

Keeping it Small: Agentic Workflows with SLMs on K8S

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01 Address challenges of Agentic workload

02 Multi-agent workflows

03 Implementation on k8s

04 Key takeaways



Part 01

Address challenges of Agentic workload



What is a Gen AI Agent



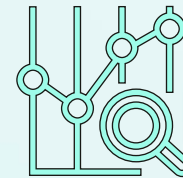
Intelligent,
autonomous
systems



Access to
enterprise
data



Plan,
reason,
and act



Ability to
use tools

AI

... it leads to challenges

Complexity

Coding gets complicated



- Complex prompts to limit hallucinations
- Fragile, hard to maintain

Accuracy

Agent gets confused



- Calling wrong tools
- Passing wrong arguments
- Inconsistent responses

Cost

Agent gets slower and more expensive



- Frontier models needed
- Prompt sizes grow
- Agents retry steps

Hints to improve cost, performance and accuracy

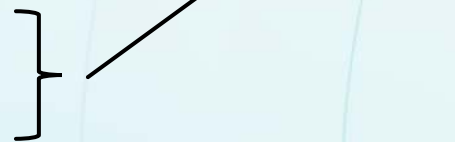
- Shorter prompts
- Smaller LLMs



Task Decomposition
Technique



- Control over workflow
- Concise prompts / context



- Cost effective chipsets
- Fast Chipsets



AI Chips Choice

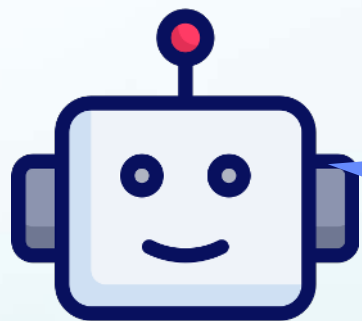


Part 02

Multi-agent workflows

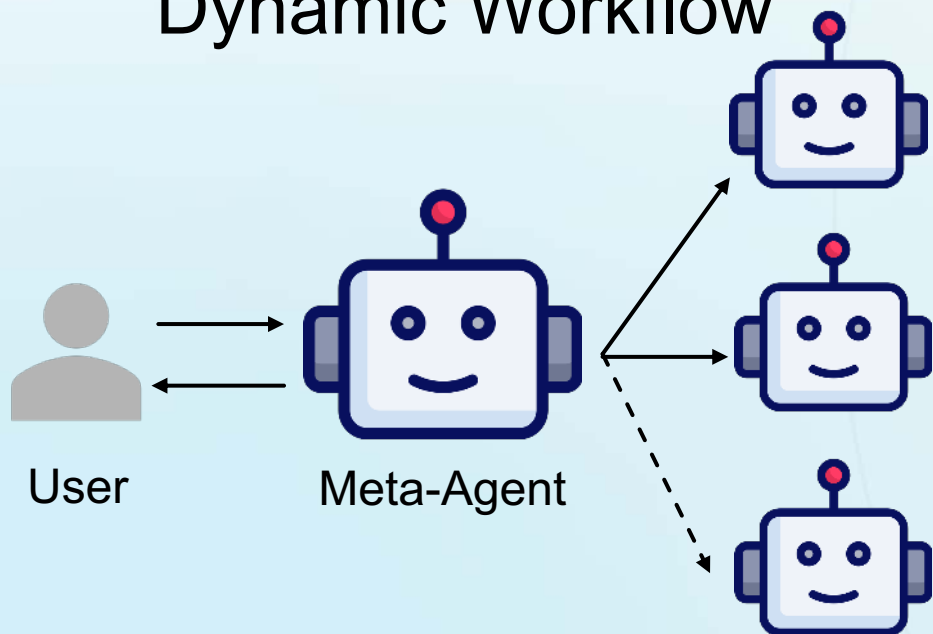


Agent & workflows

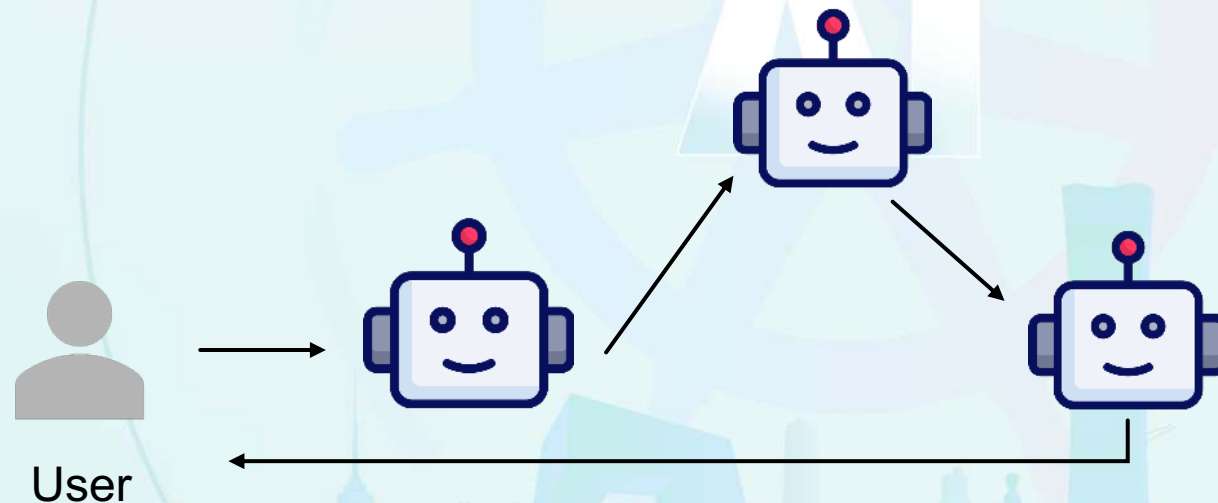


I am an “agent” specializing in a task or just to coordinate (again specialized!)

Dynamic Workflow



Static Workflow



Why Small Language Model?



Resource



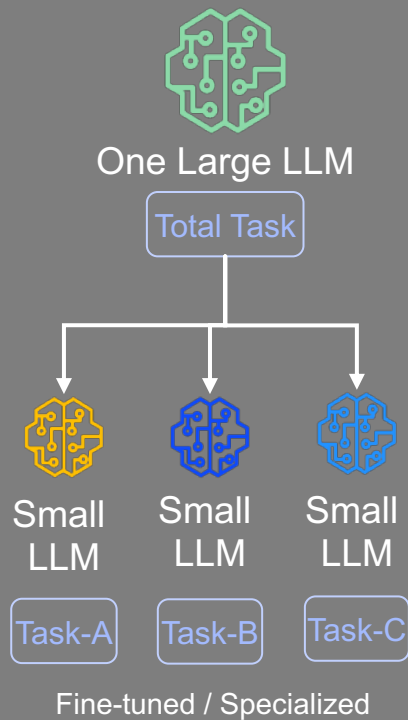
Speed



Customize

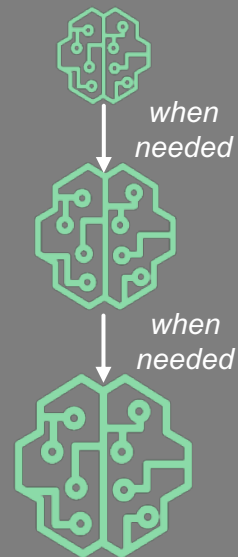
Frugal Architecture Design Patterns

Task Decomposition

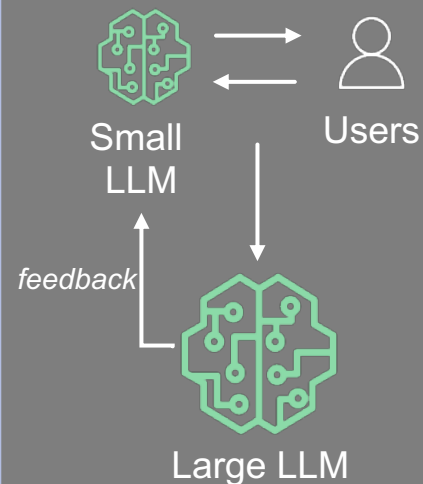


Using Small and Large LLMs Frugally

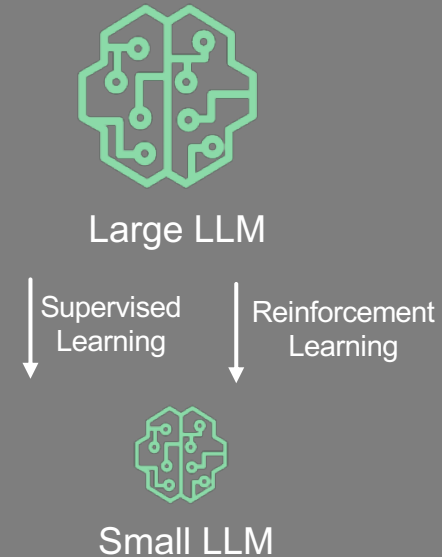
Cascaded LLM



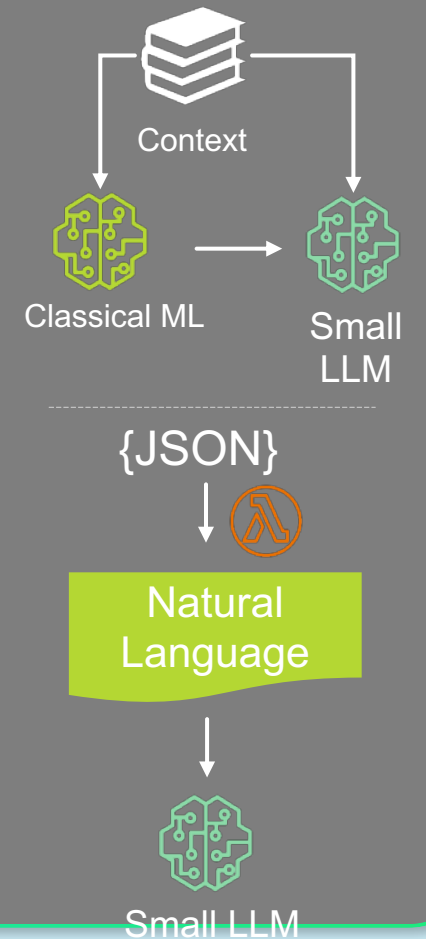
RLAIF with Large LLMs



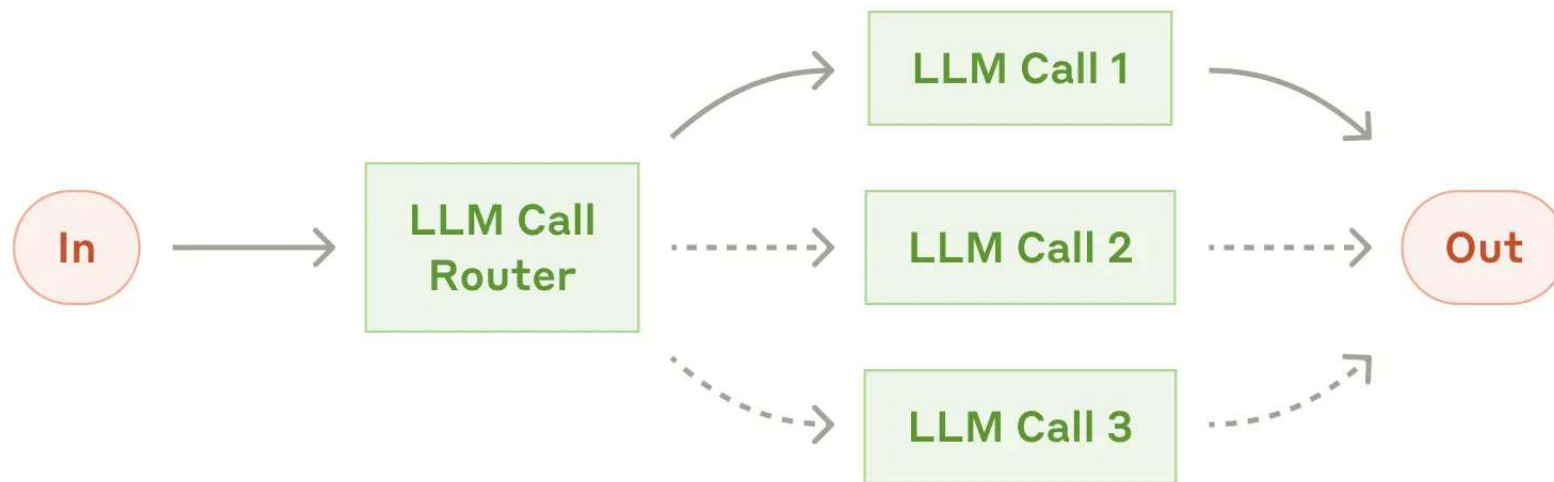
Teacher-Student LLM



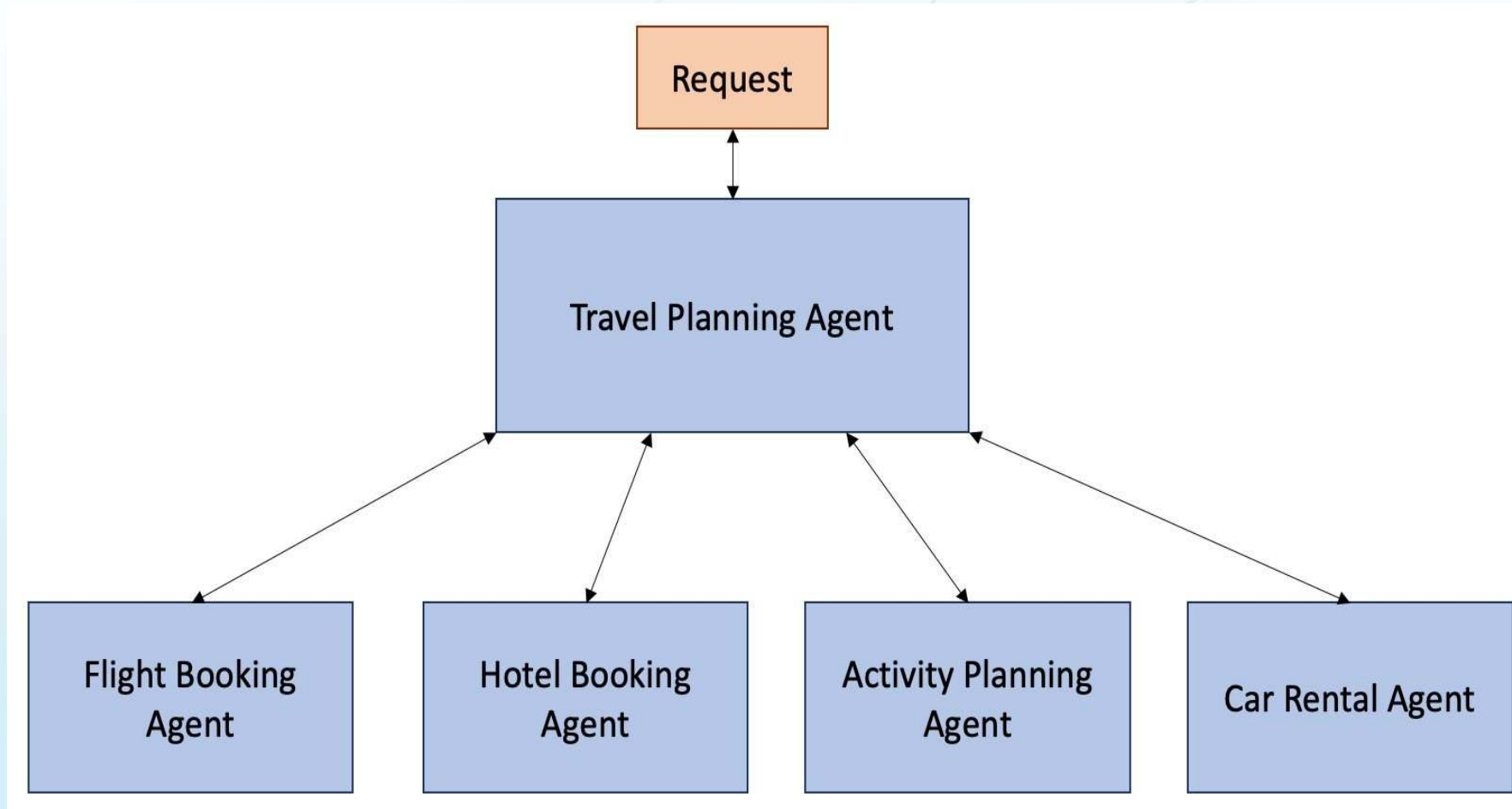
Pre-post processing for Task Simplification



Workflow : Routing



Example – Routing



Part 03

Implementation on K8S



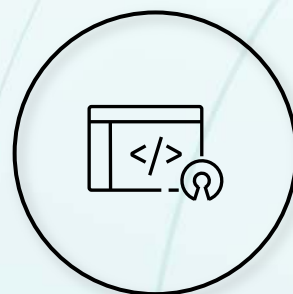
Why self-host Language Model on K8S



Data privacy
and security



Connecting
to
data sources



Customizing



Accessing
multiple
models
and newer
versions

Running Agentic workload on K8S



 **LangGraph**

 **Amazon Bedrock Agents**

AI Gateway

Observability

Runtime


 **CUDA Runtime**


 **Ray**


 **vLLM**

 **Neuron Runtime**

Compute

 **NVIDIA GPUs**

 **Inferentia**

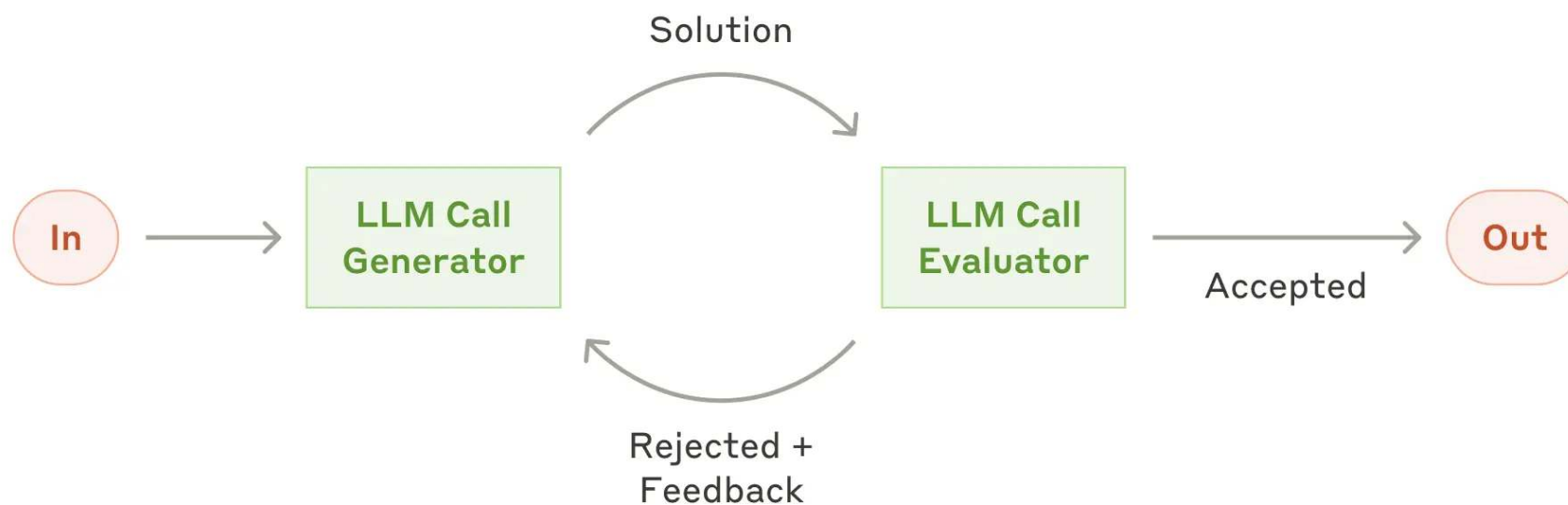
 **Trainium**

Graviton

 **Karpenter**



Workflow: Evaluator-optimizer



LiteLLM



Observability - Langfuse



Langfuse v3.2.0

View-only [demo project](#)

Dashboard

Tracing >

Evaluation Beta >

Users

Prompts

> Playground

Datasets

Feedback

PROD-EU Langfuse Demo / langfuse-docs

Dashboard

Dec 10, 24 : 00:01 - Dec 17, 24 : 00:01 7 days

Request Chart

Traces

288 Total traces tracked

qa275

dataset-run-item-cm4lr7

dataset-run-item-cm4hn6

Model costs

\$0.090574 Total cost ⓘ

Model	Tokens	USD
gpt-4o-mini	218.05K	\$0.045409
gpt-4o	5.9K	\$0.044935
text-embedding-ada-002	2.3K	\$0.00023

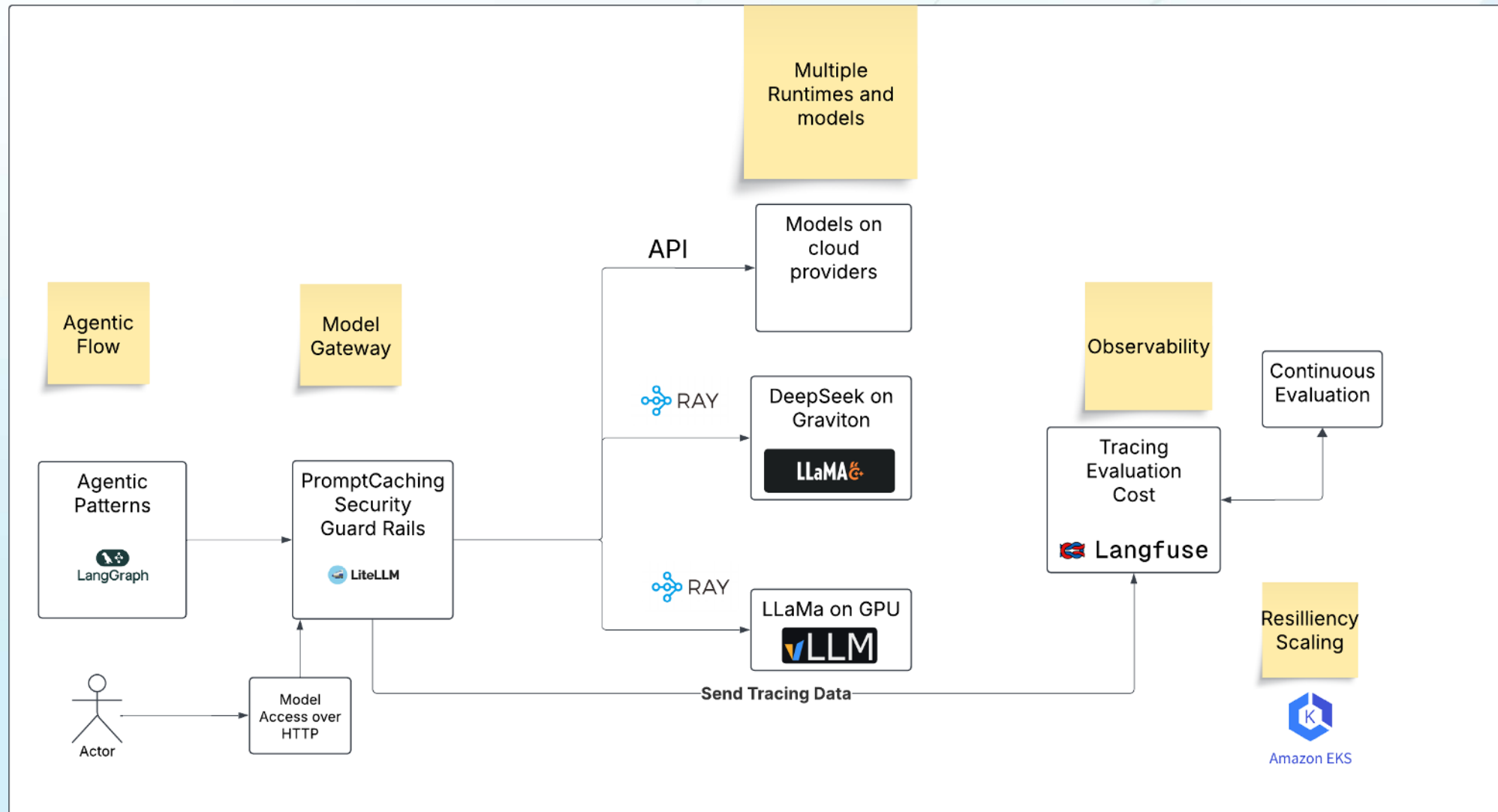
Scores

1.93K Total scores tracked

Name	#	Avg	0	1
# hallucination (eval)	388	0.17	252	28
# language-detector (eval)	370	0.89	42	328
# conciseness-v1 (eval)	361	0.46	57	63
# contextrelevance (eval)	347	0.34	194	84
# toxicity-v2 (eval)	335	0	335	0

Show all

Implementation on K8S



Part 04

Key takeaways

AI

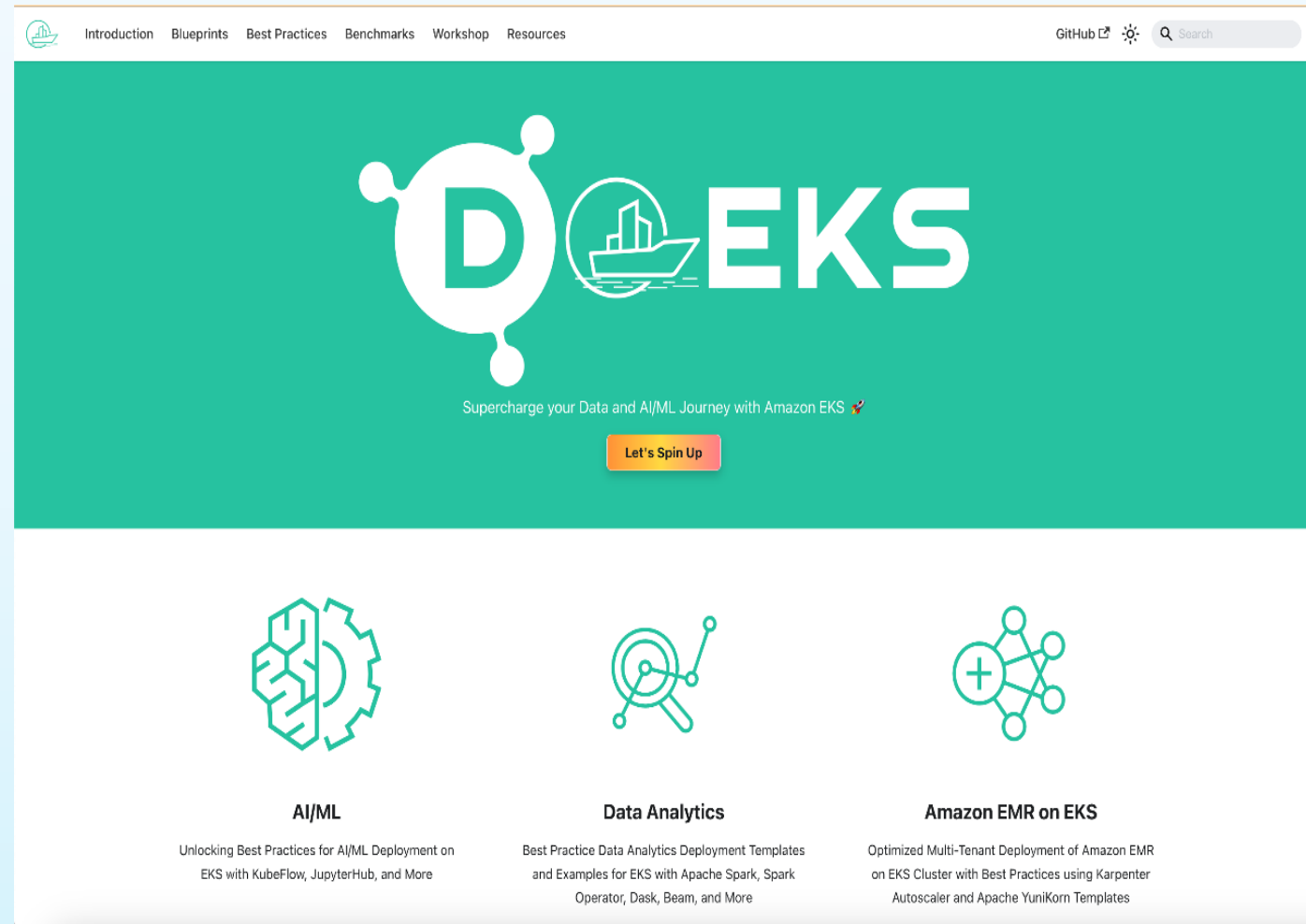


Key Takeaways

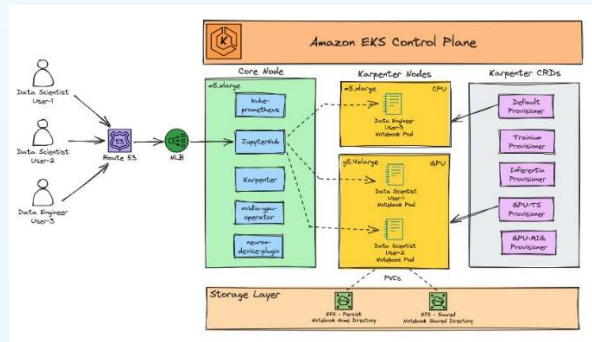
- Choice of Language Models
- Deploy models on AWS Trainium , Inferentia and Graviton
- Multi-agent workflow patterns
- Improve user experience – Centralized Security and observability

Data on EKS project

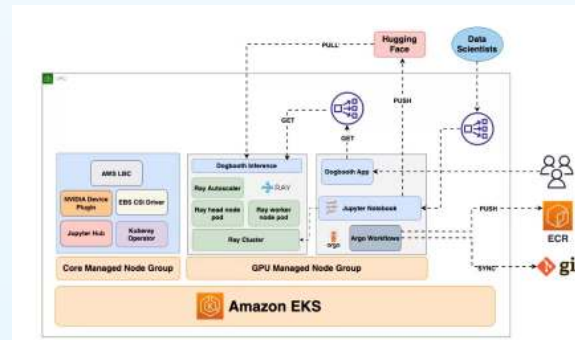
- Gen AI patterns
- IaC templates
- Best practices
- Engage with us on GitHub



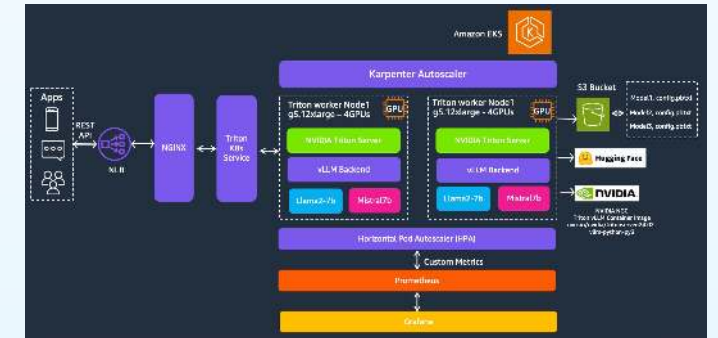
Reference patterns for gen AI on EKS



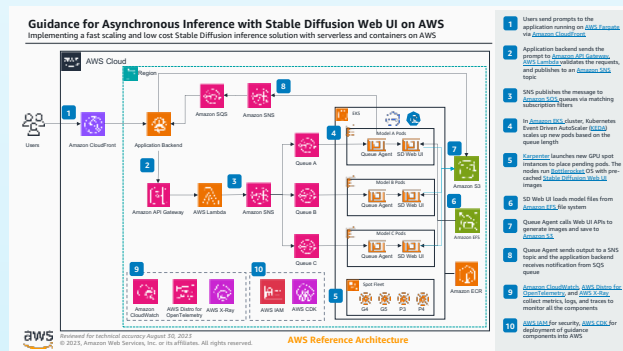
Multi-tenant JupyterHub platform



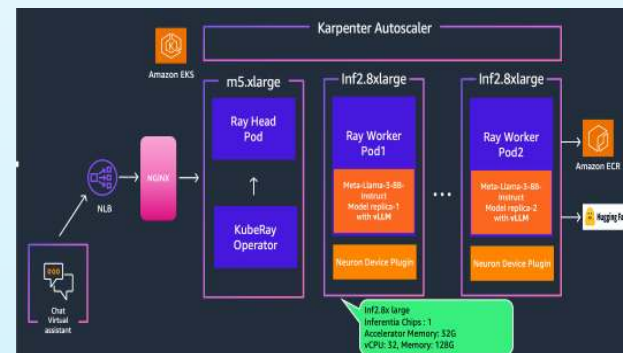
End-end generative AI orchestration platform



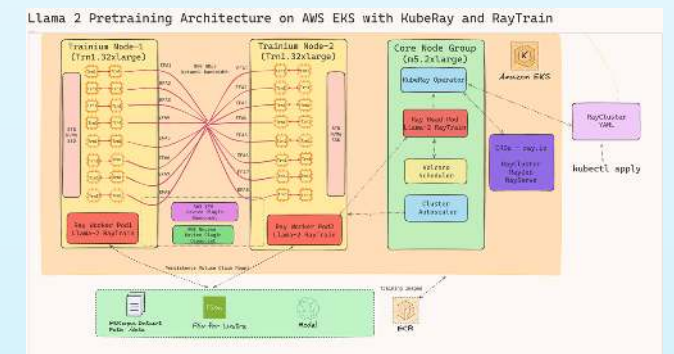
High performance inference platform on EKS (NVIDIA Triton with vLLM)



Async inference with Stable Diffusion on EKS



Inference of Llama-3-8B with RayServe/vLLM on EKS



Llama2 distributed pretraining on Trn1 with RayTrain

Thanks.

