

In This Video

- Service Introduction
- Overview
- Usage of Amazon Rekognition
- Demonstration

Searchable Image Library

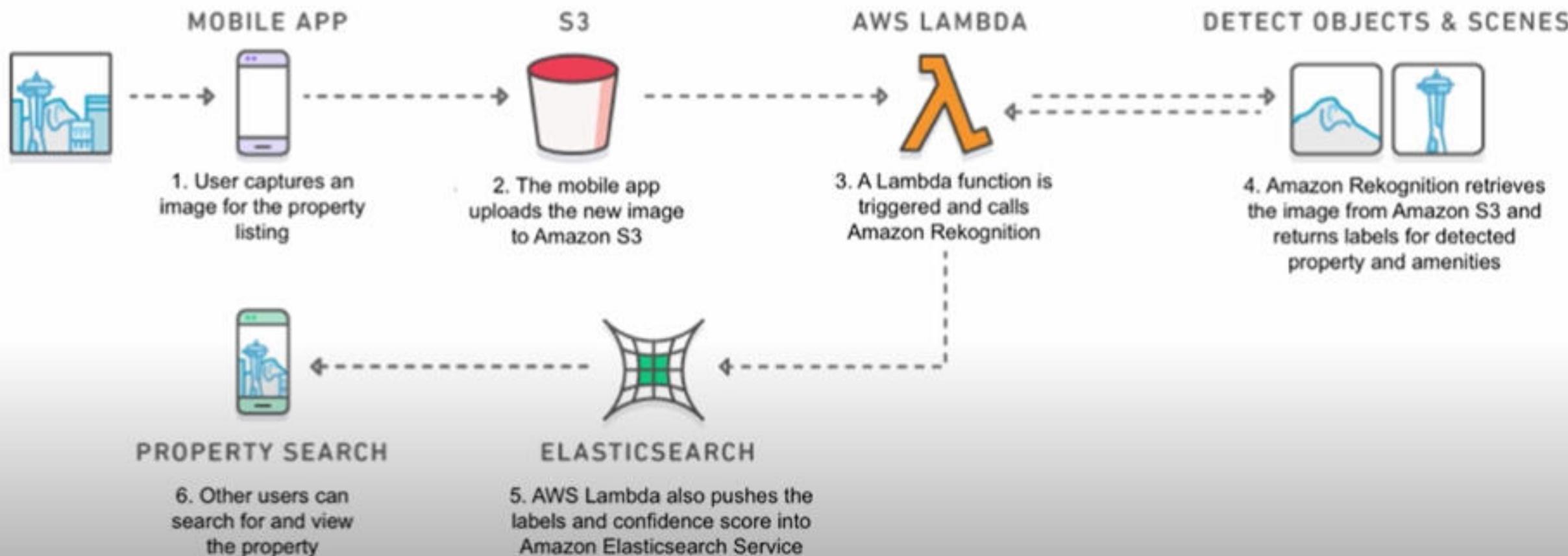
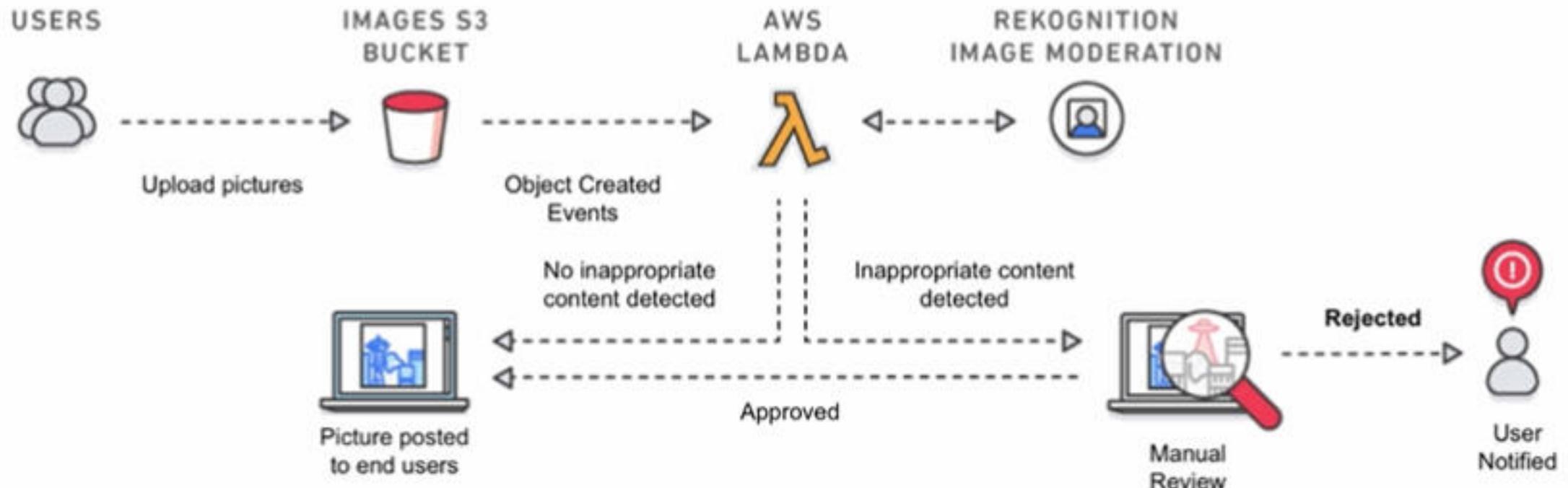
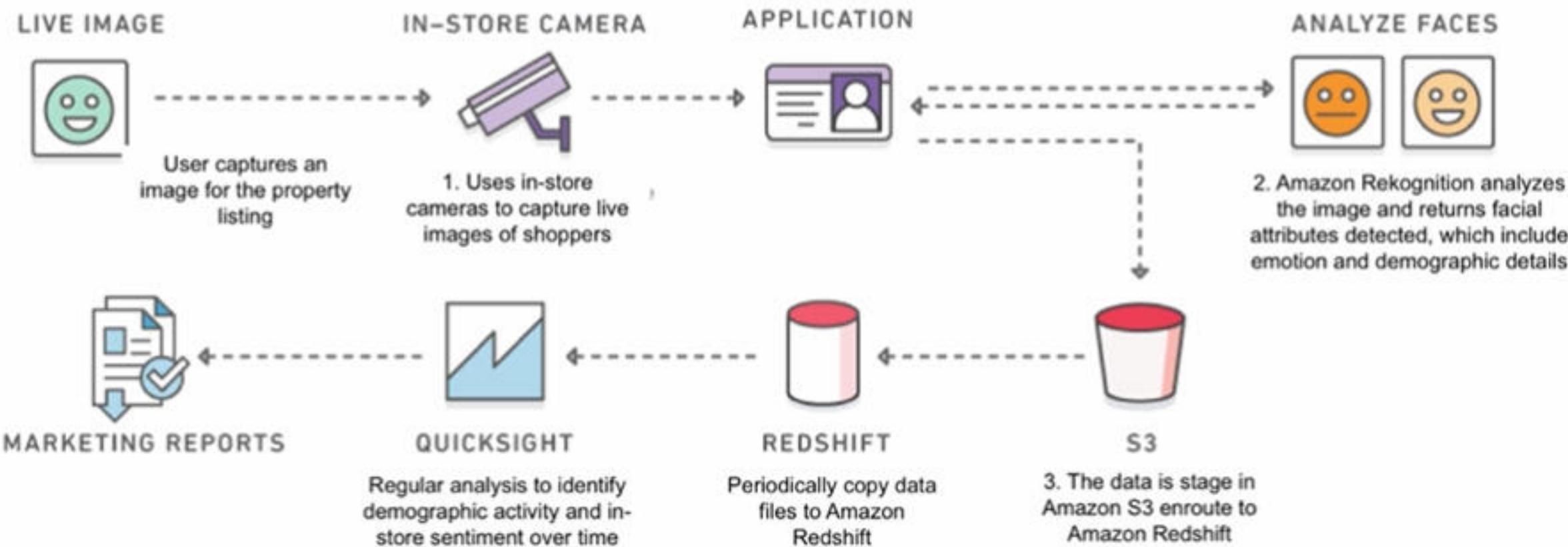


Image Moderation



Sentiment Analysis



AWS services

Find a service by name or feature (for example, EC2, S3 or VPC, storage).



Recently visited services

[Rekognition](#)[CloudWatch](#)[EC2](#)

All services

Build a solution

Get started with simple wizards and automated workflows.



Launch a virtual machine
With EC2 or Lightsail
~1-2 minutes



Build a web app
With Elastic Beanstalk
~8 minutes



Host a static website
With S3, CloudFront, Route 53
~5 minutes



Connect an IoT device
With AWS IoT
~5 minutes



Start a development project
With CodeStar
~3 minutes



Register a domain
With Route 53
~3 minutes

[See more](#)

Helpful tips

[Manage your costs](#)

Get real-time billing alerts based on your cost and usage budgets. [Start now](#)

[Create an organization](#)

Use AWS Organizations for policy-based management of multiple AWS accounts. [Start now](#)

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Build Applications with AWS Lambda

Run and scale code for Python, Node.js, Java, or C# without provisioning or managing servers. [Learn more](#) ⓘ

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Fast and flexible NoSQL database service for any scale. [Learn more](#) ⓘ

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| History | | Find a service by name or feature (for example, EC2, S3 or VM, storage) | | | | | | | |
|-------------|-------------------------------|---|-----------------------|------------------|-----------------------------|-----|--|--|-----------------------|
| | | Group | | | | A-Z | | | |
| rekognition | A | C | E | M | S | | | | View your cost |
| CloudWatch | Amazon Chime | Certificate Manager | EC2 | Machine Learning | S3 | | | | |
| EC2 | Amazon Connect | CloudFormation | EC2 Container Service | Managed Services | Server Migration Service | | | | |
| | Amazon GameLift | CloudFront | EFS | Mobile Analytics | Service Catalog | | | | |
| | Amazon Macie | CloudHSM | Elastic Beanstalk | Mobile Hub | Simple Email Service | | | | |
| | Amazon Polly | CloudSearch | Elastic Transcoder | | Simple Notification Service | | | | based on usage, Start |
| | Amazon Redshift | CloudTrail | ElastiCache | O | Simple Queue Service | | | | |
| | API Gateway | CloudWatch | Elasticsearch Service | OpsWorks | Snowball | | | | |
| | Application Discovery Service | CodeBuild | EMR | | Step Functions | | | | |
| | AppStream 2.0 | CodeCommit | | | Storage Gateway | | | | |
| | Artifact | CodeDeploy | G | P | SWF | | | | |
| | Athena | CodePipeline | Glacier | Pinpoint | T | | | | |
| | AWS Glue | CodeStar | I | Q | Trusted Advisor | | | | : for deep |
| | AWS Greengrass | Cognito | IAM | QuickSight | V | | | | |
| | AWS IoT | Config | J | R | VPC | | | | |
| | AWS Migration Hub | | K | RDS | | | | | |
| | B | D | Kinesis | Rekognition | W | | | | |
| Batch | Data Pipeline | L | Route 53 | Route 53 | WAF & Shield | | | | |
| | Database Migration Service | Lambda | | | WorkDocs | | | | |
| | Device Farm | Lex | | | WorkMail | | | | |
| | Direct Connect | Lightsail | | | WorkSpaces | | | | any scale. |
| | Directory Service | | | | X | | | | |
| | DynamoDB | | | | X-Ray | | | | See products |

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| | AWS Greengrass | Cognito | Inspector | R | V | | | | |
| | AWS IoT | Config | Kinesis | RDS | VPC | | | | |
| | AWS Migration Hub | D | Lambda | Rekognition | | X, or CWL | | | |
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| | | Direct Connect | | | WorkMail | | | | |
| | | Directory Service | | | WorkSpaces | | | | |
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| | | | | | X-Ray | re products | | | |

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Amazon Rekognition

Deep learning-based image recognition service

Search, verify, and organize millions of images

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Get Context and Information from Images

Amazon Rekognition enables your applications to gain a high-level understanding of the content of images. With Rekognition, your applications can detect objects, scenes, and faces in images. With this rich context and information, you can enable visual search, discovery, and authentication in your applications.



Scalable Deep Learning-based Image Analysis

Amazon Rekognition is designed to use deep learning technology to analyze billions of images daily. The APIs find and compare faces, detect thousands of objects. We continue to add new objects and make improvements to facial analysis so you can focus on building applications.



Integrated with AWS Services

Amazon Rekognition is designed to work seamlessly with other AWS services. Rekognition integrates directly with Amazon S3 and AWS Lambda so you can build scalable, affordable, and reliable image analysis applications. You can start analyzing images stored in Amazon S3 without moving any data. Support for Identity and Access Management (IAM) makes it easy to assign unique security credentials to each user and control their access to Amazon Rekognition APIs.

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Services - Resource Groups - N. Virginia - Support

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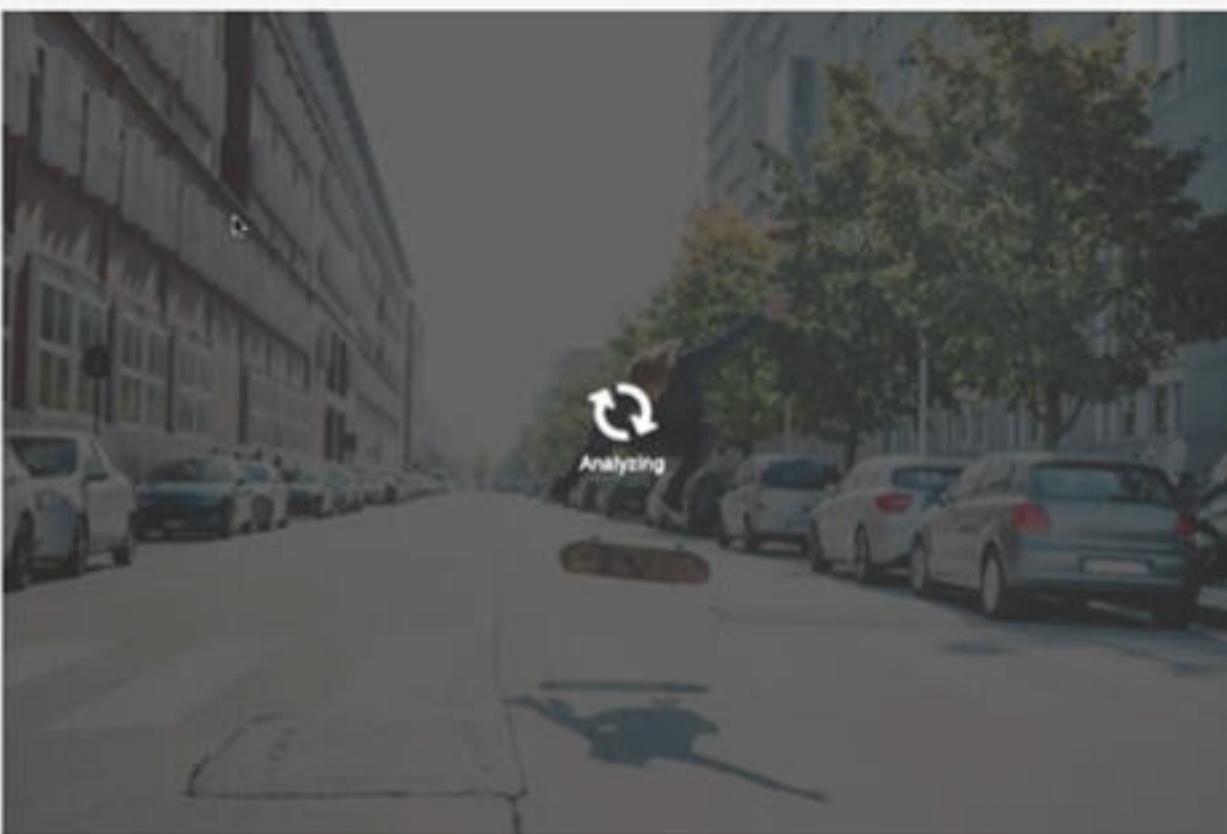
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Object and scene detection

Rekognition automatically labels objects, concepts and scenes in your images, and provides a confidence score. (Your images aren't stored.)



Choose a sample image



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▶ Response

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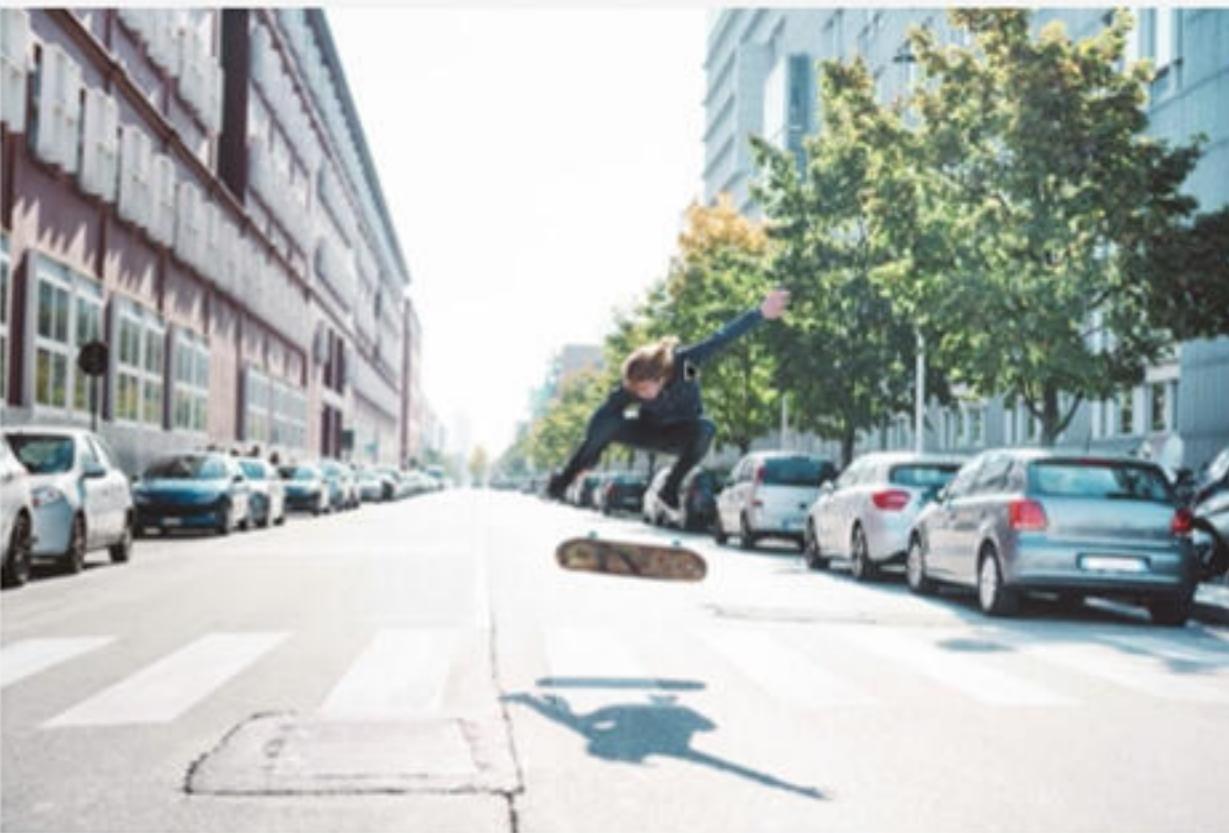
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Results

| | |
|------------|-------|
| Skateboard | 99.2% |
| Sport | 99.2% |
| People | 99.2% |
| Person | 99.2% |
| Human | 99.2% |
| Parking | 97.4% |

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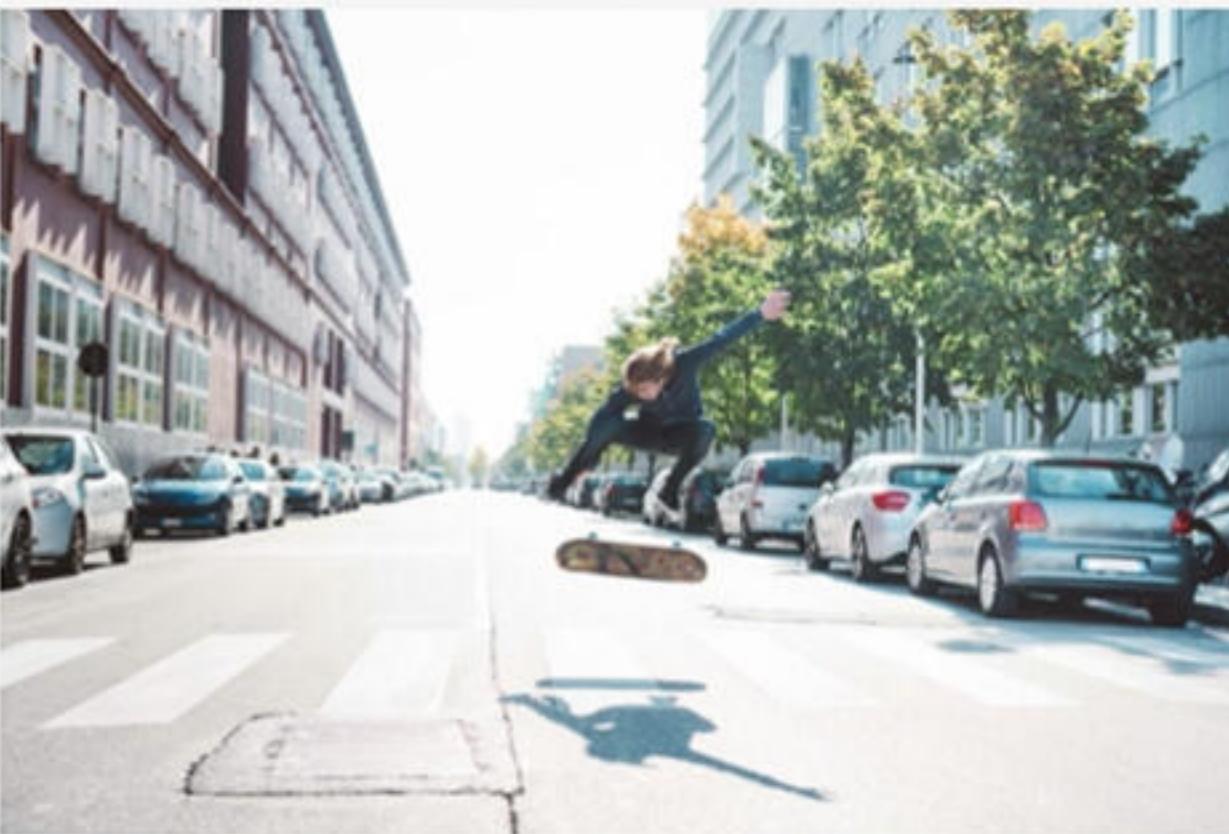
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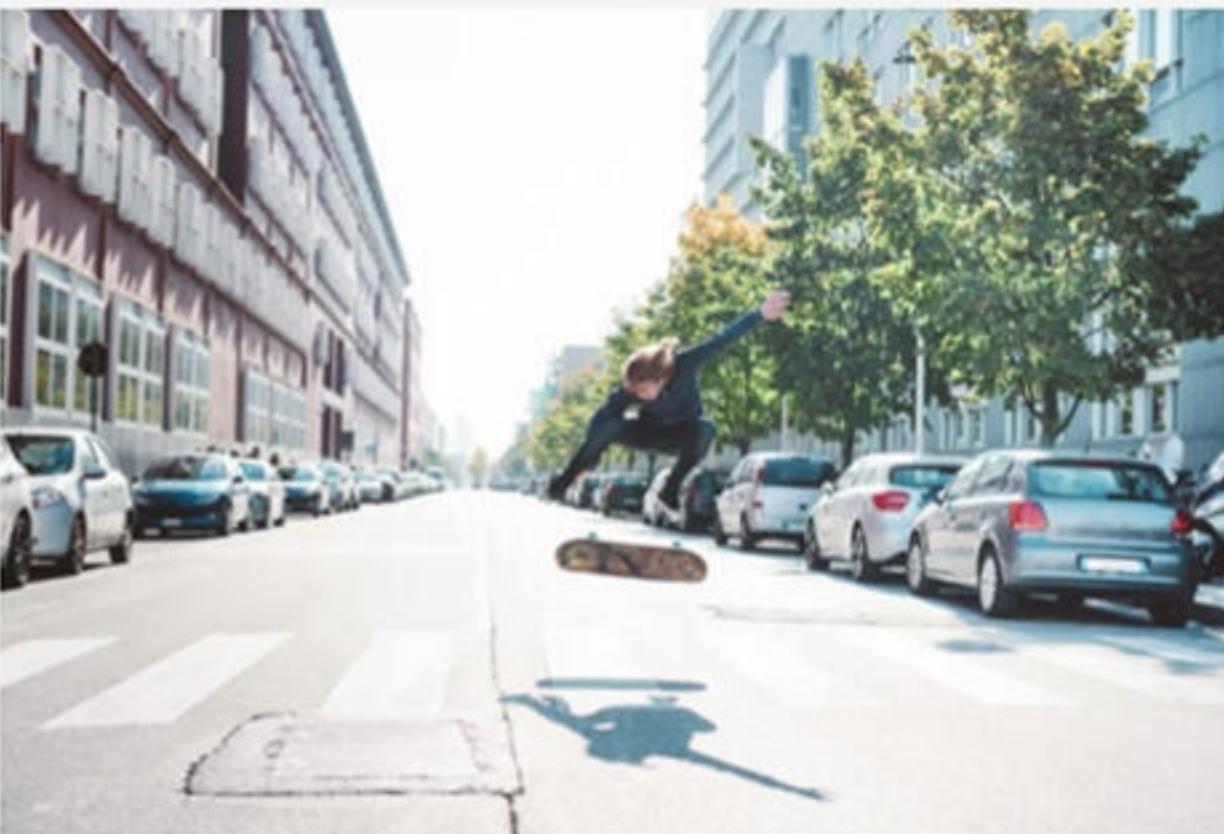
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| Parking Lot | 97.4% |
| Automobile | 91.5% |
| Car | 91.5% |
| Vehicle | 91.5% |
| Intersection | 76.8% |
| Road | 76.8% |
| Boardwalk | 76.2% |
| Path | 76.2% |
| Pavement | 76.2% |
| Sidewalk | 76.2% |
| Walkway | 76.2% |

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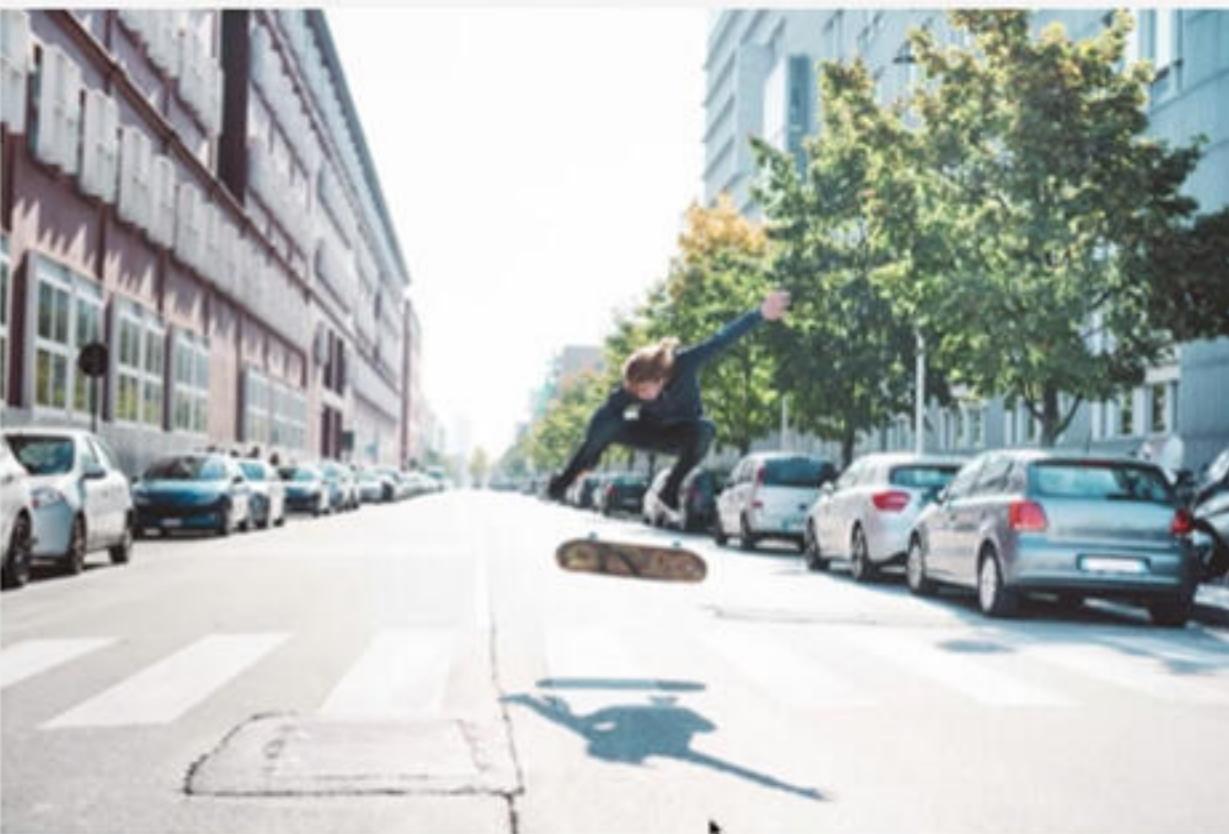
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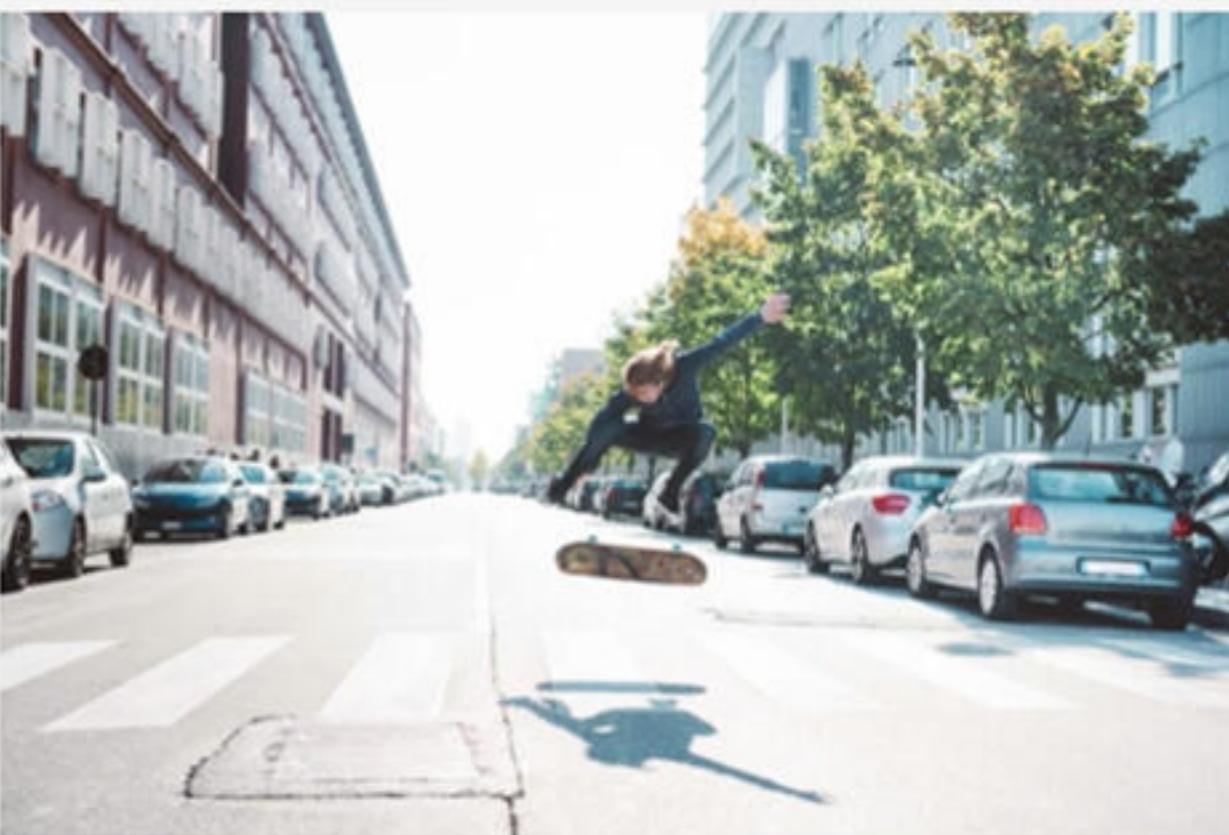
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| City | 98.3% |
| Downtown | 98.3% |
| High Rise | 98.3% |
| Skyscraper | 98.3% |
| Metropolis | 86% |

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Rekognition automatically labels objects, scenes, and faces.

Options

Format: Image Files

Cancel

Close



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Use your own image



or drag and drop

Use image URL:

Go

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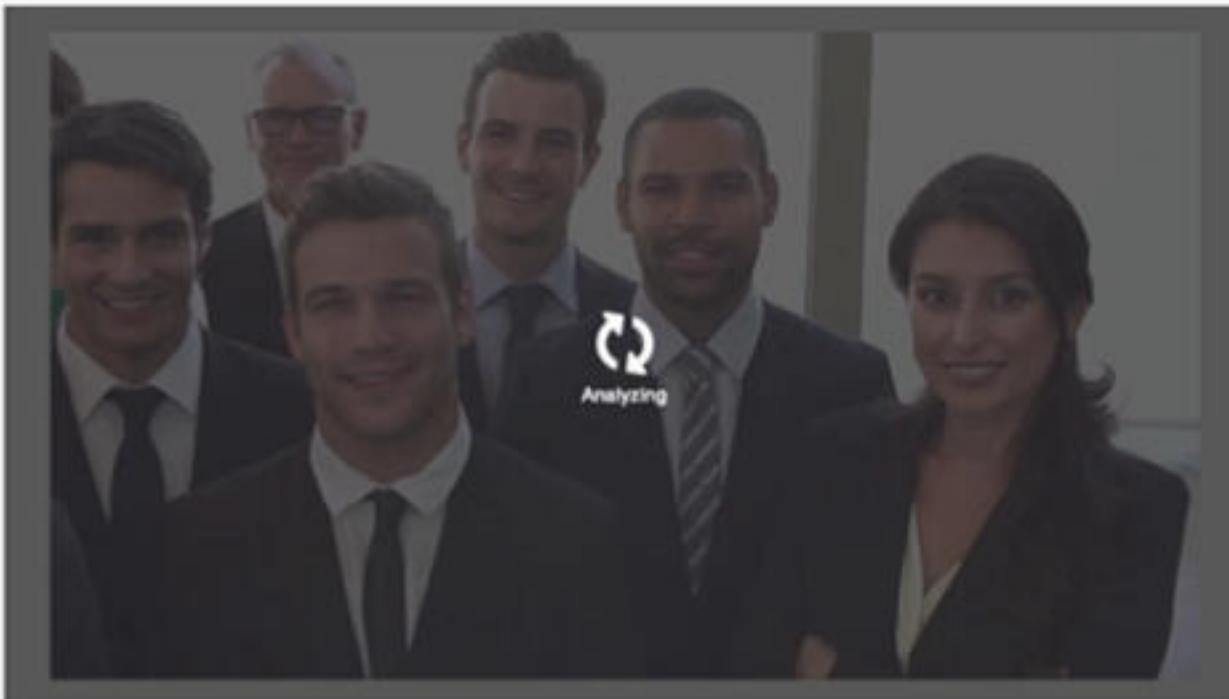
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Object and scene detection

Rekognition automatically labels objects, concepts and scenes in your images, and provides a confidence score. (Your images aren't stored.)



Choose a sample image



Use your own image

 Upload or drag and drop

Use image URL:

Go

Done with the demo?

[Download SDKs](#)

Results

| | |
|----------|-------|
| People | 99.2% |
| Person | 99.2% |
| Human | 99.2% |
| Clothing | 93.9% |
| Overcoat | 93.9% |
| Suit | 93.9% |

[Show more](#)

Request

Response

Amazon Rekognition

Metrics

Demos

Object and scene detection

Image moderation

Facial analysis

Celebrity recognition

Face comparison

Additional Resources

Getting started guide

Download SDKs

Developer resources

Pricing

FAQ

Forum

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[Show more](#)

Request

Response

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Developer resources

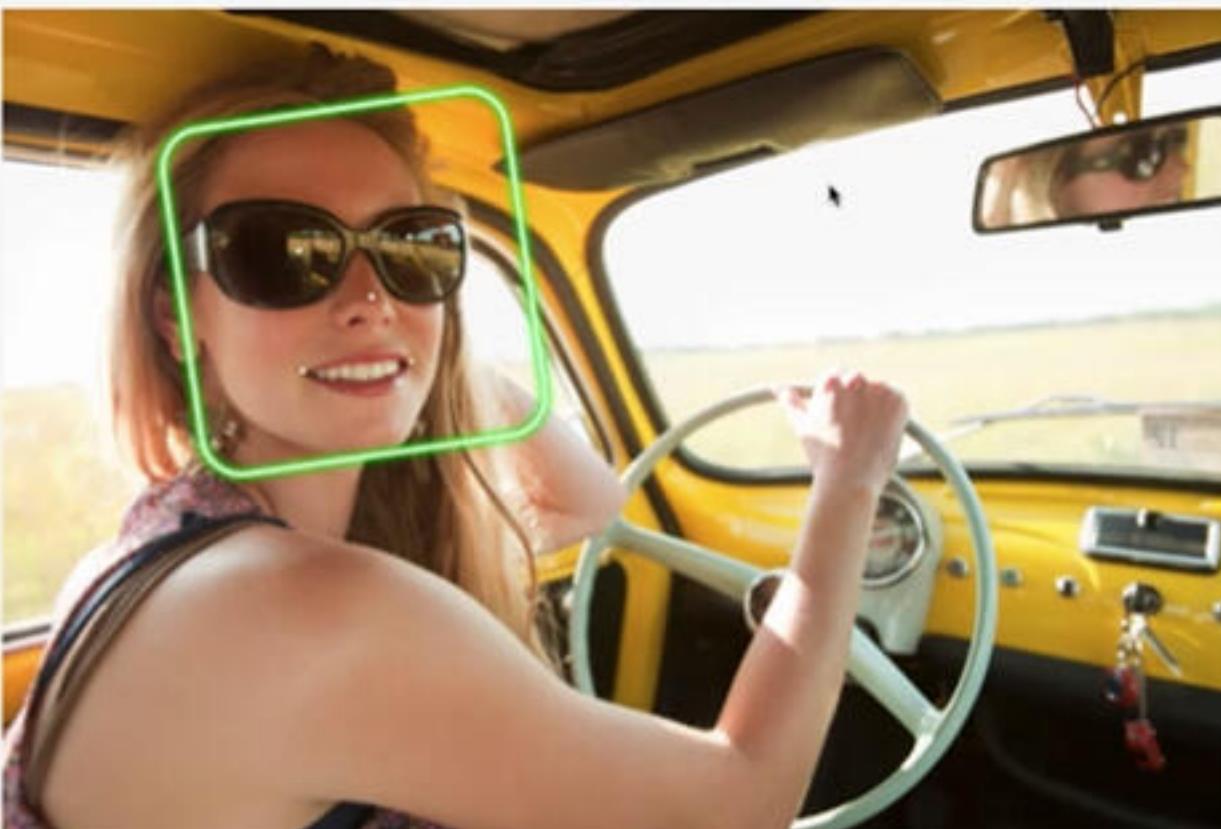
Pricing

FAQ

Forum

Facial analysis

Get a complete analysis of facial attributes, including confidence scores. (Your images aren't stored.)



Choose a sample image



Use your own image

Upload

or drag and drop

Use image URL

Go

Done with the demo?

[Download SDKs](#)

Results



| | |
|----------------------|-------------------|
| looks like a face | 99.8% |
| appears to be female | 100% |
| age range | 23 - 38 years old |
| smiling | 99.4% |
| appears to be happy | 93.2% |
| wearing glasses | 99.9% |

[Show more](#)

Request

Response

Amazon Rekognition

Metrics

Demos

Object and scene detection

Image moderation

Facial analysis

Celebrity recognition

Face comparison

Additional Resources

Getting started guide

Download SDKs

Developer resources

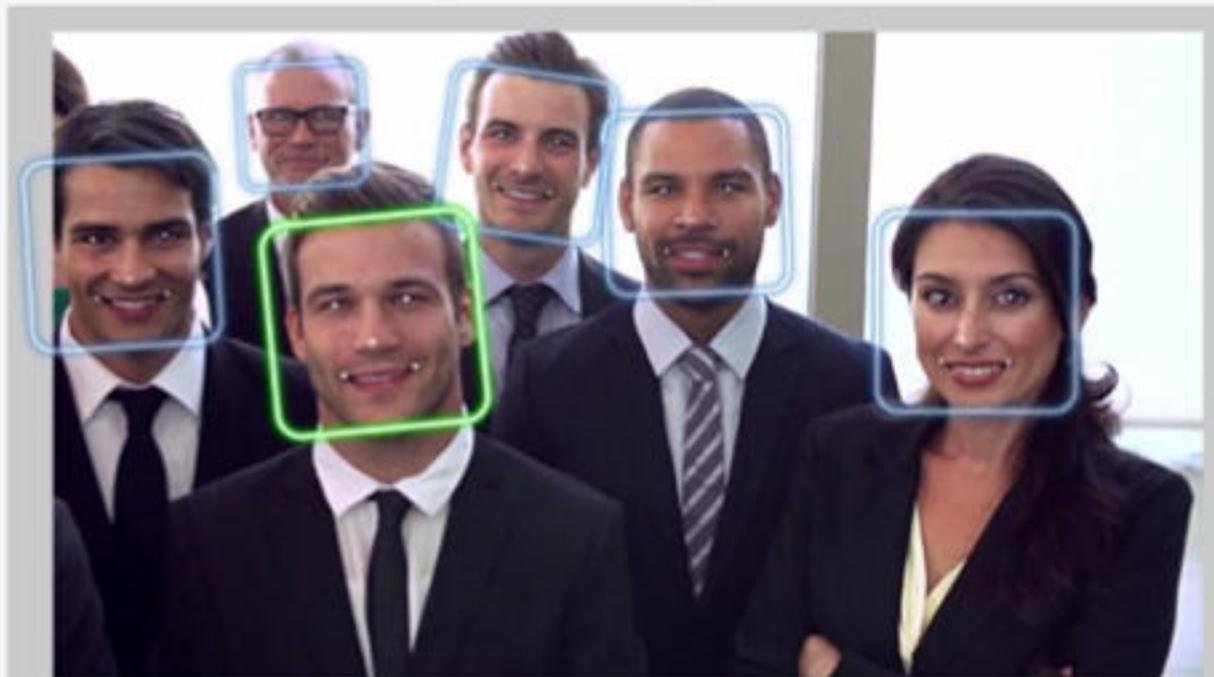
Pricing

FAQ

Forum

Facial analysis

Get a complete analysis of facial attributes, including confidence scores. (Your images aren't stored.)



Choose a sample image



Use your own image

or drag and drop

Done with the demo?

[Download SDKs](#)

Results



| | |
|---------------------|-------------------|
| looks like a face | 100% |
| appears to be male | 99.9% |
| age range | 26 - 43 years old |
| smiling | 97.4% |
| appears to be happy | 99.9% |
| not wearing glasses | 99.8% |

[Show more](#)

Request

Response

Amazon Rekognition

[Metrics](#)[Demos](#)[Object and scene detection](#)[Image moderation](#)[Facial analysis](#)**Celebrity recognition**[Face comparison](#)[Additional Resources](#)[Getting started guide](#)[Download SDKs](#)[Developer resources](#)[Pricing](#)[FAQ](#)[Forum](#)

Celebrity recognition

Rekognition automatically recognizes celebrities in images and provides confidence scores (your images aren't stored.)



Done with the demo?

[Download SDKs](#)

▶ Results

▶ Request

▶ Response

Choose a sample image



Use your own image

[Upload](#)

or drag and drop

[Use image URL](#)

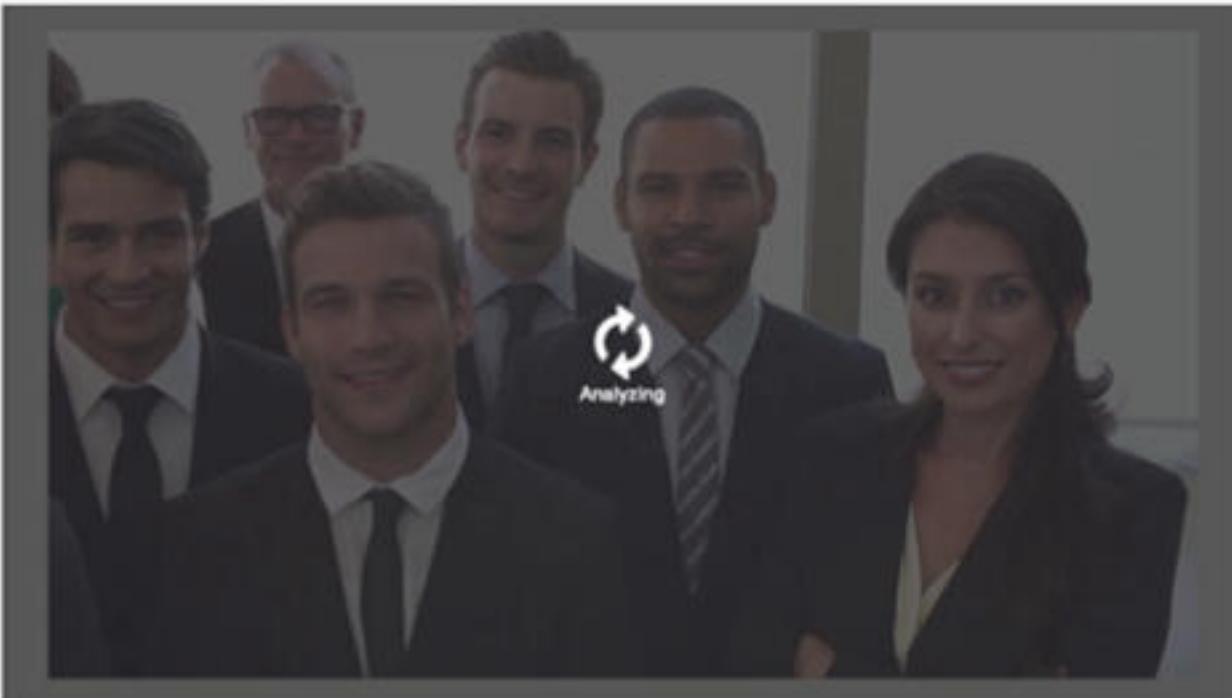
[Go](#)

Amazon Rekognition

[Metrics](#)[Demos](#)[Object and scene detection](#)[Image moderation](#)[Facial analysis](#)**Celebrity recognition**[Face comparison](#)[Additional Resources](#)[Getting started guide](#)[Download SDKs](#)[Developer resources](#)[Pricing](#)[FAQ](#)[Forum](#)

Celebrity recognition

Rekognition automatically recognizes celebrities in images and provides confidence scores (your images aren't stored.)

[Done with the demo?](#)[Download SDKs](#)[Results](#)[Request](#)[Response](#)[Choose a sample image](#)[Use your own image](#)

[Upload](#) or drag and drop

[Use image URL](#)[Go](#)

[Amazon Rekognition](#)[Metrics](#)[Demos](#)[Object and scene detection](#)[Image moderation](#)[Facial analysis](#)[Celebrity recognition](#)[Face comparison](#)[Additional Resources](#)[Getting started guide](#)[Download SDKs](#)[Developer resources](#)[Pricing](#)[FAQ](#)[Forum](#)

Celebrity recognition

Rekognition automatically recognizes celebrities in images and provides confidence scores (your images aren't stored.)

[Choose a sample image](#)[Use your own image](#) [Upload](#)

or drag and drop

[Use image URL](#)[Go](#)[Done with the demo?](#)[Download SDKs](#)

Results

No celebrity faces recognized

Request

Response



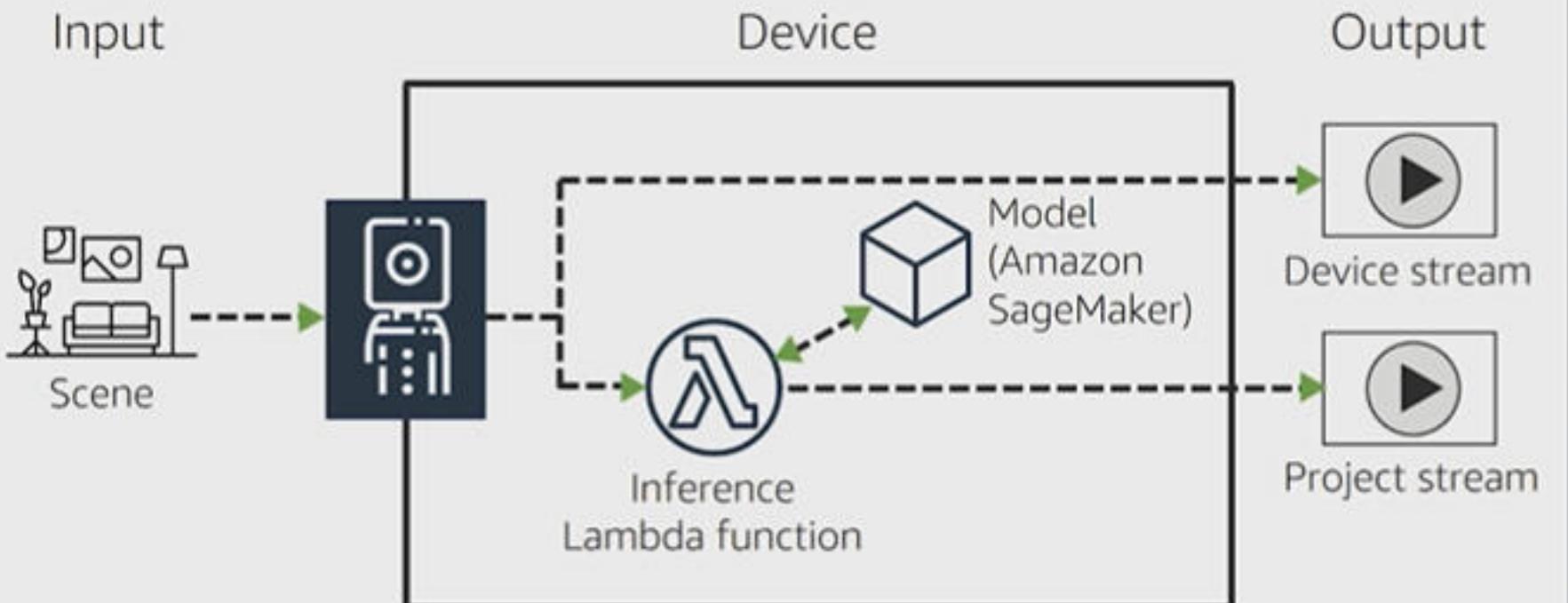
Thanks for watching!

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AWS DeepLens workflow

aws training and certification

Input

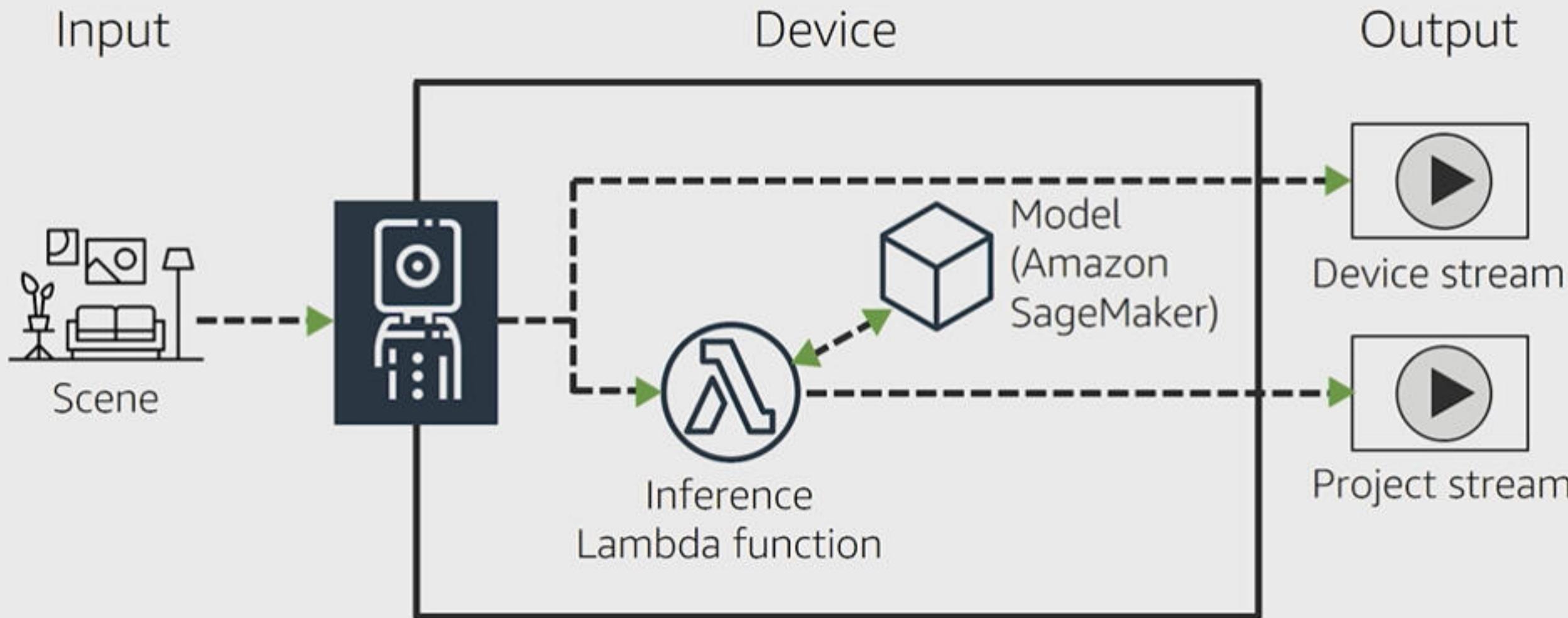


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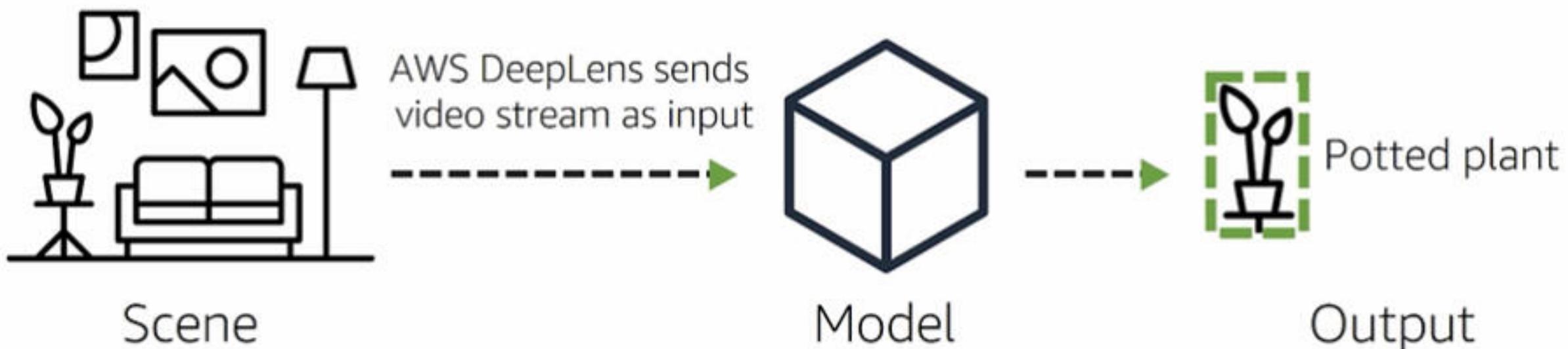


AWS DeepLens workflow

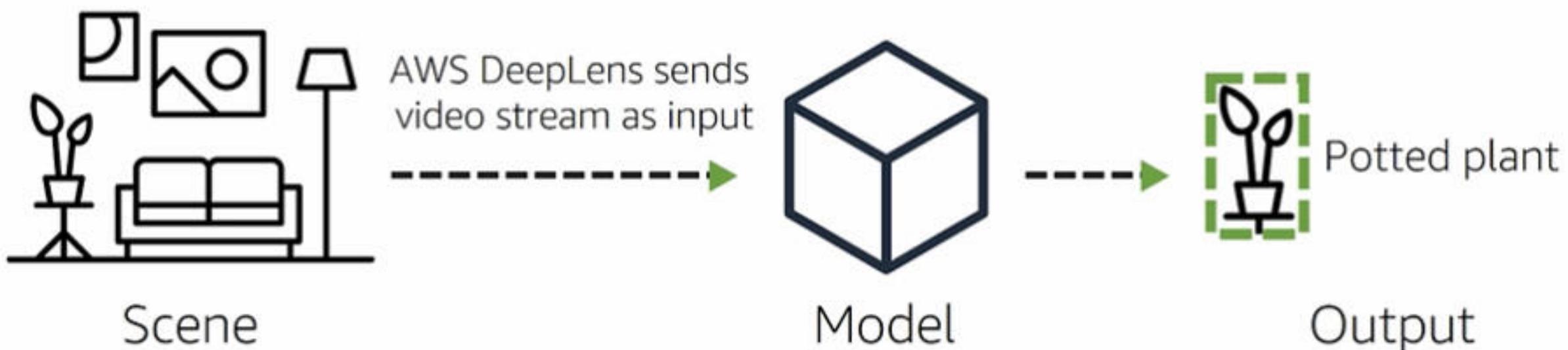
aws training and certification



Demo overview: Object Detection



Demo overview: Object Detection



AWS Management Console X + https://console.aws.amazon.com/console/home?region=us-east-1

aws Services Resource Groups 🔍 Jyothi N. Virginia Support

AWS services

Find a service by name or feature (for example, EC2, S3 or VM, storage) I

- Recently visited services
 - AWS DeepLens
 - IoT Core
 - S3
 - Lambda
 - Amazon SageMaker
- All services

Build a solution

Get started with simple wizards and automated workflows.

- Launch a virtual machine With EC2 ~2-3 minutes
- Build a web app With Elastic Beanstalk ~6 minutes
- Build using virtual servers With Lightsail ~1-2 minutes
- Connect an IoT device With AWS IoT ~5 minutes
- Start a development project With CodeStar ~5 minutes
- Register a domain With Route 53 ~3 minutes

See more See all ↗

Learn to build AWS Fargate Runs Containers for You

Helpful tips

Manage your costs Monitor your AWS costs, usage, and reservations using AWS Budgets. [Start now](#)

Create an organization Use AWS Organizations for policy-based management of multiple AWS accounts. [Start now](#)

Explore AWS

Machine Learning with Amazon SageMaker The fastest way to build, train, and deploy machine learning models. [Learn more ↗](#)

Amazon Relational Database Service (RDS) RDS manages and scales your database for you. RDS supports Aurora, MySQL, PostgreSQL, MariaDB, Oracle, and SQL Server. [Learn more ↗](#)

AWS Management Console https://console.aws.amazon.com/console/home?region=us-east-1

Services Resource Groups Jyothi N. Virginia Support

AWS services

deeplens

AWS DeepLens Deep Learning Enabled Video Camera

AWS DeepLens Lambda All services

IoT Core Amazon SageMaker

Helpful tips

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Build a solution

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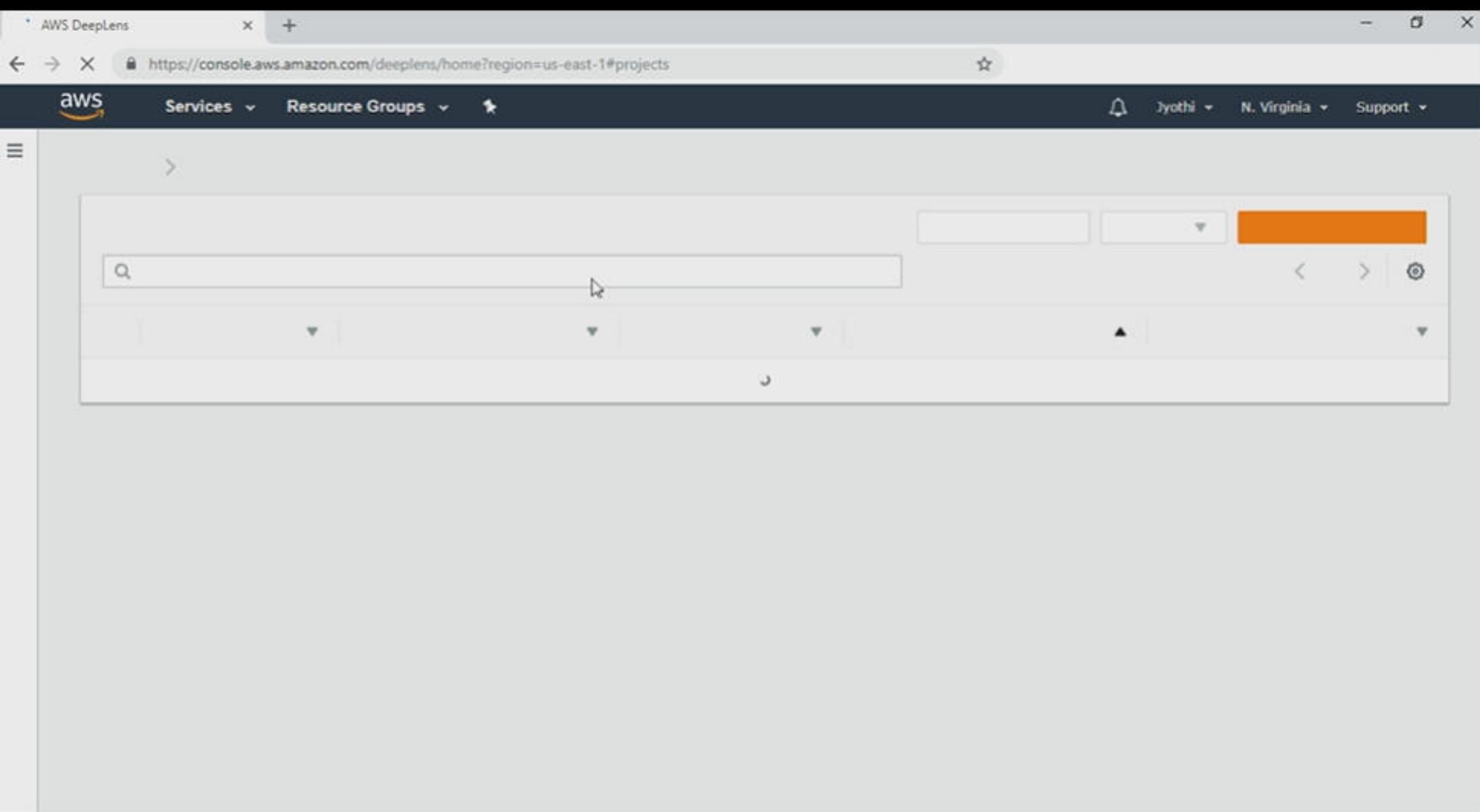
See more See all ↗

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AWS Fargate Runs Containers for You



AWS DeepLens

https://console.aws.amazon.com/deeplens/home?region=us-east-1#projects

aws Services Resource Groups

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AWS DeepLens Share your feedback

Resources Projects Devices Models Helpful links

DeepLens > Projects

Projects (0)

Deploy to device Actions Create new project

Search projects

Name Description Version Creation time Last updated

< 1 > ⌂

There are no projects in this account. To get started, choose Create a new project.

https://console.aws.amazon.com/deeplens/home?region=us-east-1#projects

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AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#projects

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DeepLens > Projects Share your feedback

Projects (0)

Search projects

Deploy to device Actions Create new project

Name Description Version Creation time Last updated

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Feedback English (US)

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AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#projects/create

aws Services Resource Groups

DeepLens > Projects > Create project Share your feedback

Step 1 Choose project type

Step 2 Specify project details

Choose project type

Project type

Choose an option

Use a project template
Test a preconfigured project to deploy a solution quickly, or customize the templates for your own use.

Create a new blank project
Choose models and functions or create new logic for a custom use case.

Project templates





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Choose an option

 Use a project template

Test a preconfigured project to deploy a solution quickly, or customize the templates for your own use.

 Create a new blank project

Choose models and functions or create new logic for a custom use case.

Project templates



Object detection

Detect 20 popular objects.



Artistic style transfer

Make your surroundings look like Van Gogh's paintings.



AWS DeepLens

https://console.aws.amazon.com/deeplens/home?region=us-east-1#projects/create

aws Services Resource Groups Jyothi N. Virginia Support



Cat and dog recognition
Locate and recognize your cat or dog.



Action recognition
Recognize more than 30 kinds of actions.



Head Pose detection
Detect 9 different head pose angles

Cancel **Next**

AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#projects/create

Services Resource Groups

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```
graph LR; Objects[Objects] --> Inference[Lambda function]; Inference --> Model[Object recognition model]; Model --> DeviceStream[Device stream]; Model --> ProjectStream[Project stream]
```

The sample project uses the [Single Shot MultiBox detector \(SSD\)](#) framework for object detection with pre-trained resnet_50 network. The network has been trained on the [Pascal VOC](#) dataset and is capable of recognizing 20 different kinds of objects. The model takes the video stream from your AWS DeepLens as input and labels it with identified objects. To get you started, we have prefilled this sample project with a pretrained optimized model ready to deploy to AWS DeepLens. After the model is deployed you can enjoy watching AWS DeepLens recognize objects around you.

Project content

We automatically associate a model and Lambda function with your project template. You can associate another model or function later.

Cancel Previous Create

AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#projects/create

Services Resource Groups

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```
graph LR; Objects[Objects] --> Inference[Lambda function]; Inference --> Model[Object recognition model]; Model --> DeviceStream[Device stream]; Model --> ProjectStream[Project stream]
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Create

AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#projects

aws Services Resource Groups

AWS DeepLens Share your feedback

Resources Projects Devices Models Helpful links

DeepLens > Projects

Projects (1)

Deploy to device Actions Create new project

Search projects < 1 > ⌂

| Name | Description | Version | Creation time | Last updated |
|------------------|---------------------------|---------|------------------------|------------------------|
| Object-detection | Detect 20 popular objects | | 11/12/2018, 3:51:35 PM | 11/12/2018, 3:51:35 PM |

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AWS DeepLens Share your feedback

Resources Projects Devices Models Helpful links

DeepLens > Projects

Projects (1)

Deploy to device Actions Create new project

Search projects < 1 > ⌂

| Name | Description | Version | Creation time | Last updated |
|------------------|---------------------------|---------|------------------------|------------------------|
| Object-detection | Detect 20 popular objects | | 11/12/2018, 3:51:35 PM | 11/12/2018, 3:51:35 PM |

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AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#project-deploy/Object-detection

aws Services Resource Groups

DeepLens > Projects > Object-detection > Deploy project Share your feedback

Step 1 Target device Step 2 Review and deploy

Target device

Choose the device you want to deploy your project to.

| Devices (1) | | | | |
|-------------|---------|---------------------|------------------|------------------------|
| Name | Project | Registration status | Device status | Creation time |
| demo | - | Registered | Update available | 11/12/2018, 2:31:34 PM |

Cancel Review

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10 07:06 / 08:40

AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#project-deploy/Object-detection

aws Services Resource Groups

DeepLens > Projects > Object-detection > Deploy project Share your feedback

Step 1 Target device Step 2 Review and deploy

Target device

Choose the device you want to deploy your project to.

| Devices (1) | | | | |
|-------------|---------|---------------------|------------------|------------------------|
| Name | Project | Registration status | Device status | Creation time |
| demo | - | Registered | Update available | 11/12/2018, 2:31:34 PM |

Cancel Review

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AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#project-deploy/Object-detection

aws Services Resource Groups

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DeepLens > Projects > Object-detection > Deploy project Share your feedback

Step 1 Target device

Step 2 Review and deploy

Review and deploy

Deployment check

AWS DeepLens will deploy the project below to your device. Choose Deploy to continue.

New project: Object-detection

| Type | Name |
|----------|--------------------------------------|
| Function | deeplens-object-detection/versions/4 |
| Model | deeplens-object-detection |

 Deployment will incur costs
AWS DeepLens uses various services to help deploy a project to your device. Costs will be aggregated and itemized for review in AWS Billing.

Cancel Previous Deploy

AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#device-details/demo

AWS Services Resource Groups

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AWS DeepLens

DeepLens > Devices > demo

Share your feedback

demo

Edit Deregister

Resources

Projects

Devices

Models

Helpful links

Device status

| | | |
|---------------------|------------------------|----------------|
| Registration status | Device status | Device version |
| Registered | Deployment in progress | 1.3.19 |

Current project

Remove project Deploy a project

You are now ready to deploy a machine learning project. Choose **Deploy a project** to get started.

Project output

You can use AWS IoT console to view a JSON-formatted output of the project deployed to your AWS DeepLens device. [Info](#)

To view MQTT messages:

1. Copy the topic that is unique to your registered device: `$aws/things/deplens_xSDWZfrQJ6Fk30PTVM0fQ/infer` [Copy](#)
2. Go to [AWS IoT console](#), paste the topic in the Subscription topic input field and choose **Subscribe to topic**.

View the video output

Waiting for console.aws.amazon.com... (1/1)

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AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#device-details/demo

aws Services Resource Groups

AWS DeepLens

C Deployment of project Object-detection, version 0 is in progress.
11/12/2018, 4:01:56 PM: Creating a Greengrass deployment

Resources Projects Devices Models

DeepLens > Devices > demo Share your feedback

demo Edit Deregister

Device status

| | | |
|---------------------|------------------------|----------------|
| Registration status | Device status | Device version |
| Registered | Deployment in progress | 1.3.19 |

Current project Remove project Deploy a project

You are now ready to deploy a machine learning project. Choose Deploy a project to get started.

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To view MQTT messages:

AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#device-details/demo

AWS Services Resource Groups

AWS DeepLens

Current project

Name: Object-detection Description: Detect 20 popular objects Version: -

Remove project Deploy a project

Project output

You can use AWS IoT console to view a JSON-formatted output of the project deployed to your AWS DeepLens device. [Info](#)

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View the video output

Device details

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AWS DeepLens

https://console.aws.amazon.com/deplens/home?region=us-east-1#device-details/demo

AWS Services Resource Groups

AWS DeepLens

Current project

Name: Object-detection Description: Detect 20 popular objects Version: -

Remove project Deploy a project

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1. Copy the topic that is unique to your registered device: [Copy](#)
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View the video output

Device details

Update

https://console.aws.amazon.com/iotv2/home?region=us-east-1#/test

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AWS DeepLens X AWS IoT X +

https://console.aws.amazon.com/iotv2/home?region=us-east-1#/test

aws Services Resource Groups

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AWS IoT

MQTT client ?

Connected to Device Gateway on client ID 'iotconsole-1542067400302-0'.

Subscriptions

Subscribe to a topic

Publish to a topic

Subscribe

Devices publish MQTT messages on topics. You can use this client to subscribe to a topic and receive these messages.

Subscription topic

Specify a topic Subscribe to topic

object_detection_rule

Saws/things/deeplens_OAX0A_635-eeRtYYeTVuWg/infer

Quality of Service ?

0 - This client will not acknowledge to the Device Gateway that messages are received

1 - This client will acknowledge to the Device Gateway that messages are received

MQTT payload display

Auto-format JSON payloads (improves readability)

Display payloads as strings (more accurate)

AWS DeepLens X AWS IoT X +

https://console.aws.amazon.com/iotv2/home?region=us-east-1#/test

aws Services Resource Groups

AWS IoT Jyothi N. Virginia Support

MQTT client ?

Connected as iotconsole-...

Monitor Onboard Manage Greengrass Secure Defend Act Test

Subscriptions

Subscribe to a topic

Publish to a topic

Subscribe
Devices publish MQTT messages on topics. You can use this client to subscribe to a topic and receive these messages.

Subscription topic
 Subscribe to topic

Max message capture ?
100

Quality of Service ?
 0 - This client will not acknowledge to the Device Gateway that messages are received
 1 - This client will acknowledge to the Device Gateway that messages are received

MQTT payload display
 Auto-format JSON payloads (improves readability)
 Display payloads as strings (more accurate)

AWS DeepLens AWS IoT

https://console.aws.amazon.com/iotv2/home?region=northamerica-east-1#/test

Services ▾ Resource Groups ▾

AWS IoT

Monitor

Onboard

Manage

Greengrass

Secure

Defend

Act

Test

Sensors/Things/deeplearning-SD... ▾

Publish to a topic

Specify a topic and a message to publish with a QoS of 0.

Topic: Sensors/Things/deeplearning-SD.../test
Message: {"message": "Hello From AWS IoT console"}

Publish to topic

Export Hide

{
 "pottedplant": 0.2724689375,
 "tvmonitor": 0.487548828125
}

Export Hide

{
 "tvmonitor": 0.3388671875
}

Export Hide

Feedback English (US)

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AWS DeepLens AWS IoT

https://console.aws.amazon.com/iotv2/home?region=us-east-1#/test

Services Resource Groups

AWS IoT

Monitor

Onboard

Manage

Greengrass

Secure

Defend

Act

Test

Publish to a topic

Saws/things/deeplens_-x5D... X

Specify a topic and a message to publish with a QoS of 0.

1 {
2 "message": "Hello from AWS IoT console"
3 }

Publish to topic

Saws/things/deeplens_-x5D... Nov 12, 2018 4:15:28 PM -0800 Export Hide

{
 "pottedplant": 0.377197265625,
 "tvmonitor": 0.26123046875
}

Saws/things/deeplens_-x5D... Nov 12, 2018 4:15:27 PM -0800 Export Hide

{
 "pottedplant": 0.353515625,
 "bottle": 0.328125
}

Saws/things/deeplens_-x5D... Nov 12, 2018 4:15:27 PM -0800 Export Hide

AWS DeepLens AWS IoT

https://console.aws.amazon.com/iotv2/home?region=us-east-1#/test

Services Resource Groups

AWS IoT

Monitor

Onboard

Manage

Greengrass

Secure

Defend

Act

Test

Publish to a topic

Saws/things/deeplens_-x5D... X

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1 {
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3 }

Publish to topic

Saws/things/deeplens_-x5D... Nov 12, 2018 4:15:28 PM -0800 Export Hide

{
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}

Saws/things/deeplens_-x5D... Nov 12, 2018 4:15:27 PM -0800 Export Hide

{
 "pottedplant": 0.353515625,
 "bottle": 0.328125
}

Saws/things/deeplens_-x5D... Nov 12, 2018 4:15:27 PM -0800 Export Hide



Services

Resource Groups



Jyothi

N. Virginia

Support



AWS IoT

Monitor

Onboard

Manage

Greengrass

Secure

Defend

Act

Test

Software

Settings

Learn

Publish to a topic

\$aws/things/deeplens_-x5D... X

Specify a topic and a message to publish with a QoS of 0.

```
1 {  
2   "message": "Hello from AWS IoT console"  
3 }
```

Publish to topic**\$aws/things/deeplens_-x5D... Nov 12, 2018 4:17:40 PM -0800 Export Hide**{
 "bottle": 0.29541015625
}**\$aws/things/deeplens_-x5D... Nov 12, 2018 4:17:39 PM -0800 Export Hide**{
 "person": 0.289794921875
}**\$aws/things/deeplens_-x5D... Nov 12, 2018 4:17:39 PM -0800 Export Hide**{
 "bottle": 0.1972265625

Thank You

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Introduction to Semantic Segmentation



Introduction to Semantic Segmentation

In This Course



- Computer Vision & Localization
- Semantic Segmentation
 - Deep Learning Problem
 - Architecture
 - Practical Application



Amazon AI

Things to Know



- Image processing
- Computer Vision Basics
- Deep Learning Fundamentals
- Convolutional Neural Networks

Computer Vision

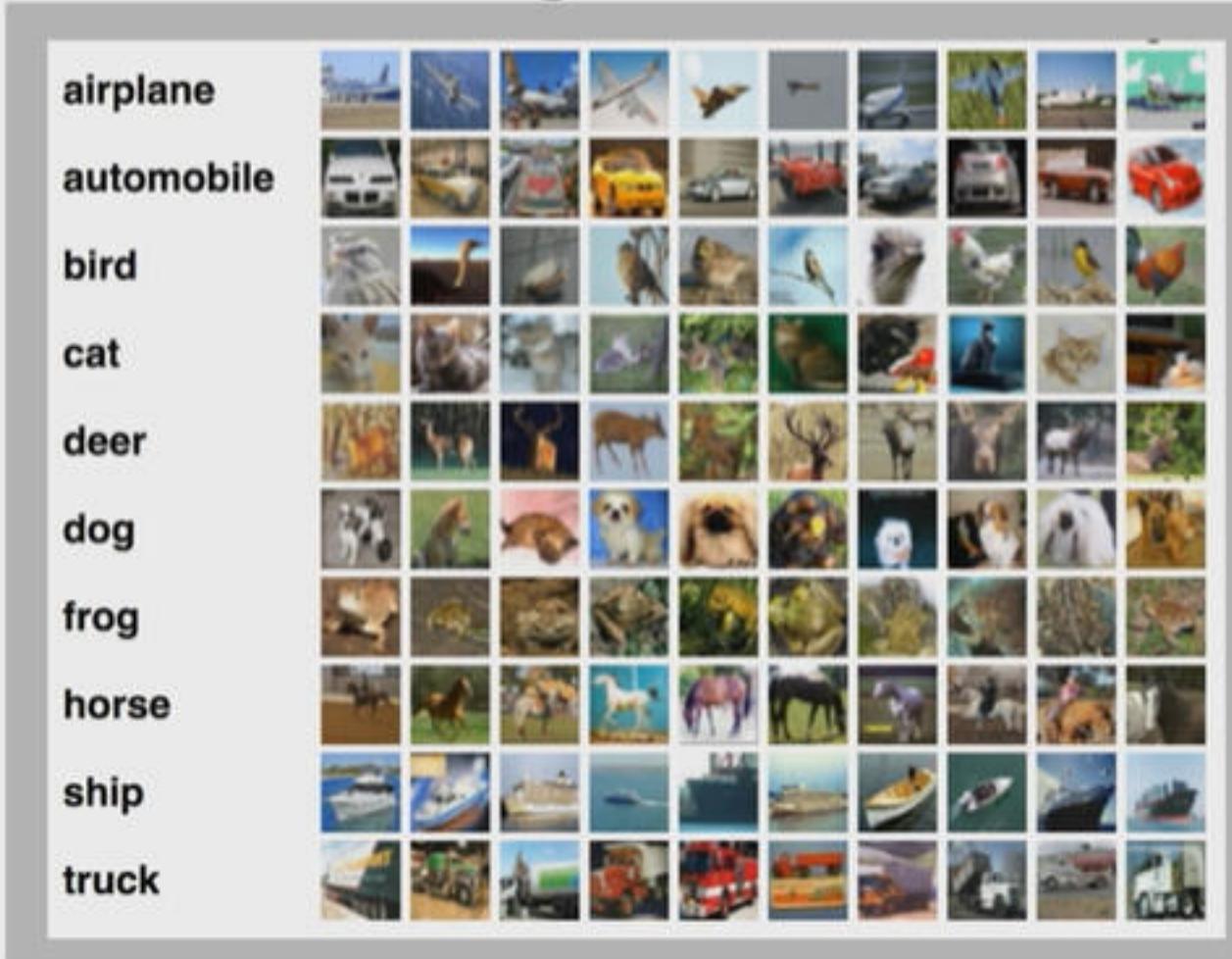


“Acquiring, processing, analyzing and understanding digital images”

Computer Vision Contd.



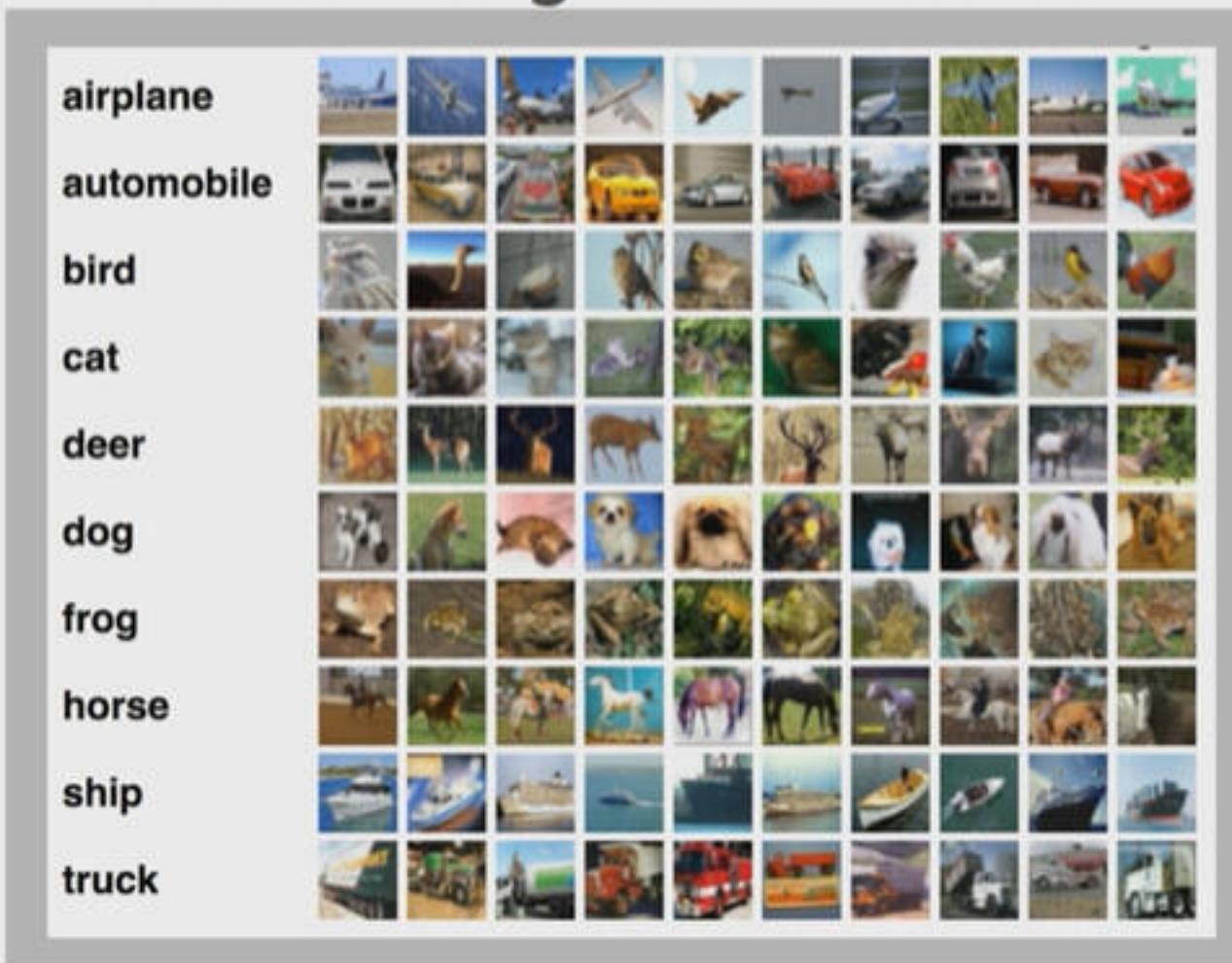
Global Image Classification



Computer Vision Contd.



Global Image Classification



Computer Vision Contd.



Image classification...



...usually not enough

Computer Vision Contd.

aws training and certification

Image classification...



...usually not enough

Computer Vision Contd.



Image classification...



...usually not enough

Localization Techniques



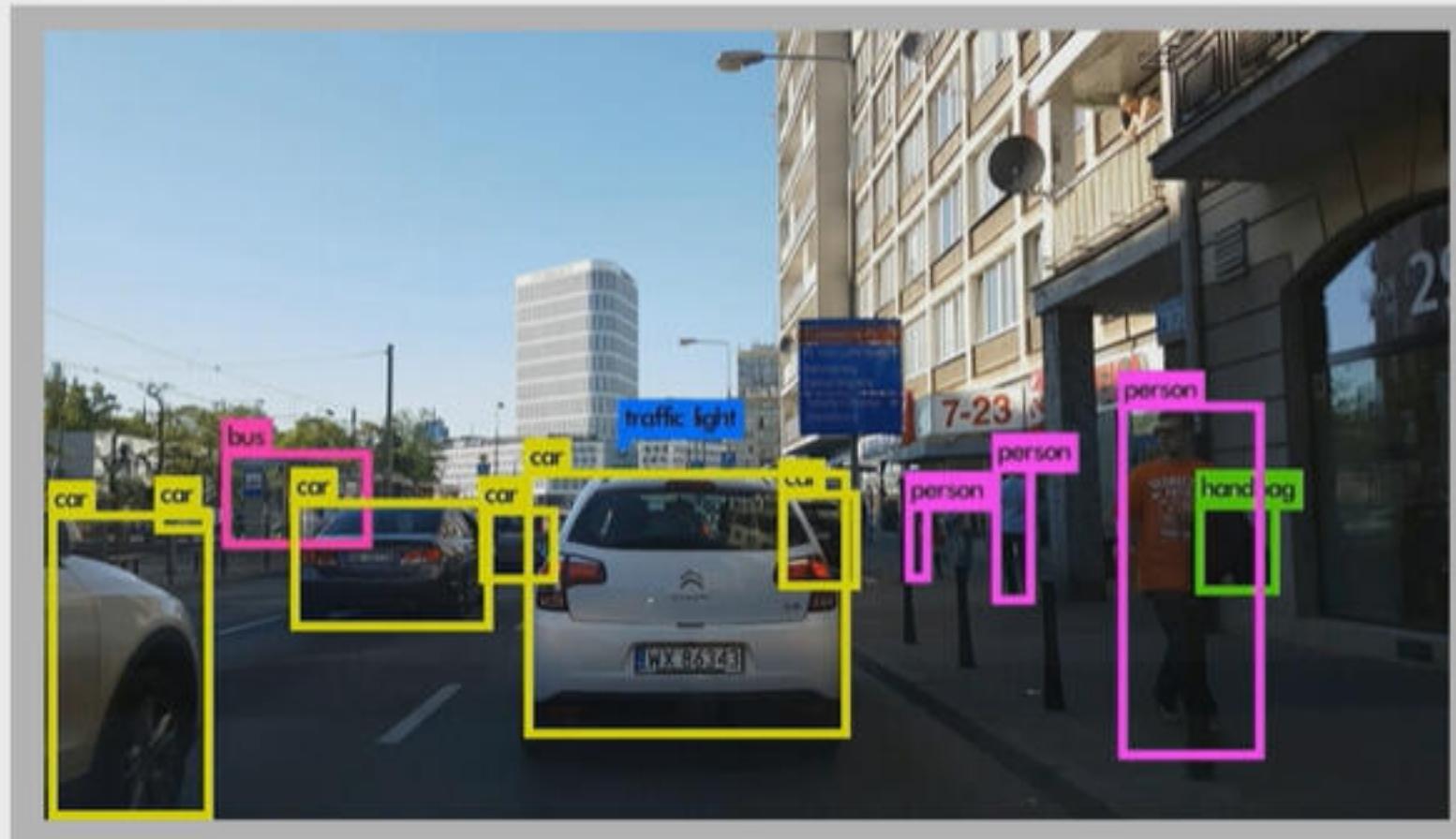
Tells us local information about an image:

- What?
- Where?
- How many?

Course Localization

aws training and certification

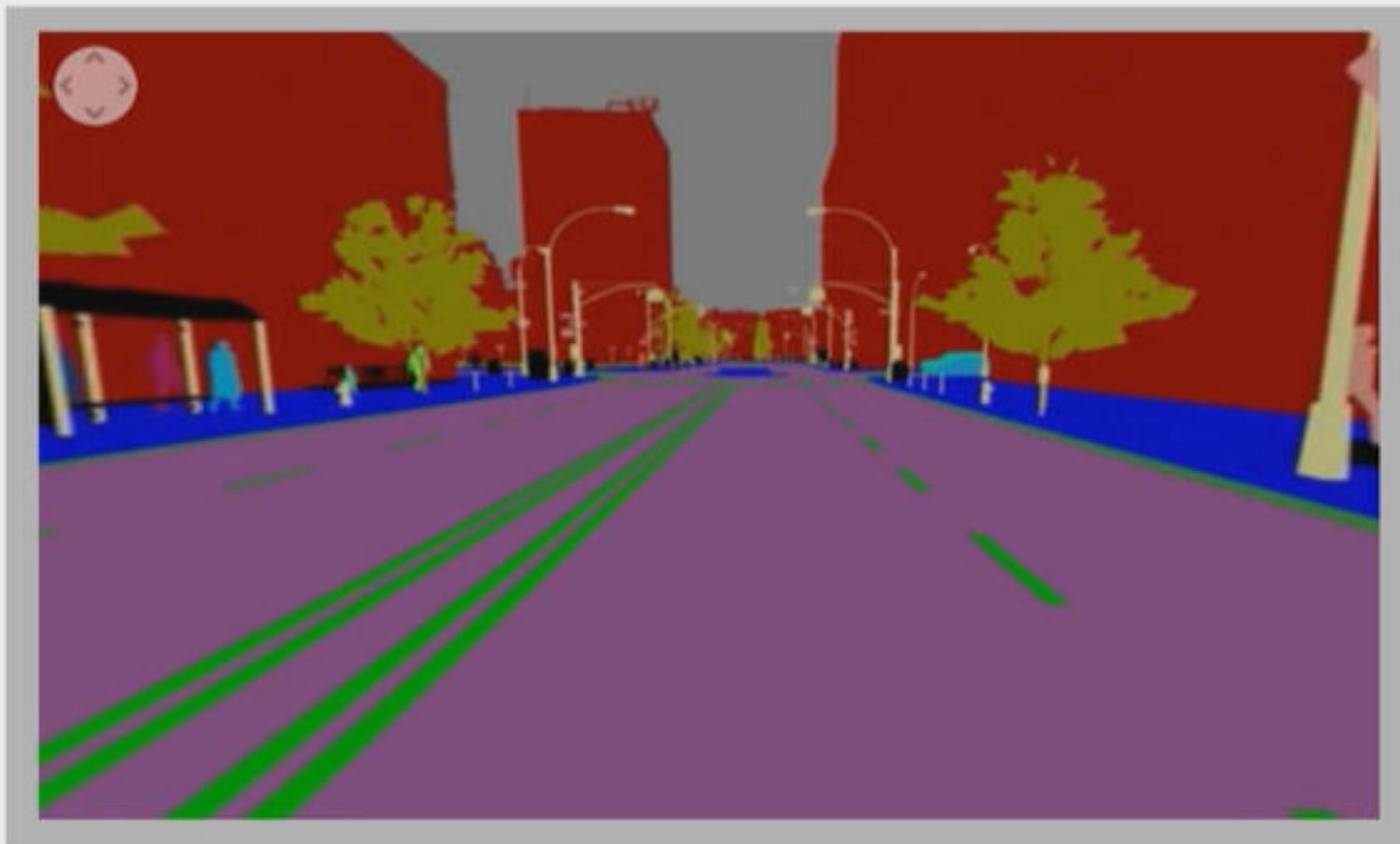
Object Detection



Fine Localization

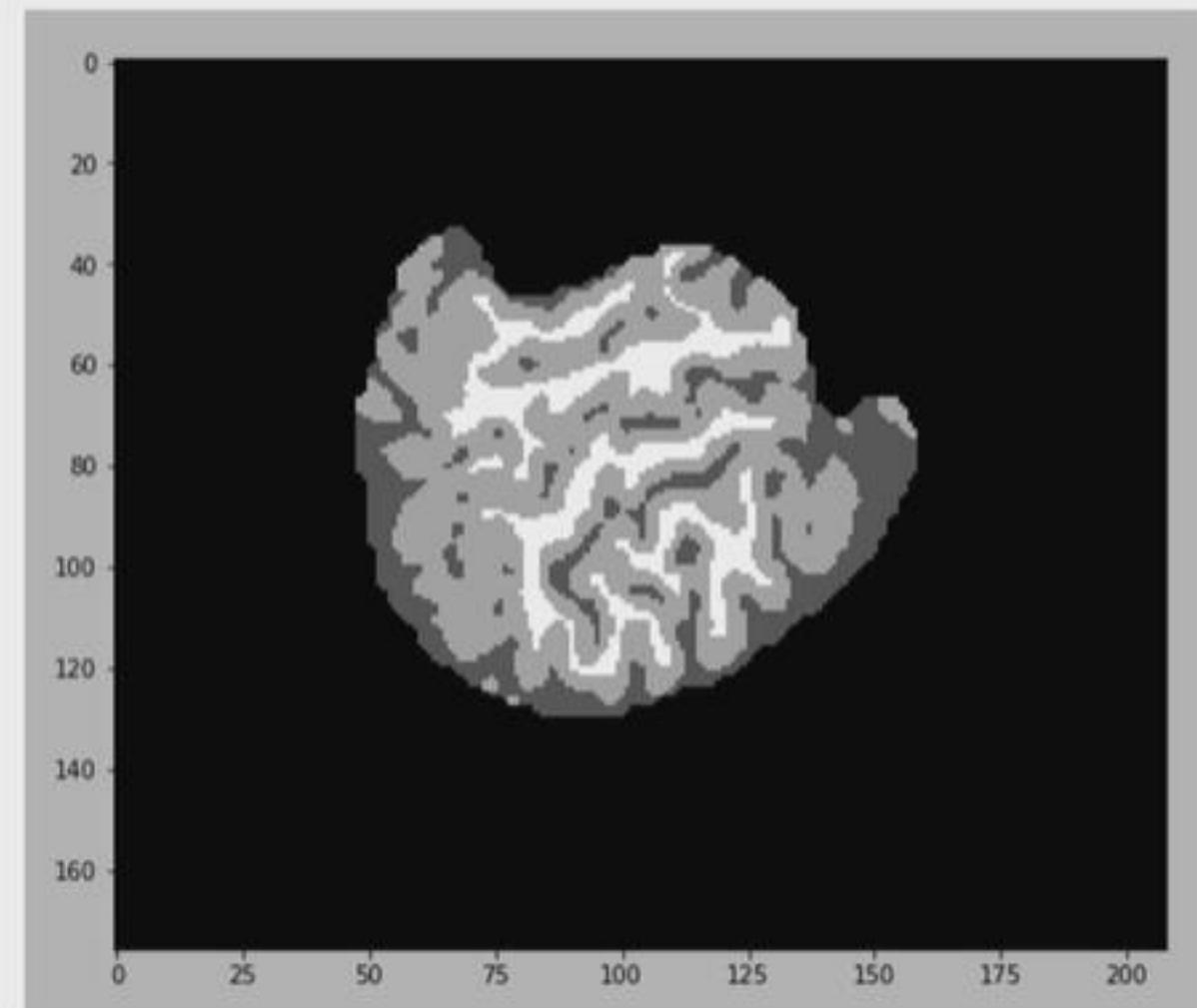
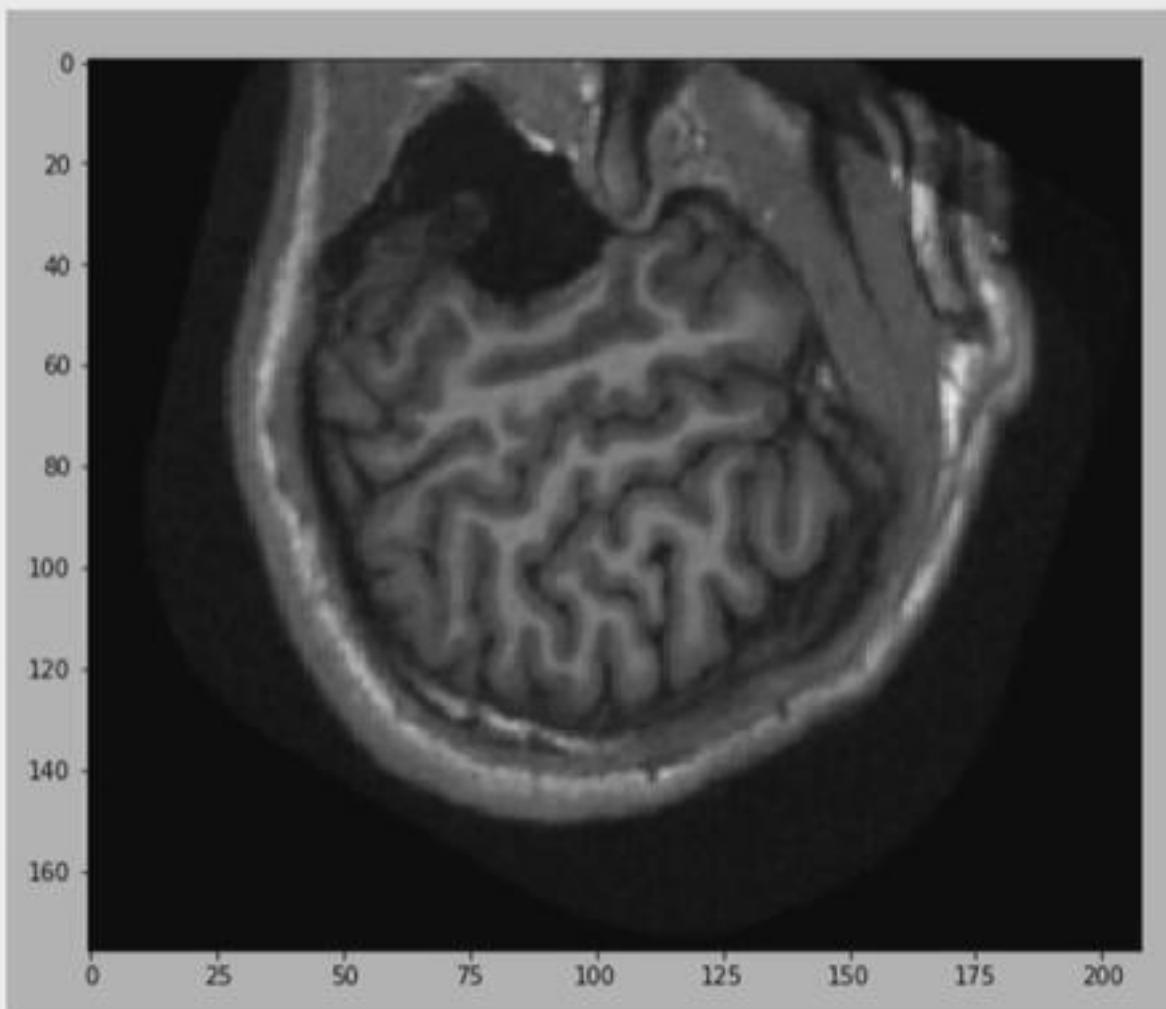
aws training and certification

Semantic Segmentation



Localization Techniques

aws training and certification



Semantic Segmentation and Deep Learning

Semantic Segmentation



Anatomy of a Deep Learning Task:

- Task
- Data
- Architecture
- Loss

Per-pixel Classification

aws training and certification



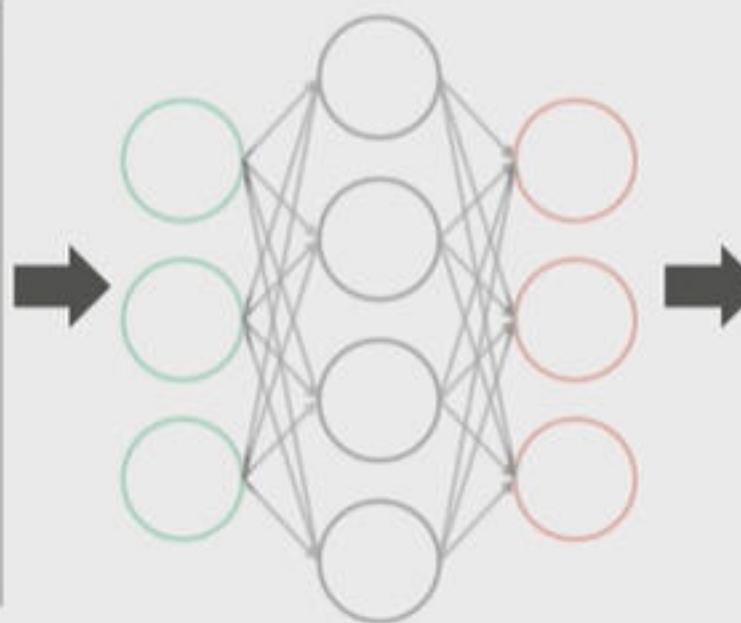
Convolutional Neural Networks

aws training and certification

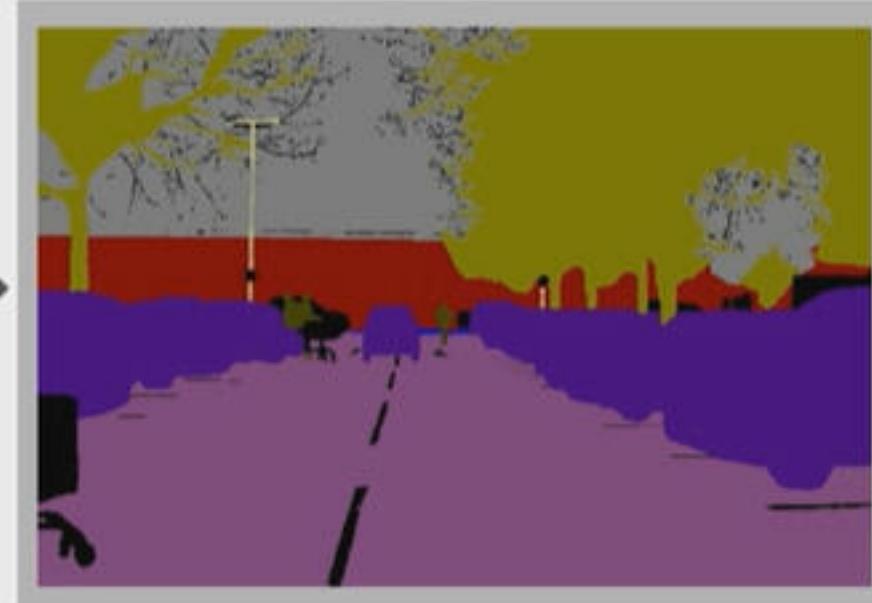
Input



CNN



Output



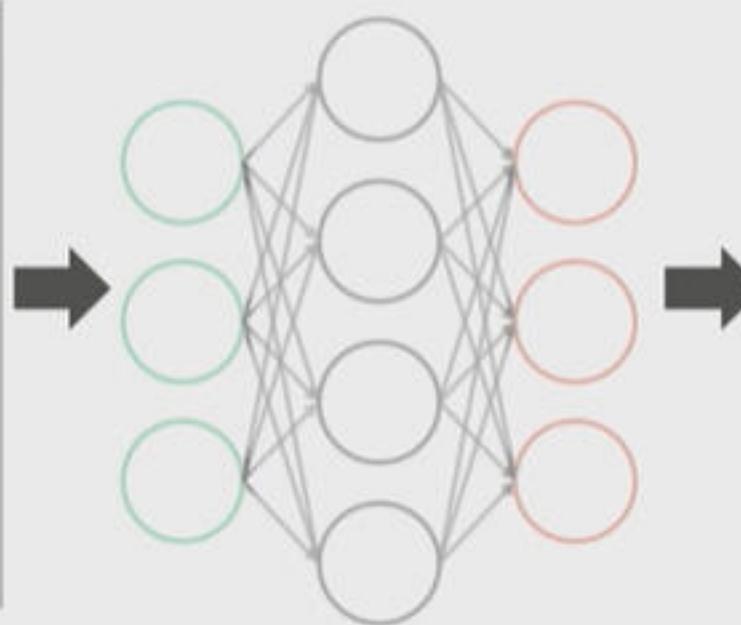
Convolutional Neural Networks

aws training and certification

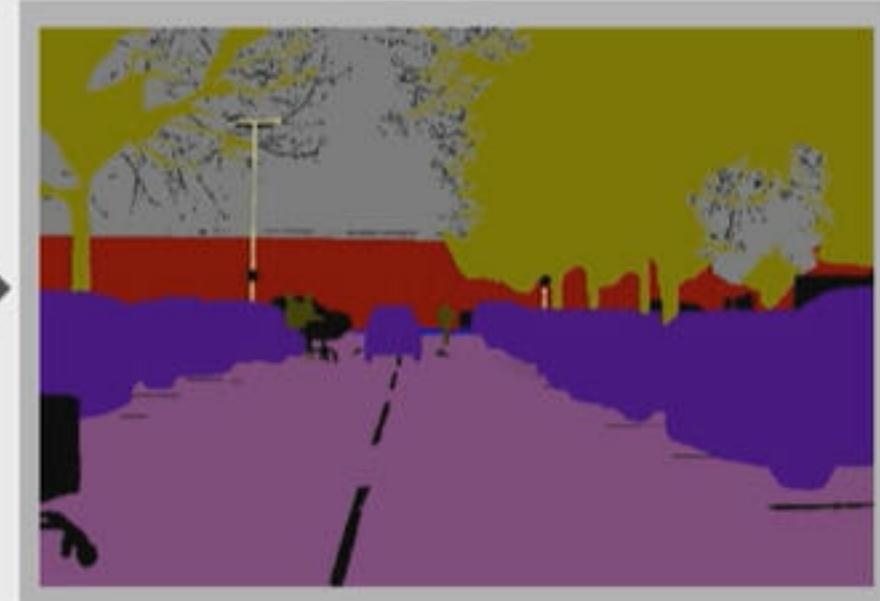
Input



CNN



Output



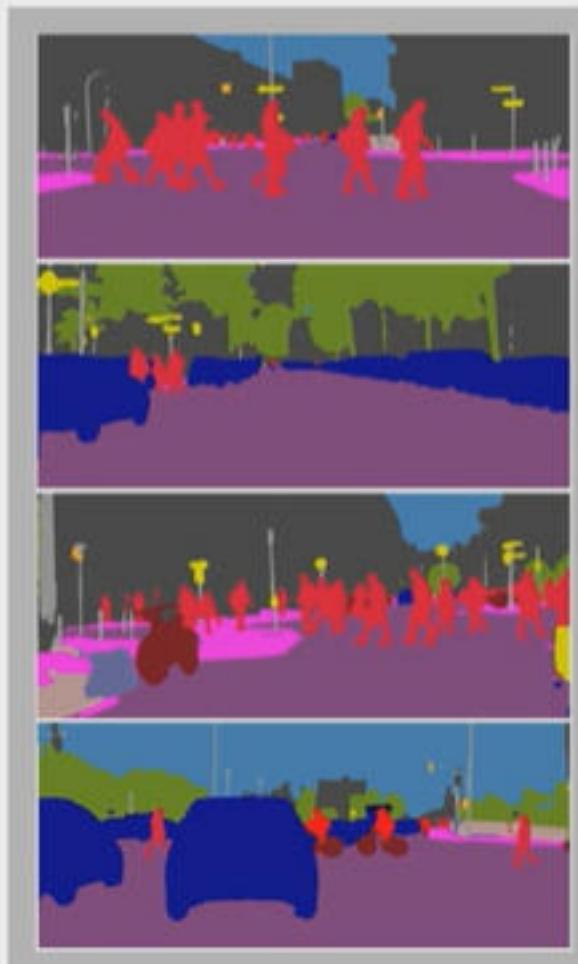
Semantic vs. Instance Segmentation

aws training and certification

Image



Semantic Segmentation



Instance Segmentation



Semantic Segmentation: Data

Data

aws training and certification

Input: Image



Data Contd.



Input: Image array

- Shape: (H x W x C)
- C = 1 for greyscale, 3 for RGB
- Unsigned integer (0-255)

```
array([[ [122,    97,    93],
         [121,    97,    93],
         [120,    99,    96],
         ...,
         [232,   233,   237],
         [226,   227,   231],
         [220,   224,   227]],

        [[120,    95,    91],
         [118,    94,    90],
         [117,    96,    93],
         ...,
         [232,   233,   237],
         [226,   227,   231],
         [220,   224,   227]],

        [[116,    91,    87],
         [114,    90,    86],
         [113,    89,    85],
         ...,
         [232,   233,   237],
         [226,   227,   231],
         [220,   224,   227]],

        ...]]
```

Data Contd.



Output?

For classification,
class labels are integers
or “one-hot encodings”,
represented as vectors.

Label-integer mapping:
{Car:0, Street:1, Tree:2}

| Label | Car | Street | Tree |
|-------|-----|--------|------|
| 1 | 0 | 1 | 0 |
| 0 | 1 | 0 | 0 |
| 0 | 1 | 0 | 0 |
| 2 | 0 | 0 | 1 |

One-hot
→

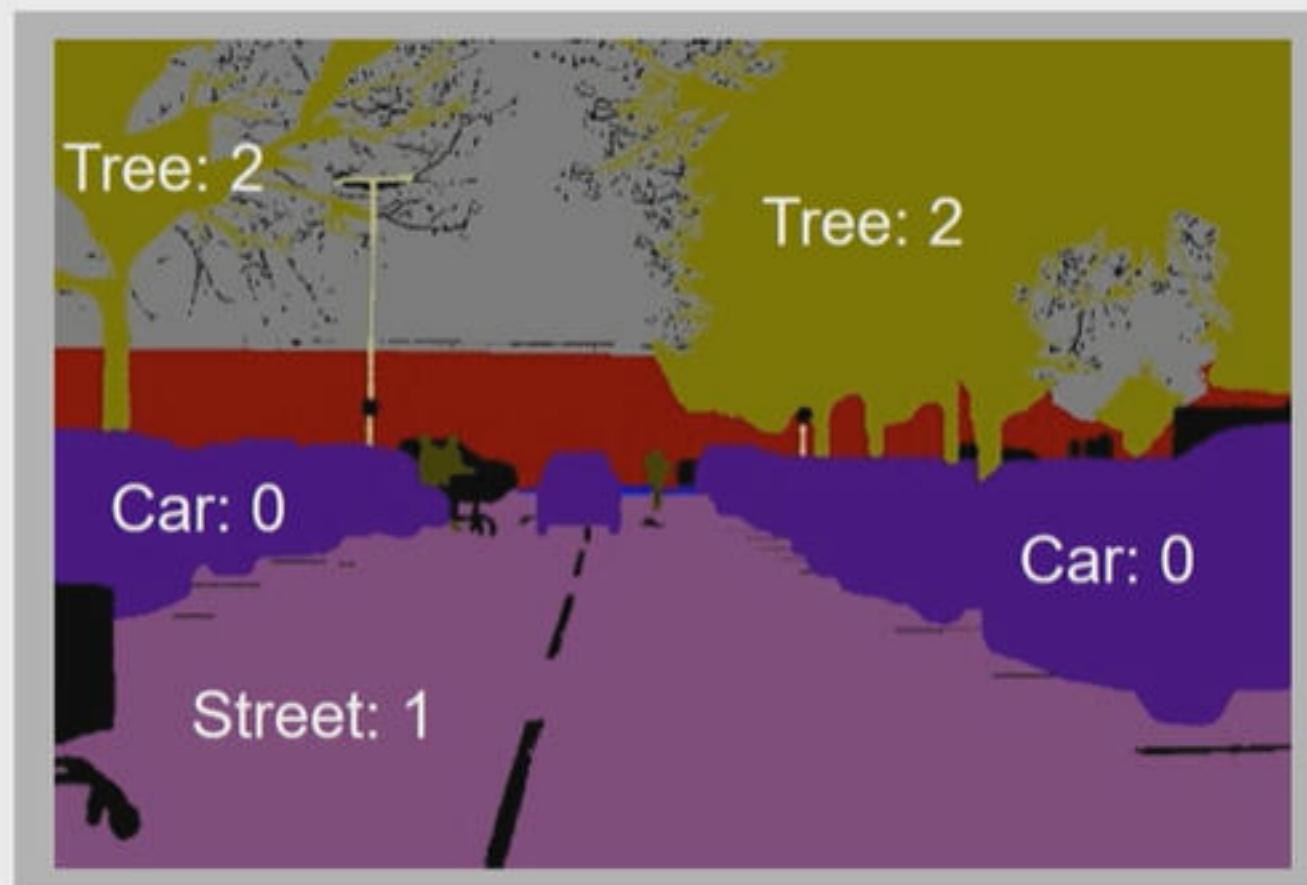
```
array([[0, 1, 0],  
       [1, 0, 0],  
       [1, 0, 0],  
       [0, 0, 1]])
```

Data Contd.

Output: Mask array

- Same spatial resolution (HxW)
- Integer values for pixel class
- Can also be one-hot encoded:
 - (HxWxN), N = # of classes

Note: Each color in this example is a unique class integer.

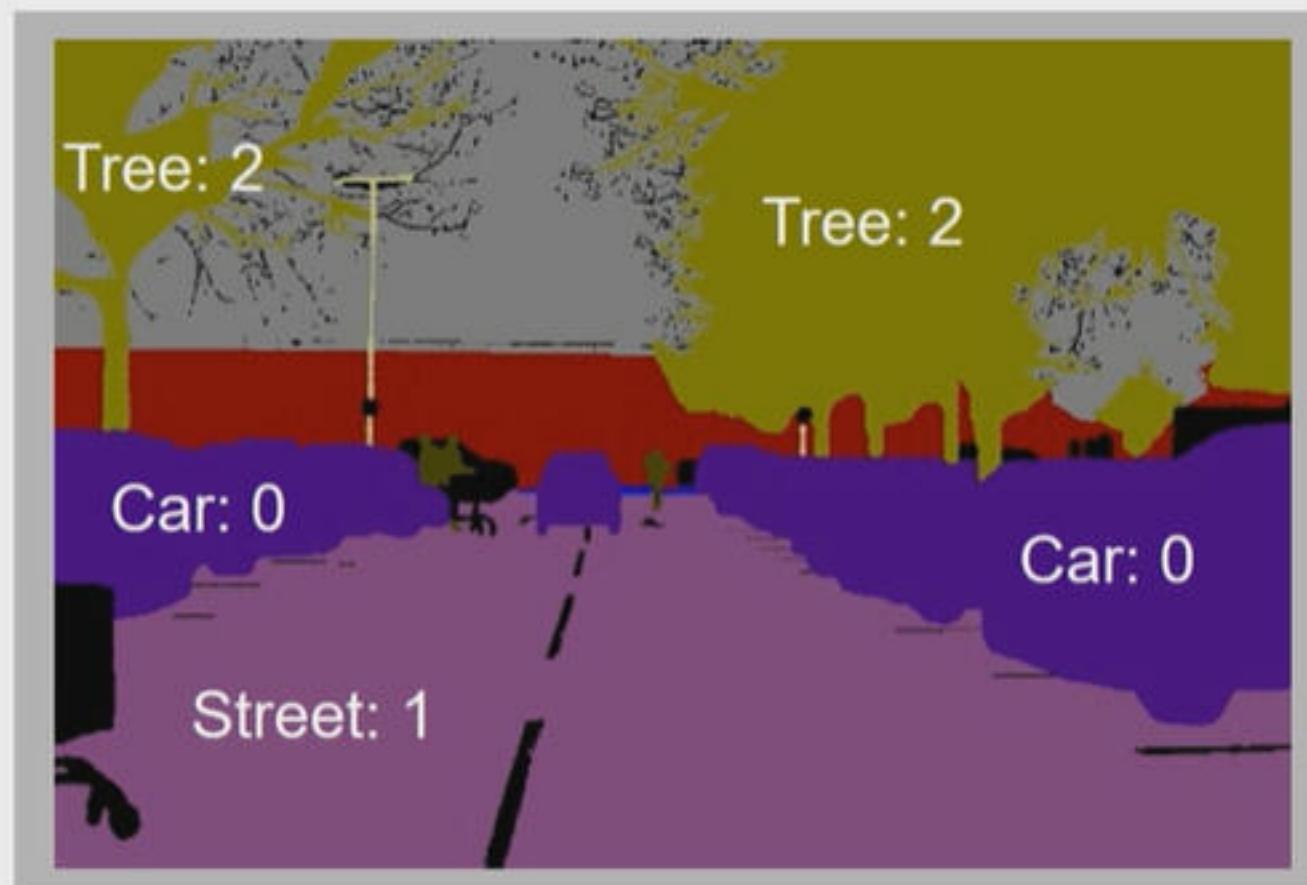


Data Contd.

Output: Mask array

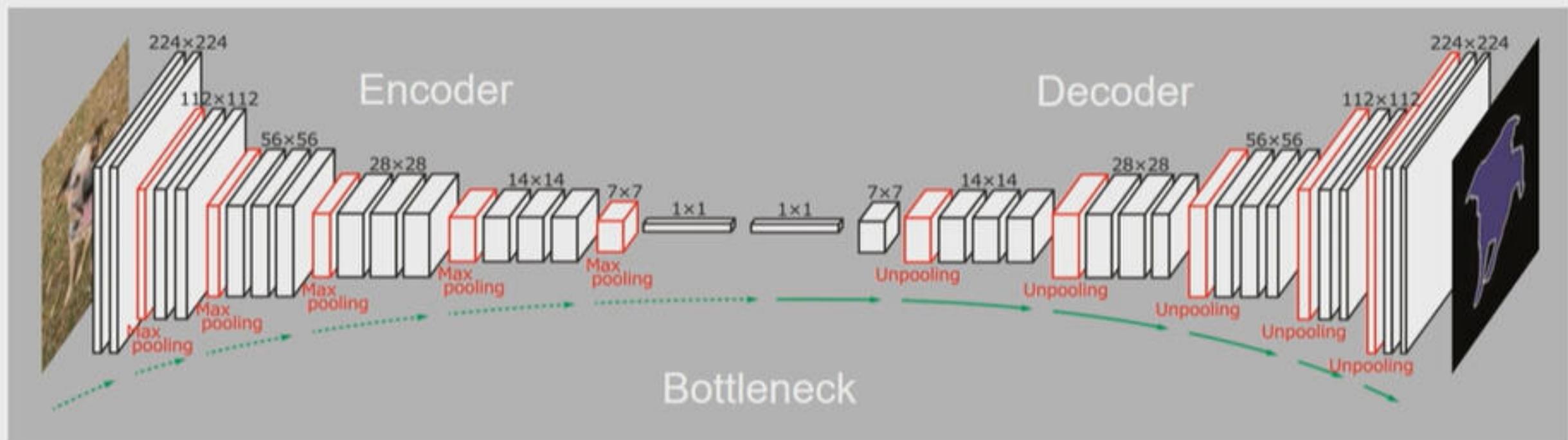
- Same spatial resolution (HxW)
- Integer values for pixel class
- Can also be one-hot encoded:
 - (HxWxN), N = # of classes

Note: Each color in this example is a unique class integer.



Semantic Segmentation: Architecture

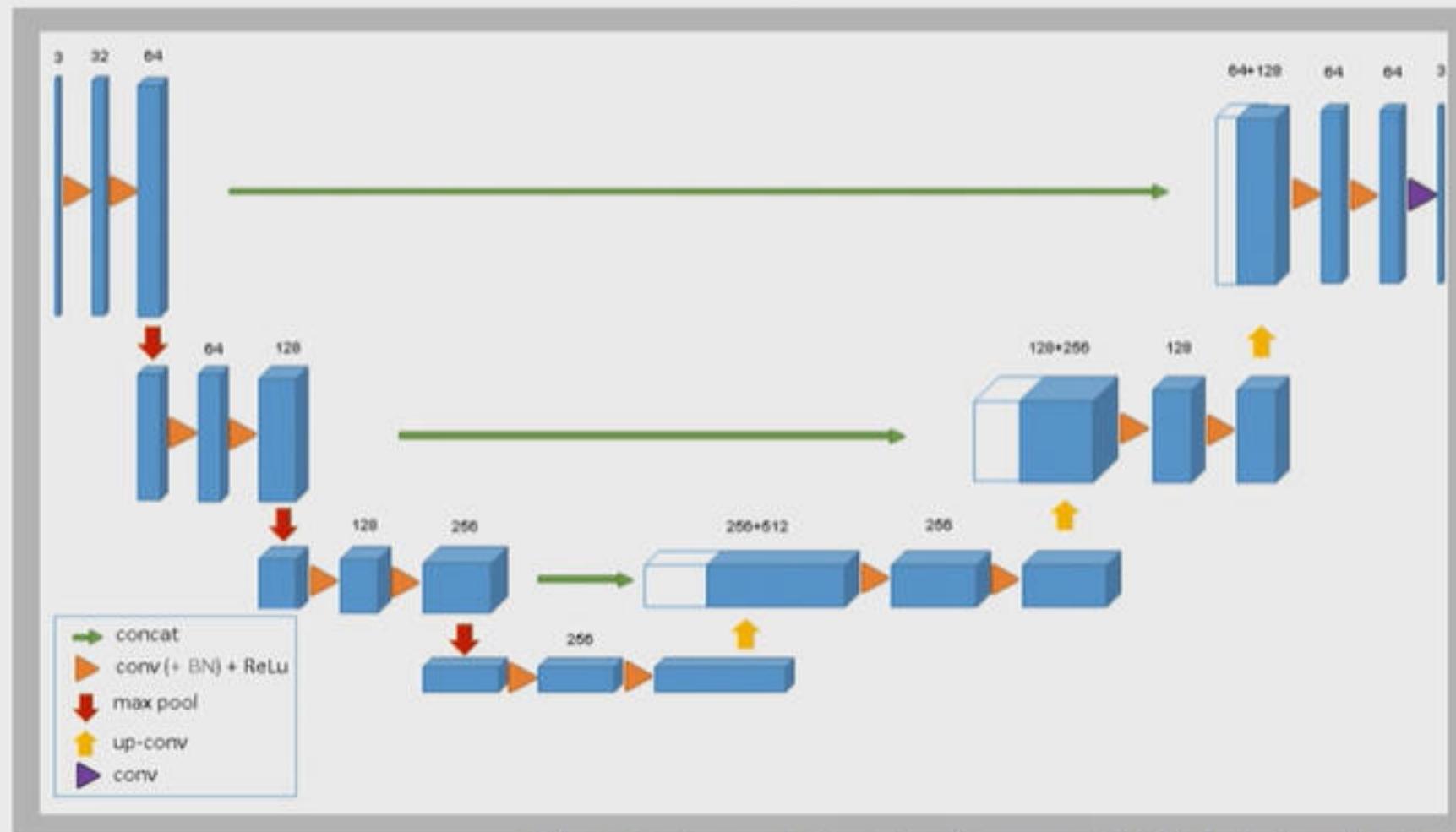
Semantic Segmentation: Architecture



U-Net

aws training and certification

A Fully Convolutional Autoencoder with Skip Connections



Skip Connections



Why Skip Connections?

- Easier training
 - “Vanishing gradient”
- Intuition
 - Coarse features fill in gaps

LOSS



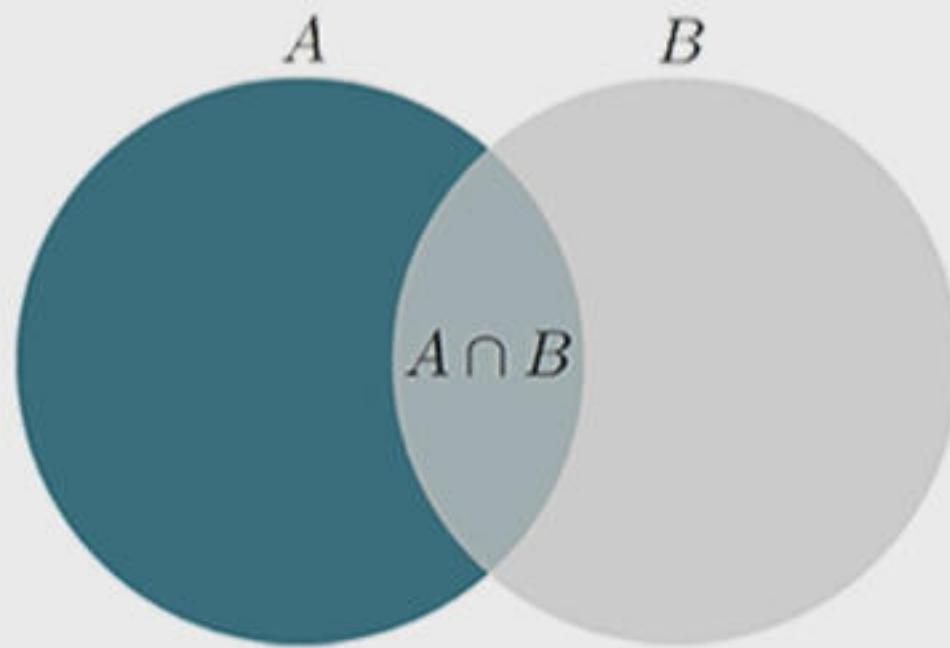
Classification: Categorical Cross-entropy

Semantic Segmentation: Average Categorical Cross-entropy

$$CE = - \sum_x p(x) \log q(x)$$

Loss Contd.

Negative dice-coefficient



Dice-Coefficient:

$$\frac{2 * |A \cap B|}{|A| + |B|}$$

- 0 if no overlap
- 1 if total overlap
- Maximize overlap by minimizing the negative
- Typically written as 1-Dice-coef
- Averaged over classes

Resizing Your Input

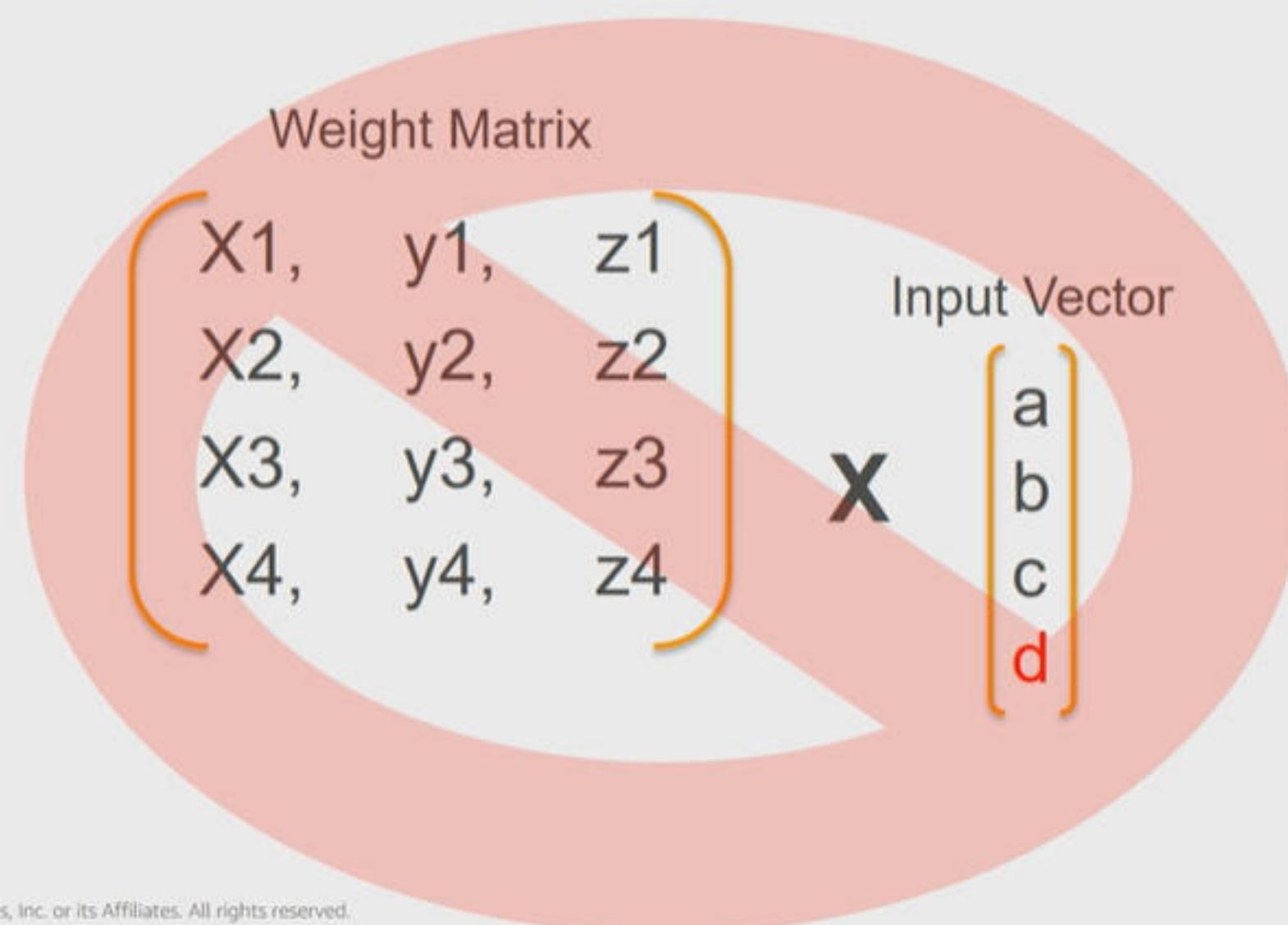


Weight Matrix

$$\begin{matrix} X_1, & y_1, & z_1 \\ X_2, & y_2, & z_2 \\ X_3, & y_3, & z_3 \\ X_4, & y_4, & z_4 \end{matrix} \times \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

The diagram illustrates the multiplication of a weight matrix and an input vector. On the left, a 4x3 matrix is labeled "Weight Matrix". It contains four rows of three elements each: (X_1, y_1, z_1) , (X_2, y_2, z_2) , (X_3, y_3, z_3) , and (X_4, y_4, z_4) . To the right of the matrix is the symbol " \times ". To the right of " \times " is a vertical vector labeled "Input Vector" above it. The vector has three components: a , b , and c .

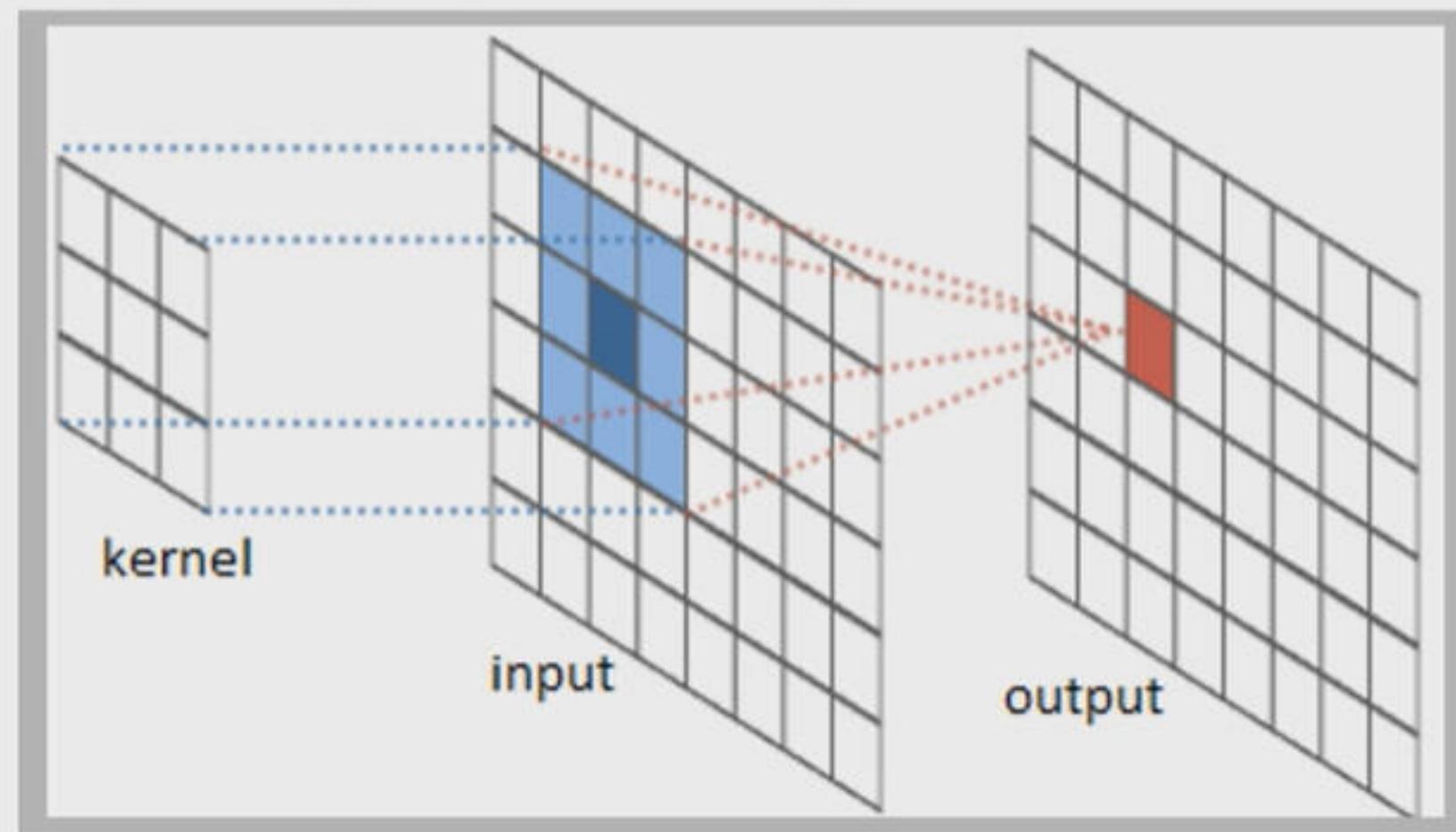
Resizing Your Input Contd.



Cropping

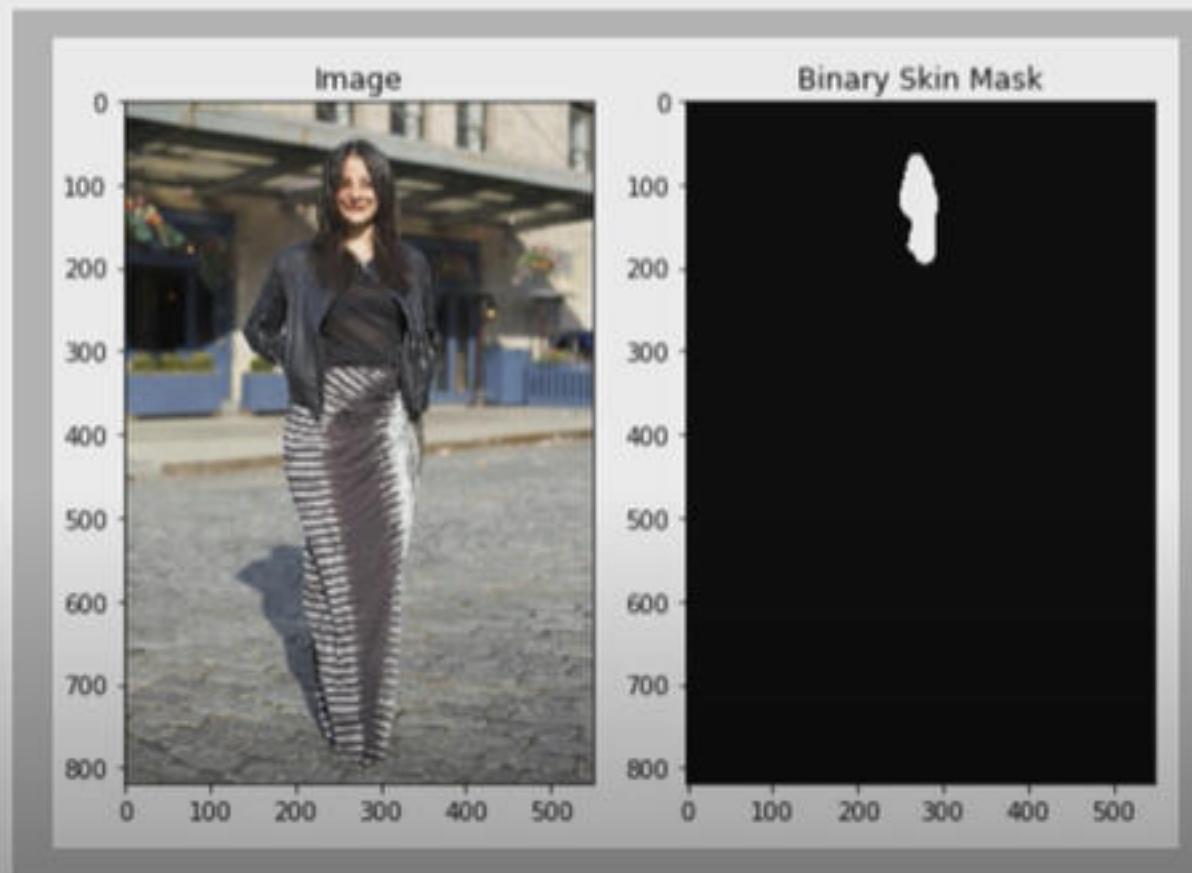


FCN's: Cropped input, same output



Class Imbalance

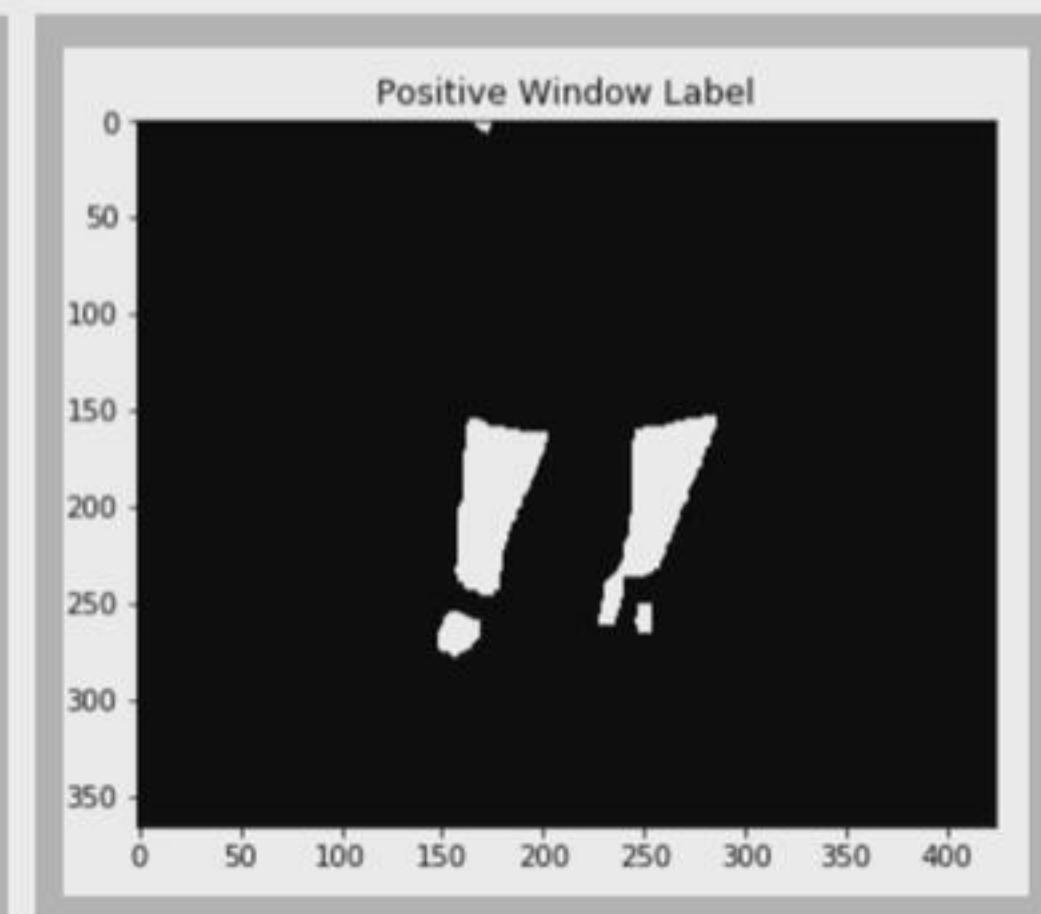
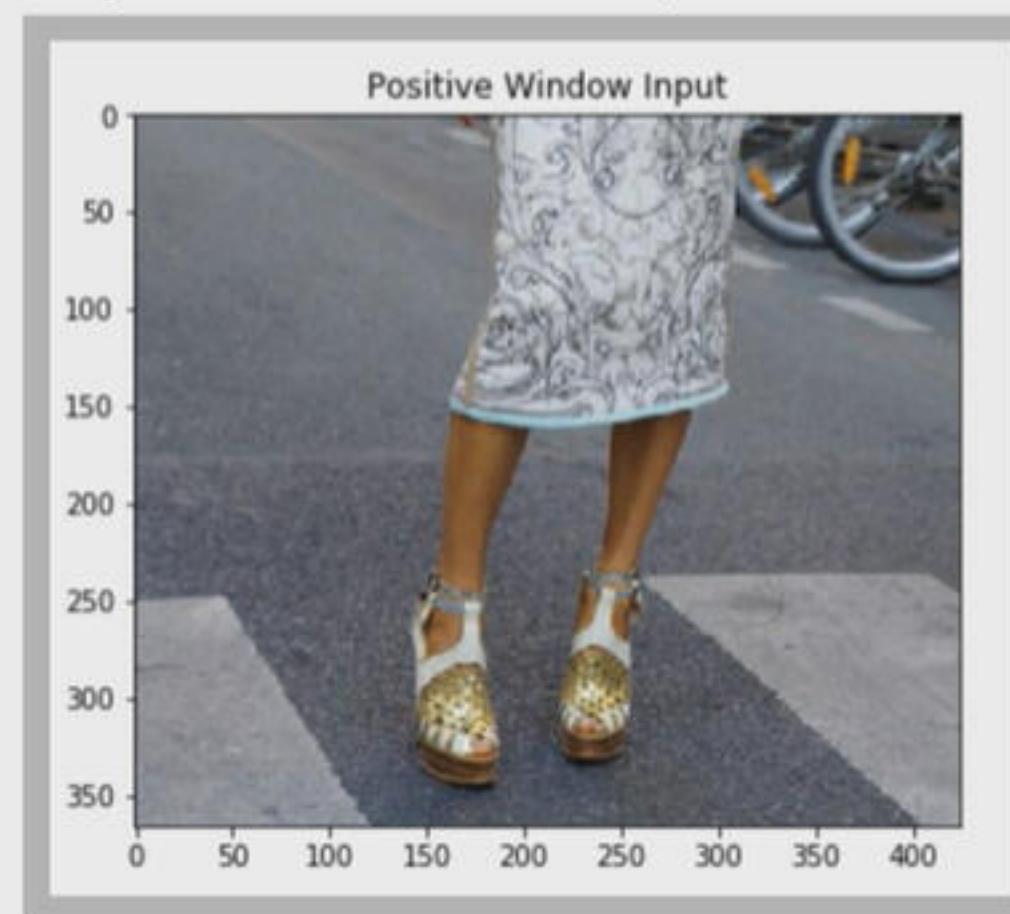
- Weighting loss function helps imbalance in general
- Segmentation also has spatial imbalance



Class Imbalance Contd.



- Solution: Crop from positive windows.
- Every window has a full positive sample.



Next Steps: Hands on!!!



github.com/aws-samples

Attributes returned by DetectFaces()

```
{  
  "FaceDetails": [ {  
    "AgeRange": {"High": ... "Low": ...},  
    "Beard": {...},  
    "BoundingBox": {"Height": ... "Left": ... "Top": ... "Width": ...},  
    "Eyeglasses": {...},  
    "EyesOpen": {...},  
    "Gender": {...},  
    "MouthOpen": {...},  
    "Mustache": {...},  
    "Smile": {...},  
    "Sunglasses": {...}  
  } ]}
```

Attributes returned by DetectFaces()

```
{  
  "FaceDetails": [ {  
    "AgeRange": {"High": ... "Low": ...},  
    "Beard": {...},  
    "BoundingBox": {"Height": ... "Left": ... "Top": ... "Width": ...},  
    "Eyeglasses": {...},  
    "EyesOpen": {...},  
    "Gender": {...},  
    "MouthOpen": {...},  
    "Mustache": {...},  
    "Smile": {...},  
    "Sunglasses": {...}  
  } ]}
```

Loop: Capture an image & draw rectangles

```
while(True):  
  
    # Capture frame  
    frame = capture.read(...)  
  
    # Resize image for faster rekognition  
    image = capture.resize(frame, 0.15, ...)  
  
    # Detect faces in image  
    faces = rekognition.detect_faces(Image={'Bytes':image}, ...)  
  
    # Draw rectangle around faces  
    for face in faces['FaceDetails']:  
        is_smiling = face['Smile']['Value']  
        draw_rectangle(image, face['BoundingBox'], is_smiling)
```

Loop: Capture an image & draw rectangles

```
while(True):\n\n    # Capture frame\n    frame = capture.read(...)\n\n    # Resize image for faster rekognition\n    image = capture.resize(frame, 0.15, ...)\n\n    # Detect faces in image\n    faces = rekognition.detect_faces(Image={'Bytes':image}, ...)\n\n    # Draw rectangle around faces\n    for face in faces['FaceDetails']:\n        is_smiling = face['Smile']['Value']\n        draw_rectangle(image, face['BoundingBox'], is_smiling)
```



Loop: Capture an image & draw rectangles

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while(True):  
  
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    # Detect faces in image  
    faces = rekognition.detect_faces(Image={'Bytes':image}, ...)  
  
    # Draw rectangle around faces  
    for face in faces['FaceDetails']:  
        is_smiling = face['Smile']['Value']  
        draw_rectangle(image, face['BoundingBox'], is_smiling)
```

Attributes returned by DetectFaces()

```
{  
  "FaceDetails": [ {  
    "AgeRange": {"High": ... "Low": ...},  
    "Beard": {...},  
    "BoundingBox": {"Height": ... "Left": ... "Top": ... "Width": ...},  
    "Eyeglasses": {...},  
    "EyesOpen": {...},  
    "Gender": {...},  
    "MouthOpen": {...},  
    "Mustache": {...},  
    "Smile": {...},  
    "Sunglasses": {...}  
  } ]}
```

Loop: Find face in collection

```
while(True):  
  
    # Detect faces in image  
    faces = rekognition.search_faces_by_image(  
        CollectionId='...', Image={'Bytes':image}, ...)  
  
    # Show name and confidence  
    for matches in faces['FaceMatches']:  
  
        if 'ExternalImageId' in matches['Face'].keys():  
            name = matches['Face']['ExternalImageId']  
            confidence = matches['Face']['Confidence']  
  
            cv2.putText(...)
```

Loop: Find face in collection

```
while(True):  
  
    # Detect faces in image  
    faces = rekognition.search_faces_by_image(  
        CollectionId='...', Image={'Bytes':image}, ...)  
  
    # Show name and confidence  
    for matches in faces['FaceMatches']:  
  
        if 'ExternalImageId' in matches['Face'].keys():  
            name = matches['Face']['ExternalImageId']  
            confidence = matches['Face']['Confidence']  
  
            cv2.putText(...)
```



Loop: Find face in collection

```
while(True):

# Detect faces in image
faces = rekognition.search_faces_by_image(
    CollectionId='...', Image={'Bytes':image}, ...)

# Show name and confidence
for matches in faces['FaceMatches']:

    if 'ExternalImageId' in matches['Face'].keys():
        name = matches['Face']['ExternalImageId']
        confidence = matches['Face']['Confidence']

    cv2.putText(...)
```



Attributes returned by `search_faces_by_image()`

```
{'SearchedFaceBoundingBox': {...},  
 'SearchedFaceConfidence': 99.9997787475586,  
 'FaceMatches': [  
     {'Face': {'BoundingBox': {...},  
             'ExternalImageId': 'John',  
             'Confidence': 99.99739837646484  
         }  
     }  
 ]  
}
```

Attributes returned by `detect_text()`

```
{u'TextDetections': [  
    {'Geometry': {'BoundingBox': ...},  
     'Confidence': 99.93452453613281,  
     'DetectedText': 'THANKS',  
     'Type': 'LINE'  
    },  
    ...
```

Also: 'Type': 'WORD'

Functions for still images

| | |
|------------------------|--|
| CompareFaces | Match face between images |
| DetectFaces | Find up to 100 faces in an image |
| DetectLabels | Find real-world entities in an image |
| DetectModerationLabels | Detect adult content |
| DetectText | Find machine-readable text |
| IndexFaces | Add 100 faces in an image to a collection |
| RegognizeCelebrities | Identify up to 100 celebrities in an image |
| SearchFaces | Find a face ID in a face collection |
| SearchFacesByImage | Find largest face in a face collection |

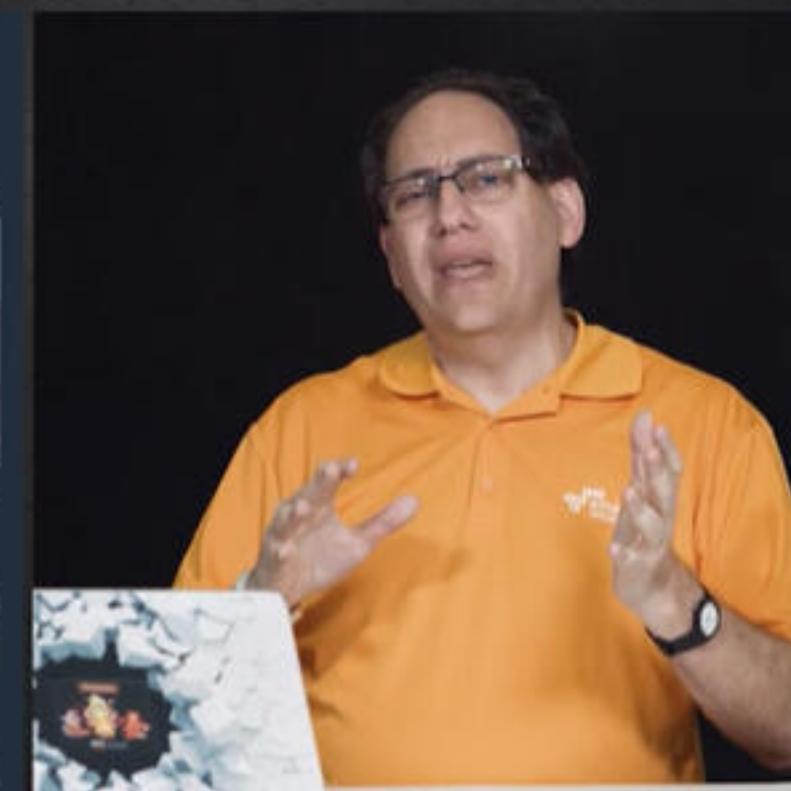
Part 2: Hands-On Rekognition Video

Let's get Hands-On!

Using Amazon Rekognition on Video Files



Find a person,
make a video of
just them



Create a Collection & Load Faces

```
$ aws rekognition create-collection  
--collection-id trainers  
  
$ aws rekognition index-faces  
--collection-id trainers  
--image "S3Object={Bucket=hands-on-rekognition,Name=John.jpg}"  
--external-image-id John
```



Create a Collection & Load Faces

```
$ aws rekognition create-collection  
--collection-id trainers
```

```
$ aws rekognition index-faces  
--collection-id trainers  
--image "S3Object={Bucket=hands-on-rekognition,Name=John.jpg}"  
--external-image-id John
```



Search for faces

```
$ aws rekognition start-face-search  
--video "S3Object={Bucket=...,Name=trainers.mp4}"  
--collection-id trainers
```



```
$ aws rekognition get-face-search --job-id...
```

Search for faces

```
$ aws rekognition start-face-search  
--video "S3Object={Bucket=...,Name=trainers.mp4}"  
--collection-id trainers
```



```
$ aws rekognition get-face-search --job-id...
```

Create a Collection & Load Faces

```
$ aws rekognition create-collection  
--collection-id trainers
```

```
$ aws rekognition index-faces  
--collection-id trainers  
--image "S3Object={Bucket=hands-on-rekognition,Name=John.jpg}"  
--external-image-id John
```



jorote — ec2-user@ip-10-0-0-213:~ — ssh - - bash — 83x24

— ec2-user@ip-10-0-0-213:~ — ssh - - bash

```
[ec2-user@ip-10-0-0-213 ~]$ aws rekognition create-collection --collection-id Trainers
{
    "CollectionArn": "awsrekognition:us-west-2:416159072693:collection/Trainers",
    "FaceModelVersion": "3.0",
    "StatusCode": 200
}
```

```
[ec2-user@ip-10-0-0-213 ~]$
```

jorote — ec2-user@ip-10-0-0-213:~ — ssh - bash — 83x24

— ec2-user@ip-10-0-0-213:~ — ssh - bash

```
[ec2-user@ip-10-0-0-213 ~]$ aws rekognition index-faces \
> --collection-id Trainers \
> --image "S3Object={Bucket=us-west-2-aws-training,Name=awsu-spl/spl-202/faces/John
.jpg}" \
> --external-image-id Johnaws rekognition index-faces \
> --collection-id Trainers \
> --image "S3Object={Bucket=us-west-2-aws-training,Name=awsu-spl/spl-202/faces/John
.jpg}" \
> --external-image-id John
```

```
jorote — ec2-user@ip-10-0-0-213:~ — ssh - bash — 83x24
```

```
— ec2-user@ip-10-0-0-213:~ — ssh - bash
```

```
},
  "Quality": {
    "Sharpness": 99.9305191040039,
    "Brightness": 39.646568298339844
  },
  "Confidence": 99.9974365234375
},
"Face": {
  "BoundingBox": {
    "Width": 0.2666666805744171,
    "Top": 0.1465798020362854,
    "Left": 0.5464788675308228,
    "Height": 0.46254071593284607
  },
  "FaceId": "487682b1-6c49-4697-b6d6-19487af206be",
  "ExternalImageId": "John",
  "Confidence": 99.9974365234375,
  "ImageId": "cc027b0d-5863-58b3-8e0a-7400fdf6f32c"
}
],
"FaceModelVersion": "3.0"
}
```

jorote — ec2-user@ip-10-0-0-213:~ — ssh · -bash — 83x24

— ec2-user@ip-10-0-0-213:~ — ssh · -bash

```
> --external-image-id John
{
  "FaceRecords": [
    {
      "FaceDetail": {
        "BoundingBox": {
          "Width": 0.2666666805744171,
          "Top": 0.1465798020362854,
          "Left": 0.5464788675308228,
          "Height": 0.46254071593284607
        },
        "Landmarks": [
          {
            "Y": 0.3537498116493225,
            "X": 0.6374950408935547,
            "Type": "eyeLeft"
          },
          {
            "Y": 0.3529587686061859,
            "X": 0.720091700553894,
            "Type": "eyeRight"
          },
          {
            "Y": 0.46254071593284607
          }
        ]
      }
    }
  ]
}
```

jorote — ec2-user@ip-10-0-0-213:~ — ssh - bash — 83x24

— ec2-user@ip-10-0-0-213:~ — ssh - bash

```
        "X": 0.7185091376304626,
        "Type": "mouthRight"
    },
],
"Pose": {
    "Yaw": 0.40988653898239136,
    "Roll": -0.7546926140785217,
    "Pitch": -1.0719298124313354
},
"Quality": {
    "Sharpness": 99.9305191040039,
    "Brightness": 39.646568298339844
},
"Confidence": 99.9974365234375
},
"Face": {
    "BoundingBox": {
        "Width": 0.2666666805744171,
        "Top": 0.1465798020362854,
        "Left": 0.5464788675308228,
        "Height": 0.46254071593284607
},
"FaceId": "487682b1-6c49-4697-b6d6-19487af206be",
"FaceAttributes": {
    "EyesOpenProbability": 0.9999999999999999,
    "EyesClosedProbability": 0.00010000000000000002,
    "MouthOpenProbability": 0.9999999999999999,
    "MouthClosedProbability": 0.00010000000000000002
}
}
```

Search for faces

```
$ aws rekognition start-face-search  
--video "S3Object={Bucket=...,Name=trainers.mp4}"  
--collection-id trainers
```



```
$ aws rekognition get-face-search --job-id...
```

jorote — ec2-user@ip-10-0-0-213:~ — ssh - bash — 83x24

— ec2-user@ip-10-0-0-213:~ — ssh - bash

```
[ec2-user@ip-10-0-0-213 ~]$ JOB=`aws rekognition start-face-search \
> --video "S3Object={Bucket=aws-tc-largeobjects,Name=SPLs/202/technical-trainers.mp
4}" \
> --collection-id Trainers \
> --query JobId \
> --output text`  
[ec2-user@ip-10-0-0-213 ~]$  
[ec2-user@ip-10-0-0-213 ~]$ echo $JOB  
f1ef0d93e690fdc15ce4a3eb2b6276d9ed4a30b8ad672d06f9c136d893b526dd  
[ec2-user@ip-10-0-0-213 ~]$
```

jorote — ec2-user@ip-10-0-0-179:~ — ssh • -bash — 83x24

— ec2-user@ip-10-0-0-179:~ — ssh • -bash

```
[ec2-user@ip-10-0-0-179 ~]$ aws rekognition get-face-search --job-id $JOB
```

```
jorote@ec2-user:~$ ssh -l jorote ip-10-0-0-179:~$ bash -c "cat < /dev/null" > /tmp/pose.json
jorote@ec2-user:~$ cat /tmp/pose.json
{
    "Timestamp": 38440,
    "Person": {
        "BoundingBox": {
            "Width": 0.5765625238418579,
            "Top": 0.06388889253139496,
            "Left": 0.3765625059604645,
            "Type": "mouthRight"
        }
    },
    "Pose": {
        "Yaw": -13.82411003112793,
        "Roll": -12.427720069885254,
        "Pitch": 5.478622913360596
    },
    "Quality": {
        "Sharpness": 99.68138885498047,
        "Brightness": 43.79252243041992
    },
    "Confidence": 99.99217987060547
}
}
{
}
```

jorote — ec2-user@ip-10-0-0-179:~ — ssh - bash — 83x24

— ec2-user@ip-10-0-0-179:~ — ssh - bash

```
"Person": {  
    "BoundingBox": {  
        "Width": 0.61328125,  
        "Top": 0.1027777910232544,  
        "Left": 0.14140625298023224,  
        "Height": 0.894444465637207  
    },  
    "Index": 6  
}  
}  
],  
"NextToken": "+2DWHimFsAma4Xnu0PaYorz02jgUUCz+040yThh3AF/8+cCTN1yWtZtPaDUq3wC2q  
jQ9GHpOnN7f",  
"JobStatus": "SUCCEEDED",  
"VideoMetadata": {  
    "Format": "QuickTime / MOV",  
    "FrameRate": 25.0,  
    "Codec": "h264",  
    "DurationMillis": 197080,  
    "FrameHeight": 720,  
    "FrameWidth": 1280  
}  
}
```

```
jorote — ec2-user@ip-10-0-0-179:~ — ssh - bash — 83x24
— ec2-user@ip-10-0-0-179:~ — ssh - bash

    "Height": 0.8958333134651184
},
    "Index": 6
}
},
{
    "Timestamp": 56560,
    "FaceMatches": [
        {
            "Face": {
                "BoundingBox": {
                    "Width": 0.3111110031604767,
                    "Top": 0.11671099811792374,
                    "Left": 0.134443998336792,
                    "Height": 0.42998701333999634
                },
                "FaceId": "f6def7fd-3ab1-4e81-a0ae-6b5eb49a7228",
                "ExternalImageId": "Edward",
                "Confidence": 100.0,
                "ImageId": "f867286b-c780-5aa9-b44c-aff7ff5170d4"
            },
            "Similarity": 99.4621810913086
        }
    ]
}
```

```
jorote — ec2-user@ip-10-0-0-179:~ — ssh -bash — 83x24
— ec2-user@ip-10-0-0-179:~ — ssh -bash

        "Height": 0.8958333134651184
    },
    "Index": 6
}
{
    "Timestamp": 56560,
    "FaceMatches": [
        {
            "Face": {
                "BoundingBox": {
                    "Width": 0.311110031604767,
                    "Top": 0.11671099811792374,
                    "Left": 0.134443998336792,
                    "Height": 0.42998701333999634
                },
                "FaceId": "f6def7fd-3ab1-4e81-a0ae-6b5eb49a7228",
                "ExternalImageId": "Edward",
                "Confidence": 100.0,
                "ImageId": "f867286b-c780-5aa9-b44c-aff7ff5170d4"
            },
            "Similarity": 99.4621810913086
        }
    ]
}
```

FaceSearch() returns Timestamps and Faces

```
{ "Persons": [ { "Timestamp": 7360, "FaceMatches": [ { "Face": { "ExternalImageId": "John", "Confidence": 99.99750518798828 ... } }, { "Timestamp": 7560, ... }
```

Clip Stitching with Amazon Elastic Transcoder

Timestamps:

[99800, 99840, 100000, 100040, ...]

Convert into individual scene start/end times:

[(99800, 101480), (127520, 131760), ...]

Convert into **clips** for Amazon Elastic Transcoder:

```
[  
    {'Key': 'trainers.mp4',  
     'TimeSpan': {'StartTime': '99.8', 'Duration': '1.68'}},  
    {'Key': 'trainers.mp4',  
     'TimeSpan': {'StartTime': '127.52', 'Duration': '4.24'}},  
    ...  
]
```

Result:



AWS Machine Learning Blog

Automated video editing with YOU as the star!

by John Rotenstein | on 10 APR 2018 | in [Amazon Rekognition Video](#) | [Permalink](#) | [Comments](#) | [Share](#)

Have you ever wanted to find a specific person among hours of video footage?

Perhaps you're preparing a video for a 21st birthday celebration, wanting to find happy memories of your birthday child. Maybe you are scouring video footage, looking to see what a specific employee did on their last day at work. Or it may be that you want to produce a highlight reel of your own efforts at the [Nathan's Hot Dog Eating Contest](#).

In this blog post, you will learn how to combine the capabilities of Amazon Rekognition Video and Amazon Elastic Transcoder to automatically convert a long video into a highlight video showing all footage of a given person.

Result:



AWS Machine Learning Blog

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In this blog post, you will learn how to combine the capabilities of Amazon Rekognition Video and Amazon Elastic Transcoder to automatically convert a long video into a highlight video showing all footage of a given person.

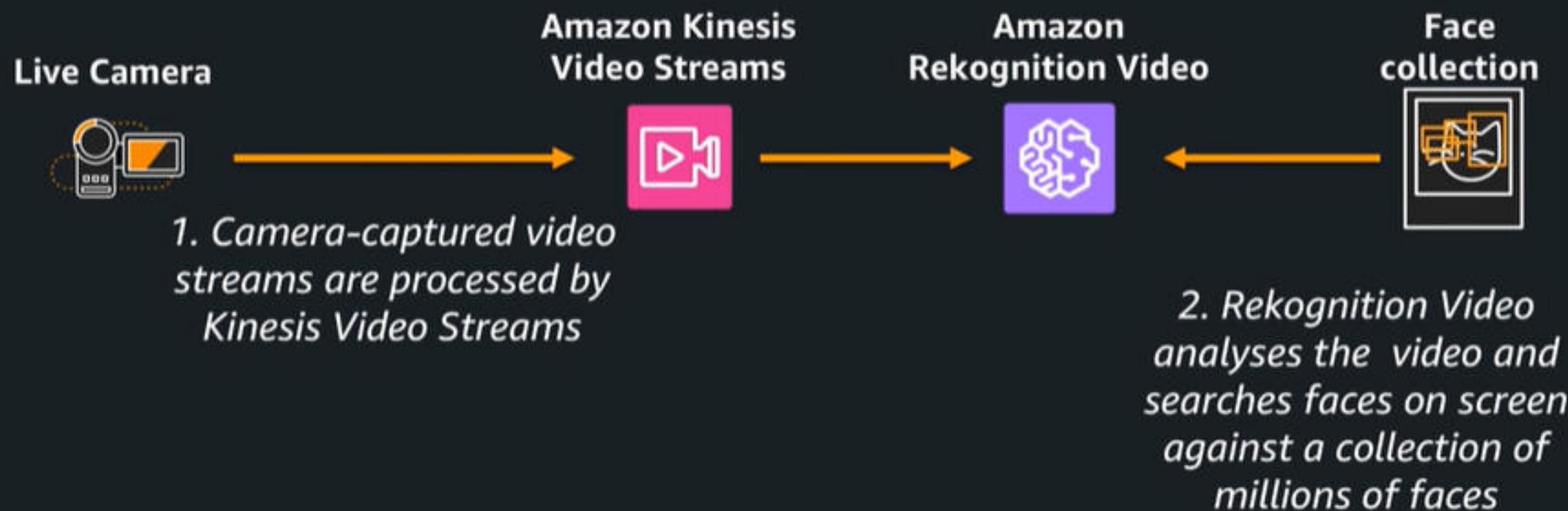
Functions for stored videos

| | |
|---------------------------|---|
| StartCelebrityRecognition | Recognize celebrities in a stored video |
| StartContentModeration | Detect adult content in a stored video |
| StartFaceDetection | Detect faces in a stored video |
| StartFaceSearch | Search for faces in a collection that match faces of persons detected in a stored video |
| StartLabelDetection | Detect real-world entities in a video |
| StartPersonTracking | Track persons in a stored video |

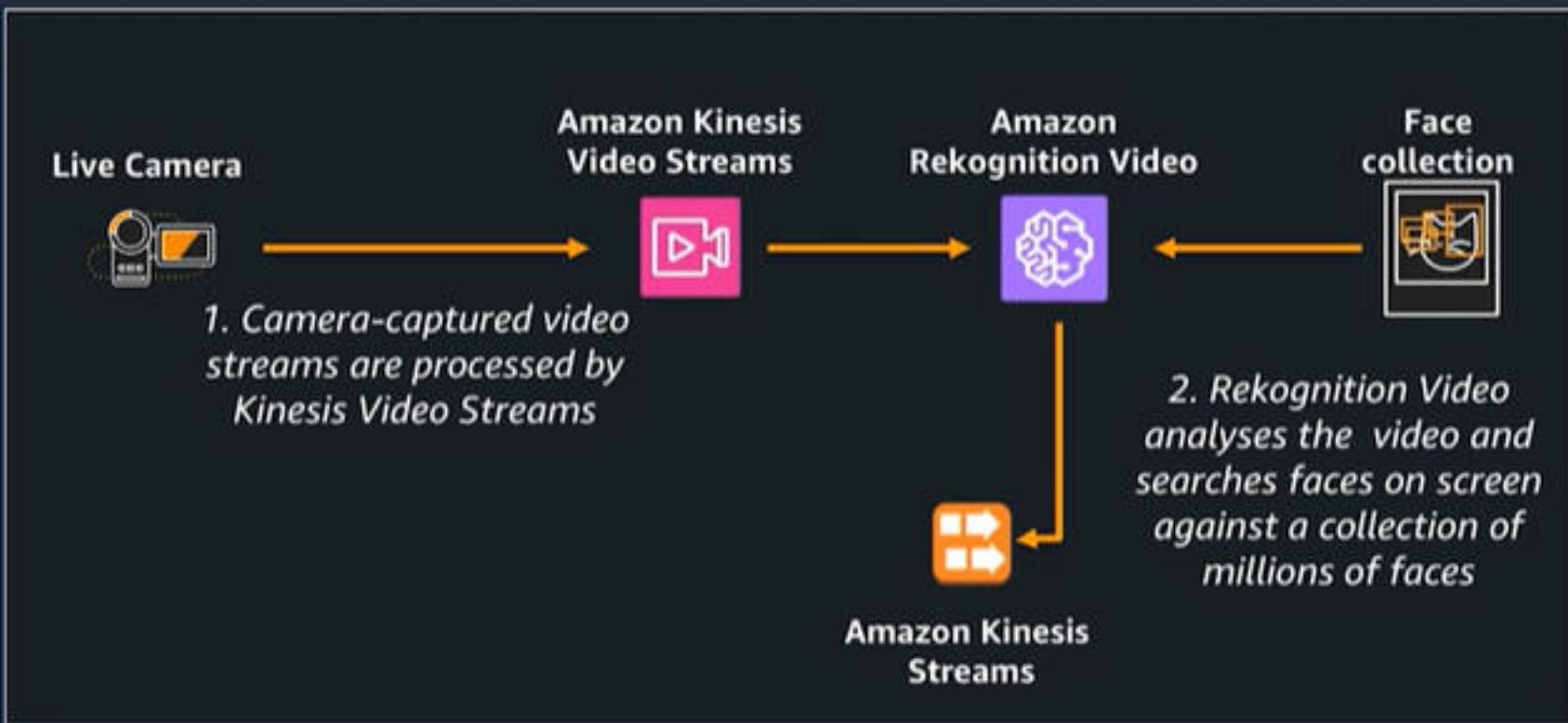
Part 3: Rekognition Streaming Video

Face Detection on streaming video

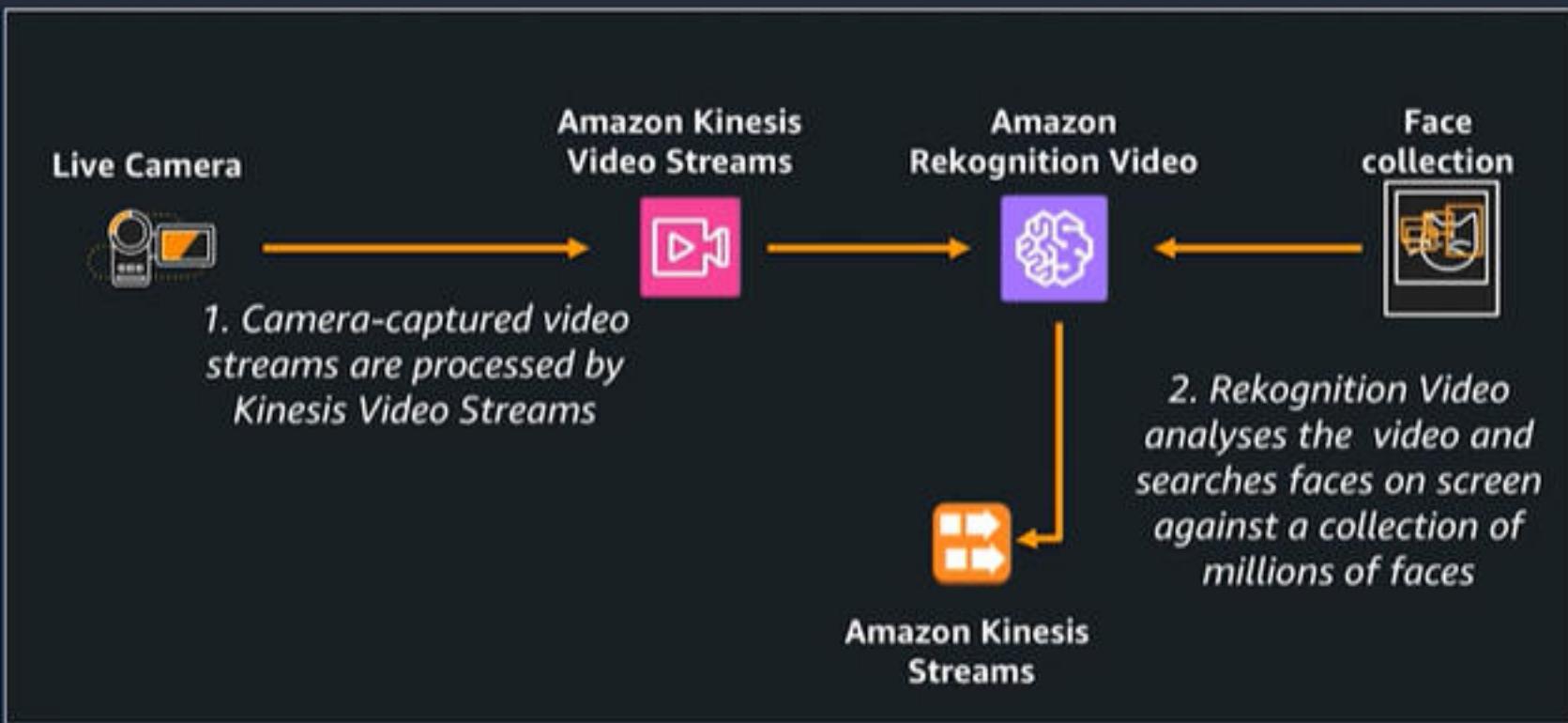
Architecture diagram



Architecture diagram



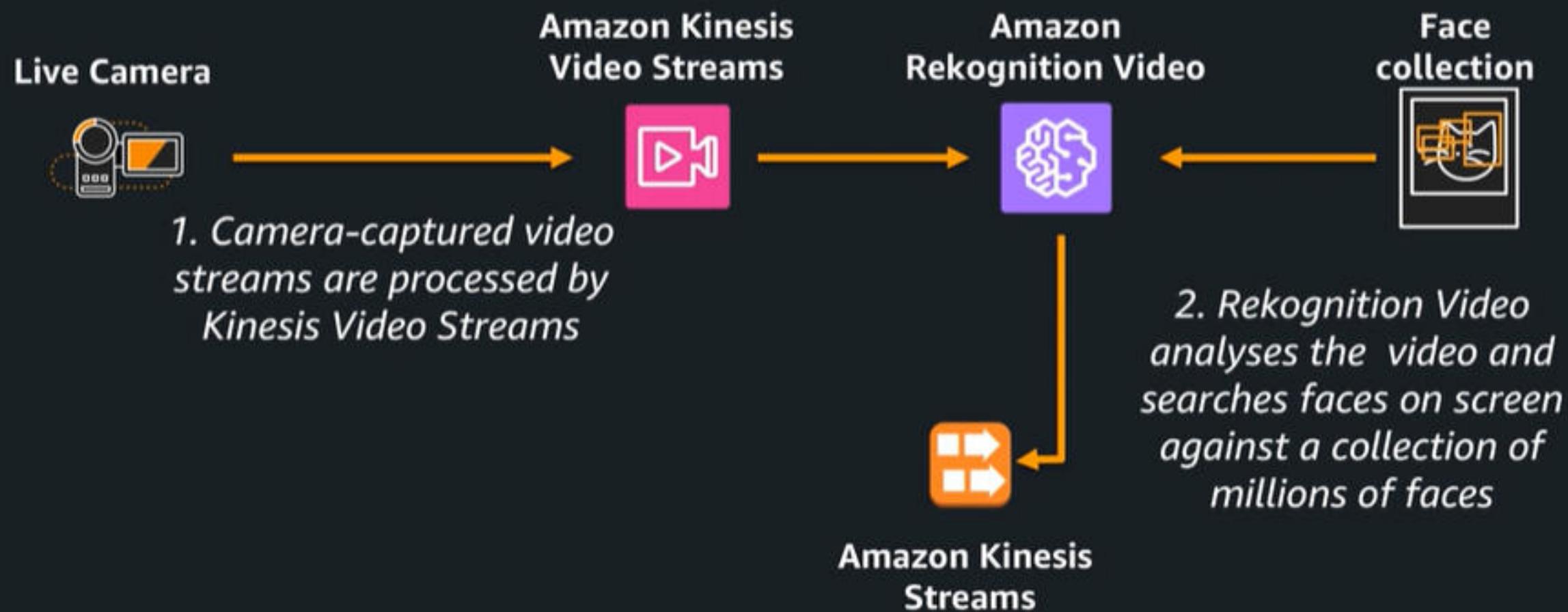
Architecture diagram



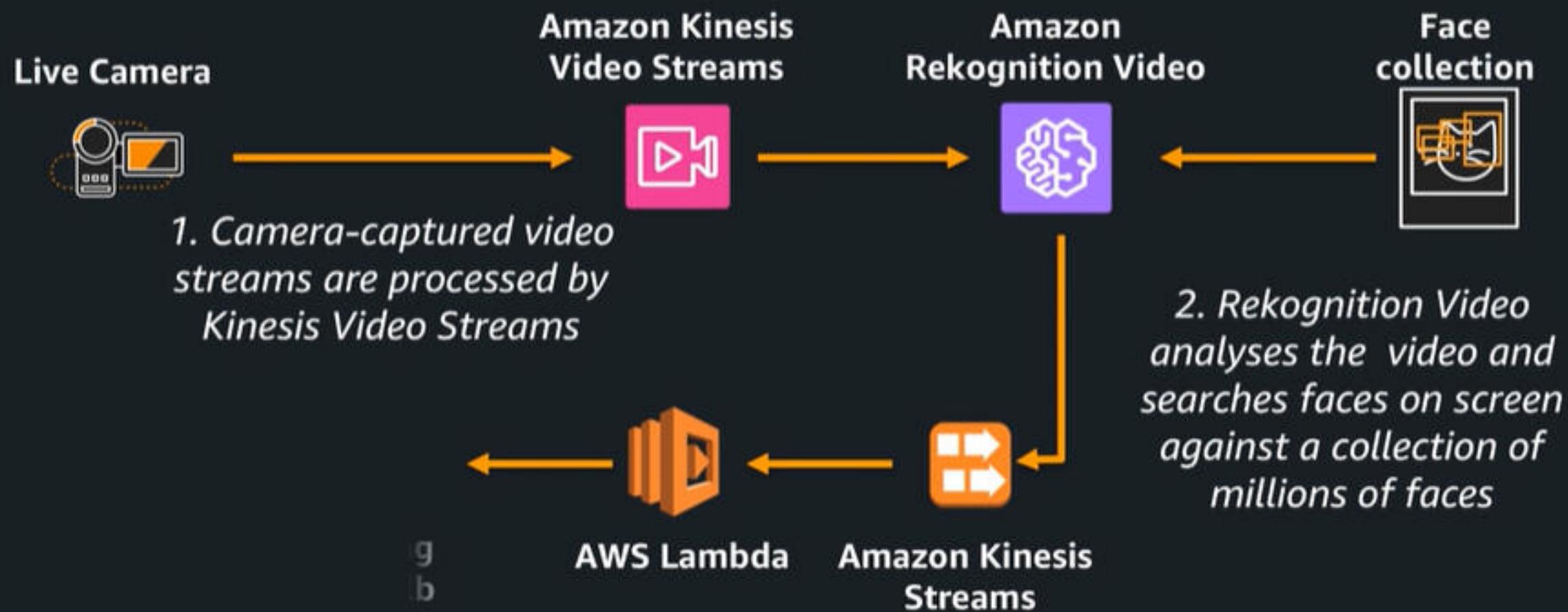
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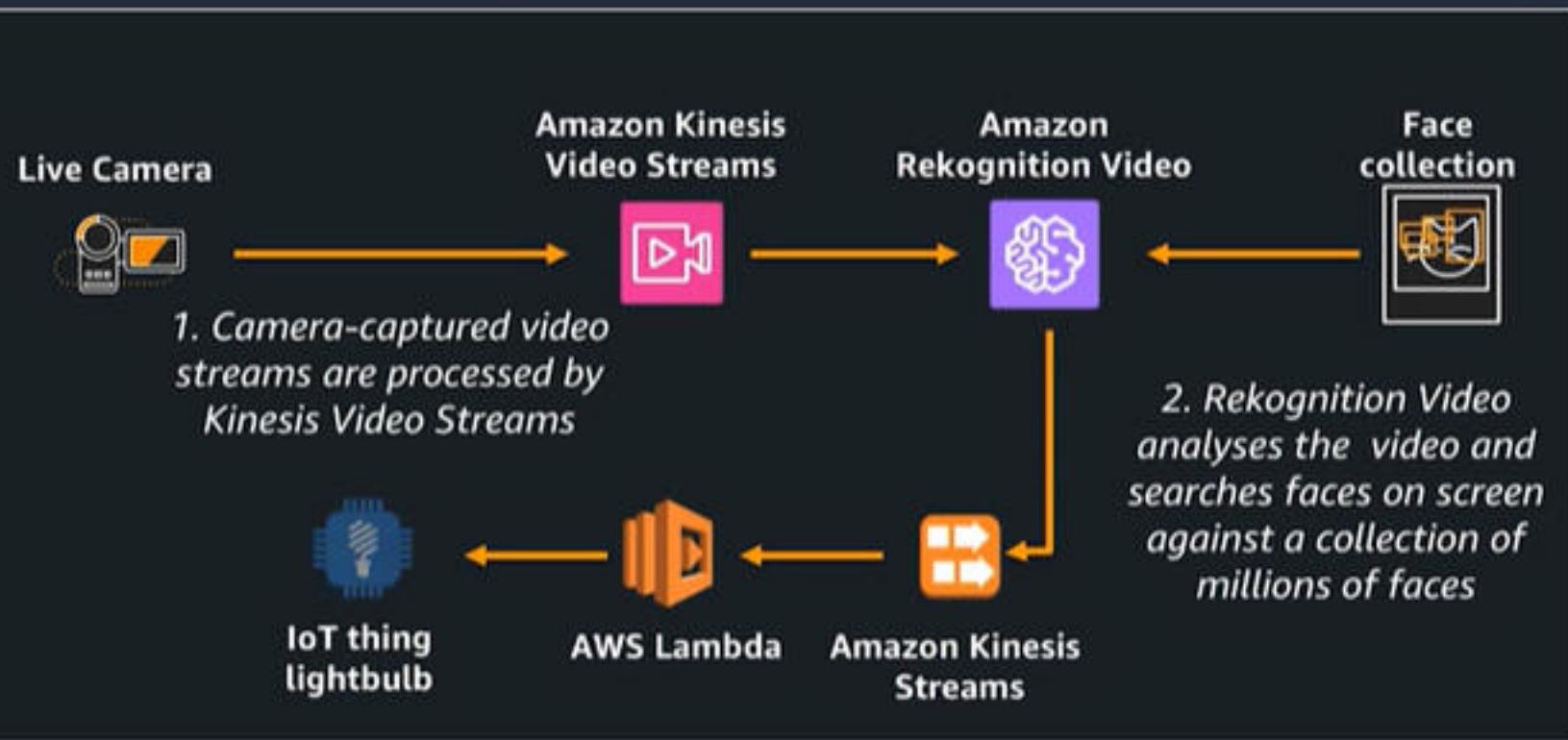
Architecture diagram



Architecture diagram



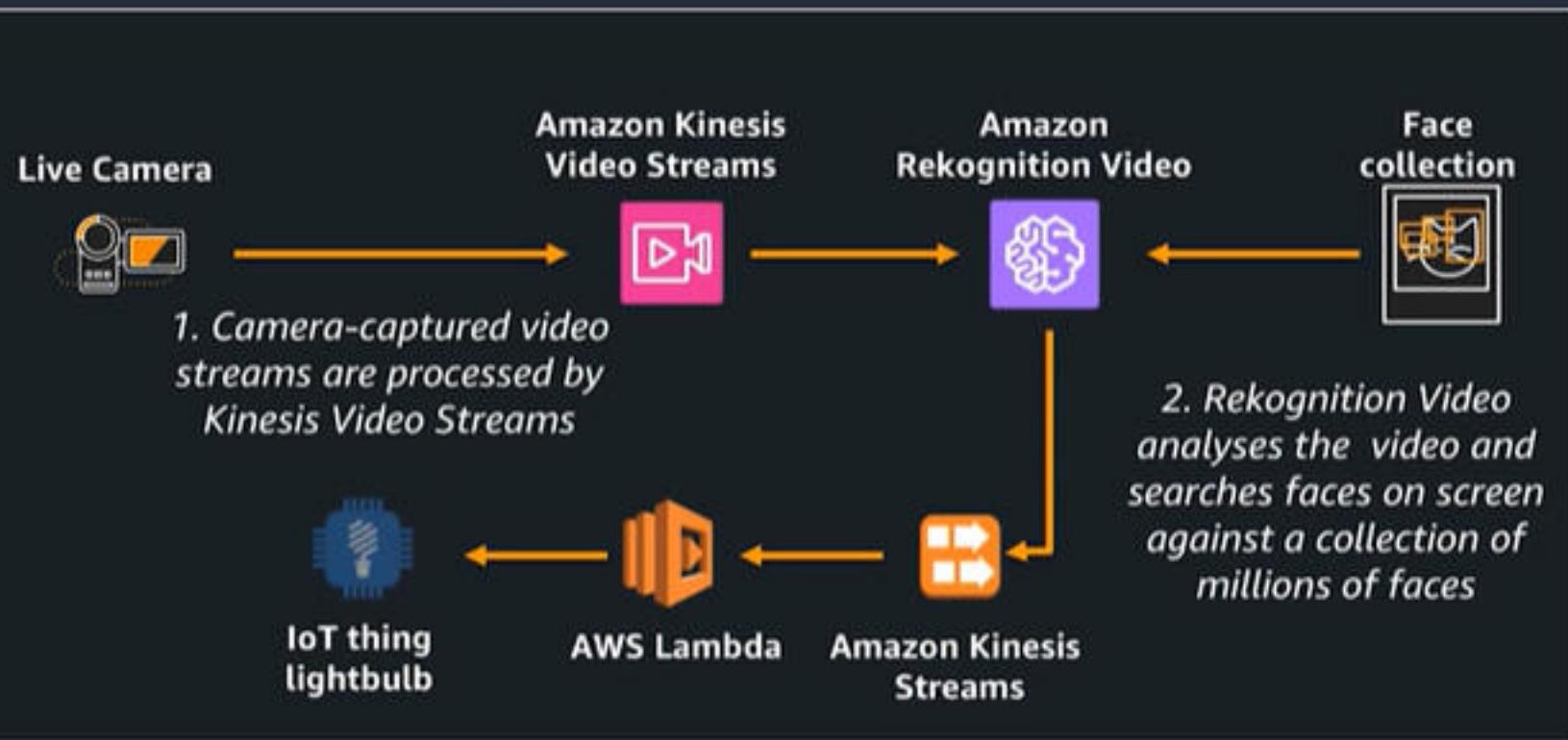
Architecture diagram



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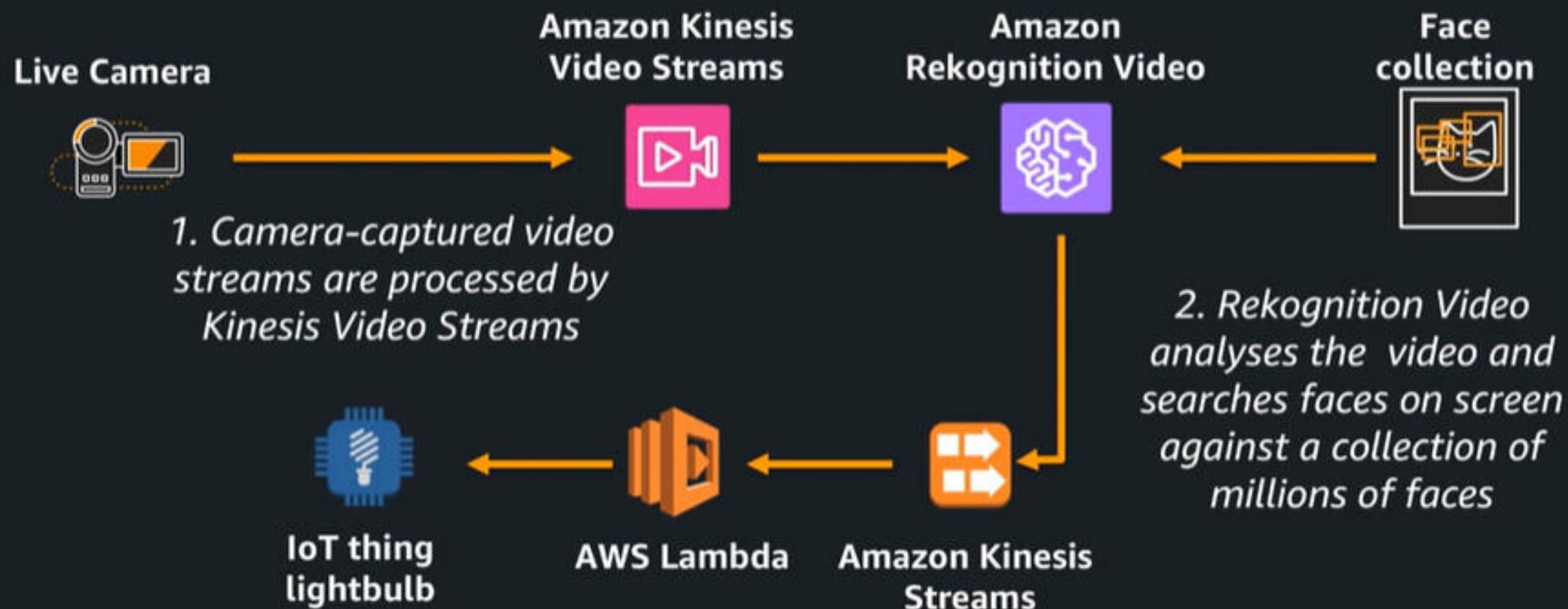
Architecture diagram



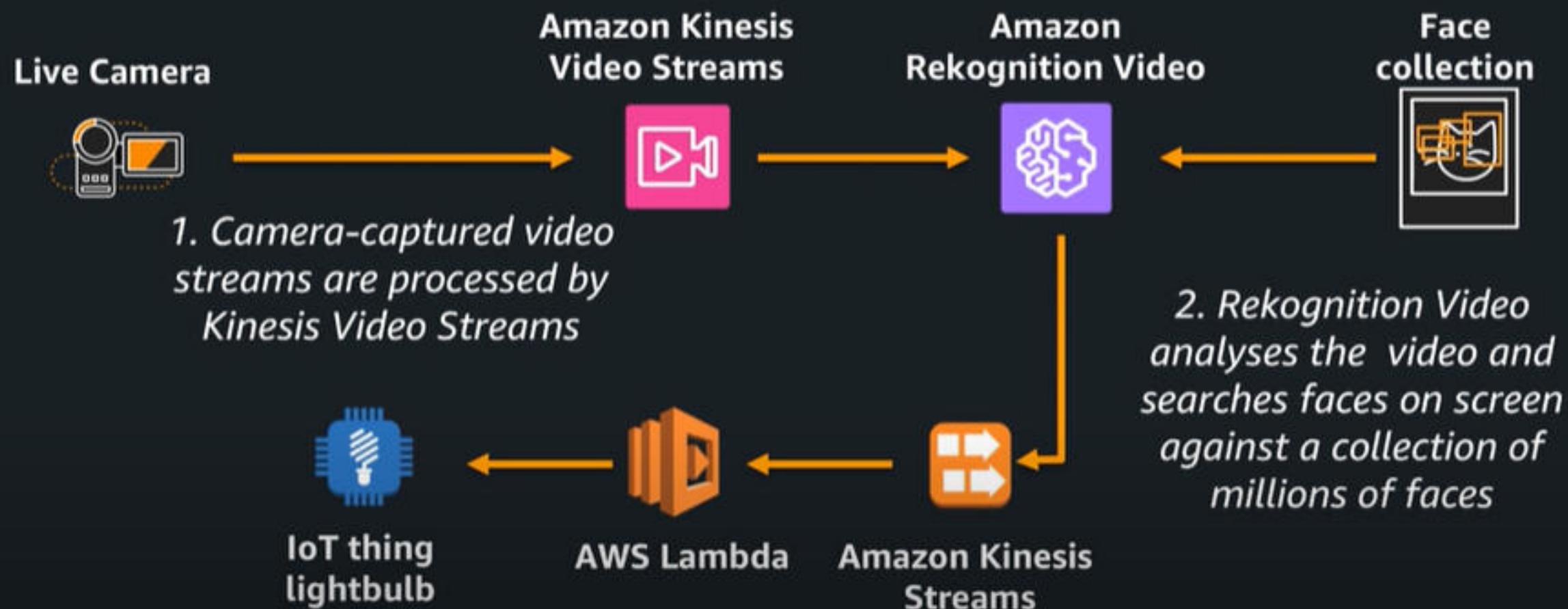
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Architecture diagram



Architecture diagram



Agenda

- Computer Vision
- Amazon Rekognition
- Amazon Rekognition API
- Use cases and Reference Architectures
- Best Practices

Computer Vision

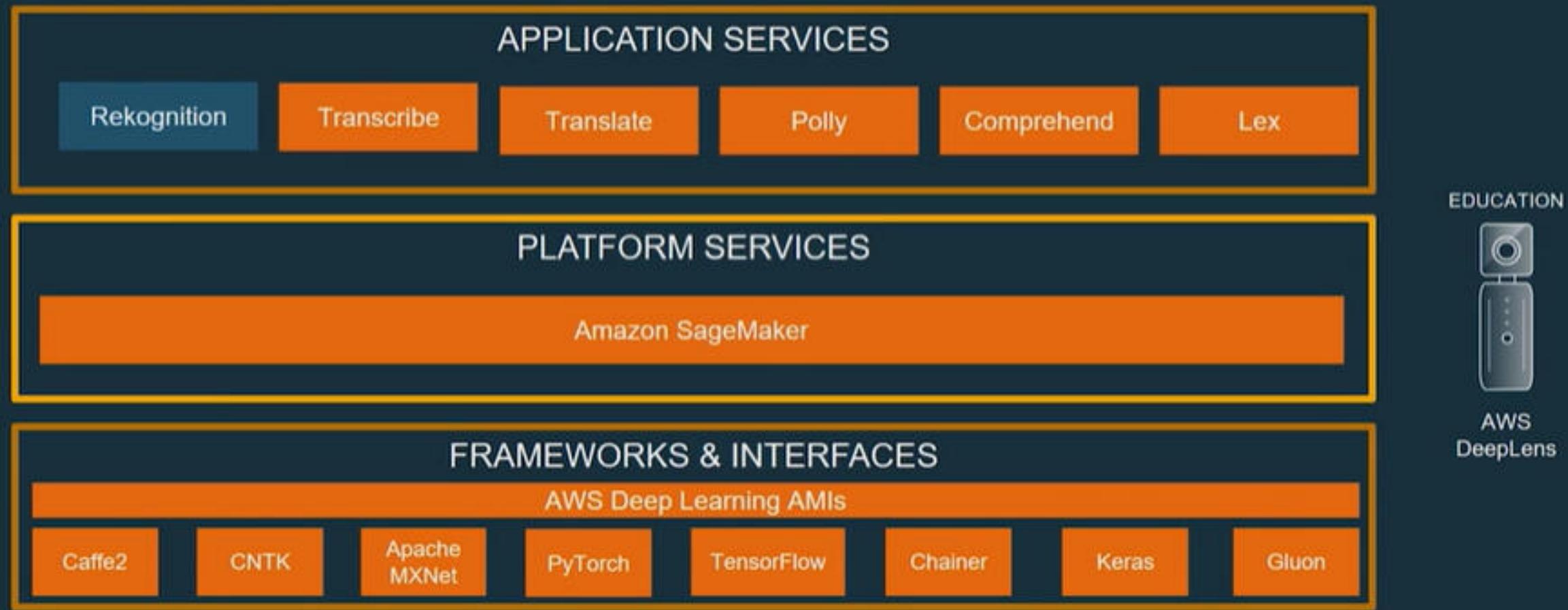


Computer Vision



```
[110 31 84 62 37 33 65 183 88 97 121 31 114 31 124 181]
[48 104 64 67 77 99 122 93 98 67 71 68 49 103 31 125]
[186 125 36 187 68 110 119 86 31 112 44 125 44 113 34 53]
[89 54 114 98 43 126 48 59 114 59 108 36 68 54 99 38]
[55 54 123 97 32 128 27 97 113 106 73 53 28 115 98 58]
[66 63 96 51 126 72 98 115 67 25 44 125 71 55 37 119]
[33 111 88 73 114 115 62 118 42 72 25 40 101 58 78 95]
[62 27 34 110 88 29 118 96 94 89 32 52 41 118 82 102]
[87 51 56 26 123 65 81 93 63 69 57 31 67 100 93 102]
[48 123 49 117 52 48 72 97 56 115 53 44 105 92 76 39]
[92 43 95 41 111 101 119 49 68 117 27 118 69 68 61 93]
[125 69 64 82 41 68 74 123 73 87 124 36 48 122 106 128]
[56 86 97 57 118 68 44 66 82 81 27 92 114 123 87 109]
[58 75 37 52 112 88 44 68 117 81 61 53 47 95 44 72]
[74 115 59 40 117 94 38 58 76 81 108 68 38 111 96 71]
[92 53 103 120 73 116 95 31 78 53 96 113 38 86 26 40]
[34 26 63 77 117 128 79 45 45 61 118 91 31 96 114 65]
[99 95 65 55 64 126 64 118 68 47 74 36 91 106 103 106]
[51 101 42 56 123 107 71 114 56 117 79 67 122 116 65 28]
[33 74 68 73 36 58 96 44 78 71 89 88 42 103 109 65]
[83 91 101 96 93 85 38 27 37 37 116 46 62 46 55 98]
[115 52 122 70 47 88 92 111 118 99 42 93 98 72 115 41]
[49 47 124 93 90 111 121 106 43 47 82 57 50 96 66 25]
[108 115 59 76 99 31 110 28 59 96 121 111 110 87 26 32]]
```

Amazon Machine Learning Stack



Amazon Rekognition Customers

Openinfluence



C-SPAN



QQ SPOKEO



HYPETAP



scrippsnetworks
interactive



MARINUS
ANALYTICS



Influential

SmugMug

THORN 1

witlee

THE TAKE

make.tv

HYPR

POPSUGAR

Limbik

zmagS



SOCIAL
SOUP

wia

SOUL

Mobilink

SEN



每日新聞

Sygic Travel

Artfinder

butterfleye



campsite
aws

Amazon Rekognition Image

Deep learning-based image analysis service



Object & Scene Detection



Facial Analysis



Face Recognition



Celebrity Recognition



Unsafe Image Detection



Text in Image

Amazon Rekognition Video

Deep learning-based video analysis service



Object & **Activity** Detection



Pathing



Face Detection & Recognition



Celebrity Recognition



Unsafe Video Detection



Real-time Live Stream

Amazon Rekognition Video

Deep learning-based video analysis service



Object & **Activity** Detection



Pathing



Face Detection & Recognition



Celebrity Recognition



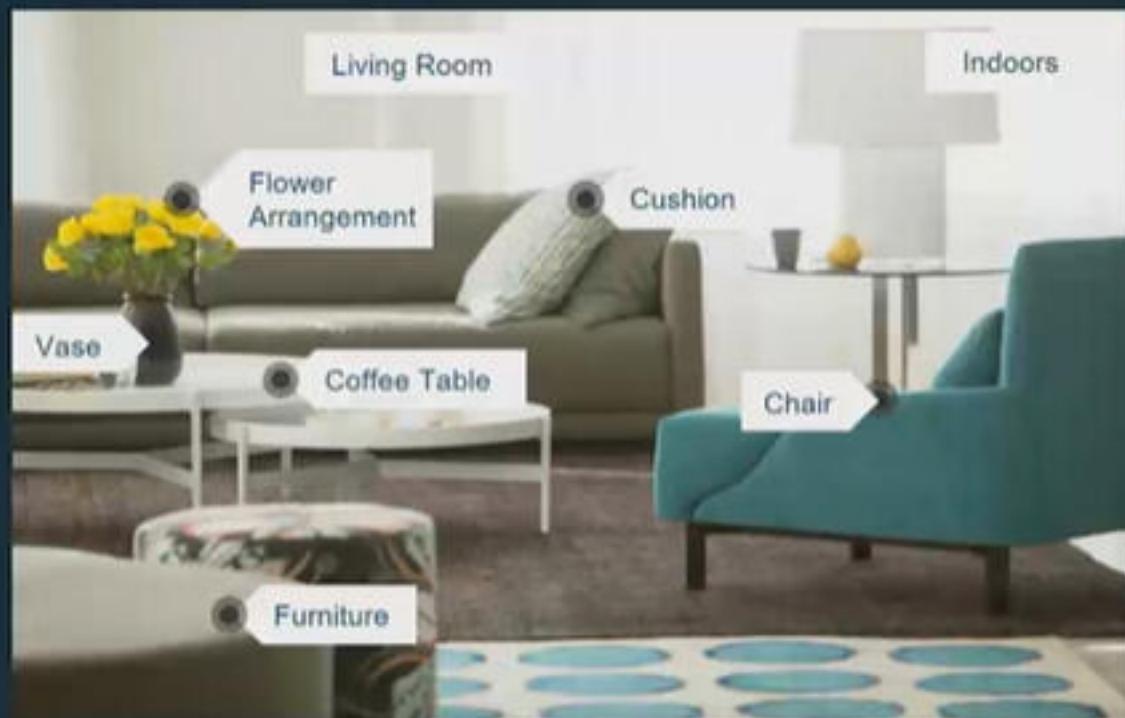
Unsafe Video Detection



Real-time Live Stream

Object & Scene Detection

Identify objects and scenes with confidence scores



Object and scene detection makes it easy for you to add features that search, filter, and curate large image libraries.

Facial Analysis

Analyze facial characteristics in multiple dimensions

Demographic Data

Emotion Expressed

Facial Landmarks

General Attributes

Image Quality

Facial Pose

Facial Analysis

Analyze facial characteristics in multiple dimensions

Demographic Data

Age Range 29-45
Gender:Male 96.5%

Facial Landmarks

EyeLeft, EyeRight, Nose
RightPupil, LeftPupil
MouthRight, LeftEyeBrowUp
Bounding Box...

Image Quality

Brightness 23.6%
Sharpness 99.9%



Emotion Expressed

Happy 83.8%
Surprised 0.65%

General Attributes

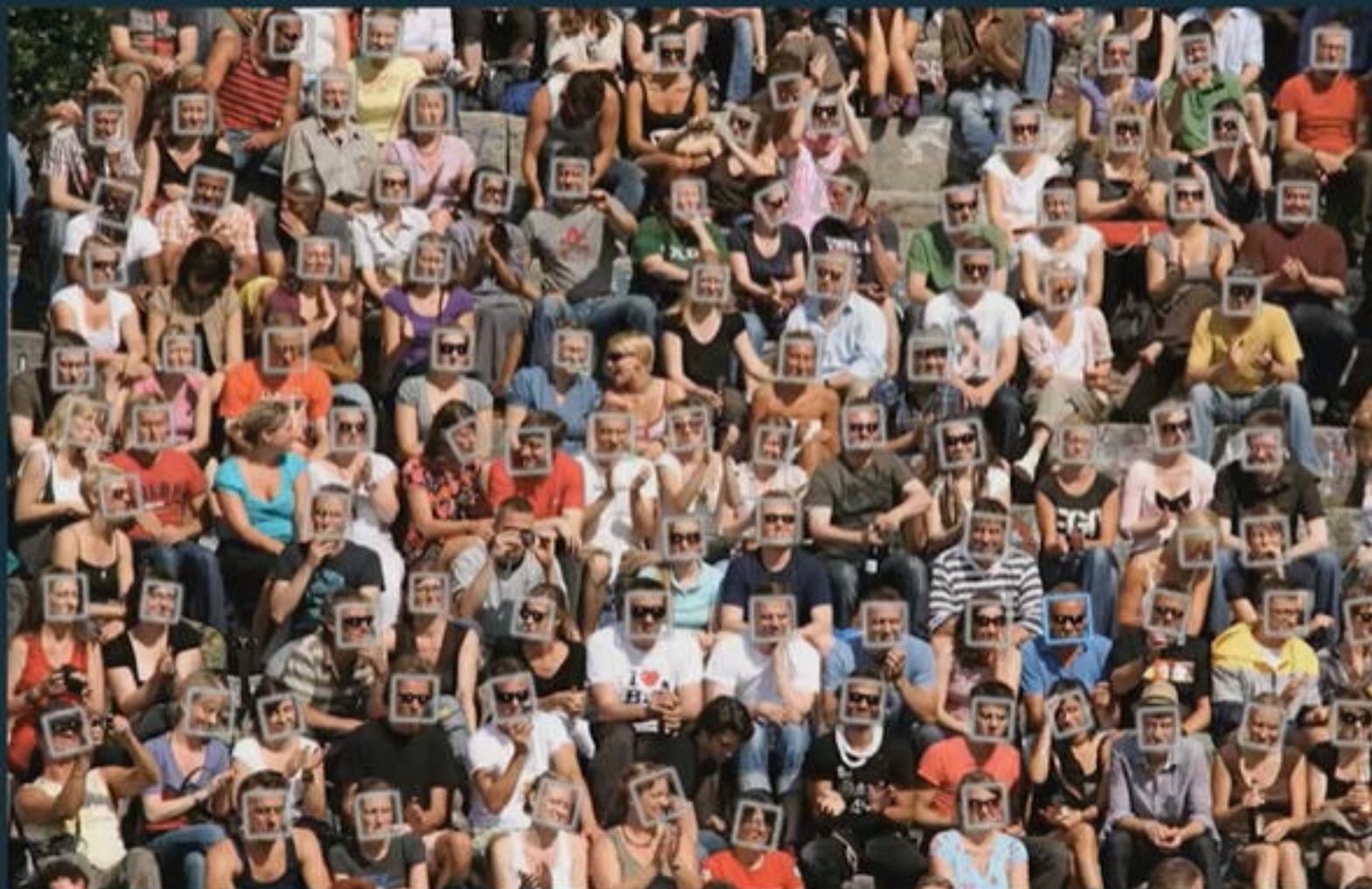
Smile:True 23.6%
EyesOpen:True 99.8%
Beard:True 99.5%
Mustache:True 99.9%...

Facial Pose

Pitch 1.446
Roll 5.725
Yaw 4.383

Facial Detection and Analysis - Crowd

Support for up to 100 faces



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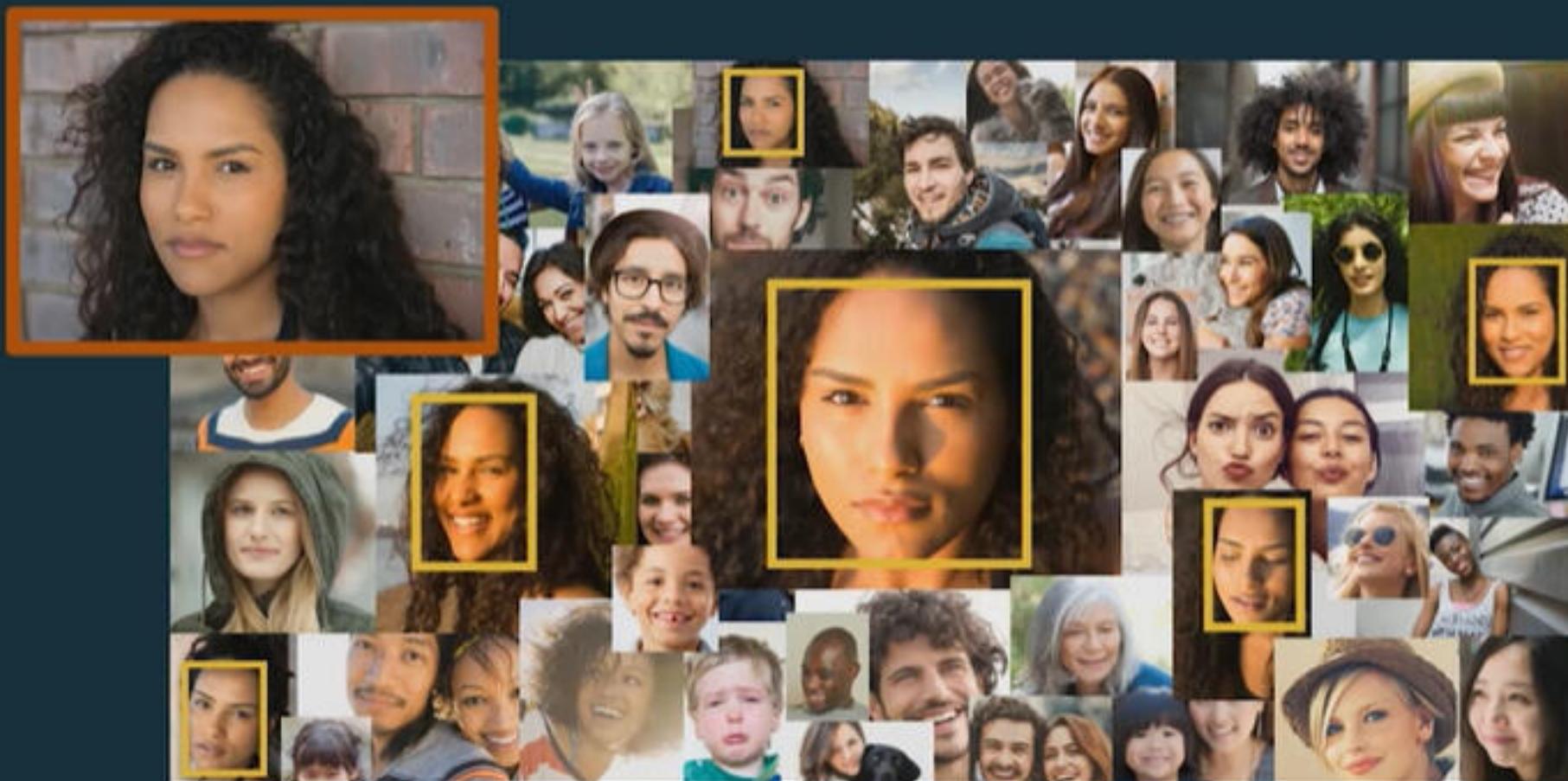
Face Comparison

Measure the likelihood that faces are of the same person



Facial Recognition

Identify a person in a photo or video using your private repository of face images



Unsafe Content Detection

Detect explicit and suggestive content



Explicit and Suggestive Content Labels

| Top-Level Category | Second-Level Category |
|--------------------|------------------------------|
| Explicit Nudity | Nudity |
| | Graphic Male Nudity |
| | Graphic Female Nudity |
| | Sexual Activity |
| | Partial Nudity |
| Suggestive | Female Swimwear or Underwear |
| | Male Swimwear or Underwear |
| | Revealing Clothes |

Explicit and Suggestive Content Labels

| Top-Level Category | Second-Level Category |
|--------------------|------------------------------|
| Explicit Nudity | Nudity |
| | Graphic Male Nudity |
| | Graphic Female Nudity |
| | Sexual Activity |
| | Partial Nudity |
| Suggestive | Female Swimwear or Underwear |
| | Male Swimwear or Underwear |
| | Revealing Clothes |

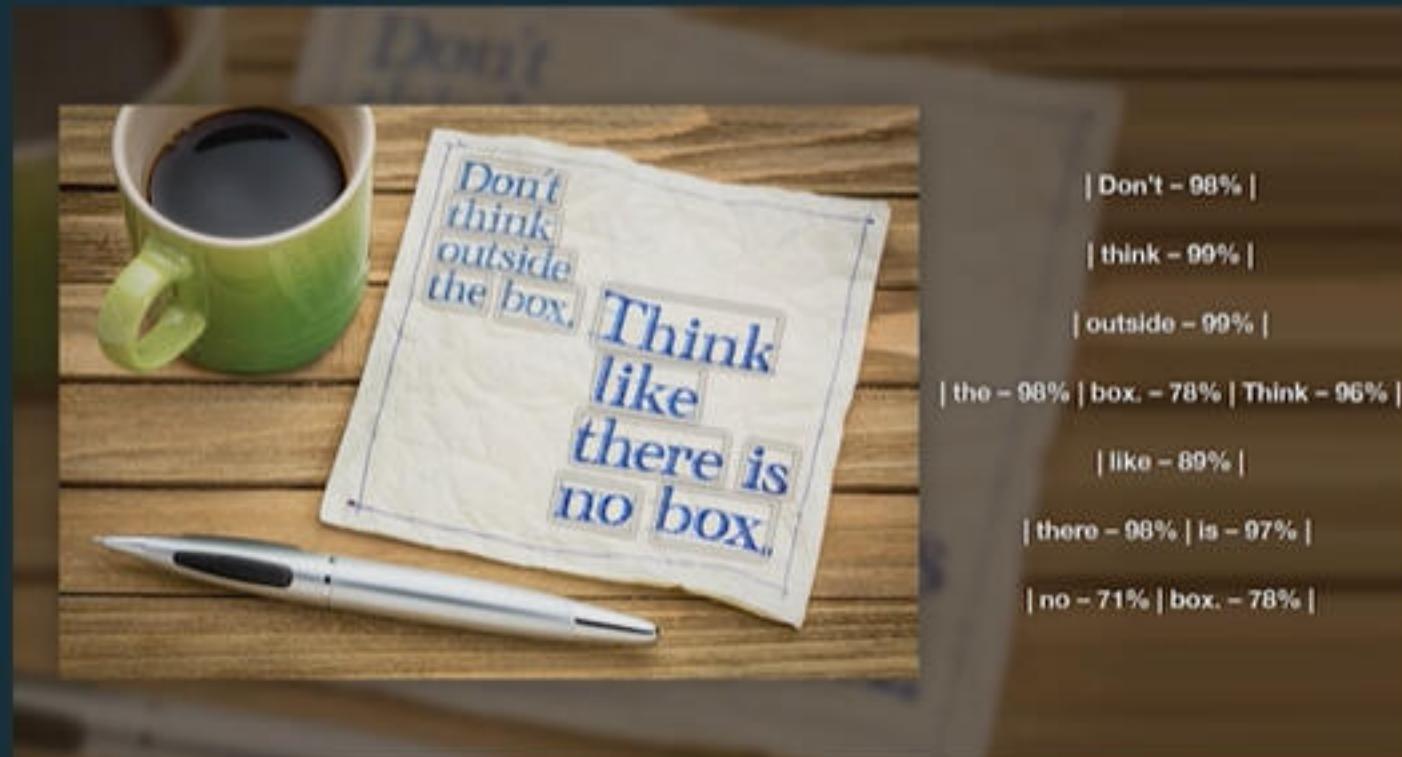
Celebrity Recognition

Recognize thousands of famous individuals



Detecting Text

Detect and recognize text from images



Pathing

Capture the path of people in the scene



Amazon Rekognition API



Object & Scene Detection – Image API



```
{  
  "Image": {  
    "Bytes": blob,  
    "S3Object": {  
      "Bucket": "string",  
      "Name": "string",  
      "Version": "string"  
    }  
  },  
  "MaxLabels": number,  
  "MinConfidence": number  
}
```



DetectLabels

Object & Scene Detection – Image API



```
{  
  "Image": {  
    "Bytes": blob,  
    "S3Object": {  
      "Bucket": "string",  
      "Name": "string",  
      "Version": "string"  
    }  
  },  
  "MaxLabels": number,  
  "MinConfidence": number  
}
```

DetectLabels

```
{  
  "Labels": [  
    {  
      "Confidence": number,  
      "Name": "string"  
    }  
  ],  
  "OrientationCorrection": "string"  
}
```

Object & Scene Detection – Video API

```
{  
    "ClientRequestToken": "string",  
    "JobTag": "string",  
    "MinConfidence": number,  
    "NotificationChannel": {  
        "RoleArn": "string",  
        "SNSTopicArn": "string"  
    },  
    "Video": {  
        "S3Object": {  
            "Bucket": "string",  
            "Name": "string",  
            "Version": "string"  
        }  
    }  
}
```



StartLabelDetection

Object & Scene Detection – Video API

```
{  
    "ClientRequestToken": "string",  
    "JobTag": "string",  
    "MinConfidence": number,  
    "NotificationChannel": {  
        "RoleArn": "string",  
        "SNSTopicArn": "string"  
    },  
    "Video": {  
        "S3Object": {  
            "Bucket": "string",  
            "Name": "string",  
            "Version": "string"  
        }  
    }  
}
```



StartLabelDetection

Object & Scene Detection – Video API

```
{  
    "ClientRequestToken": "string",  
    "JobTag": "string",  
    "MinConfidence": number,  
    "NotificationChannel": {  
        "RoleArn": "string",  
        "SNSTopicArn": "string"  
    },  
    "Video": {  
        "S3Object": {  
            "Bucket": "string",  
            "Name": "string",  
            "Version": "string"  
        }  
    }  
}
```



StartLabelDetection



Object & Scene Detection – Video API



```
{  
    "ClientRequestToken": "string",  
    "JobTag": "string",  
    "MinConfidence": number,  
    "NotificationChannel": {  
        "RoleArn": "string",  
        "SNSTopicArn": "string"  
    },  
    "Video": {  
        "S3Object": {  
            "Bucket": "string",  
            "Name": "string",  
            "Version": "string"  
        }  
    }  
}
```

StartLabelDetection

JobId

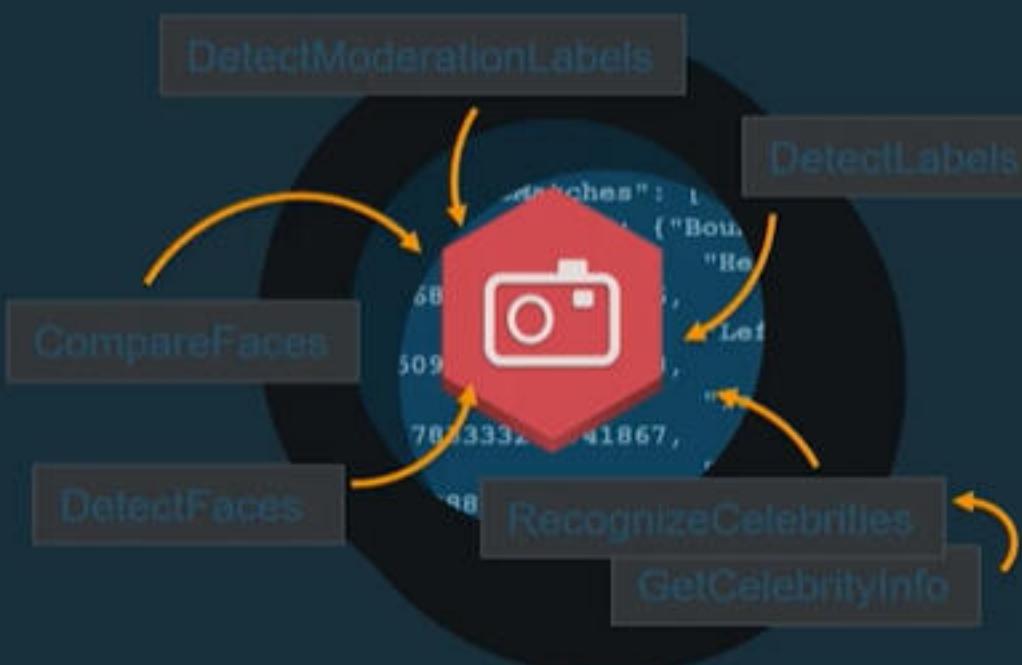
GetLabelDetection

```
{  
    "JobStatus": string,  
    "StatusMessage": string,  
    "VideoMetadata": {  
        "Format": string,  
        "Codec": string,  
        "DurationMillis": number,  
        "FrameRate": float,  
        "FrameWidth": number,  
        "FrameHeight": number  
    },  
    "NextToken": string,  
    "Labels": [  
        {  
            "Timestamp": number,  
            "Label": {  
                "Name": string,  
                "Confidence": float  
            }  
        }  
    ],  
    ...  
}
```

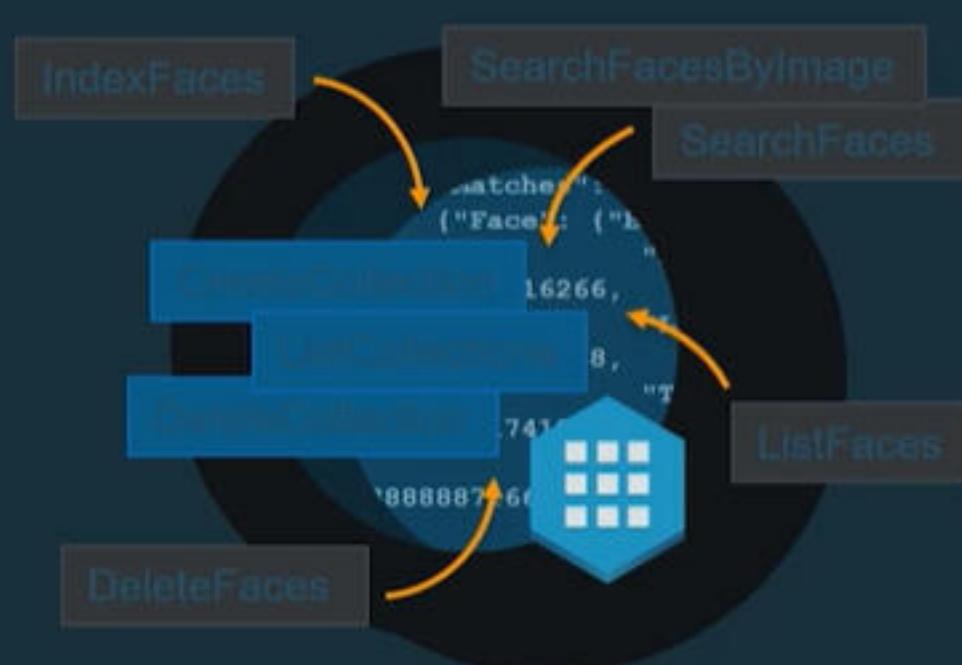


Rekognition APIs – Overview

Non-storage API Operations



Storage-based API Operations



Rekognition's computer vision API operations can be grouped into Non-storage API operations, and Storage-based API operations

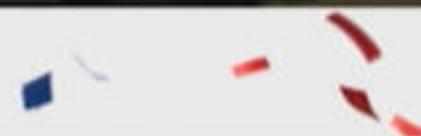
Celebrity Guests at the Royal Wedding



- “Who’s Who Live” function
- Celebrity guests identified in live stream
- On Screen captions of relation to the royal couple



PROFILES



Princess Michael of Kent

The wife of Prince Michael, the Queen's cousin, was born in 1945 in what was part of Germany, now the Czech Republic. She was an interior designer before becoming an author.

[Watch arrival](#)



Will Greenwood



Caroline Greenwood



Duchess of York



Norma Major



Sir John Major



Mike Tindall



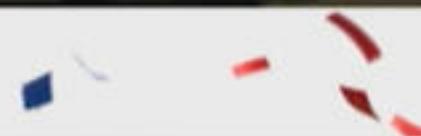
Princess Michael of Kent



Autumn Phillips



PROFILES



Princess Michael of Kent

The wife of Prince Michael, the Queen's cousin, was born in 1945 in what was part of Germany, now the Czech Republic. She was an interior designer before becoming an author.

[Watch arrival](#)



Will Greenwood



Caroline Greenwood



Duchess of York



Norma Major



Sir John Major



Mike Tindall



Princess Michael of Kent



Autumn Phillips

sky news



sky news

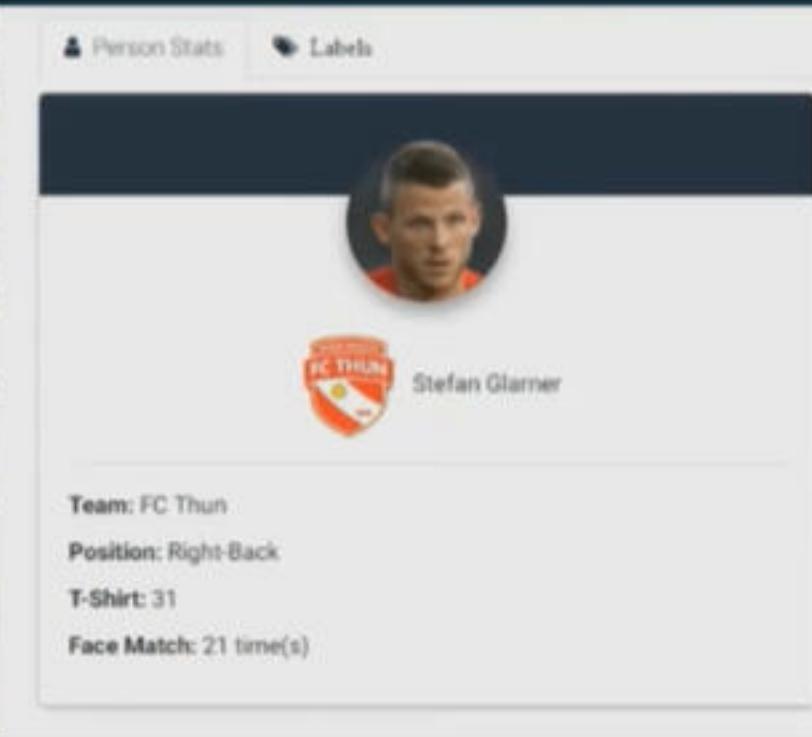


Amazon Rekognition and sports media tagging

- Player recognition
- Motion path tracking
- Objects, activities, and event detection



The image shows a soccer match in progress on a green field. Several players are tracked by orange boxes labeled with their names and IDs: Person 14, Person 24, Person 13, Person 19, Person 20, Person 18, Person 17, and Person 11. Motion paths are shown as dashed lines connecting these boxes across the frame. A player in a white jersey is highlighted in yellow.



A detailed player profile for Stefan Glarner from FC Thun. It includes:

- Person Stats
- Labels
- Stefan Glarner (Profile Picture)
- FC THUN (Team Logo)
- Team: FC Thun
- Position: Right-Back
- T-Shirt: 31
- Face Match: 21 time(s)

Combatting Human Trafficking on AWS

ML supporting law enforcement and victim rescue



THORN

- Machine learning and analytics platforms for law enforcement
- Match photos of exploited children to those on the dark web
- Reduces the time and effort to identify and rescue victims

Combatting Human Trafficking on AWS

ML supporting law enforcement and victim rescue



THORN

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Combatting Human Trafficking on AWS

ML supporting law enforcement and victim rescue



THORN

- Machine learning and analytics platforms for law enforcement
- Match photos of exploited children to those on the dark web
- Reduces the time and effort to identify and rescue victims

Analyze User Generated Content



Person 99.2%
Dog 95.7%

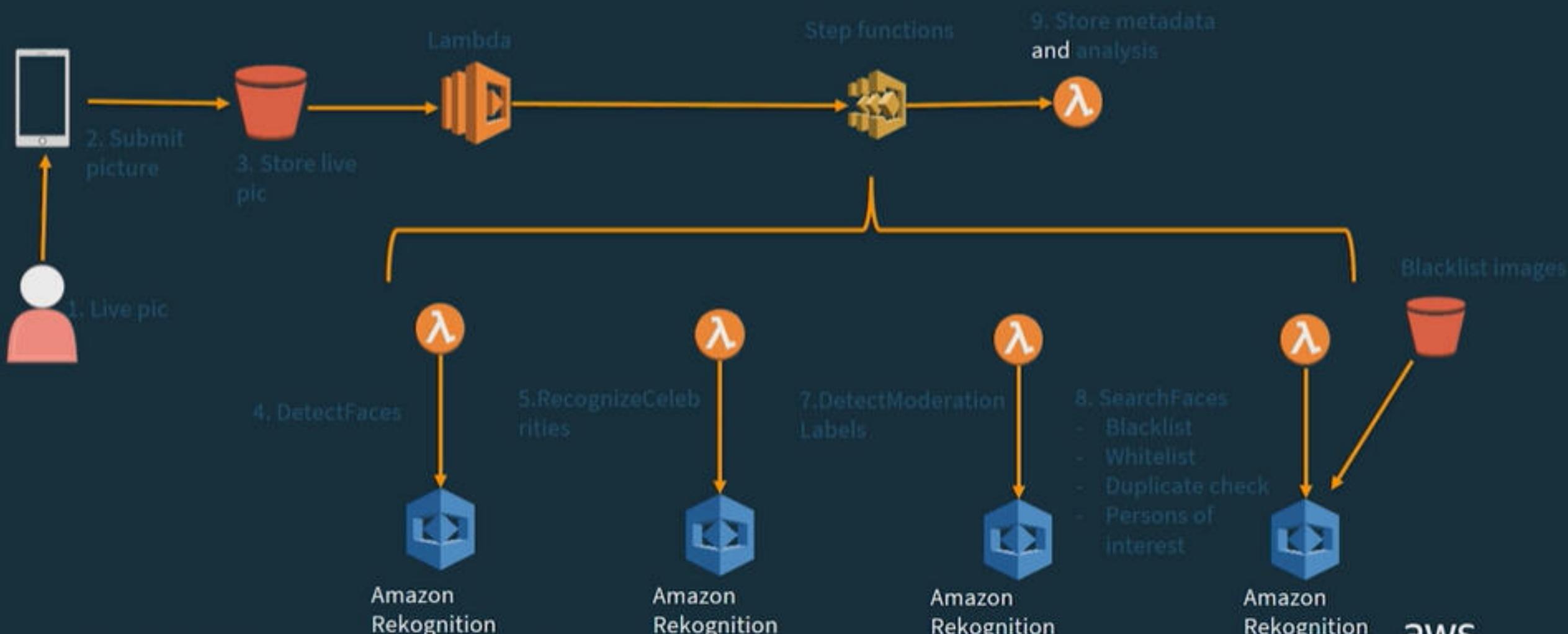


Person 99.2%
Snowboarding 98.1%

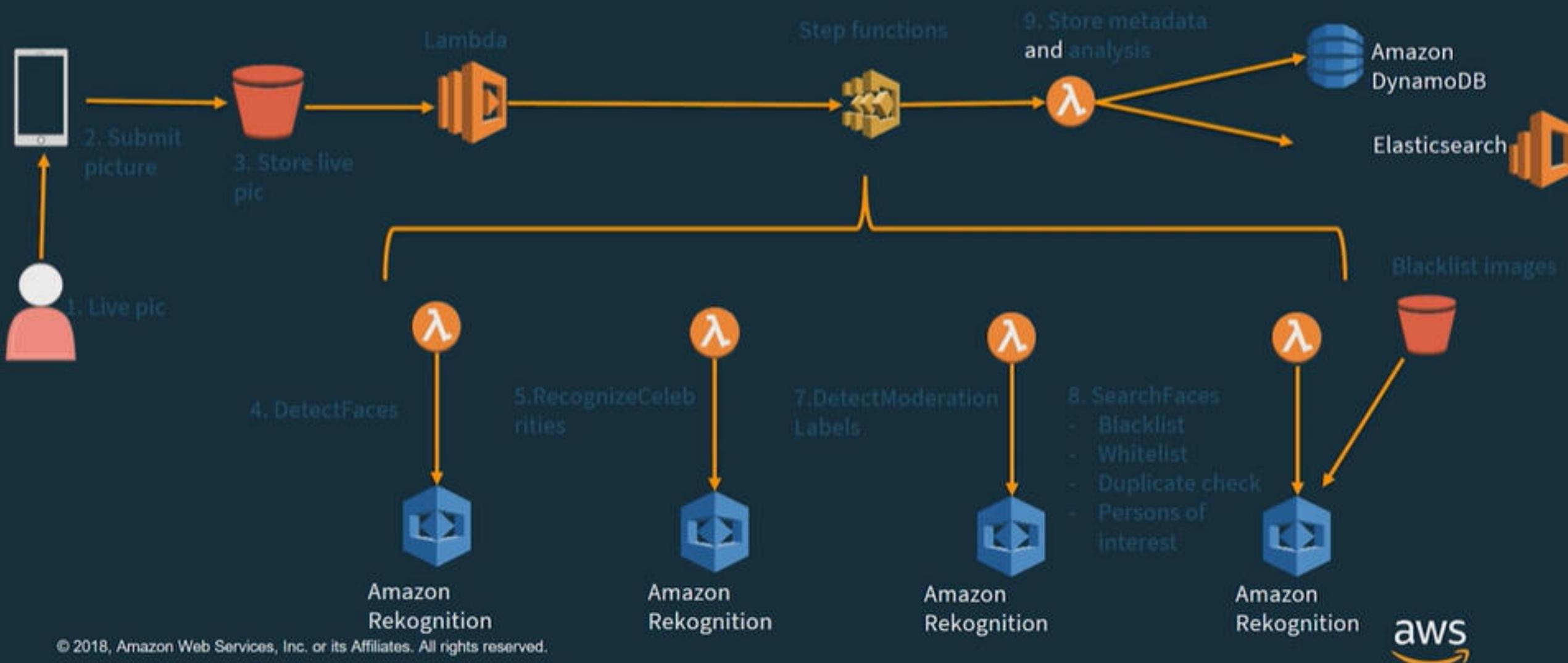


No Moderation labels
detected

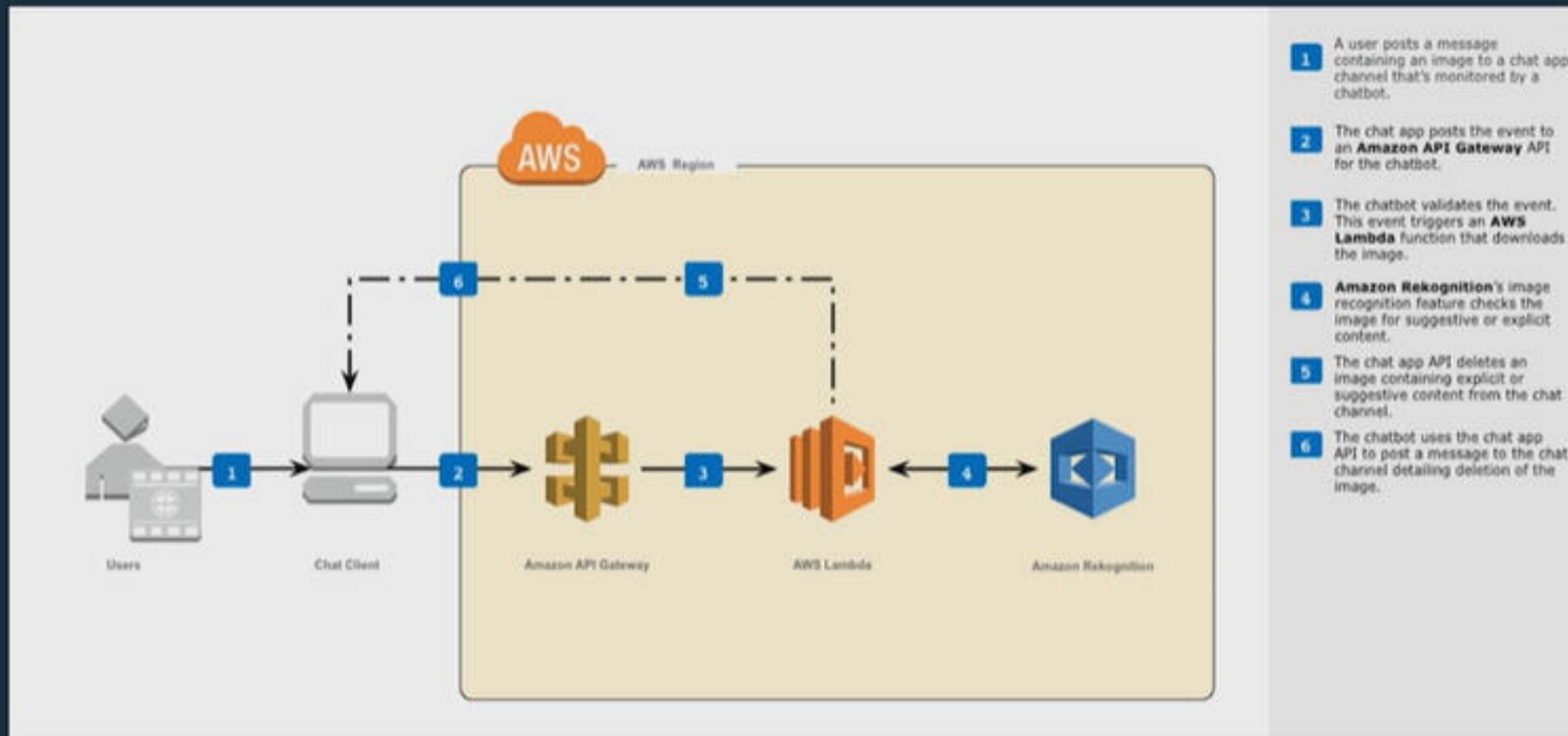
AWS solution



AWS solution



Chat moderation



<https://github.com/aws-samples/lambda-rekognition-image-moderation-chatbot>

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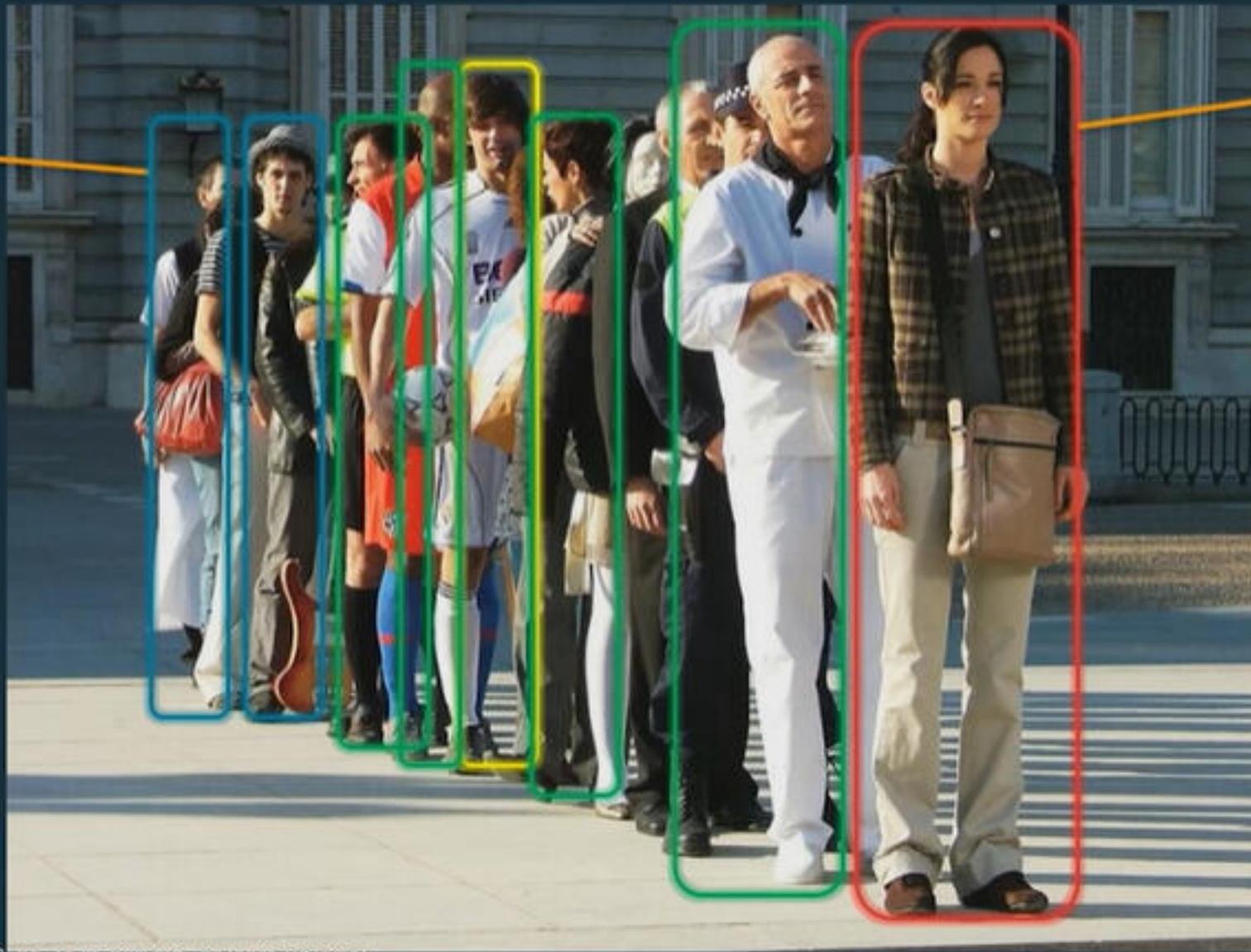
How long people have to wait in the line?



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aws
Amazon

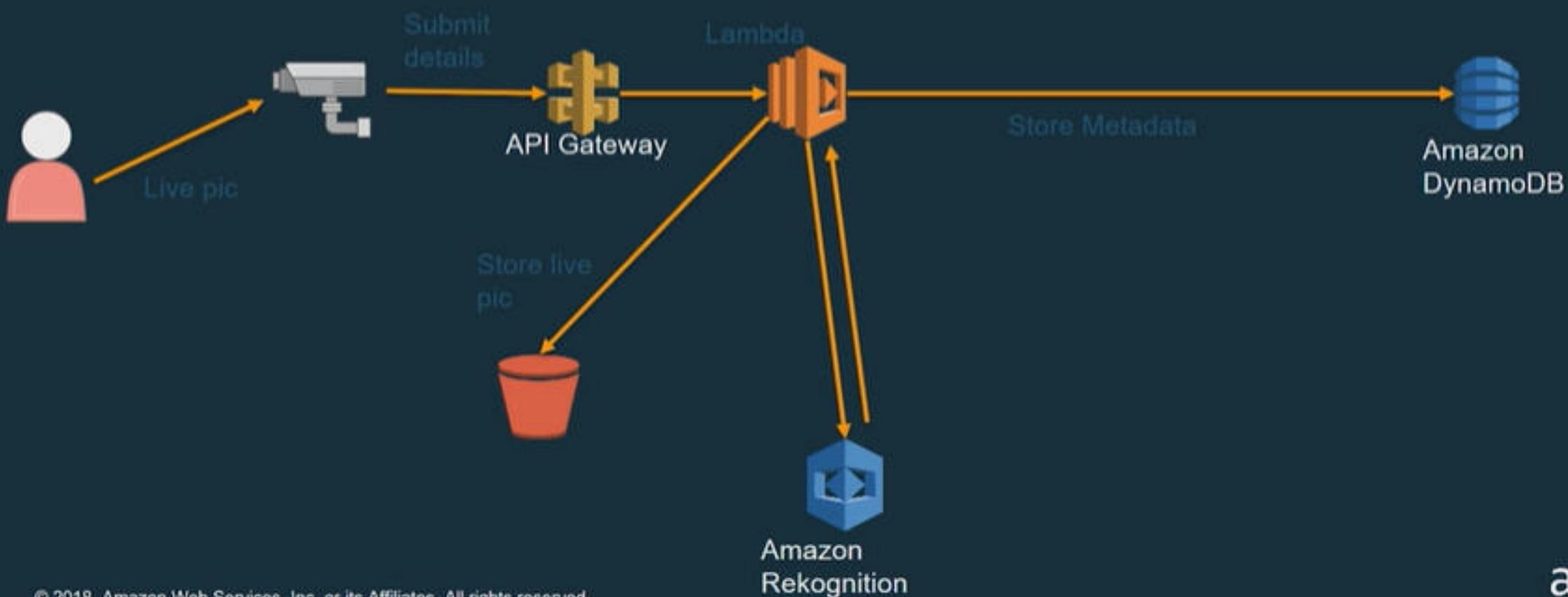
How long people have to wait in the line?



Live demographic analysis



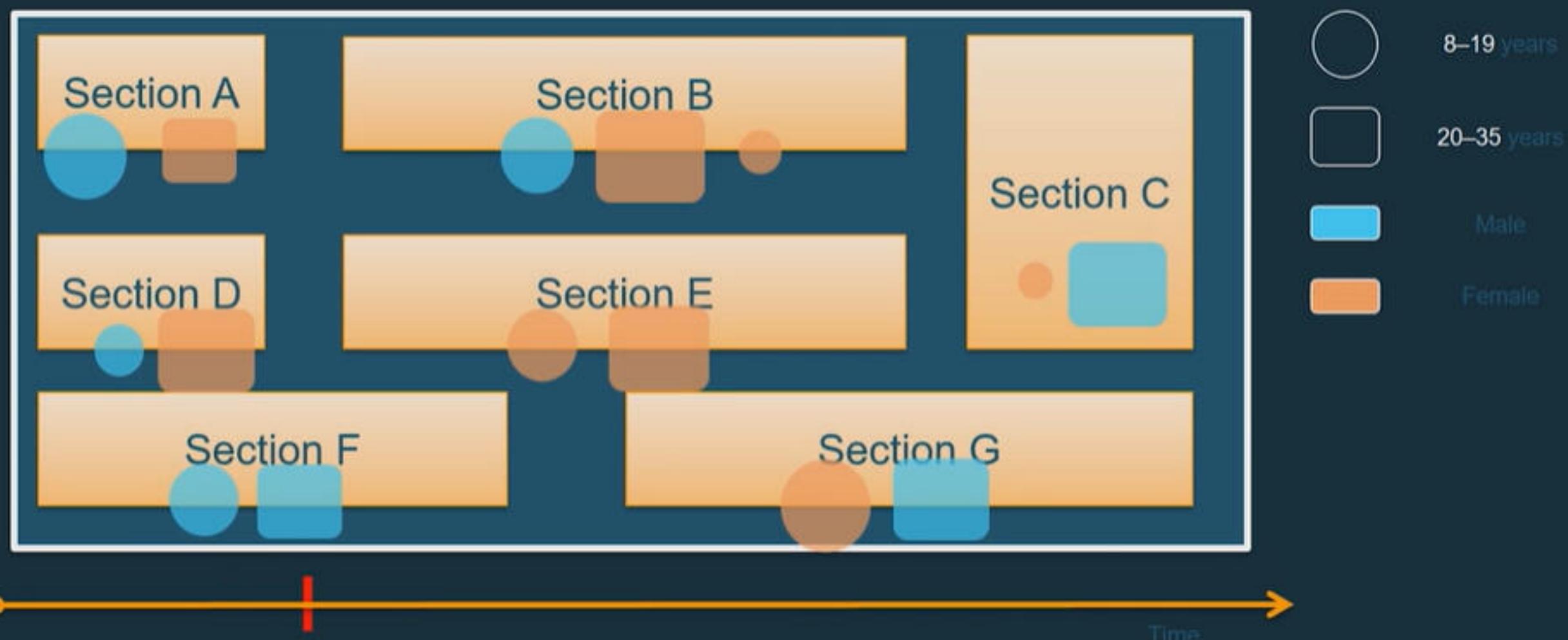
AWS solution



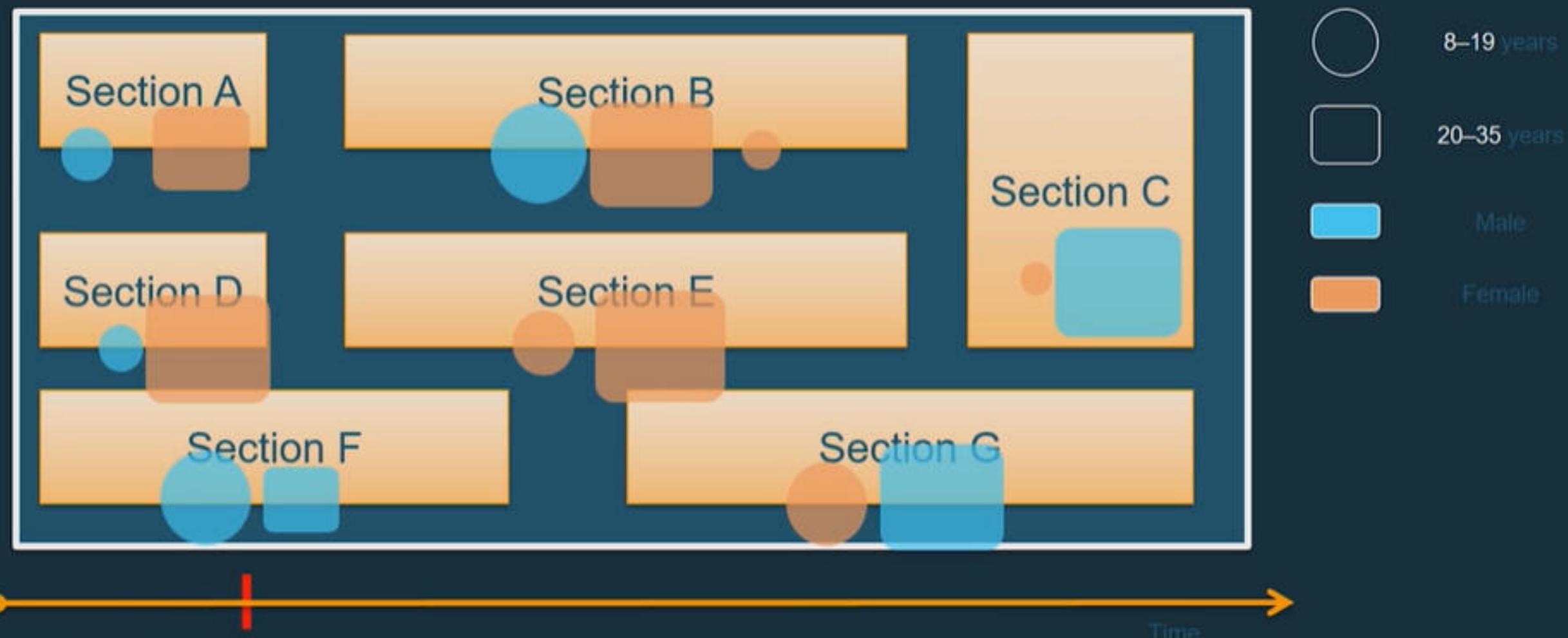
Live demographic analysis



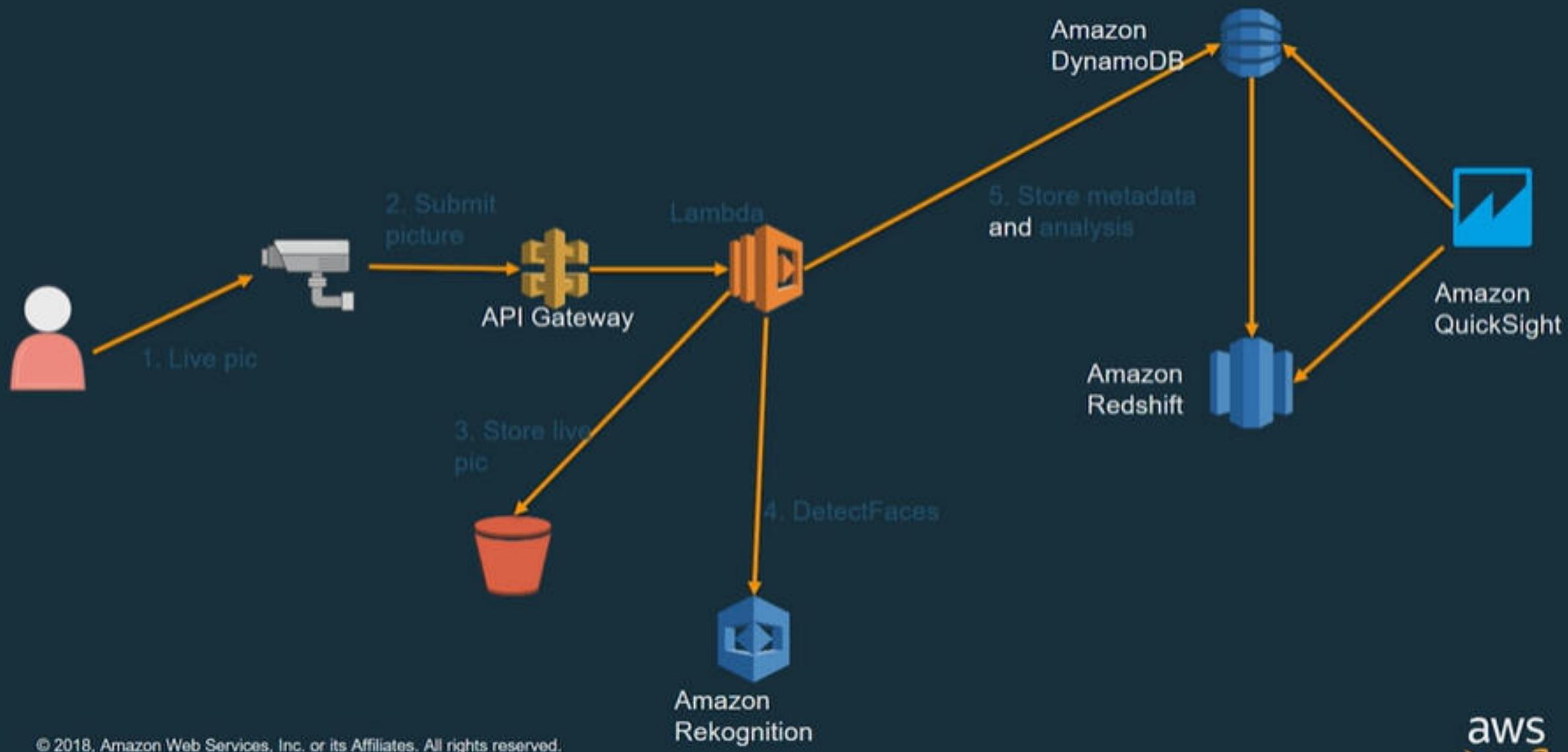
Real-time store heat map



Real-time store heat map

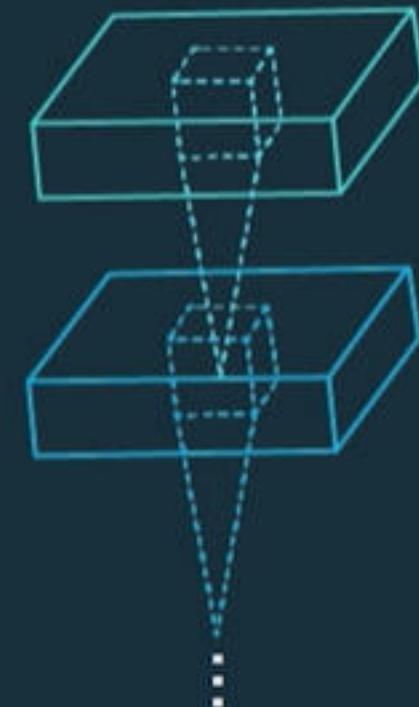


AWS solution



Best practices: Interfacing with Amazon Rekognition

- Image format
 - PNG or JPEG
- Max image size
 - Amazon S3 : 15 MB
 - API calls: 5 MB (base64 encoded)
- Video format
 - mp4, mov
- Video Codec
 - H264
- Max video size
 - 8 GB



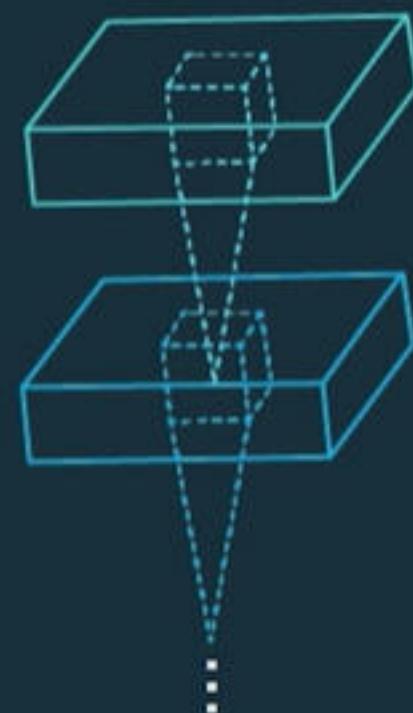
Best practices: Interfacing with Amazon Rekognition

- Image resolution
 - Min 80 px
- Collections are for faces! (not cats, cartoons, ...)
- Max number of faces in a single face collection is 20 million
 - Latency still under a second!
- Max matching faces the search API returns = 4096
- Keep your images to re-index collection for future versions



Best practices: Interfacing with Amazon Rekognition

- IndexFaces detects largest 100 faces in the input image and adds them to the specified collection
- SearchFacesByImage detects the largest face in the image, and then searches the specified collection for matching faces



Media Analysis Solution

Media Analysis Solution Upload Browse Settings

introducing_amazon_translate.mp4

The screenshot shows a video player interface for a file named "introducing_amazon_translate.mp4". The video frame displays a presentation slide about Amazon Transcribe, featuring a man speaking and several icons representing features like multi-language support and support for multiple speakers. The interface includes a "Currently tracking: Persons" section with control buttons for Play, Pause, Restart, and Captions. Below this is a "Click to track:" section with buttons for Persons, Faces, Celebrities, and Known Faces, where "Known Faces" is highlighted. At the bottom, there are tabs for Labels, Facial Attributes, Known Faces, Celebrities, Transcript, Entities, and Phrases, followed by a list of detected entities.

Labels Facial Attributes Known Faces Celebrities Transcript Entities Phrases

transcribe longform audio files text customers the world several different languages transcribed audiotext a lot
different languages the way people this problem translation agencies your expensive and time

<https://aws.amazon.com/answers/media-entertainment/media-analysis-solution/>

References

- [Amazon Rekognition](#)
- [Amazon Rekognition on AWS AI blog](#)
- [Media Analysis Solution](#)
- [Rekognition-Lambda-SQS](#)
- [Custom Celebrities](#)
- [Serverless Image Recognition processing backend](#)
- [Policing user content](#)
- [API Reference](#)
 - [Actions](#)
 - [Data Types](#)



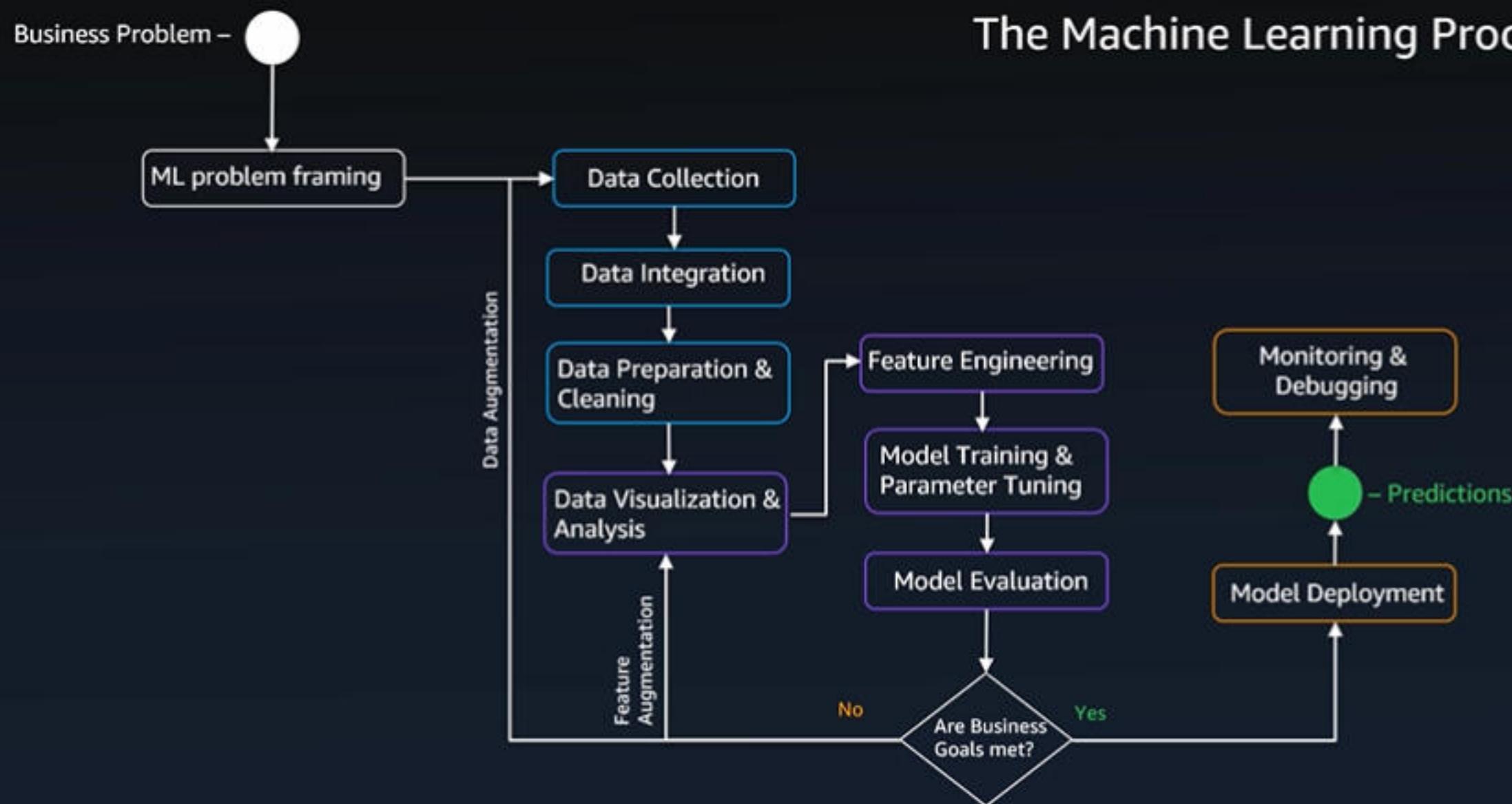
Amazon SageMaker

Satellite Image Classification In SageMaker

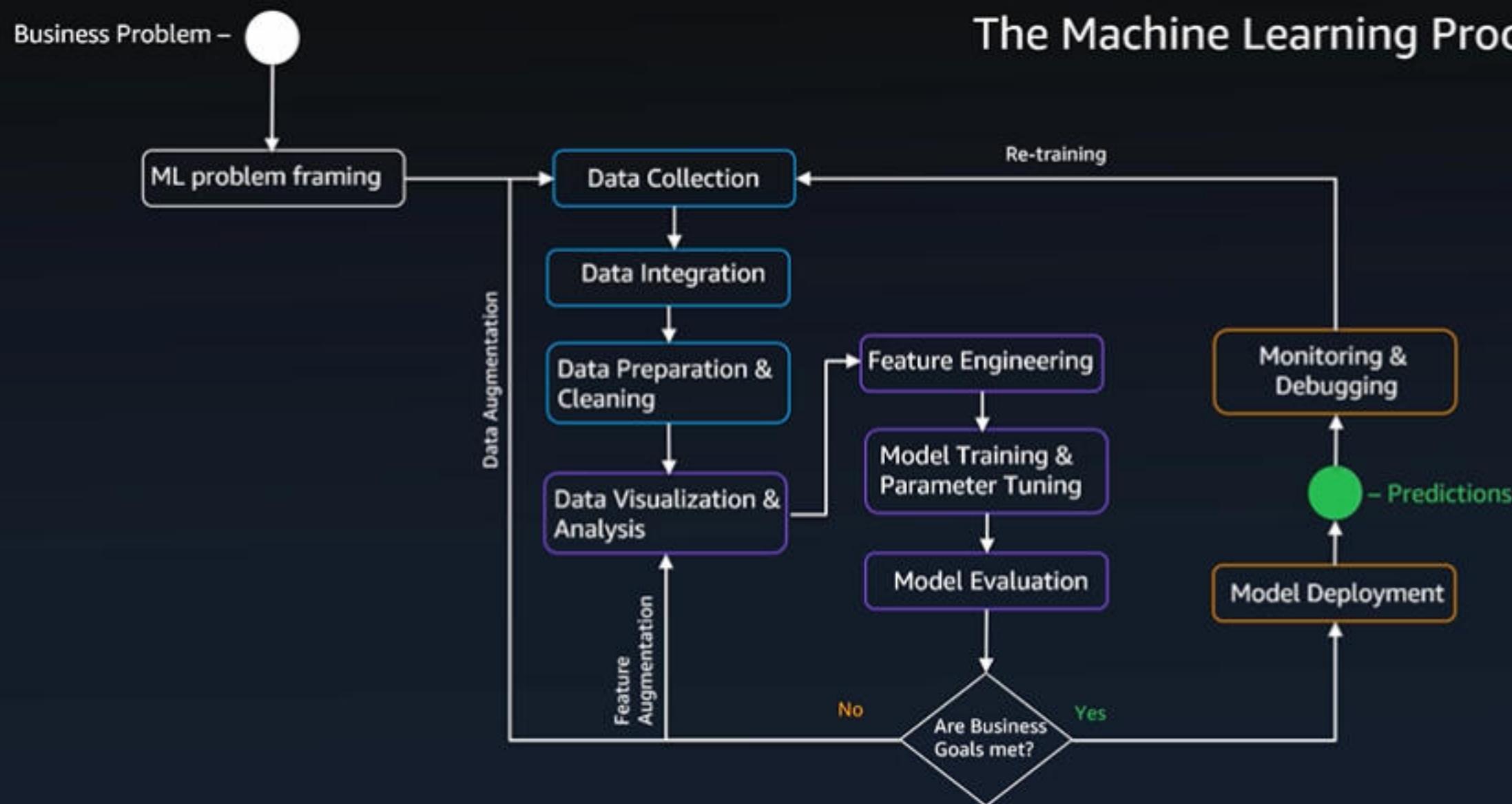
Bardia Nikpourian – Sr. Specialist TAM – Artificial Intelligence

Varun Jain - Global Accounts Solutions Architect

The Machine Learning Process



The Machine Learning Process



Satellite Image Recognition: The Business Problem

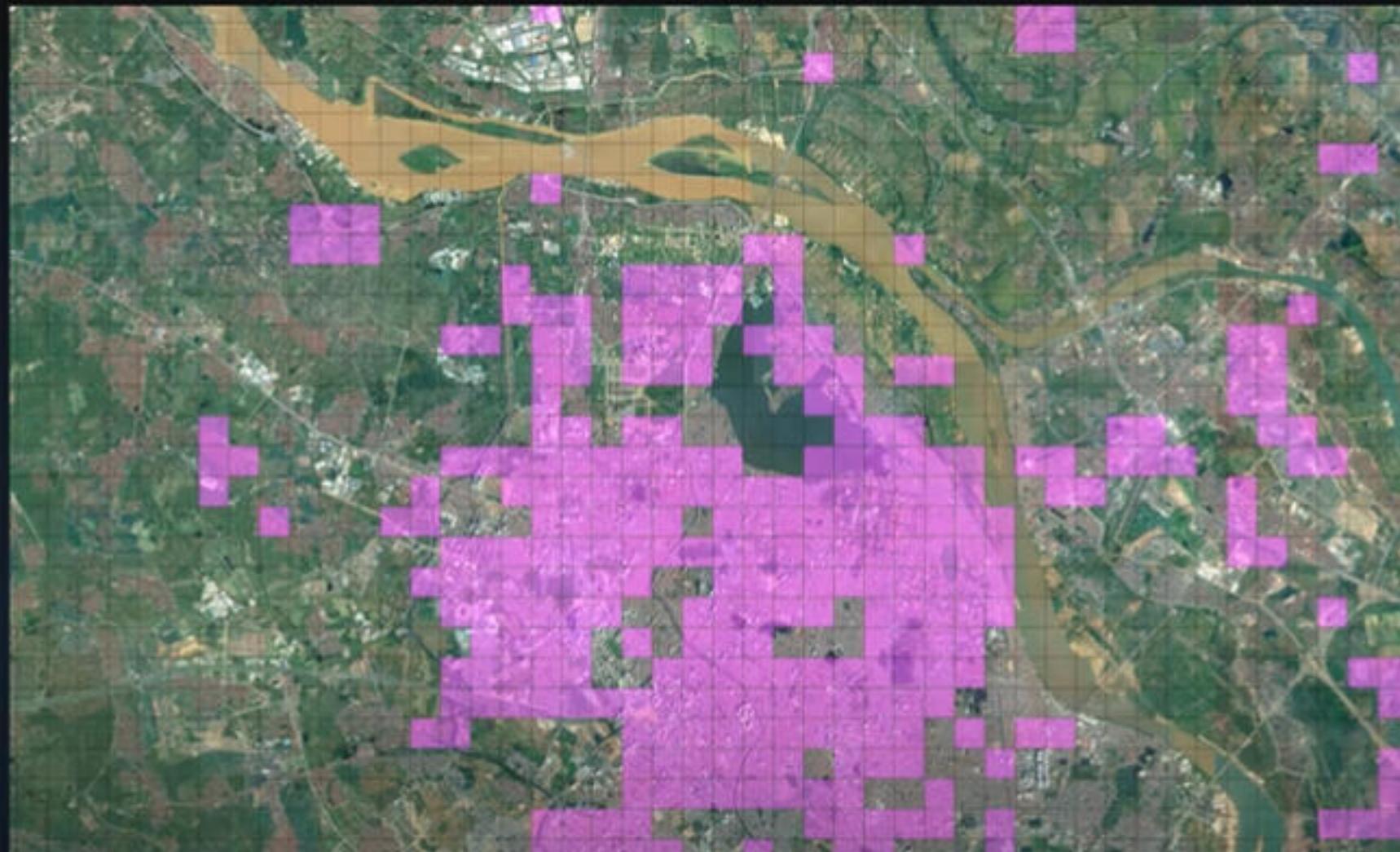
Problem Statement :

- Identifying and counting the number of specific items in an image is a human cognition task that's both costly and inefficient.

ML Problem Framing:

- Create a machine learning model capable of *classifying* objects such as roads, automobiles, and buildings in images that can be operated in production and be retrained as new data and locales become available.

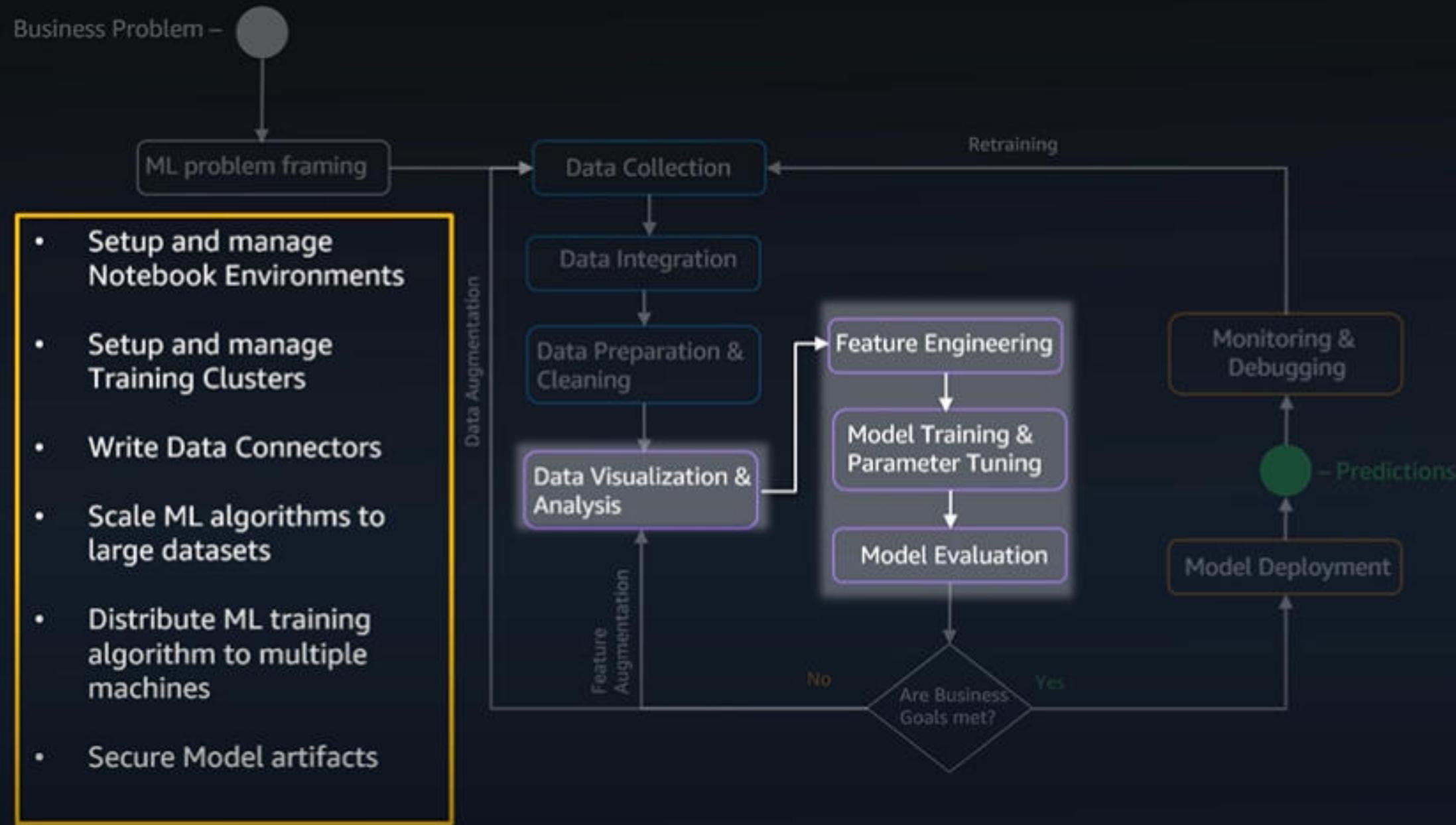
Data Preparation : Labeling Digital Globe Images



<https://github.com/developmentseed/label-maker>

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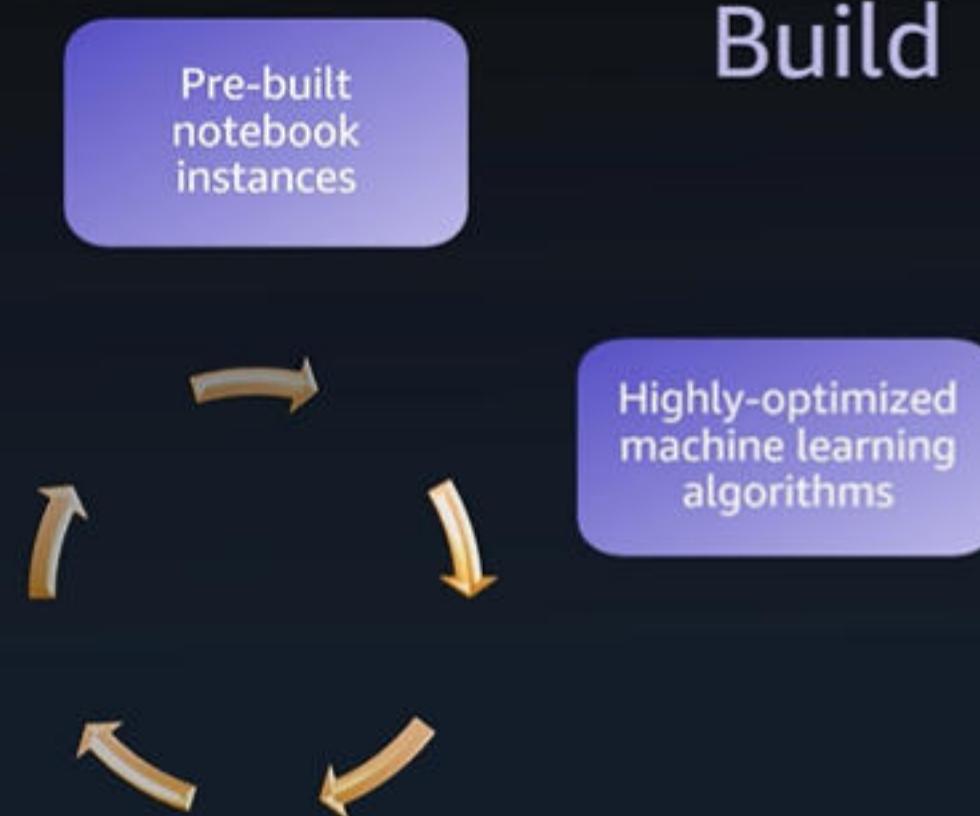


Amazon SageMaker

A **fully managed service** that enables **data scientists** and **developers** to quickly and easily **build** machine-learning based models **into production** smart applications.

Amazon SageMaker

Build



Amazon SageMaker

Build

Pre-built
notebook
instances

Highly-optimized
machine learning
algorithms



Train

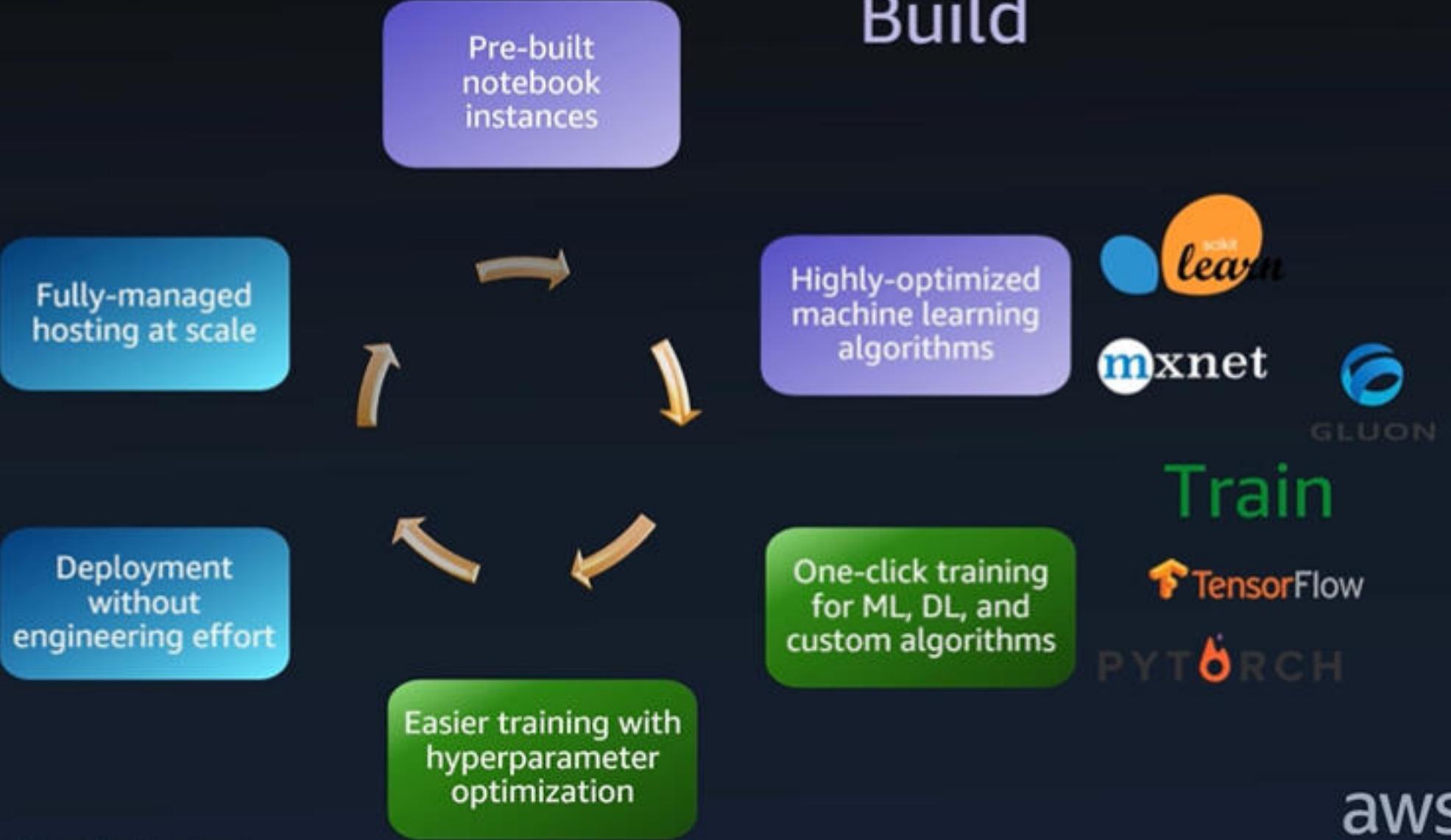
One-click training
for ML, DL, and
custom algorithms

Easier training with
hyperparameter
optimization

Amazon SageMaker

Deploy

Build



Amazon SageMaker: Launch Customers



As the world's leading provider of high-resolution Earth imagery, data and analysis, DigitalGlobe works with enormous amounts of data every day. DigitalGlobe is making it easier for people to find, access, and run compute against our entire 100PB image library, which is stored in AWS's cloud, to apply deep learning to satellite imagery. We plan to use Amazon SageMaker to train models against petabytes of Earth observation imagery datasets using hosted Jupyter notebooks, so DigitalGlobe's Geospatial Big Data Platform (GBDX) users can just push a button, create a model, and deploy it all within one scalable distributed environment at scale.

“

- Dr. Walter Scott, CTO of Maxar Technologies and founder of DigitalGlobe

Convolution Fundamentals

Filters detect patterns, edges, shapes, textures etc. A filter is a small matrix with a given number of columns and rows and the same depth as the input data because a filter is computing a dot product with a chunk of the image.

Low Level Features:

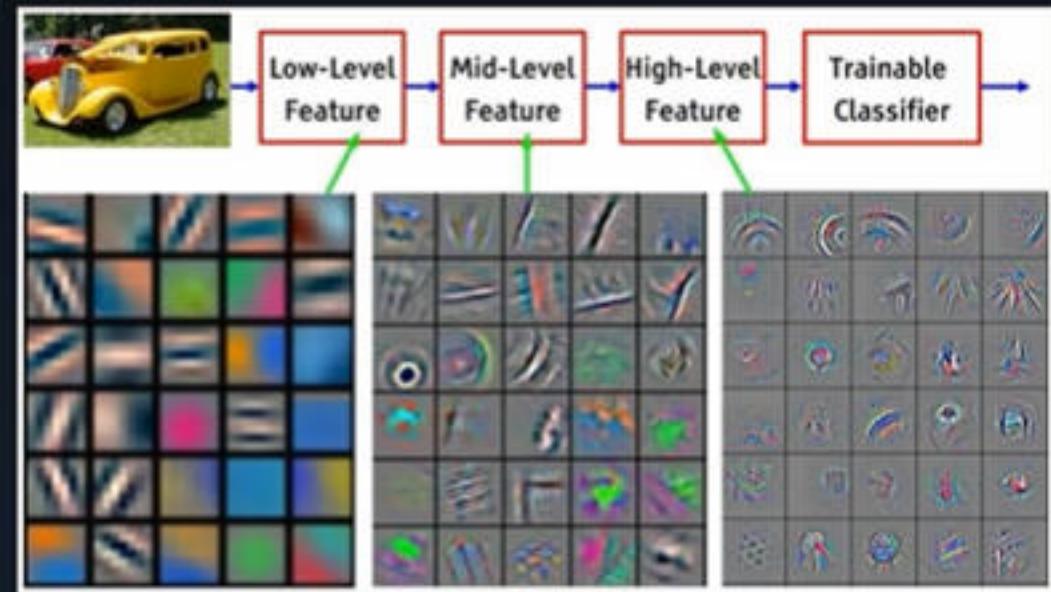
- Edge Detection

Mid Level Features:

- Shapes (Circles, Squares)

High Level Features:

- Buildings, Roads, Automobiles

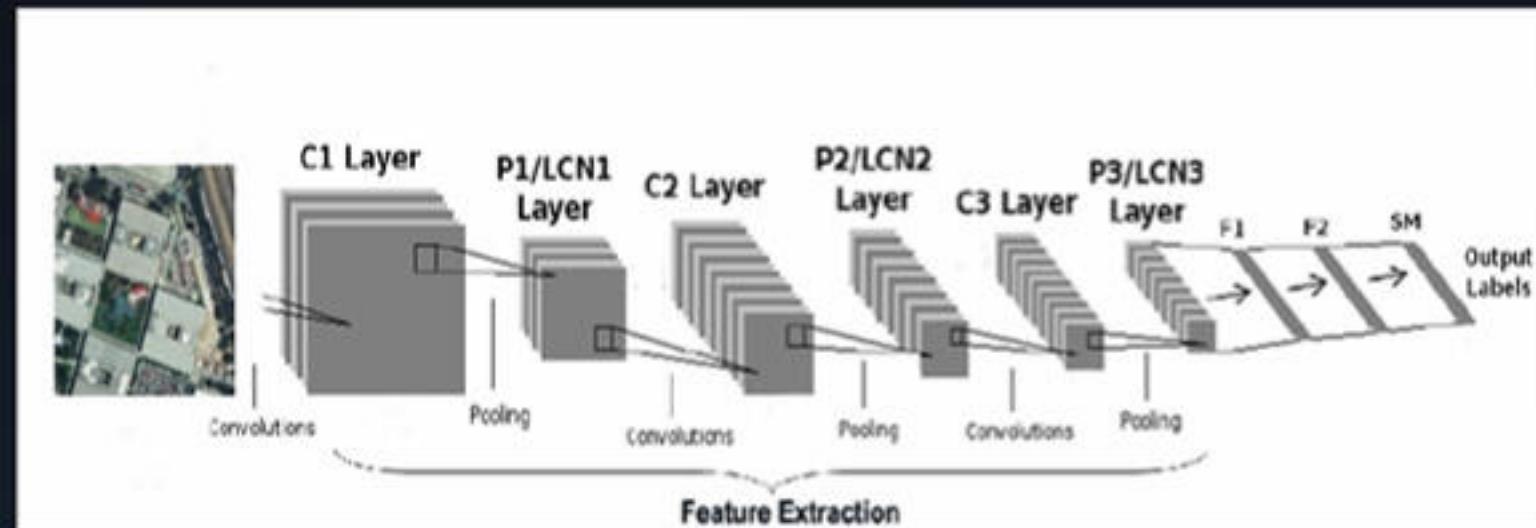


Example:

<http://scs.ryerson.ca/~aharley/vis/conv/>

Le-Net

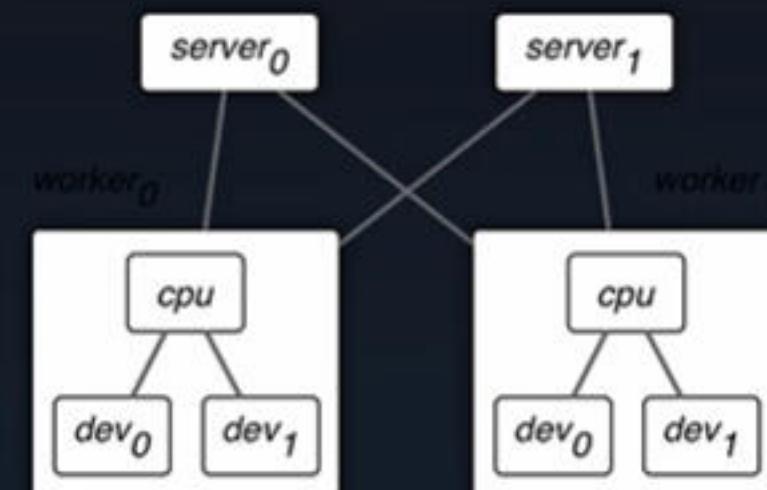
- Network Architecture
 - 3 Conv Layers
 - 3 Pooling Layers
 - 2FC Layers
 - 1 Classifier SoftMax
- Optimizer
 - SGD → ADAM
- Hyperparameters
 - learning_rate: 0.1
 - momentum: 0.9
- Batch Size: 16 , Epochs: 500
- TTT = 13m



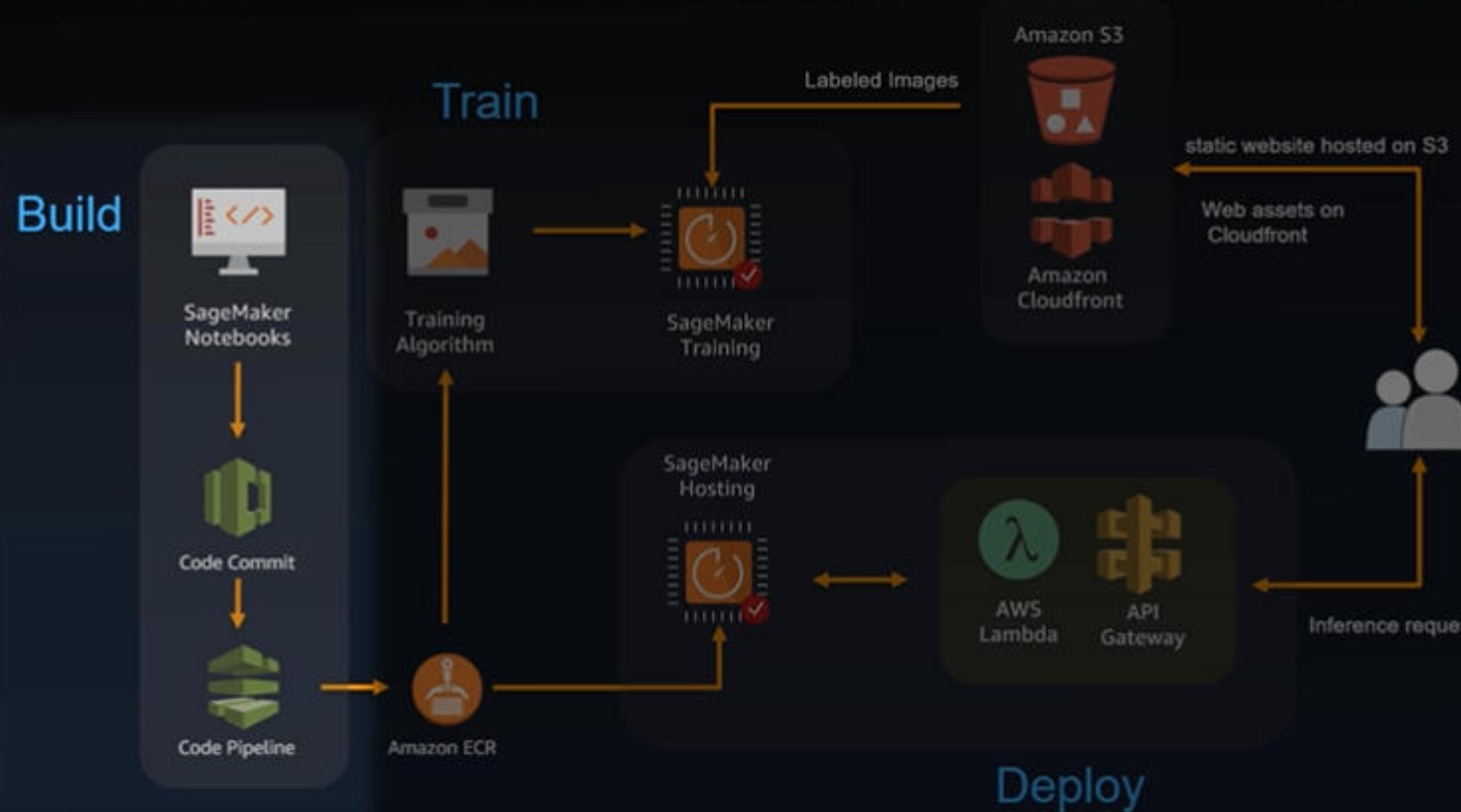
Training Across Multiple GPUs with SageMaker and MXNet

You can run a multi-machine, distributed MXNet training using the MXNet Estimator.

- When you run multi-machine training, SageMaker will import your training script and invoke train on each host in the cluster.
- When using an MXNet Estimator, SageMaker automatically starts MXNet kvstore server and scheduler processes on hosts in your training job cluster.



SageMaker Sample End-to-End Architecture: SAR Image Classification



Amazon SageMaker x Home x SageMaker_mx3mln0 x

Secure | https://us-west-2.console.aws.amazon.com/sagemaker/home?region=us-west-2#/notebook-instances

aws Services Resource Groups

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Amazon SageMaker X

Amazon SageMaker > Notebook instances

Notebook instances

Open Start Update settings Actions Create notebook instance

Search notebook instances

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| Name | Instance | Creation time | Status | Actions |
|-----------|---------------|------------------------|-----------|-------------|
| AWorkshop | ml.m4.4xlarge | May 30, 2018 18:55 UTC | InService | Open Stop |
| Jupyter | ml.p2.xlarge | Dec 04, 2017 21:03 UTC | InService | Open Stop |

Feedback English (US)

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Amazon SageMaker Home SageMaker_mx_lenet(2) Trusted | conda_mxnet_p36

[https://jupyter.notebook.us-west-2.sagemaker.aws/notebooks/SageMaker_mx_lenet\(2\).pynb](https://jupyter.notebook.us-west-2.sagemaker.aws/notebooks/SageMaker_mx_lenet(2).pynb)

jupyter SageMaker_mx_lenet(2) Last Checkpoint: 04/12/2018 (autosaved)

```
File Edit View Insert Cell Kernel Widgets Help
Code
```

```
    num_gpus: apply to the same rule above
    ...
    train_iter, val_iter = prep_data(data_path)
    lenet = mx.lenet()
    lenet_model = mx.mod.Module(
        symbol=lenet,
        context=mx.context.get_train_context(num_cpus, num_gpus))
    logging.getLogger().setLevel(logging.DEBUG)
    lenet_model.fit(train_iter,
                    eval_data=val_iter,
                    kvstore=kvstore,
                    optimizer='ADAM',
                    optimizer_params={'learning_rate': 0.1, 'momentum': 0.9},
                    eval_metric='acc',
                    batch_end_callback=mx.callback.Speedometer(batch_size, 10),
                    num_epoch=100)
    return lenet_model

def get_train_context(num_cpus, num_gpus):
    """
    Define the model training instance.
    Parameters
    -----
    num_cpus: If train the model on an AWS GPU machine, num_cpus = 0 and num_gpus = 1, vice versa.
    num_gpus: apply to the same rule above
    """
    if num_gpus > 0:
        return mx.gpu()
    return mx.cpu()

def get_train_context(num_cpus, num_gpus):
    if num_gpus > 0:
        print("It's {} instance".format(num_gpus))
        return mx.gpu()
    print("It's {} instance".format(num_cpus))
    return mx.cpu()

Overwriting mx_lenet_sagemaker.py
```

In [2]: `#time`
from sagemaker.mxlenet import MXNet
from sagemaker import get_execution_role

```
MXNetEstimator = MXNet("mx_lenet_sagemaker.py",
                       role=get_execution_role(),
                       train_instance_type="ml.p2.xlarge",
                       train_instance_count=1)
MXNetEstimator.fit("s3://bucketexample") # give your s3 bucket URL here.
```