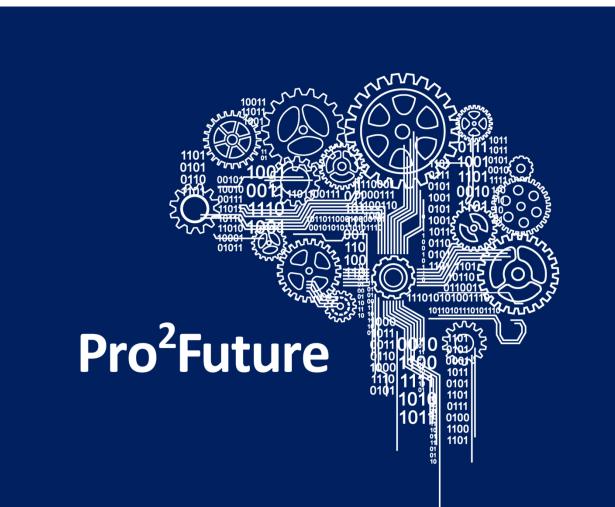
TrustinLLM

Trustworthy Digital Systems Assisted by Large Language Models (LLMs)



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MOTIVATION & GOALS

As digital infrastructures increasingly incorporate Artificial Intelligence, particularly Large Language Models (LLMs), ensuring their trustworthiness is essential. This project seeks to establish a foundational framework to strengthen trust in LLM-assisted systems by focusing on Transparency, Reliability, and Safety. We propose novel methodologies that enable the generation of factually accurate responses by explicitly integrating context from verified and reliable sources. These approaches will address both system-level attributes (e.g., architecture, robustness) and data-level qualities (e.g., integrity, bias), contributing to the development of AI systems that are dependable, transparent, and ethically aligned.

Project FactBox

Project Name TrustinLLMProject ID FFG 915295Duration 36 Months

Area 3
Area Analytics

Project LeadDI Dr. Belgin Mutlu

APPROACH

Retrieval-Augmented Generation (RAG) is a promising approach to mitigate hallucinations in LLMs by grounding responses in source documents provided by users. This enables a fast and secure mechanism for validating the correctness of responses for users. This project enhances RAG through improved pre- & post-retrieval processing, transparency and traceability measures, and UI-based interpretability features.

CONTRIBUTION

Scientific contribution

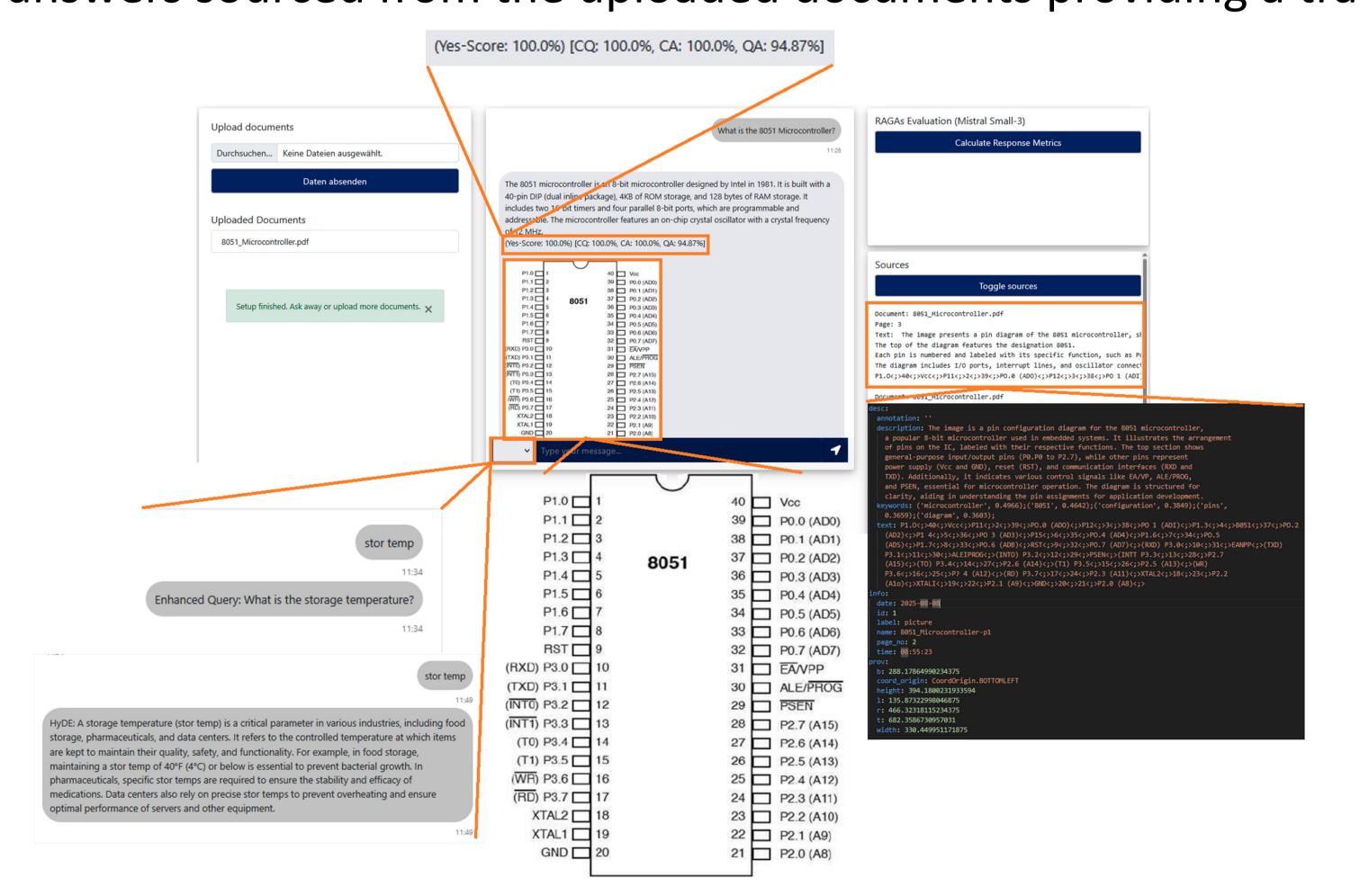
- Increasing transparency capabilities of LLM-based RAG systems
- Increased trustworthiness for AI-based systems
- Data provenance and traceability solutions to tackle rising data obfuscation
- Mitigation of hallucinations and deceiving responses from LLM inference

Economic contribution

- Lower risk of costly errors through reliable AI outputs
- Higher productivity via reduced manual oversight
- Faster adoption due to increased user trust
- Improved compliance with safety and transparency
- Competitive advantage through trustworthy AI offerings

SYSTEM ARCHITECTURE

Documents are uploaded onto a locally hosted RAG-Framework, pre-processed to extract **multimodal embedded items** and their textual content chunked into a vector store. Users can then ask questions, and the LLM-based system generates answers sourced from the uploaded documents providing a tracible way of information generation.



Chunking Generate Extract Text Document **Embeddings** Vector Extract Images 8 **Docling** Generate Descriptions **Detect & Extract YOLO** (Pin) Diagrams Pre-Retrieval: Post-Retrieval: HyDE, Summarization, LLM Context + Promp Generate Response **Embeddings** Answer Evaluation Loop

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