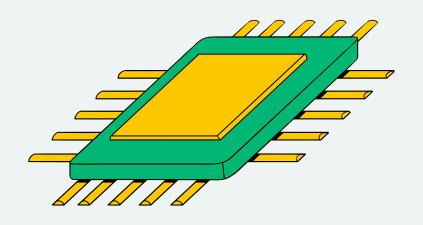


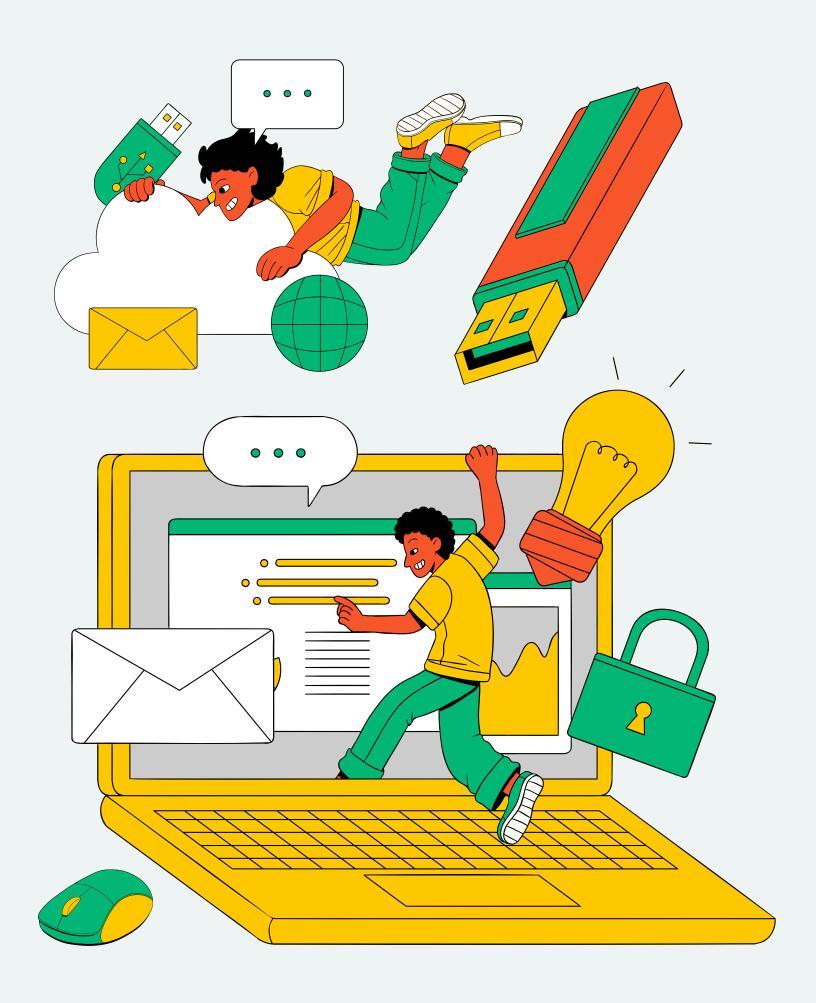
MULTI-AGENT SYSTEM FOR QUESTION ANSWERING

PRESENTATION

PRÉSENTÉ PAR:

YASSINE ZITOUNI



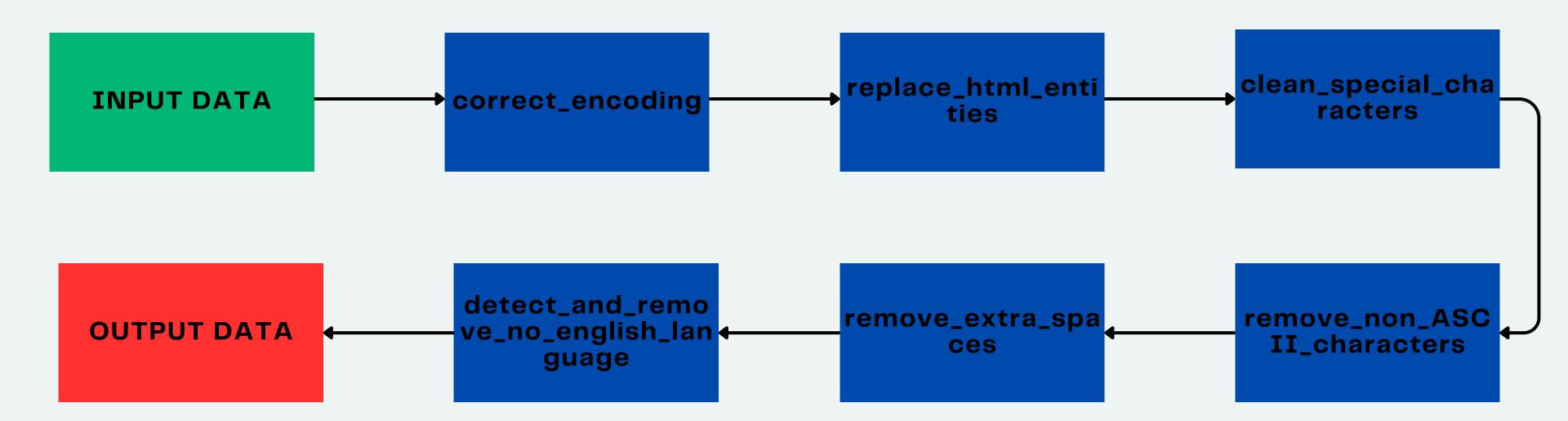


PRESENTATION OUTLINE

- Data Exploration & Preprocessing
- High-Performance Benchmark: GPT-4.5
 with External Reasoning Agents
- Comparative Study of Direct Reasoning
- Fine-Tuning a Lightweight Model: T5 in Action
- Proposed Enhancements for Low-Cost
 Performance Boost

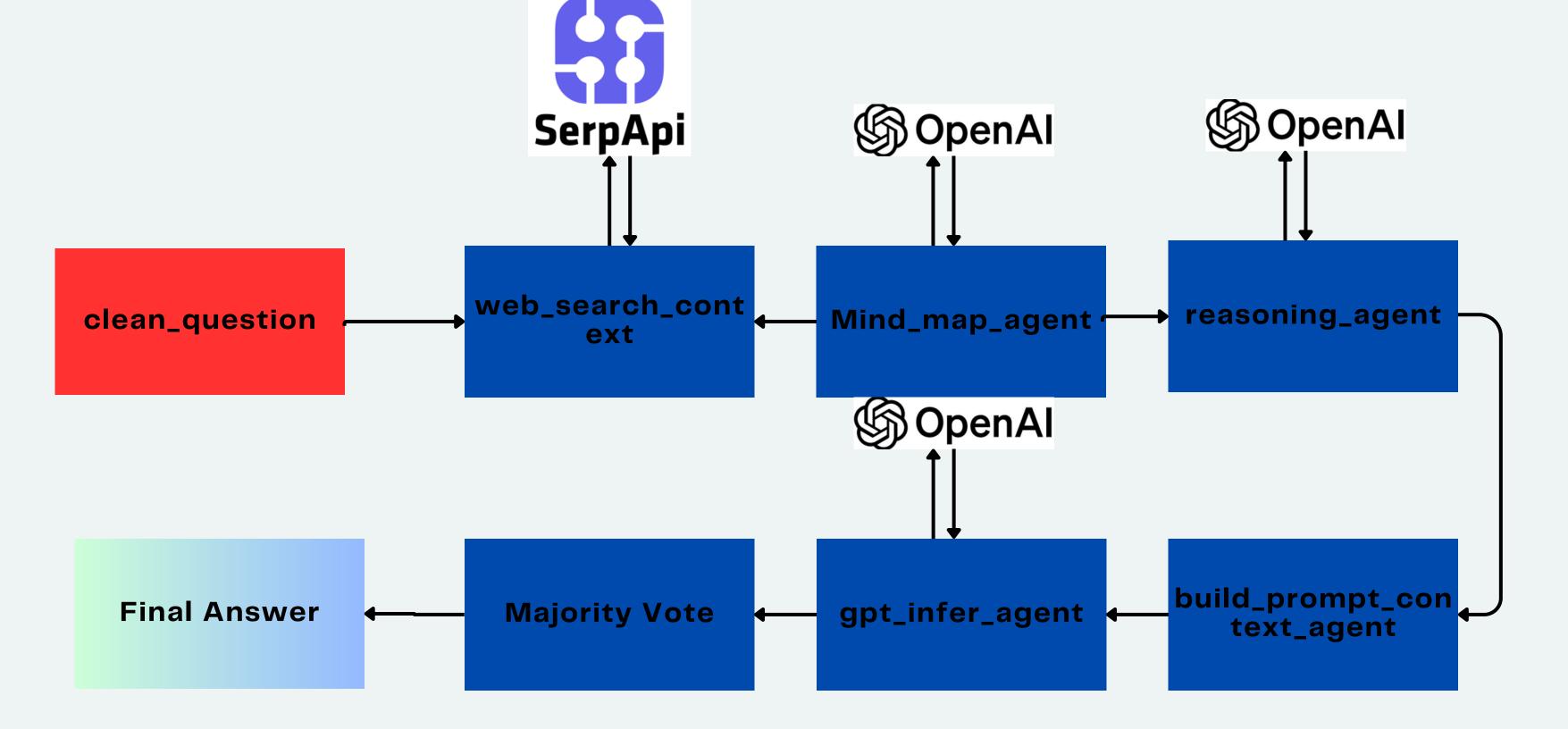


INTRODUCTION



HIGH-PERFORMANCE BENCHMARK: GPT-4.5 WITH EHTERNAL REASONING AGENTS

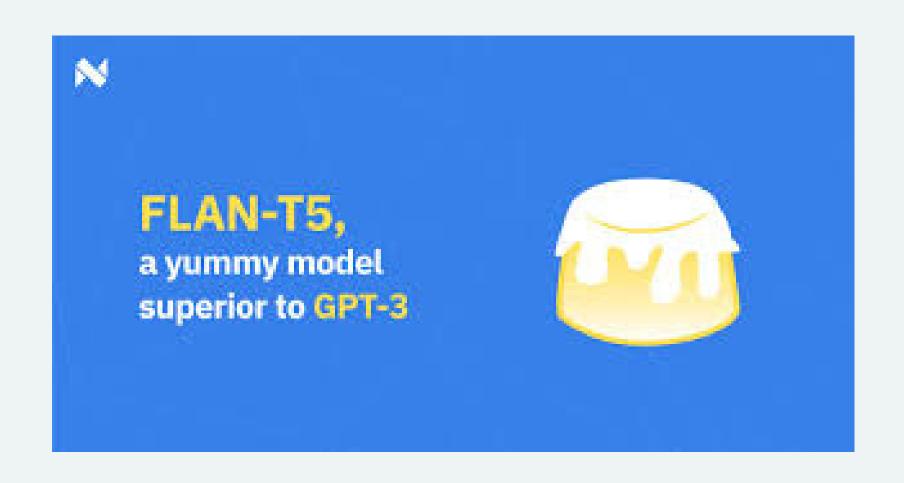
Model Name	Public score	Private score	Test Data Accuracy	
GPT4.5	0.75	0.75	0.75	



. COMPARATIVE STUDY OF DIRECT REASONING

Model Name	Train Data Accuracy	racy Public score Private score		Test Data Accuracy	
GPT-3.5-TURBO	0.44	0.75	0.4333	0.5915	
GPT-4	0.48	0.5833	0.4833	0.533	
GPT-4-TURBO	0.48	0.600	0.51666	0.558	
GPT-40-MINI	0.44	0.5888	0.5666	0.5777	
GPT-40	0.48	0.533	0.588	0.56	
Ol	0.48	0.488	0.588	0.538	

FINE-TUNING A LIGHTWEIGHT MODEL: TS IN ACTION



input data

construcut few-shot prompt

tokenisation

Fine-tuning with Seq2SeqTrainer

generate prediction

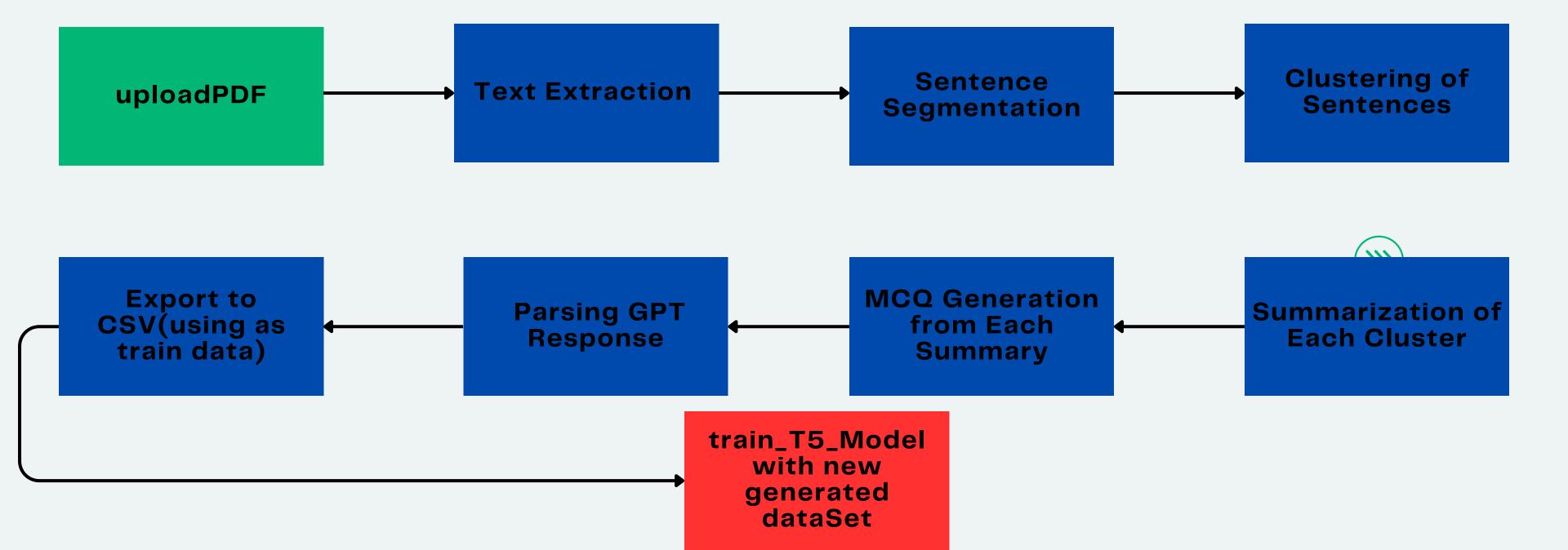


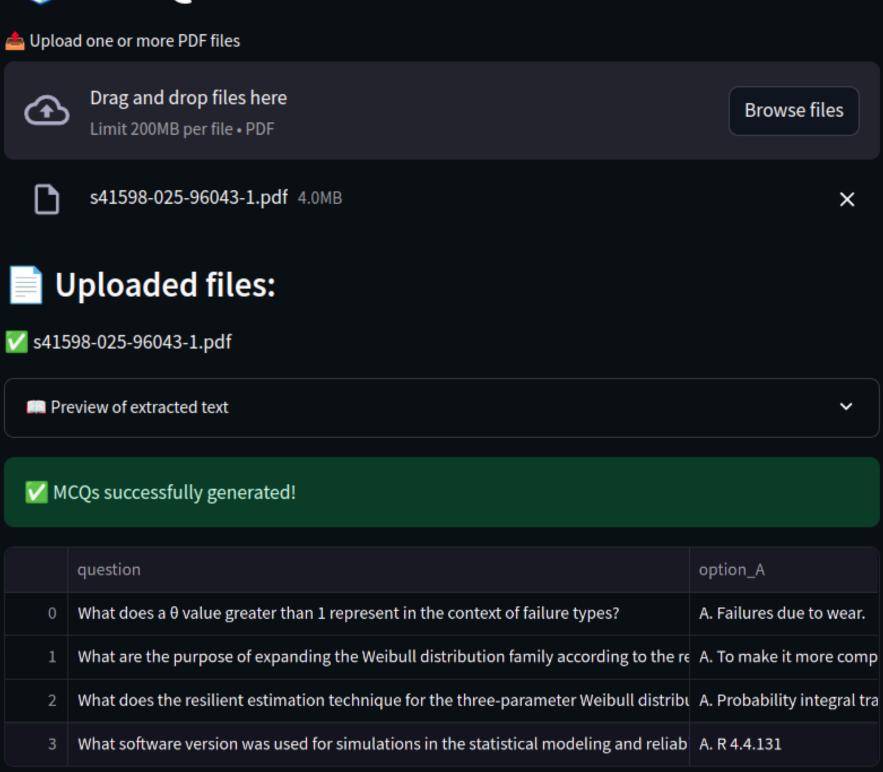
output

Modele	public score	private score	learning rate	epochs	zero shot	one shot	few shot
google/flan-t5-small	0,4166	0,5833	3,00E-05	30	no	no	yes

PROPOSED ENHANCEMENTS FOR LOW-COST PERFORMANCE BOOST







📥 Download MCQs in CSV format



Agentic Reasoning: Reasoning LLMs with Tools for Deep Research

LEVERAGING LARGE LANGUAGE MODELS FOR MULTIPLE CHOICE QUESTION ANSWERING Answering https://arxiv.org/pdf/2210.12353