

SCI for AI

Discussion points

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SCI Use Cases in AI

Use Case 1: LLM API Consumers (Prompt-based Usage)

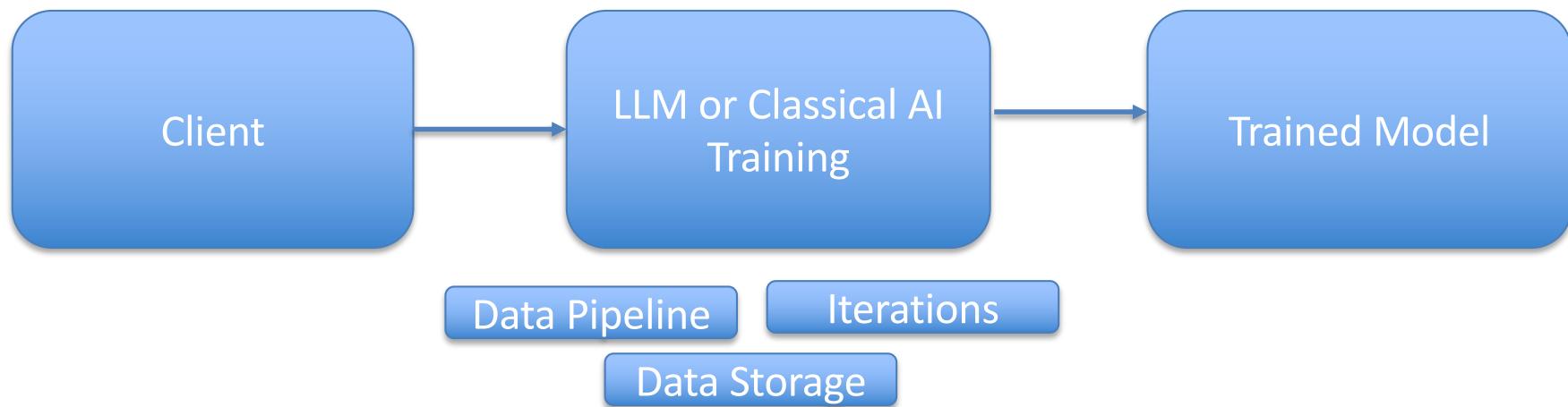


- Key Questions:

- Should training emissions (of trained model) be **amortized** into per-prompt SCI?
- Is inference cost already **included by provider**?
- How to account for **dynamic model routing**?
- Some analogy - If you are using an managed database or API gateway, do you factor development cost for database or API gateway or only runtime ?

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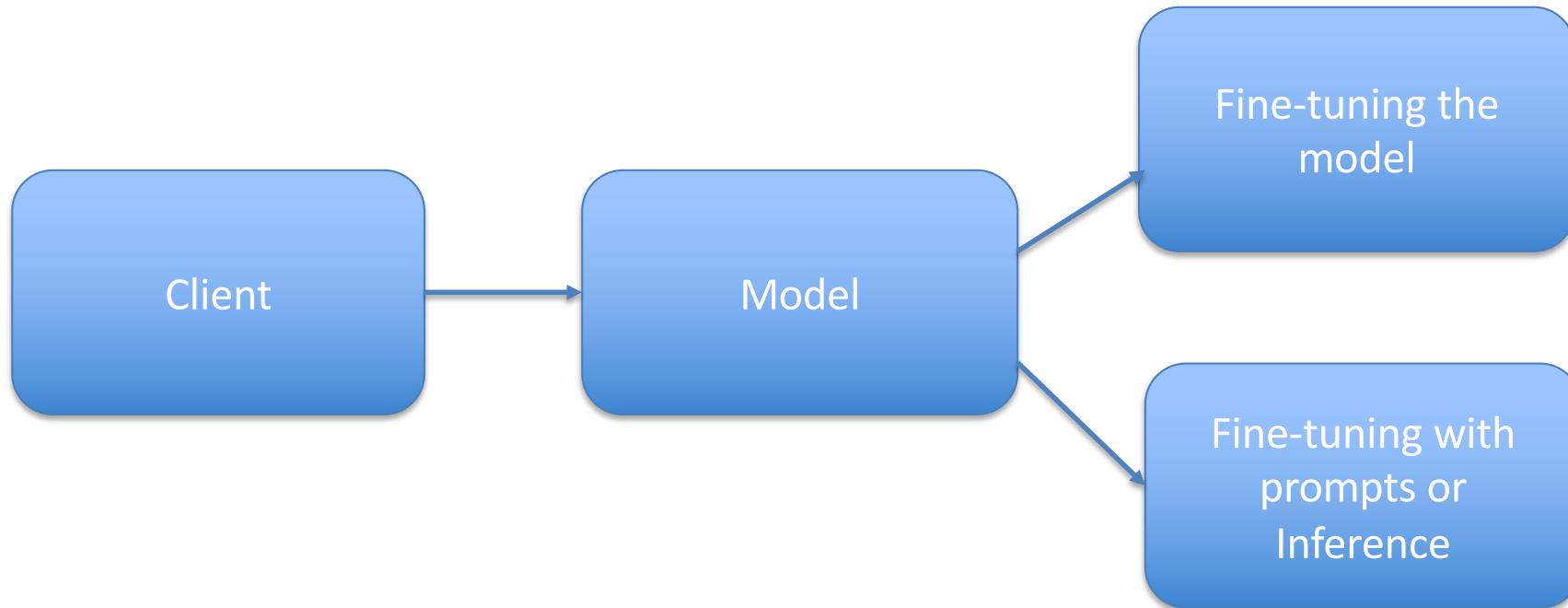
Use Case 2: Model Creators and Owners



- Boundary Considerations:
 - Include **entire data pipeline**?
 - Handle **multiple training iterations**?
 - **Versioning** and upgrades

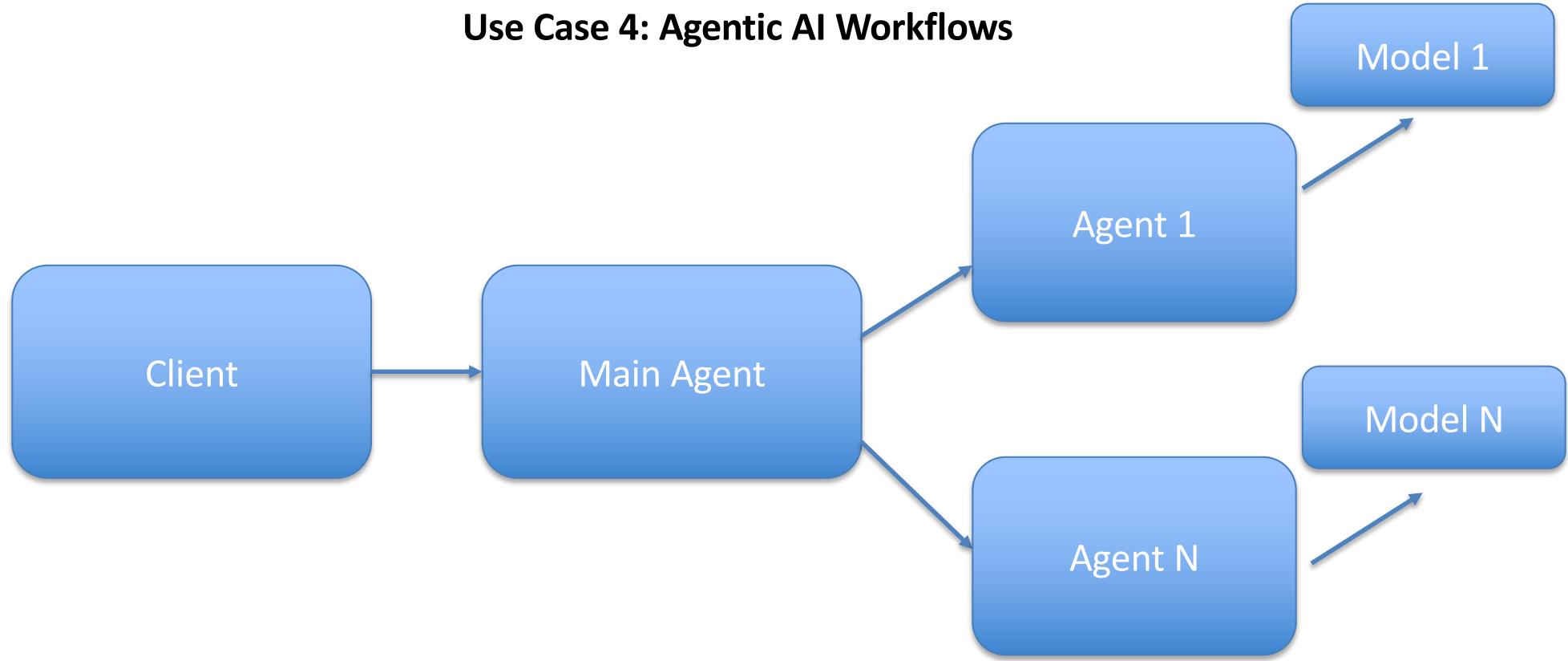
SCI Use Cases in AI

Use Case 3: Model Fine-Tuners (Pre-trained + Custom Data)



- Sits between full training and inference
- Requires **distinct SCI accounting**

SCI Use Cases in AI



- Composite workflows using **multiple models**
- Combination of API-based inference + fine-tuning
- Requires **hybrid SCI strategy**

SCI in AI - Actions

Use case	Training of Model included	Actions for reduction
LLM API Consumers (Prompt-based Usage)	No	<ul style="list-style-type: none">• Contextual prompt design to minimize response latency and reduce the number of interactions needed for accurate output.• Prompt caching and batching to avoid redundant computations and enhance efficiency at scale• Dynamic model routing to select the most efficient model based on task requirements—optimizing for carbon footprint, cost, performance, and safety• Energy-efficient hardware deployment for inference workloads, where organizations host or run LLMs internally.

SCI in AI - Actions

Use case	Training of Model included	Actions for reduction
Model Creators and Owners	Yes	<ul style="list-style-type: none">• Select energy-efficient model architectures that balance performance and training cost.• Utilize programming languages and frameworks known for computational efficiency and low overhead• Utilize custom energy-efficient chips optimized for AI training workloads.• Optimize supporting components, including energy-efficient data pipelines, storage solutions, and data augmentation processes.

SCI in AI - Actions

Use case	Training of Model included	Actions for reduction
Model Fine-Tuners (Pre-trained + Custom Data)	Yes (only for the fine-tuned model, not for the original pre-trained model)	<ul style="list-style-type: none">• Select energy-efficient model architectures that balance performance and training cost.• Utilize programming languages and frameworks known for computational efficiency and low overhead• Utilize custom energy-efficient chips optimized for AI training workloads.• Optimize supporting components, including energy-efficient data pipelines, RAG, storage solutions, and data augmentation processes.

SCI in AI - Actions

Use case	Training of Model included	Actions for reduction
Agentic AI Workflows	No	<ul style="list-style-type: none">• Choose energy-efficient languages, frameworks, and architectures to optimize multi-agent execution• Deploy custom energy-efficient chips for running agentic workflows and distributed reasoning tasks.• Design contextual prompts for effective reasoning, minimizing unnecessary computation and agent-to-agent interactions.• Implement prompt caching and batching across agents to reduce duplication and improve execution efficiency.• Optimize supporting components, including energy-efficient data pipelines, RAG, storage systems, and augmentation workflows.

SCI in AI

To define SCI for AI:

- Recognize **distinct roles** and **use cases**
- Clarify **what to include** in SCI boundaries
- Drive **consensus** across stakeholders
- Make it **simple** to implement, **explainable** and **actionable** for each of use cases to lower carbon emissions and increase energy efficiency.