



AIPI Chatbot – Fine-tuned Mistral 7B & CRAG



- Mrinoy, Leslie, Kahlia, Samyukta



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Motivation



Motivation



- High speed search and scalability
- Better integration
- Easy Hosting



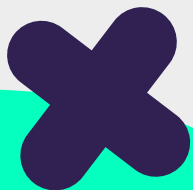
- High Benchmark results
- Efficient Architecture
- Relatively lightweight



- **CRAG** based pipeline hooked with tavily search
- Directs questions with low match to web search
- Enhances bots ability to handle out-of-context questions.



02



Dataset



Data

1. AIPI FAQ Document
2. Duke AIPI Web domain
 - <https://ai.meng.duke.edu/>
 - All subdomains attached to this master domain
3. Syllabus Information for AIPI courses
 - Excluding certain ones we did not have access to, i.e. AIPI 560



Data Preprocessing

Data Selection and Pruning

Editing files to remove:

- Information on Duke Medx, PhDs
- Remove old content that is irrelevant.
- Irrelevant headers
- Adding document context (syllabus)

Vectorization and Tokenization

WhereIsAI/UAE-Large-V1 tokenizer –

- Model Size (Million Parameters): 335
- Retrieval Average: 54.66
- Summarization Average: 32.03

Chunking Strategy

- Data Integrity and Leakage Prevention
- Chunk Overlap – 50
- Chunk Size – 500 tokens maximum
- Create embeddings solely from within same documents



Pinecone



03



Fine Tuning





Databricks Dolly 15K

01

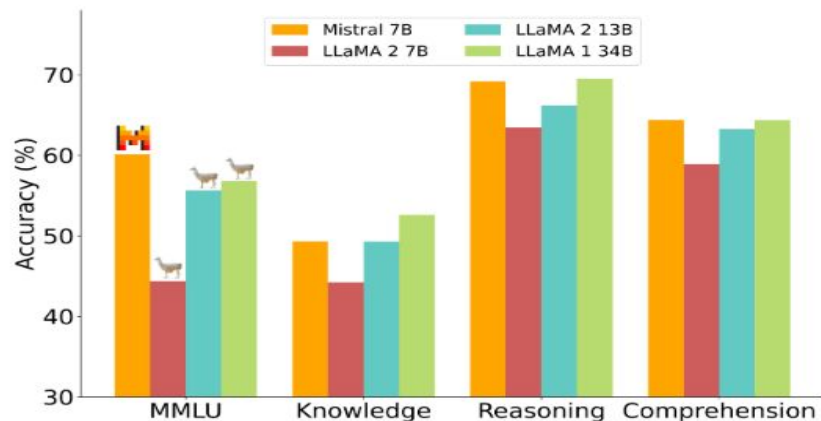
15,000 human-generated instruction corpus specifically designed for training conversational AI

02

Can be used, modified, and extended for any purpose, including academic or commercial applications

Training-test split: 80-20

Mistral-7B-v0.1



Efficient architecture with Grouped Query Attention (GQA) and Sliding Window Attention (SWA)

01



Model Configurations

01

ChatML format

- No Need for Instruct Tags
- Enhanced Format Clarity
- Optimized for RAG

02

BitsAndBytes Quantized

- Models are loaded in 4-bit precision to decrease memory usage.
- Utilizes 'torch.bfloat16' for computing, balancing performance and precision.

03

LoRA

Applies to all linear layers, enhancing the model's ability to adapt.

04

Flash Attention

- Reduced Memory Footprint
- Increased Computational Speed
- Scalability

05

Additional

- Optimized for 1 GPU - Training & Inference
- Optimizer – Adamw_bnb_8bit





Results

Training

- AWS g5.16xlarge
- GPU: Nvidia A10 (24GB)
- All experiments: 24 hrs
- Final model: 103 mins

Training Costs

- All experiments: ~\$100
- Final model: ~\$7

Evaluation

- 1000 random samples
- LLM as Judge – the LLaMa2-7B model serves as the standard, comparing the outputs generated by the Mistral bot against established ground truth data.
- Accuracy – 82.7%

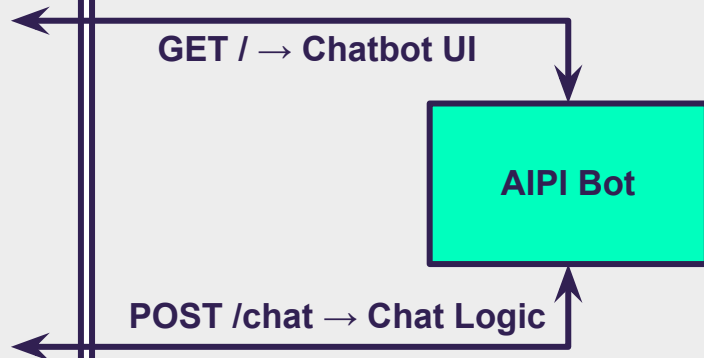
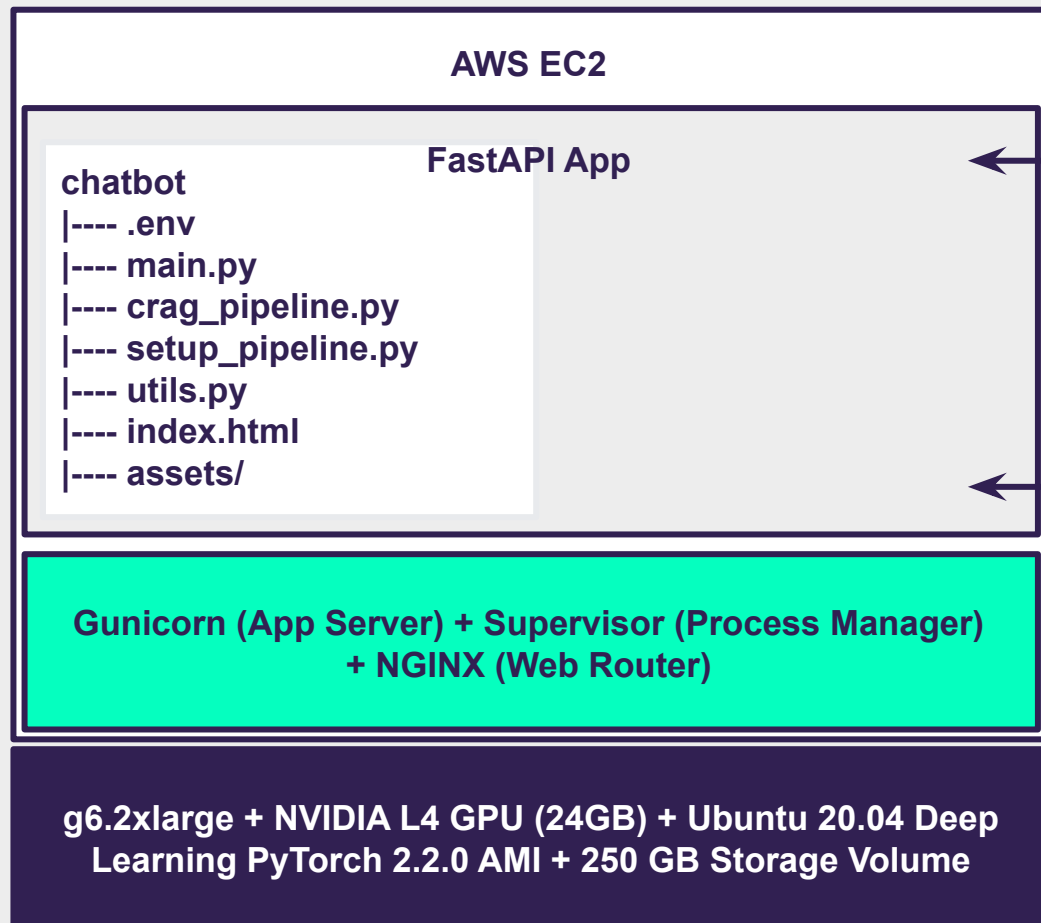


04



System Architecture

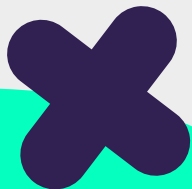




Cost: \$0.978/hr



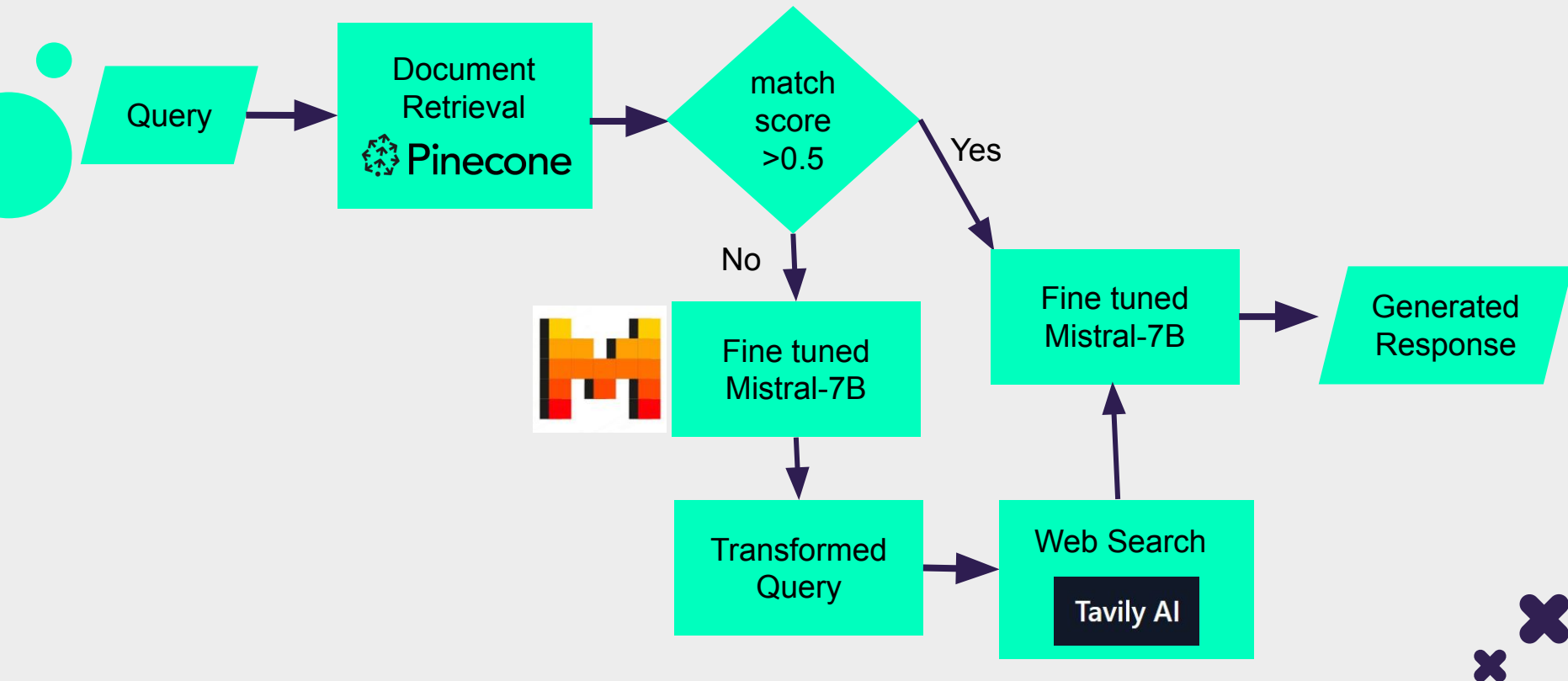
05



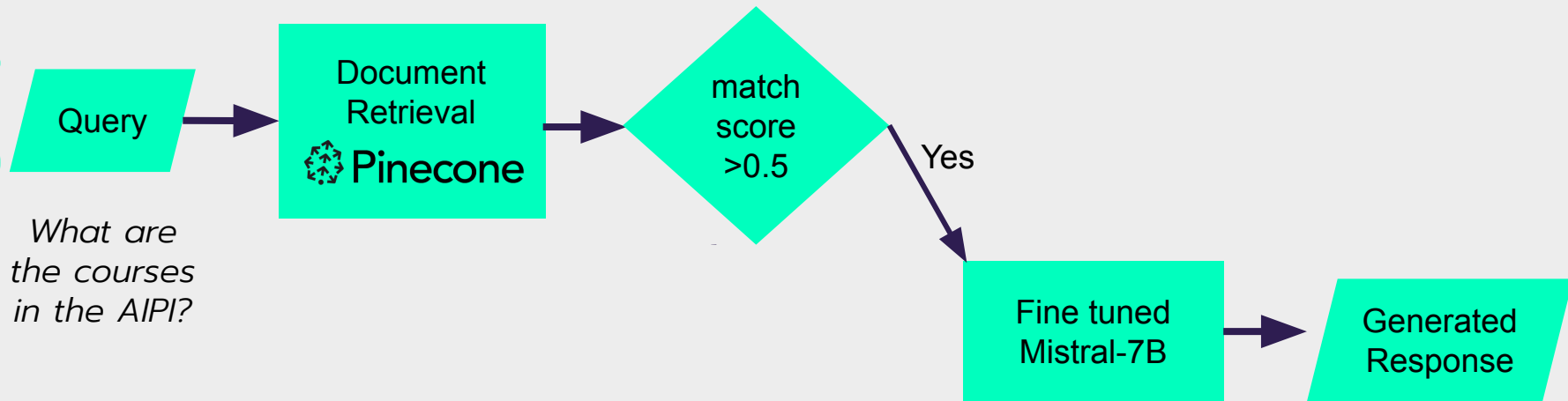
Inference



cRAG Pipeline



cRAG Pipeline



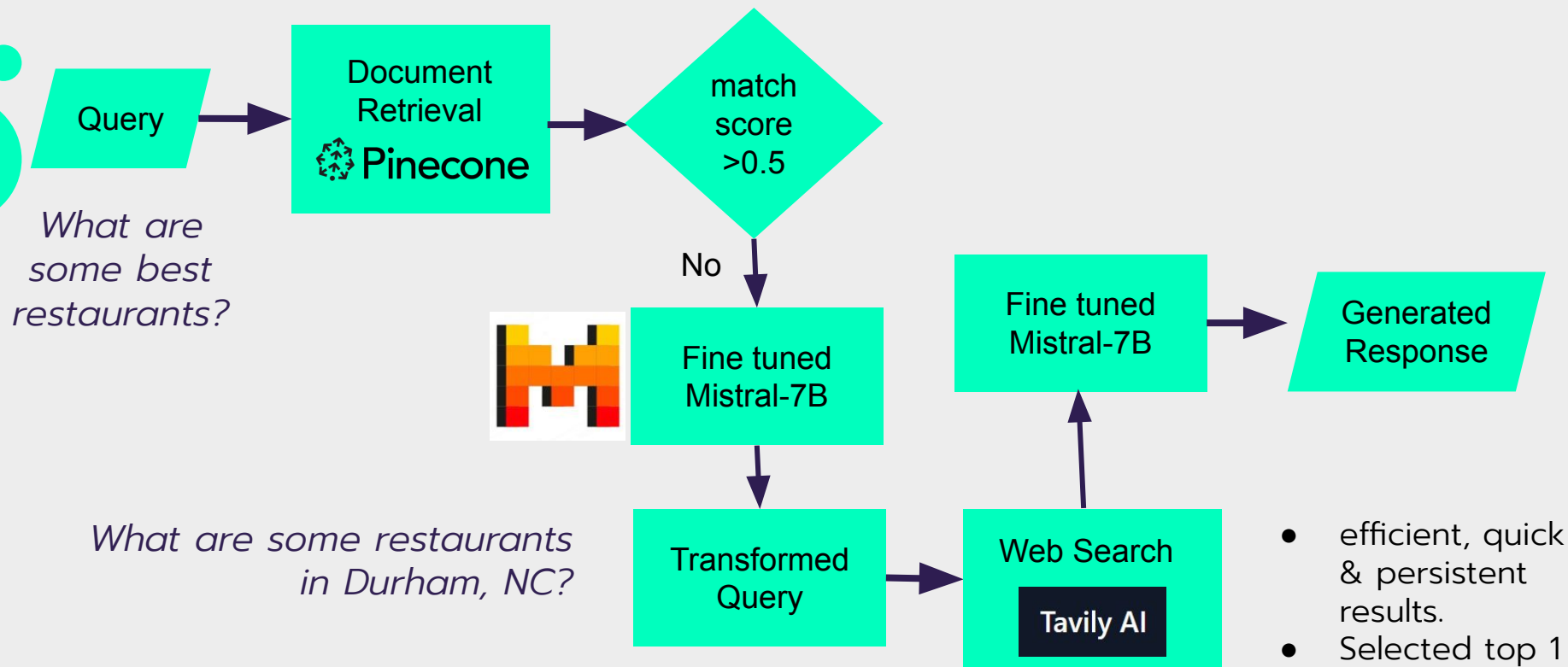
Key Points

- Pinecone: Easy to Host
- Token size = 600
- Top 2 queries were selected

Cost

- 10k queries/month
- 2k writes/month
- 10k vectors
- \$0.45/month

cRAG Pipeline



06



Evaluation



Results



Q/A

Domain Questions

- 27 catered questions to Duke AIPI
- Subset of data about broad Duke information



User Observations

- Response is based on content retrieval from Pinecone & web search
- AIPI specific questions perform better without hitting the web search
- Infrequent yet present hallucination



Evaluation

- Human-as-a-Judge
- Rated on 1-5 scale
- Judged on information correctness & response formatting
- **AIPI Questions: 2.95**
- **Duke Questions: 2.2**



07



Cost






Pricing breakdown

Experimentation:	~\$100 total
Model Fine-tuning:	~\$7 total
AWS Deployment:	~\$700/month
Pinecone DB:	~\$0.45/month

TOTAL	~\$8455/year
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Thank
you!

