieval to generate draft flowcharts and answer queries.
- Verifier (Expert): domain expert who reviews, validates, and optionally refines the Al-generated flowchart.

#### Features:

- Send Document performed by User, Verifier
- Plot and Query performed by Al Engine
- Verify and Feedback performed by Verifier
- Interact and Ask performed by User, Verifier



#### Contributions

Introduction

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#### Scientific Contribution

Formalize a modular Al pipeline that combines retrieval, prompting, and DOT visualization, and propose a metric to evaluate the effectiveness of transforming complex documents into structured flowcharts.

#### Real-world Contribution

Aid users in understanding procedural documents through interactive visual flowcharts, applicable in education, healthcare, technical documentation, and many more.



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## Advancements

- GenFlowchart applies generative AI to understand and segment flowchart diagrams.
- Lester et al. introduce multi-phase summarization for long regulatory documents.
- ChatGPT identifies procedural steps and actors using advanced NLU capabilities.
- Eraser Flowchart Maker supports text-to-diagram generation with intuitive UI



## Research Gaps

#### Research gap 1

Current Al systems lack interactive flowchart generation that preserves procedural logic and structure from unstructured documents.

#### Research gap 2

The combination of Al-guided generation with graph and vector representations for storing and retrieving steps, entities, and relationships is still under-discovered in procedural document understanding.

⇒ A new approach that builds graph-augmented prompts and renders interpretable flowcharts locally can fill this gap.



### Technical difficulties

- The amount of background knowledge required to combine graph theory, vector search, and prompt engineering is substantial
- LLMs are probabilistic, which introduces unpredictability in output consistency, especially for structured generation tasks.
- Mapping Al-generated content into structured and verifiable DOT code demands careful post-processing and validation.



- 3 Proposed Method



# System overview

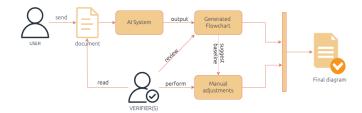
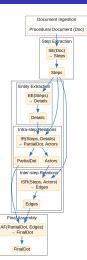


Figure: System overview of UIFLowGen Protocol



### Process - Flowchart Generation Algorithm

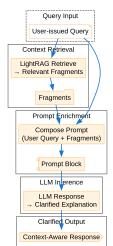


- **Goal:** convert procedural documents into structured flowcharts via automated step/entity extraction and reasoning.
- **Result:** generate a valid FinalDot representation used for rendering interactive flowcharts
- **Step:** extract steps → enrich with entities ightarrow infer relations ightarrow assemble graph.
- Scope: execute locally with LLMs, LightRAG, regex post-processing, ensuring offline reproducibility.



# Process - Query Algorithm





- **Goal:** support users and verifiers in exploring or confirming specific flowchart elements or logic.
- **Result:** generate a context-aware explanation or inference tied to gueried nodes or actions.
- **Step:** receive query → retrieve relevant content  $\rightarrow$  compose enriched prompt  $\rightarrow$ return response.
- **Scope:** acts as an interactive support layer during review.



- Result



# Prototyping

#### GitHub repository: https://github.com/Khim3/IUFlowGen



Figure: Generated flowchart from document.



Figure: Clarification query interface



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# Benchmark Results



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# Analysis

- The system provides high interpretability and transparency by visualizing extracted procedural logic in a human-readable form.
- The interactive human-in-the-loop model empowers domain experts to validate and refine Al-generated flowcharts.
- IUFlowGen adapts well to varying document complexity, demonstrating strong flexibility across synthetic and real-world texts.
- Designed to operate on local infrastructure, the system performs reliably in resource-constrained or offline environments.



#### Limitations

- Entity extraction and semantic embedding are resource-intensive, causing latency on low-power machines.
- LLM-based generation is non-deterministic, leading to occasional inconsistencies that require human correction.
- The system supports only single-user, sequential document processing, limiting scalability in multi-user scenarios.
- Performance depends on the clarity and structure of the input document, especially for domain-specific texts.
- ⇒ Despite these limitations, IUFlowGen establishes a reliable base for Al-assisted procedural understanding in secure, offline contexts.



- 6 Conclusion



# Summary

- Built for many applications: IUFlowGen enables automated, interactive flowchart generation from procedural text, suitable for use in education, healthcare, compliance, and technical documentation, etc.
- Ready for practical adoption: Designed for offline deployment, IUFlowGen is accessible to end users. organizations, and domain experts needing secure document visualization tools



## **Future Works**

- Improve UI/UX design with real-time feedback, interactive previews, and adjustable layout controls for better user experience.
- Enable direct flowchart editing, including drag-and-drop, node repositioning, and in-place corrections.
- Package the system for web or mobile deployment to improve accessibility in education, legal, and enterprise settings.



#### Demonstration

#### Process A

Visualize the flowchart from an uploaded procedural document.

#### Scenario 1

Successful generation and rendering of a valid procedural flowchart.

#### Scenario 2

Clarification query triggered on a node to retrieve related explanation.

#### Process B

Perform a query from the generated chart to aid user understanding.





