

# AWS re:Invent

DECEMBER 2 - 6, 2024 | LAS VEGAS, NV

DEV 204

# Searching images through patterns: An AI-powered serverless solution

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APPGAMBiT



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Founder, APPGAMBiT

# Agenda

<b>01</b>	Introduction	<b>05</b>	Customer use case
<b>02</b>	State of gen AI models	<b>06</b>	Serverless AI-powered solution
<b>03</b>	Recent advancements	<b>07</b>	Our learnings
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“

**The true delight is in  
finding out rather  
than in the knowing.**

**Isaac Asimov**

Writer, Professor



# It all started with this . . .



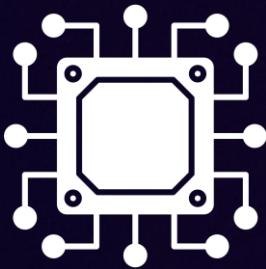
May 2024

```
{  
  "rating": 4,  
  "positives": [  
    "Well-defined heart shape in latte art",  
    "Good contrast between crema and milk",  
    "Even distribution of microfoam"  
  ],  
  "negatives": [  
    {  
      "issue": "Slight asymmetry in heart shape",  
      "solution": "Focus on pouring technique to ensure even  
      milk flow and adjust angle/position of pitcher for better  
      symmetry"  
    }  
  ]  
}
```

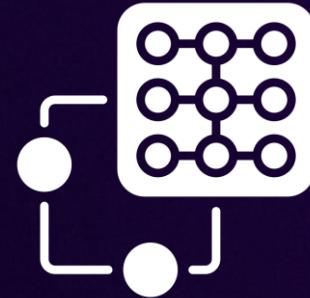
# The current state of large language models

	OpenAI	Meta	Google	Anthropic	Cohere	Mistral	Amazon	Stability AI	Microsoft
2024	o1-preview o1-mini GPT-4o	Llama 3.2 Llama 3.1	Gemini 1.5 Gemma	Haiku v2 Sonnet 3.5 v2 Sonnet 3.5 Claude 3 Haiku Sonnet Opus	Command R Command R+ Sonnet 3.5	Minstral 3B Minstral 8B Mistral Small Mistral Large Codestral Mixtral 8x22b Codestral Pixtral	Titan Text Premier	SD 3.0	Phi-3 Florence 2
2023	GPT-4	Code Llama Llama 2 Llama	Gemini PaLM 2	Claude 2	Command	Mixtral 8x7B / 7B	Titan Image Generator	XL Turbo XL 1.0	Phi-2 Phi-1

# Technology advancements



Better, faster GPUs



Improved dataset  
and  
training techniques

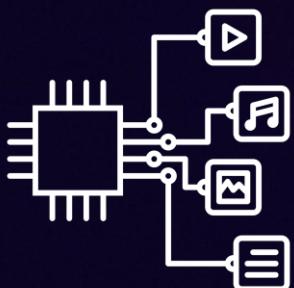
# Model advancements



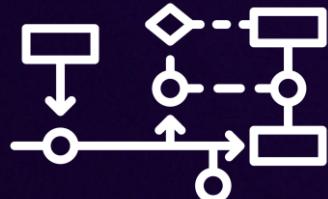
Larger context window



Better reasoning  
and inference



Multi-modality support



Agentic workflow

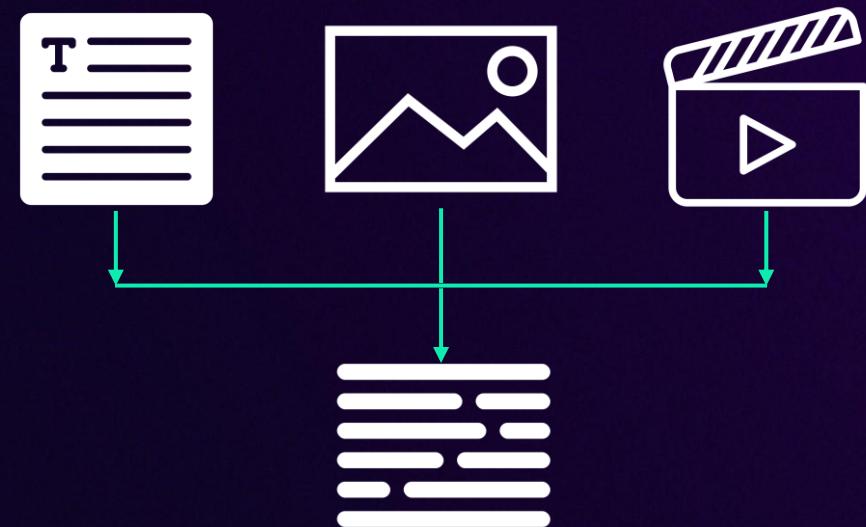
# Text models

Text-based models are designed to analyze input text (**prompt**) and generate new text information (**tokens**)



# Multimodal models

Multimodal models are designed to process information from different sources like text, image, audio, and video and generate new information (**tokens**)



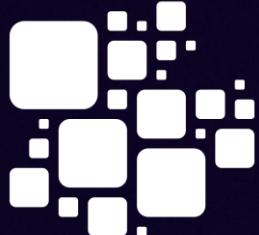
# Multimodal large language models

- Larger context window
- Improved image understanding
- Extract information from visual contents like infographics, charts, figures, tables, slides, etc.
- Suitable for business use cases
- Zero-shot prompting

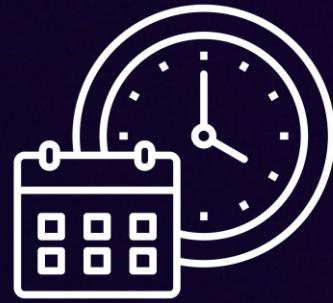
# Multimodal large language models

- Larger context window
- Improved image understanding
- Extract information from visual contents like infographics, charts, figures, tables, slides, etc.
- Suitable for business use cases
- Zero-shot prompting

# Why zero-shot prompting is important!



Extract information  
without previous  
examples



Faster time to  
implement and  
validate new use cases



Cost efficient

# Zero-shot prompting example



Model: Anthropic Haiku v1 on Amazon Bedrock

Q: Can you identify bird in this picture?

A: The bird in the image appears to be a **hummingbird**. The small, slender body with the long, thin beak are characteristic features of a hummingbird. The vibrant green and brown feathers also match the coloring of many hummingbird species. Hummingbirds are known for their incredible ability to hover in mid-air and dart rapidly between flowers, which seems to be what this hummingbird is doing on the branch.

Photo by [Cedric Fox](#) on [Unsplash](#)

# Zero-shot prompting example in Spanish



Model: Anthropic Haiku v1 on Amazon Bedrock

Q: ¿Puedes identificar el pájaro en esta imagen?

A: El pájaro en la imagen es un colibrí. Los colibríes son pequeñas aves que se caracterizan por su plumaje colorido y la capacidad de volar de forma muy ágil, batiendo sus alas rápidamente. En esta imagen, se puede apreciar el característico plumaje verde iridiscente del colibrí, así como su pico largo y delgado, típico de esta especie.

Photo by [Cedric Fox](#) on [Unsplash](#)

# Examples

- Visual question answering
- Visual interpretation
- Visual captioning
- Visual grounding

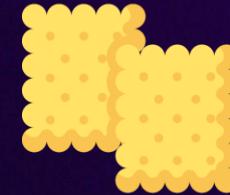
## Input

### Text

Question: Which property do these two objects have in common?

Options: A) soft B) salty

### Vision



Crackers

Fries

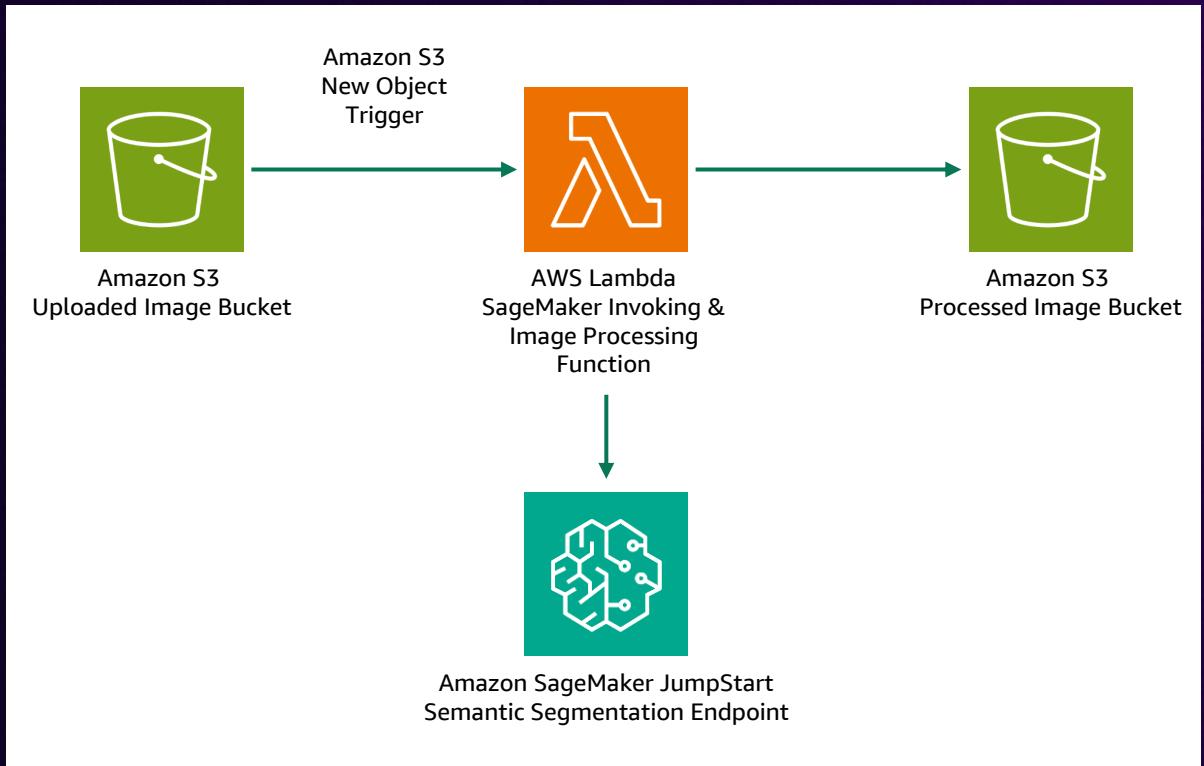
## Output

The property the two objects have in common is that they are both **salty**. The image on the left appears to show a **salty cracker** or biscuit, while the image on the right shows a serving of **french fries**, which are typically salted.

Output generated by Anthropic Claude 3 Haiku v1

# Examples

- Visual question answering
- Visual interpretation
- Visual captioning
- Visual grounding



Q: Describe the image.

A: Based on the diagram, the Uploaded Image Bucket and Processed Image Bucket serve different purposes in the image processing workflow:

**1. Uploaded Image Bucket:**

This is the **initial storage location for images that are uploaded to the system**. It's where new images are first placed, triggering the subsequent processing steps.

**2. Processed Image Bucket:**

This is the **destination for images after they have been processed by the AWS Lambda function**. It stores the results of the image processing and SageMaker operations.

# Examples

- Visual question answering
- Visual interpretation
- Visual captioning
- Visual grounding

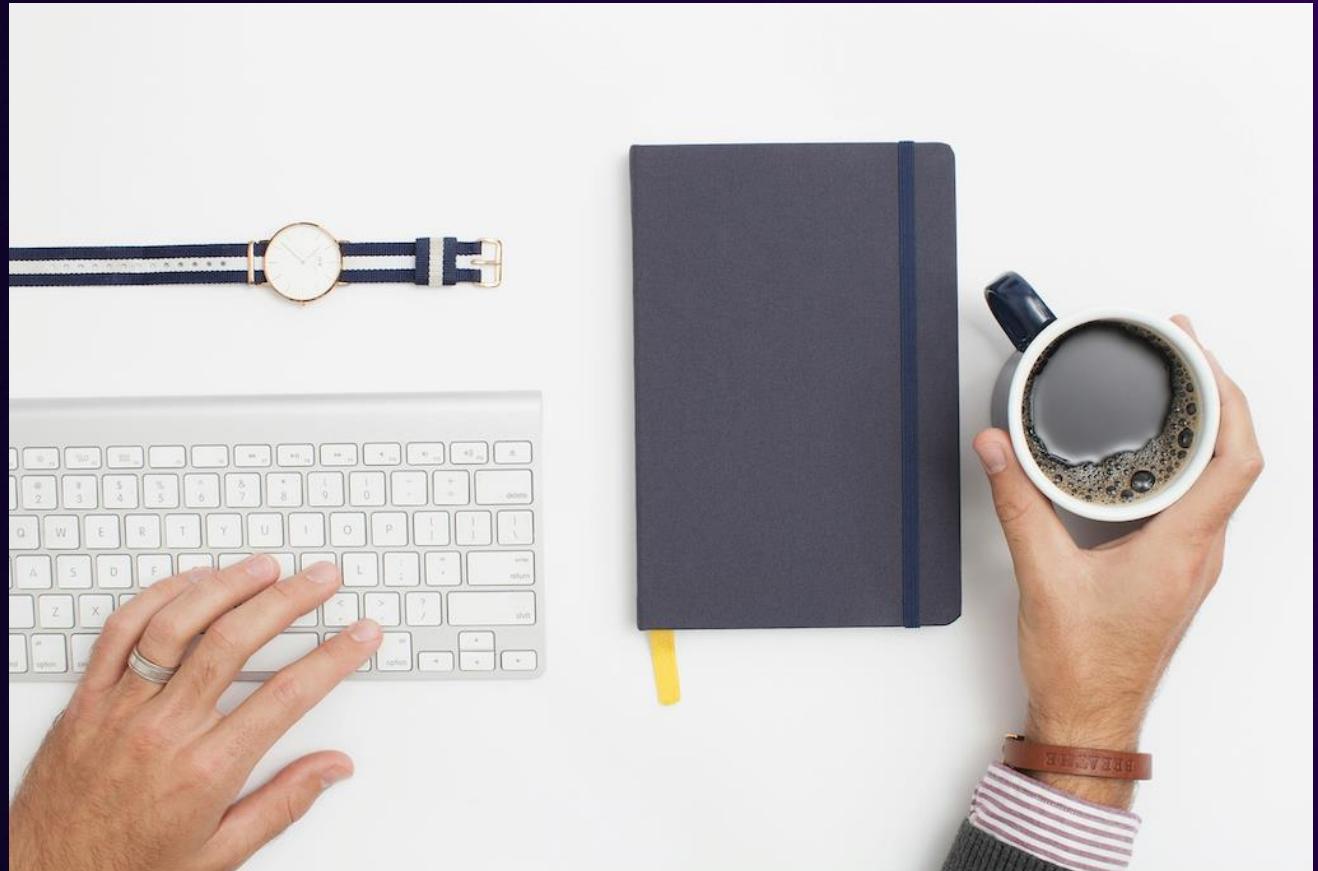


Photo by [Cathryn Lavery](#) on [Unsplash](#)

Q: Describe the image in one sentence.

A: A workspace scene featuring **hands typing on a keyboard**, **holding coffee**, **with a watch**, **notebook**, and **other office essentials** **neatly arranged on a white surface**.

# Examples

- Visual question answering
- Visual interpretation
- Visual captioning
- Visual grounding

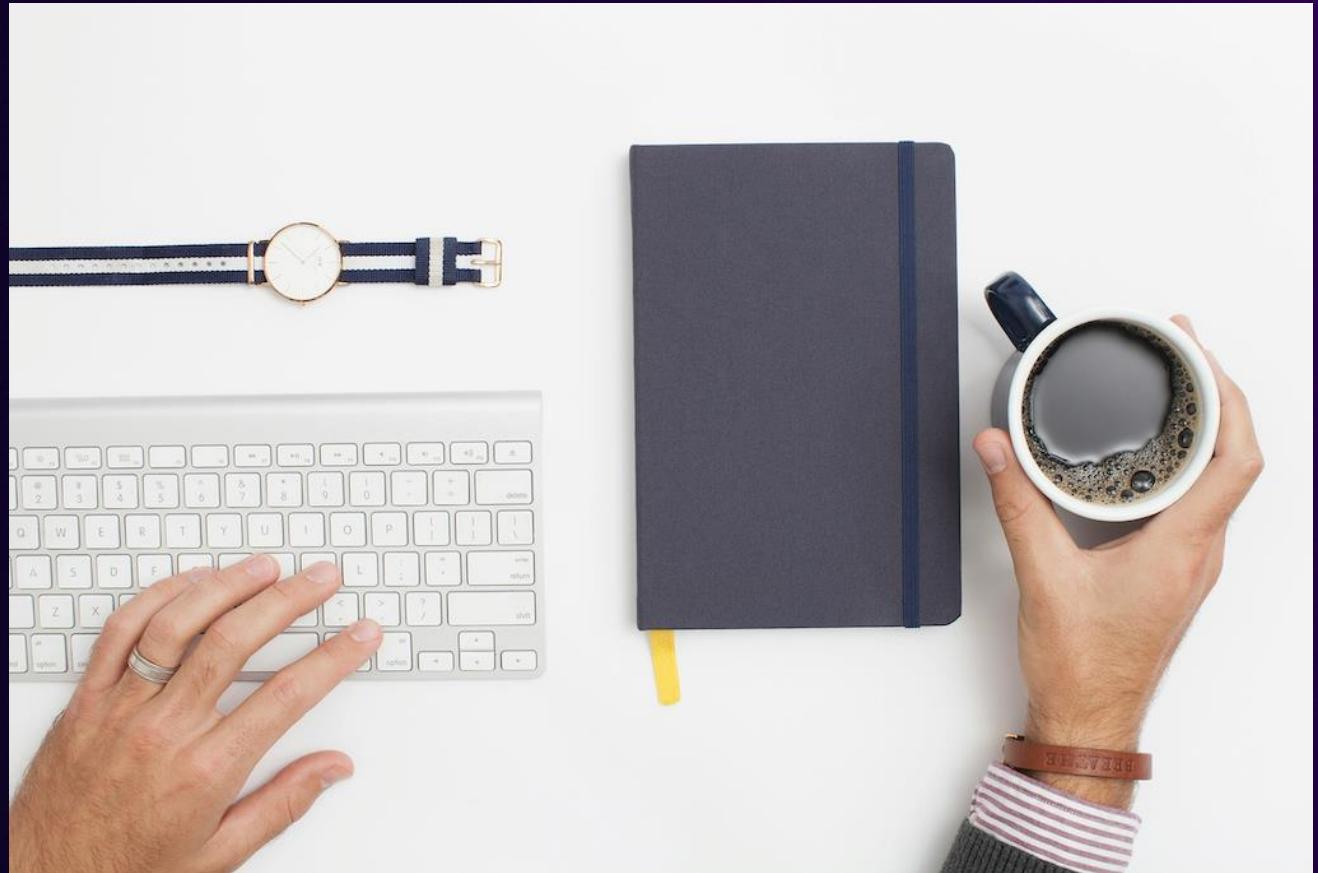


Photo by [Cathryn Lavery](#) on [Unsplash](#)

Q: Which side is the coffee cup and the wrist watch?

A: In the image, the coffee cup is on the right side, being held by a hand. The wrist watch is on the left side of the image, lying flat on the surface.

# Examples

- Visual question answering
- Visual interpretation
- Visual captioning
- Visual grounding

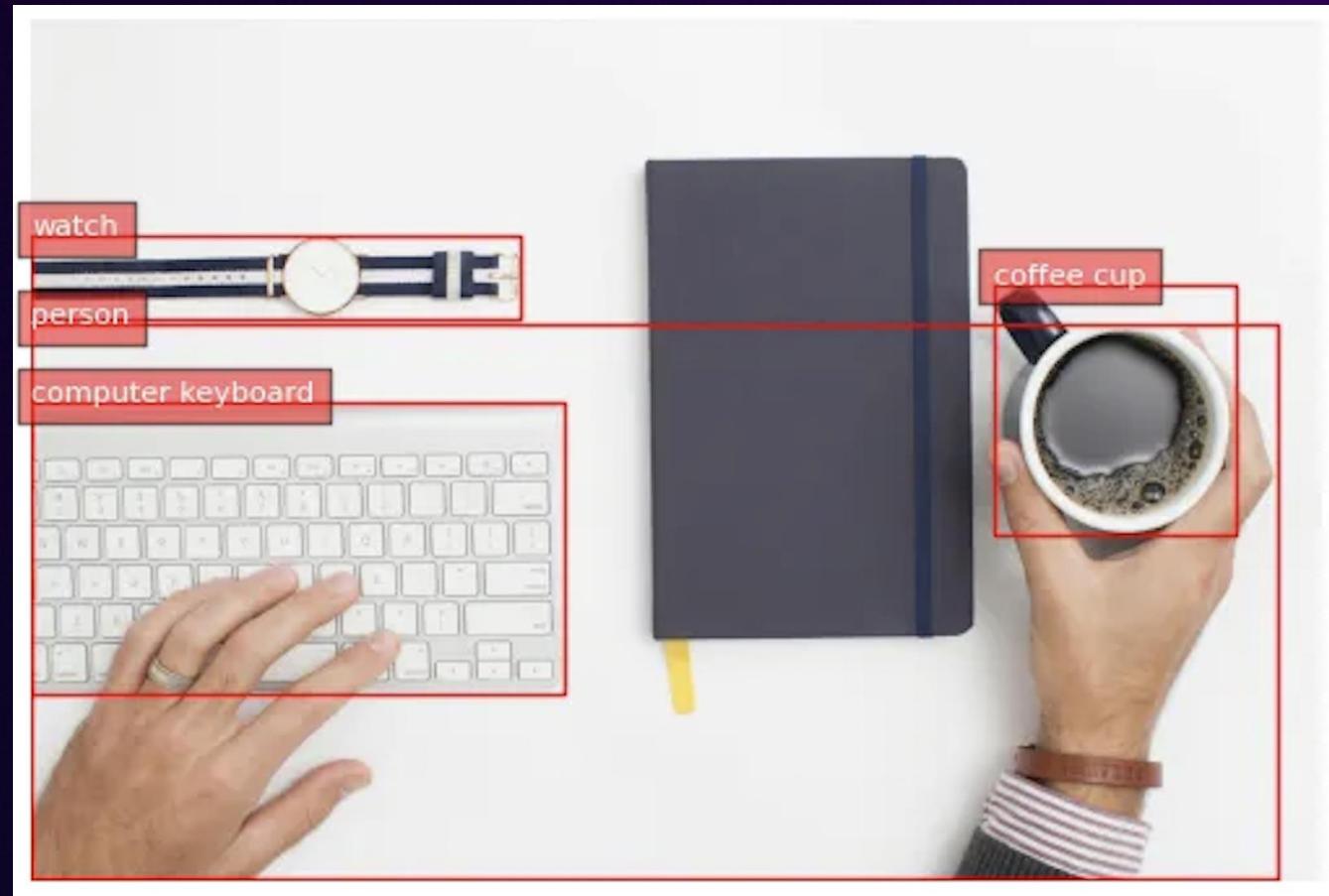


Photo by [Cathryn Lavery](#) on [Unsplash](#)

## P: Caption + Grounding

```
A: { "<OD>": { "bboxes": [[1789.2000732421875, 492, 2238, 959.2000122070312], [1.200000476837158, 711.2000122070312, 990.0000610351562, 1250.4000244140625], [1.200000476837158, 567.2000122070312, 2319.60009765625, 1597.5999755859375], [1.200000476837158, 402.3999938964844, 910.800048828125, 556]], "labels": ["coffee cup", "computer keyboard", "person", "watch"] } }
```

# Customer use case

# Digital Fabric Printing House



Raw designs created  
by designers

TIFF files



Raw material is used to  
create finished garments



Printing machine prints  
the design on fabric

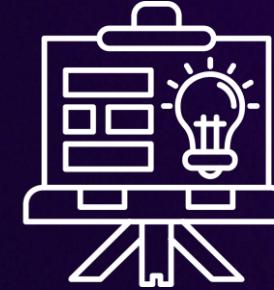
# Business process flow



Design theme  
identification

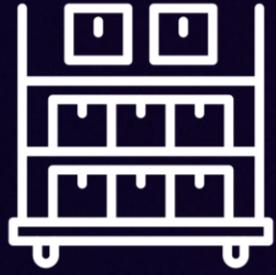


Search images for  
concept building

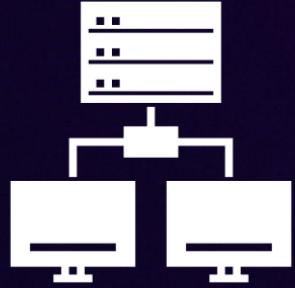


Create design  
mood boards

# Customer infrastructure



Large dataset of  
design files



Private storage  
with network  
access



Files are stored in  
RAW and preview  
format

# Resource inefficiency

- NAS is a file storage system
- Searching through NAS is a manual and human-driven process
  - Users manually search through file system and extract images
  - Time-consuming and error-prone
- Using online commercial resources
  - Paying for the images already available in local storage
  - Cost more due to time shortage

# What we needed!

- Process images with low infrastructure burden
- Generate user-friendly image attributes
- Attribute-based search and retrieval
- Cost-efficient
- Low maintenance
- Pay-as-you-use model

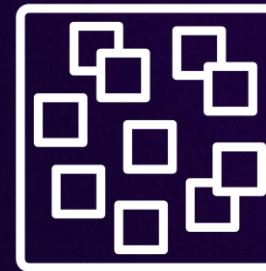
# Initial challenges



Dataset with large  
file sizes



Not all design files  
are required



Unstructured  
dataset

“

**Generative AI technology  
can be very effective in  
converting unstructured  
data into structured format.**

# AI-powered serverless solution

# Cost and process efficiency

- Experiment with budget
- Direct process impact
- Data privacy and security
- Low-cost solution

# Expected output

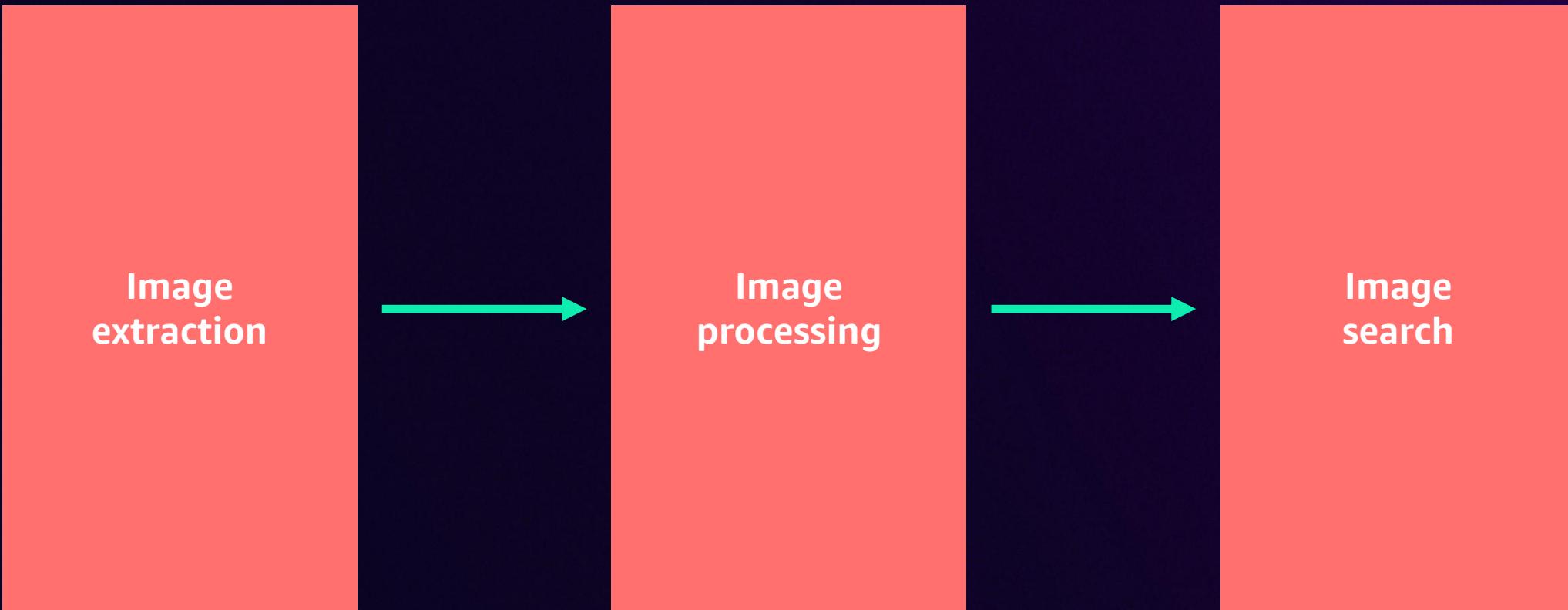
Input image



Output attributes

Patterns	Animal, Leopard
Colors	Orange, Black, Brown
Background Colors	Orange
Animals	TRUE
MultiColor	TRUE
AdultPrint	TRUE
KidsPrint	FALSE

# Solution breakdown



# Important decisions

Preprocess images  
on  
local infrastructure

Using gen AI model  
instead of  
machine learning model

# Process images on on-premise infrastructure



Low-cost solution



Original images are  
easily accessible



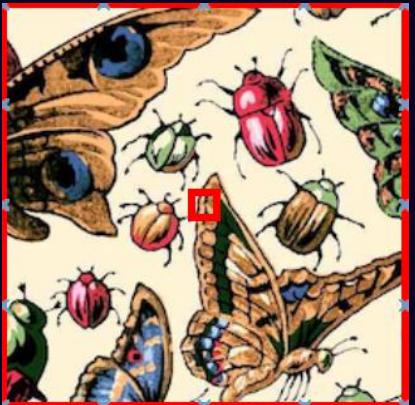
Reuse existing  
infrastructure

# Extract segments from original images



Sample segment

# Segment properties



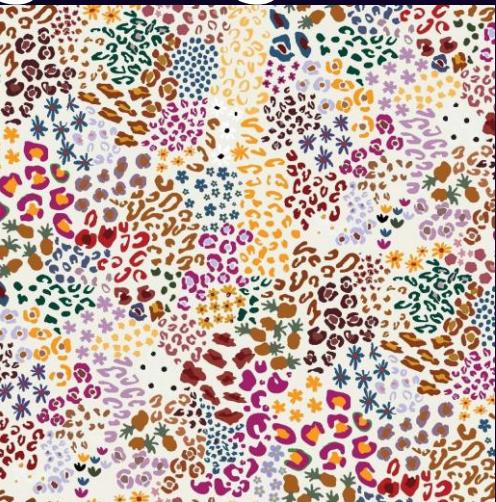
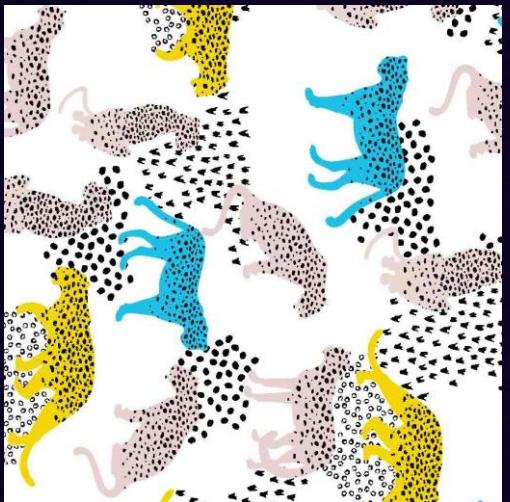
1024 x 1024  
extraction

Optimize  
storage size

Sufficient detail  
covered

Repeatable  
process

# Extracted image segments



# Amazon Rekognition

- Image recognition and video analysis with machine learning
- Use cases
  - Detect objects, scenes, and faces
  - Extract texts (OCR)
  - Recognize celebrities
  - Extract certain image properties
- Managed service
- Pay per image analysis

# Amazon Rekognition

- Image recognition and video analysis with machine learning
- Use cases
  - Detect objects, scenes, and faces
  - Extract texts (OCR)
  - Recognize celebrities
  - Extract **certain image properties**
- Managed service
- Pay per image analysis

# Amazon Rekognition with image properties



Dominant colors (In RGB format)  
#ffffe0, #2f4f4f, #808080

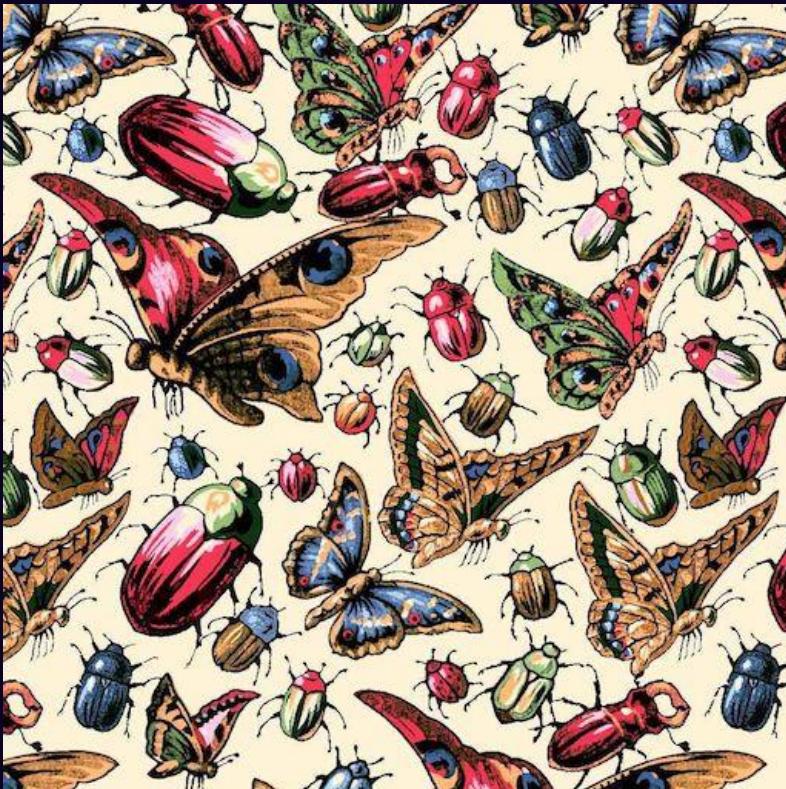
83  
Brightness

100  
Sharpness

93  
Contrast

# We wanted to achieve this

Input image



Output attributes

Patterns	Insects, Bugs, Butterflies
Colors	Red, Blue, Green, Brown, Yellow, Pink
Background Colors	Cream
Animals	TRUE
MultiColor	TRUE
AdultPrint	TRUE
KidsPrint	TRUE

“

There are applications that used to take **months to build** with higher cost, resources, and specialised skillset.

Now, people can build those applications **much faster** with large language models.

**Andrew Ng**

Opportunities in AI - 2023, Stanford Online

“

**Claude 3 models have sophisticated vision capabilities on par with other leading models.**

**They can process a wide range of visual formats, including photos, charts, graphs, and technical diagrams.**

**Claude 3 Launch Post**

4th March 2024

# Claude 3 Haiku output



```
"image_analysis": {  
  "pattern": [  
    "animal", "leopard"  
  ],  
  "colors": [  
    "orange", "black", "brown"  
  ],  
  "backgroundColors": [  
    "orange"  
  ]  
}
```

# More attributes



```
{  
  pattern: [  
    "animal", "leopard"  
,  
  colors: [  
    "orange", "black", "brown"  
,  
  backgroundColors : ["orange"],  
  hasShapes: false,  
  hasPeople: false,  
  hasAnimals: true,  
  multiColor: true,  
  adultPrint: true,  
  kidsPrint: false,  
  hasFlowers: false,  
  hasFruits: false,  
  hasStripes: false  
}
```

# Image analysis cost

Claude 3 Haiku is priced at \$0.25 per million input tokens

Image processing cost

Image tokens = (width px \* height px) / 750

Image size	# of image tokens	# of text tokens	Cost	Cost 1K images
1024 x 1024	~ 1398	~ 200	~ \$0.0004	~ \$0.4

Perfect combination of cost and performance

# High-level process flow



Upload images  
for processing

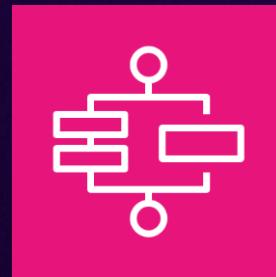


Process and save  
attributes  
into SQL database



API to search  
and retrieve images  
based on attributes

# Image processing on AWS



Images uploaded  
in Amazon S3  
Bucket

Amazon EventBridge  
Cron schedule start  
processing

AWS Step Functions  
to process images

# Amazon Bedrock

- Fully managed service
- Select from list of foundation models
  - Amazon, Cohere, Meta, Anthropic, AI21, Mistral, and Stability AI
- Private access
- Flexible pricing
  - On-demand
  - Provisioned throughput
  - Batch

# Serverless experience

```
const callAIModel = async (filePath) => {
  var prompt =
    Please analyze the attached image for fabric pattern printing.
  ...
  ;

  const base64Image = await getFile(S3_IMAGES_BUCKET, filePath);

  return invokeModel({
    imageData: base64Image,
    prompt,
    modelId: "anthropic.claude-3-haiku-20240307-v1:0"
  });
};
```

Insert image

Add prompt

Set model ID

# Amazon Aurora Serverless v1

- Serverless SQL database
- Data API support
- Auto pause and resume
- Faster startup time
- Cost-effective for our use case

# Amazon Aurora Serverless v1

- Serverless SQL database
- Data API support
- Auto pause and resume
- Faster startup time
- Cost-effective for our use case

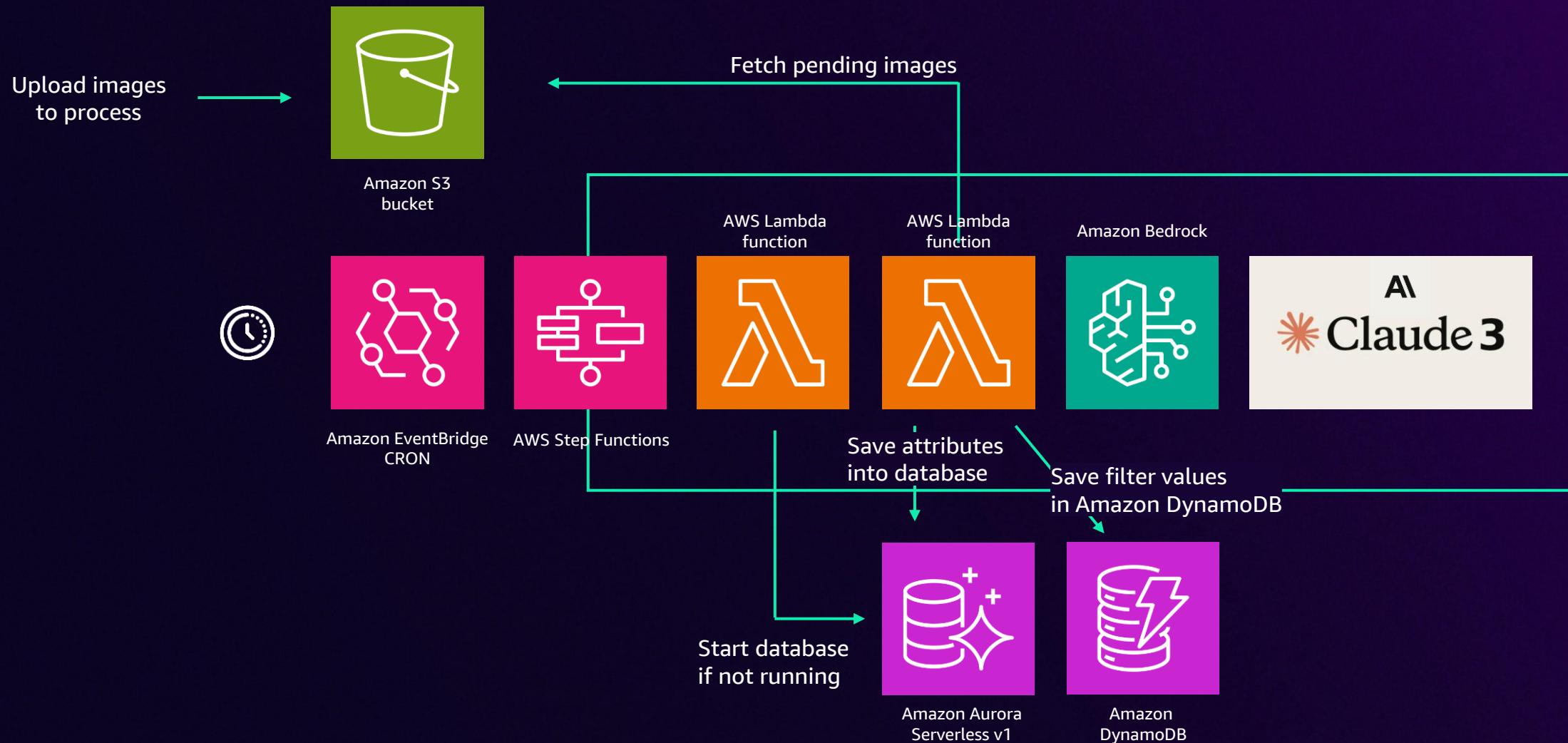
Aurora Serverless v1 is deprecated now

# Amazon Aurora Serverless v2

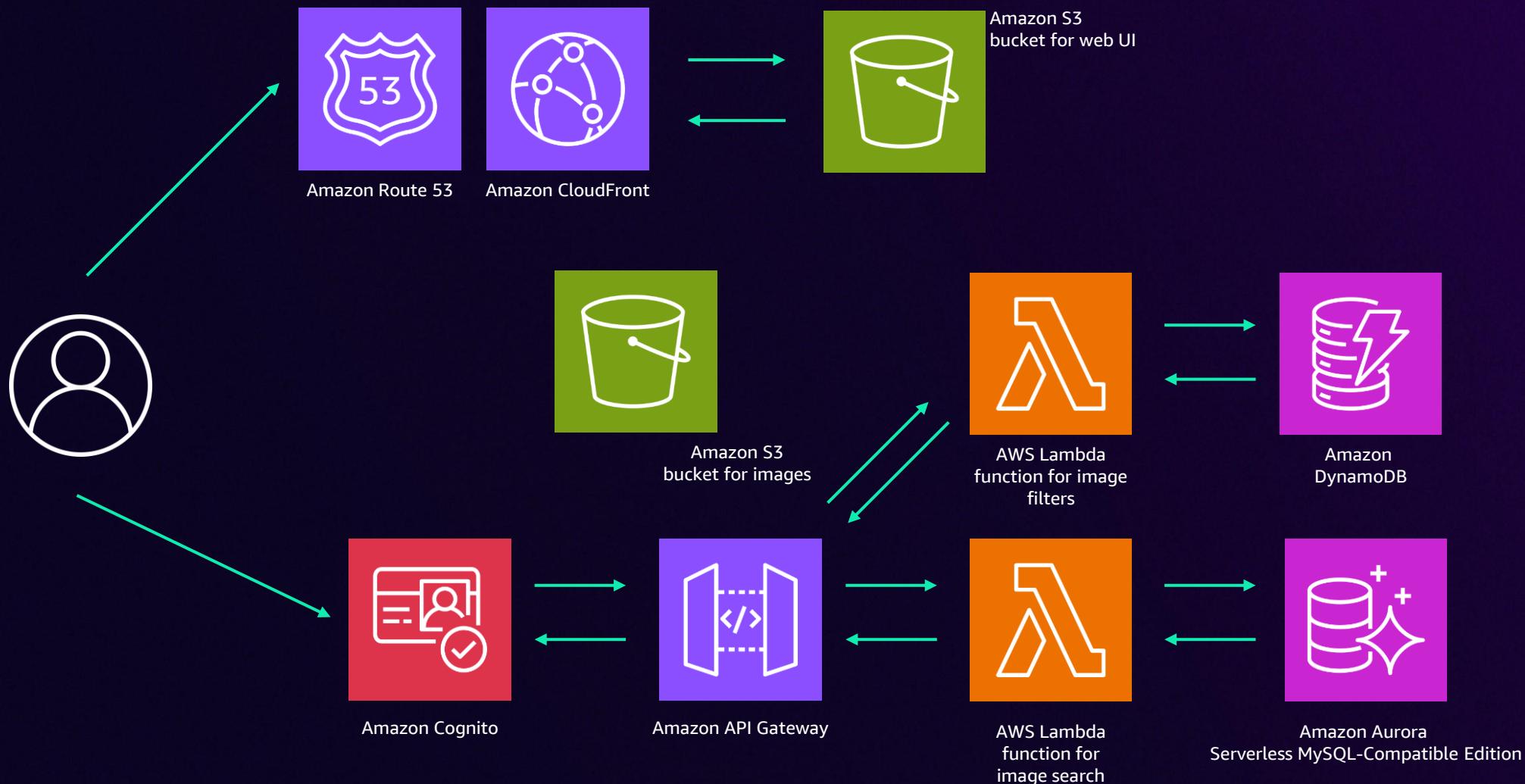
- Serverless SQL database
- Data API support
- Costly compared to Aurora Serverless v1
- Scale to zero with automation

AWS announced the Auto Pause and Resume for  
Aurora Serverless v2

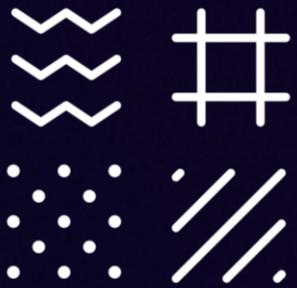
# Architecture diagram: Image processing



# Architecture diagram: Search images



# Our learnings



Identified 300+  
patterns



Speed and  
efficiency



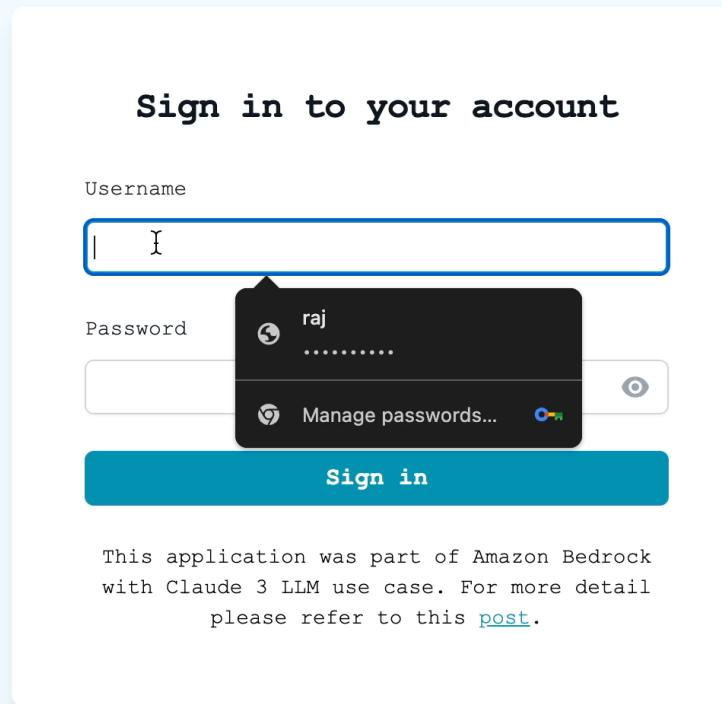
Cost-  
effective

# Our learnings

- Keep process flow simple
- Constant feedback loop
- Use low-maintenance infrastructure
- Use pay-as-you-go services

# Demo

# Pattern Search



# Pattern Search

[Logout](#)

Abstract X X | V

Colors X X | V

White X X | V

Search

Shapes  People  Animals  Fruits  Flowers  Adult  Kids

Database is starting up, please try again in few seconds.

# Summary

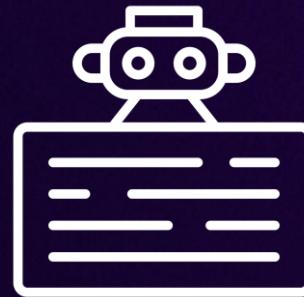
- Serverless and managed infrastructure helps in **rapid development**
- Do generative AI models work 100%?
  - Not 100%, but **very effective**
- LLMs are **becoming commodities**
- Use **pay-as-you-go model for experimentation**
- Every use case is **different**
  - Continuous experimentation helps in better understanding



Focus on the problem



Use cloud infrastructure  
intelligently



LLMs are rapidly  
advancing



Cost is a huge factor for  
many use cases

# Q&A

# Thank you!

**Dhaval Nagar**

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Please complete the session  
survey in the mobile app