

Amazon Bedrock

Michael Lin

Sr. Solutions Architect Amazon Web Services

The tipping point for generative Al

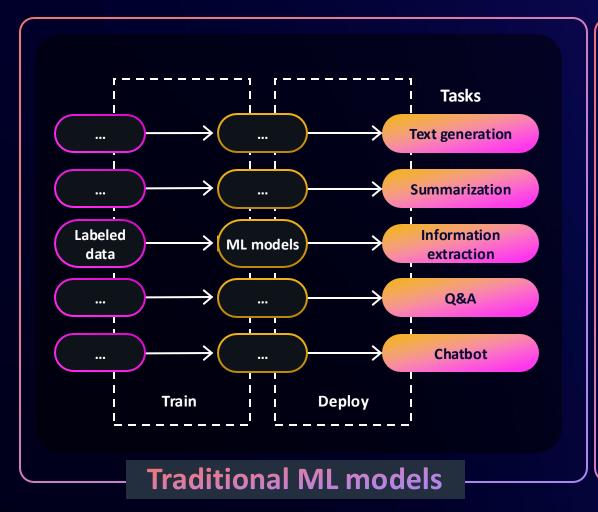
Massive proliferation of data

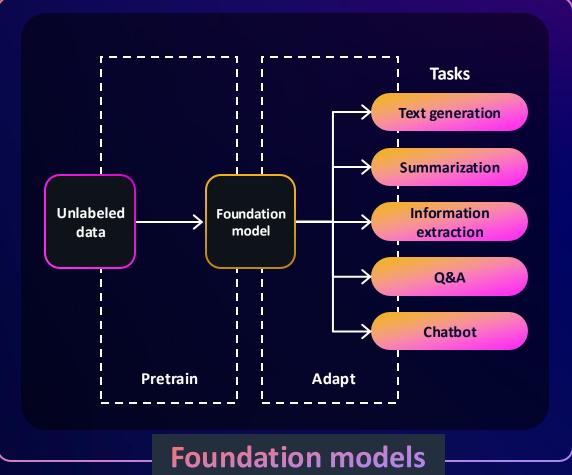
Availability of scalable compute capacity

Machine learning innovation



Generative AI is powered by foundation models (FMs)







Amazon Bedrock

The easiest way to build and scale generative Al applications with foundation models (FMs)

Choice of leading FMs through a single API

Model customization

Retrieval Augmented Generation (RAG)

Agents that execute multistep tasks

Security, privacy, and safety



Amazon Bedrock

BROAD CHOICE OF MODELS

Al21 labs	amazon	ANTHROP\C	c ohere	∞ Meta	MISTRAL AI_	stability.ai
	•	•	<u> </u>	•	•	•
Contextual answers, summarization, paraphrasing	Text summarization, generation, Q&A, search, image generation	Summarization, complex reasoning, writing, coding	Text generation, search, classification	Q&A and reading comprehension	Text summarization, text classification, text completion, code generation, Q&A	High-quality images and art
Jamba-Instruct	Amazon Titan Text Premier	Claude 3.5 Sonnet	Command	Llama 3 8B	Mistral Small	Stable Diffusion XL1.0
Jurassic-2 Ultra		Claude 3 Opus	Command Light	Llama 3 70B	Mistral Large	Stable Diffusion
Jurassic-2 Mid	Amazon Titan Text Lite	Claude 3 Sonnet	Embed English	Llama 2 13B	Mistral 7B	XL 0.8
	Amazon Titan Text Express	Claude 3 Haiku	Embed Multilingual	Llama 2 70B	Mixtral 8x7B	
		Claude 2.1	Command R+			
	Amazon Titan Text Embeddings	Claude 2	Command R			
		Claude Instant				
	Amazon Titan Text					



Embeddings V2

Amazon Titan
Multimodal Embeddings

Amazon Titan
Image Generator

Common approaches for customizing FMs

Complexity, quality, cost, time

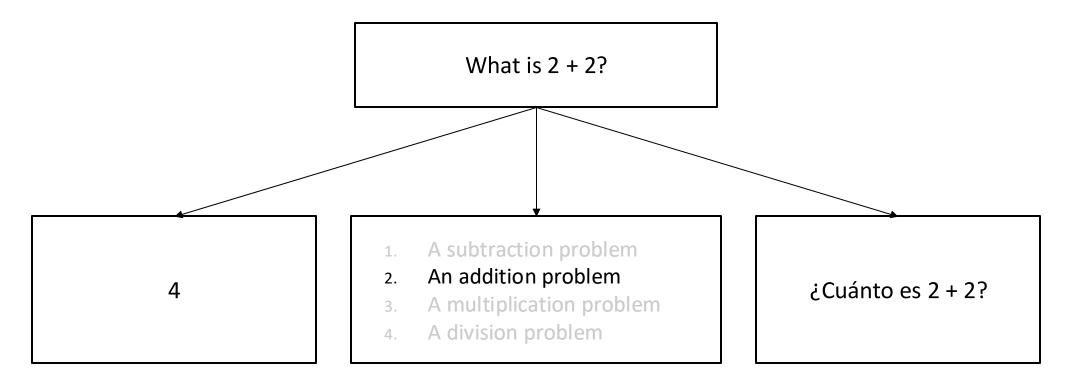
Prompt Generation engineering (RAG)

Fine-tuning

Continued pretraining



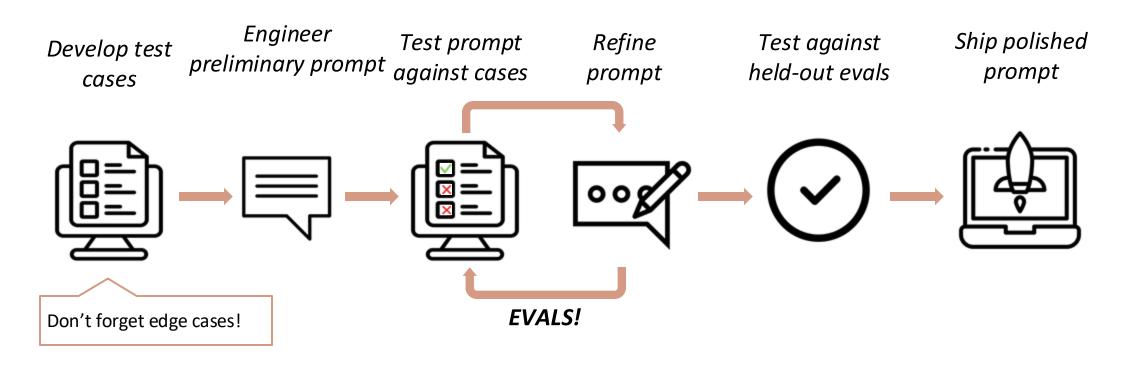
What is prompt engineering?

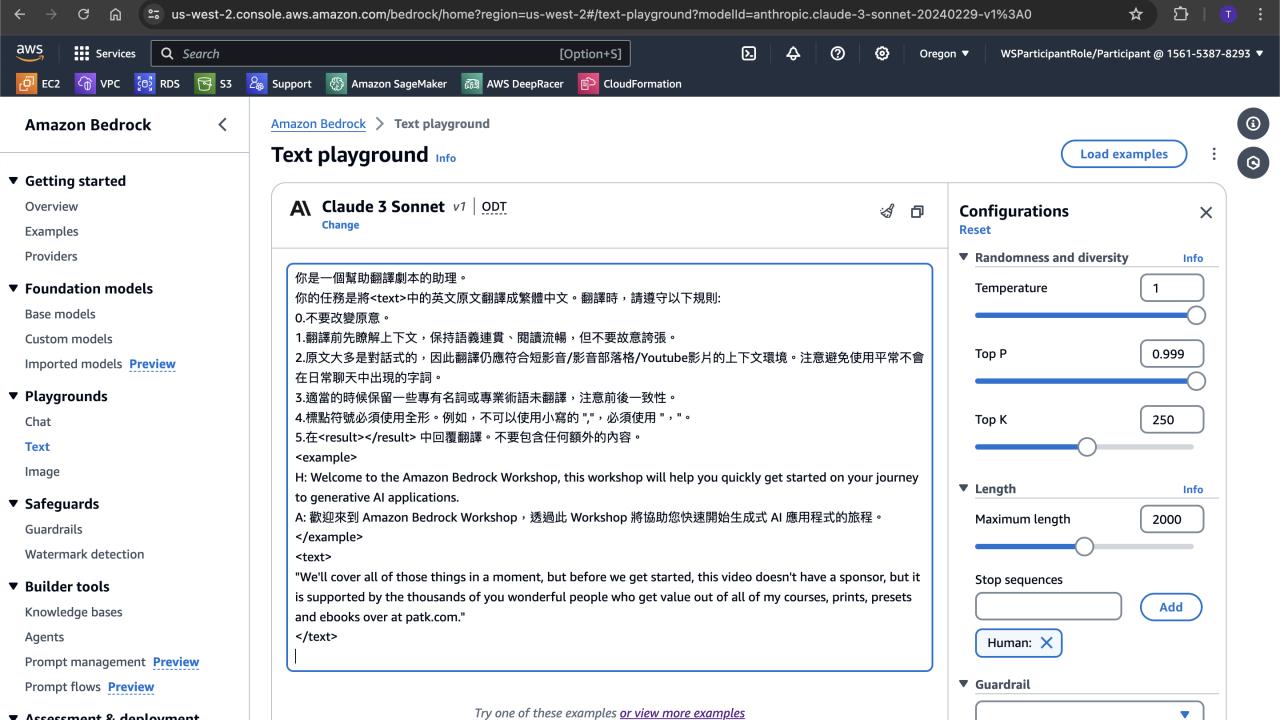


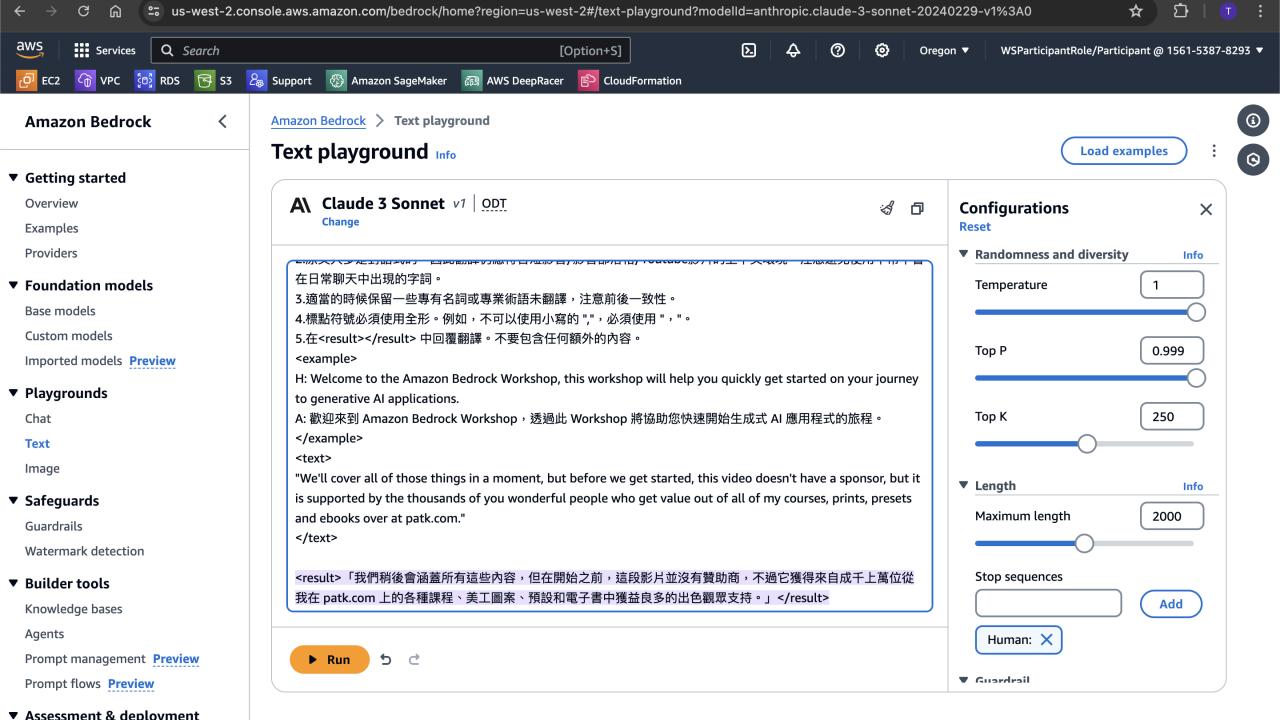
Prompt engineering is the process of controlling model behavior by optimizing your prompt to elicit high performing LLM responses (as assessed by rigorous evaluations tailored to your use case).

How to engineer a good prompt

Empirical science: always test your prompts & iterate often!







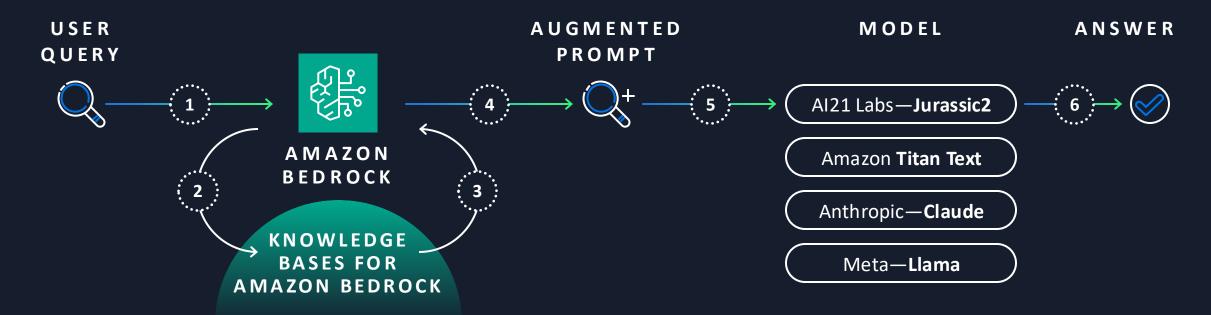
Knowledge Bases for Amazon Bedrock

NATIVE SUPPORT FOR RAG

Securely connect FMs to data sources for RAG to deliver more relevant responses Fully managed RAG workflow including ingestion, retrieval, and augmentation

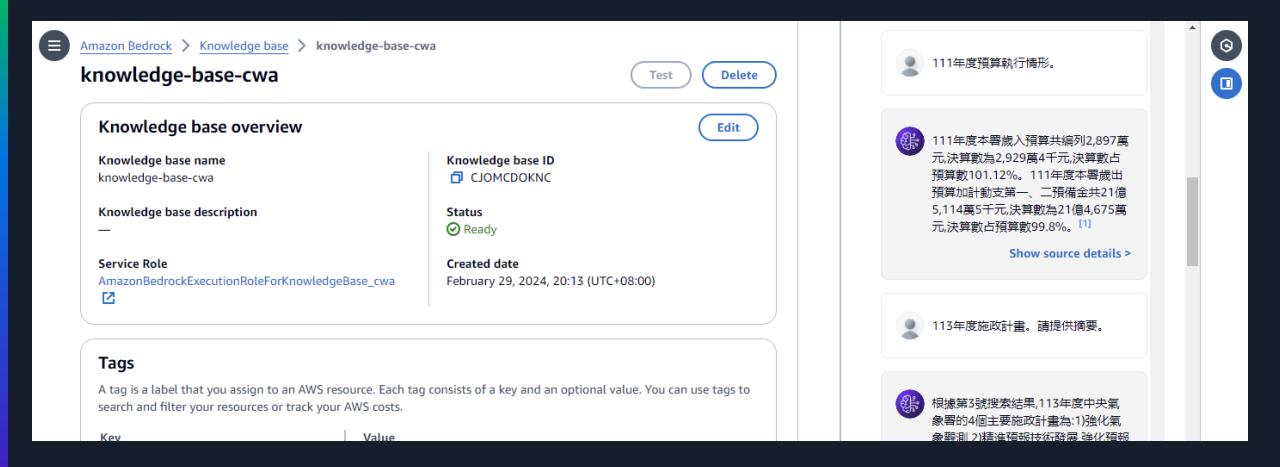
Built-in session context management for multiturn conversations

Automatic citations with retrievals to improve transparency





Bedrock Knowledge Base in Action







CWA_2024_PlanAndBudget.pdf 1 / 32 | − 100% + | 🕻 ♦

中華民國 112 年 10 月 立法院第 10 屆第 8 會期

交通部中央氣象署 113 年度

施政計畫及預算編列口頭報告

- 一、111 年度預算執行情形
 - (一)111 年度本署歲入預算共編列 2,897 萬元,決算 數為 2,929 萬 4 千元,決算數占預算數 101.12%
 - (二)111 年度本署歲出預算加計動支第一、二預備金 共 21 億 5,114 萬 5 千元,決算數為 21 億 4,675 萬元,決算數占預算數 99.8%。
- 二、112年度截至9月止預算執行情形
 - (一)112 年度本署歲入預算共編列 2,997 萬元,截至 9月止預算分配數 1,782 萬 4 千元,實收數 2,548 萬 2 千元,占預算分配數 142.96%。
 - (二)112 年度本署歲出預算(含預備金動支數)共編列 19 億 692 萬 8 千元,截至 9 月止預算分配數 12 億 2,413 萬 6 千元,執行數 11 億 7,312 萬 7 千 元,執行數占預算分配數 95.83%。

中央氣象署的署長是誰。



根据提供的内容可以知道,交通部中央氣象署署長是程家平。

9:32 PM

113年度新聞媒體從業人員氣象資訊推廣說明會,總共有幾場。 9:32 PM

9:32 PM



從上下文可以知道,113年度新聞媒體從業人員氣象資訊推廣說明會總共有4場,分別是北區、南區、東區和中區。

9:32 PM

北區場次的時間,與地點。

9:32 PM



根據提供的內容,北區場次的時間是2月29日(星期四),地點是中央氣象署617會議室(臺北市中正區公園路64號6樓)。

9:32 PM

111年度預算執行情形。

9:33 PM



據文件內容顯示,111年度本署歲入預算共編列2,897萬元, 決算數為2,929萬4千元,決算數占預算數101.12%。111年 度本署歲出預算加計動支第一、二預備金共21億5,114萬5 千元,決算數為21億4,675萬元,決算數占預算數99.8%。

9:33 PM



Customizing model responses for your business



Fine-tuning

PURPOSE

Maximizing accuracy for specific tasks

DATA NEED

Small number of labeled examples



Continued pretraining

PURPOSE

Maintaining model accuracy for your domain

DATA NEED

Large number of unlabeled datasets



Fine-Tuning in Action: Text Summarization

```
prompt = """
Summarize the simplest and most interesting part of the following conversation.
#Person1#: Hello. My name is John Sandals, and I've got a reservation.
#Person2#: May I see some identification, sir, please?
#Person1#: Sure. Here you are.
#Person2#: Thank you so much. Have you got a credit card, Mr. Sandals?
#Person1#: I sure do. How about American Express?
#Person2#: Unfortunately, at the present time we take only MasterCard or VISA.
#Person1#: No American Express? Okay, here's my VISA.
#Person2#: Thank you, sir. You'll be in room 507, nonsmoking, with a queen-size bed. Do you approve, sir?
#Person1#: Yeah, that'll be fine.
#Person2#: That's great. This is your key, sir. If you need anything at all, anytime, just dial zero.
Summary:
1111111
body = {
    "prompt": prompt,
    "temperature": 0.5.
    "top p": 0.9,
    "max gen len": 512,
```



Fine-Tuning in Action: Baseline Completion

```
response = bedrock_runtime.invoke_model(
    modelId="meta.llama2-13b-chat-v1", # compare to chat model
    body=json.dumps(body)
)

response_body = response["body"].read().decode('utf8')
print(json.loads(response_body)["generation"])
```

A man named John Sandals checks into a hotel and provides identification and a credit card. The hotel only takes M asterCard or VISA, so he uses his VISA card. He is given room 507, a nonsmoking room with a queen-size bed.



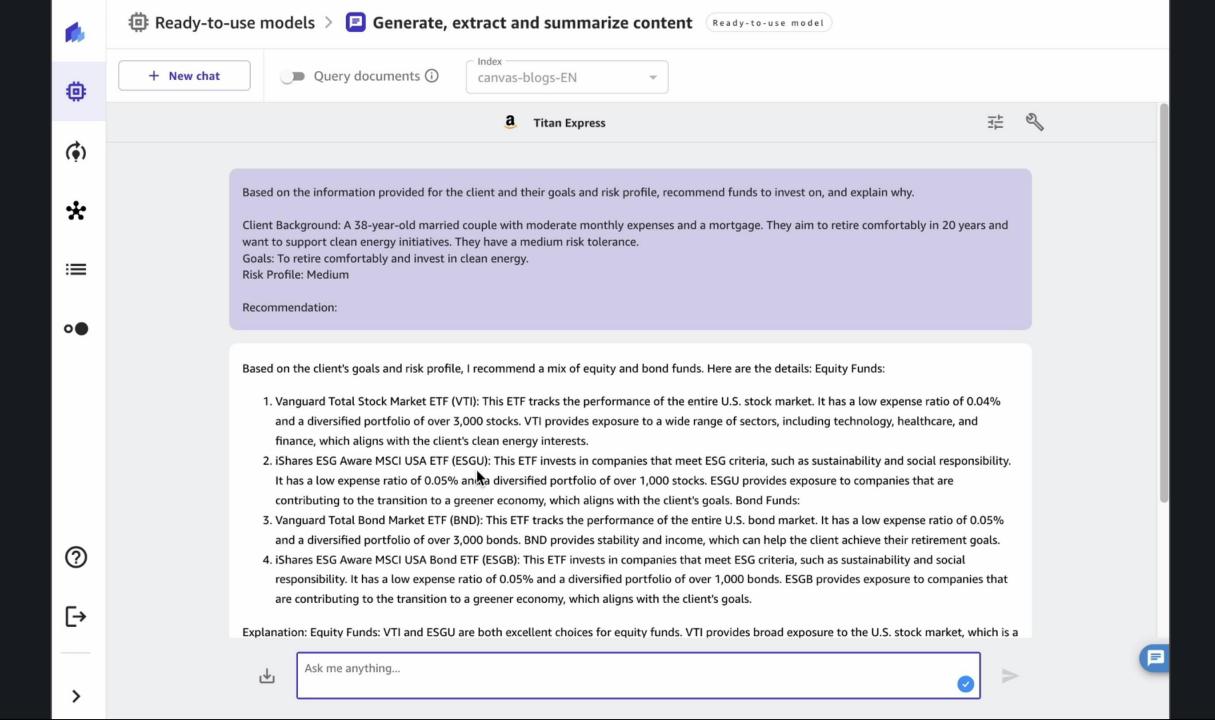
Fine-Tuning in Action: Improved Completion

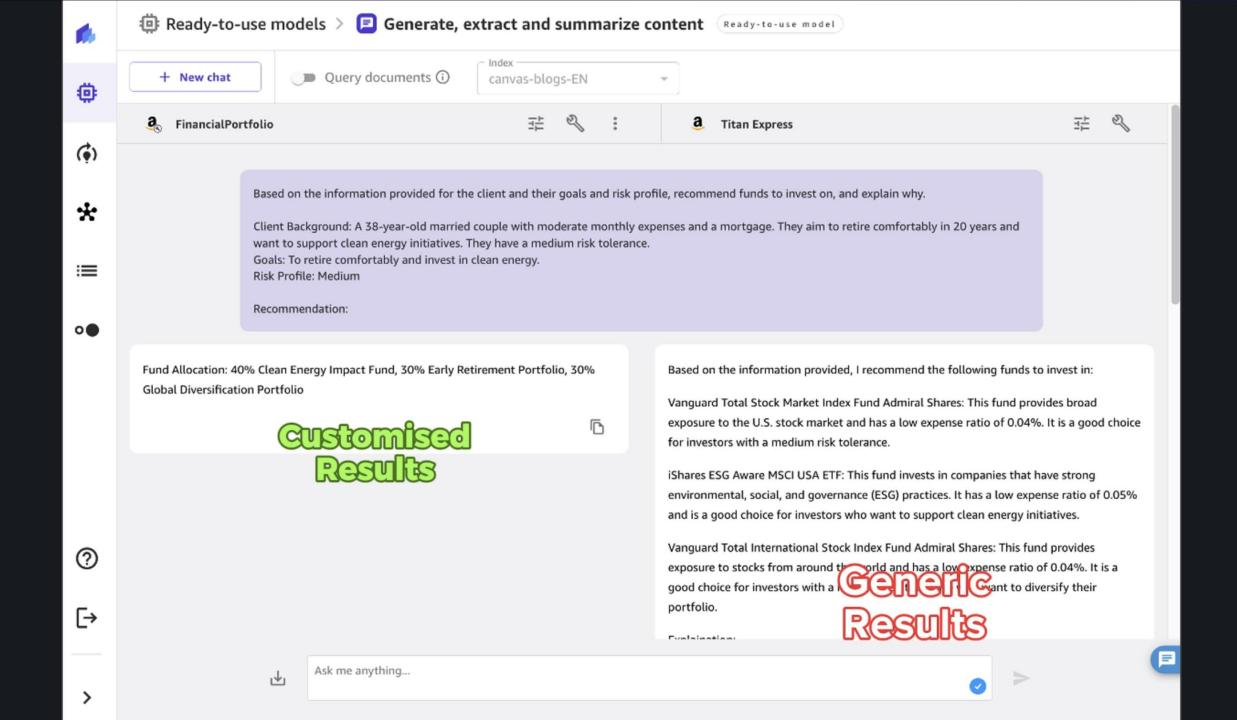
```
response = bedrock_runtime.invoke_model(
    modelId=provisioned_model_arn, # custom fine-tuned model
    body=json.dumps(body)
)

response_body = response["body"].read().decode('utf8')
print(json.loads(response_body)["generation"])
```

John Sandals checks in the hotel with VISA and is assigned room 507, nonsmoking, with a queen-size bed.







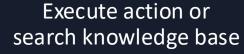
Agents for Amazon Bedrock

ENABLE GENERATIVE AI APPLICATIONS TO EXECUTE MULTISTEP TASKS USING COMPANY SYSTEMS AND DATA SOURCES



Decompose into steps using available actions and Knowledge Bases for Amazon Bedrock







Observe results



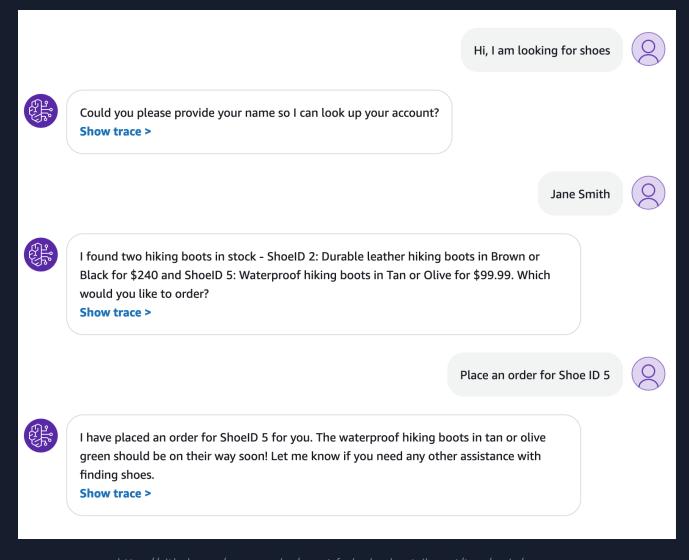
Think about next step



Until final answer



Agent in Action





Thank you!

Michael Lin

linmicht@amazon.com