



TITLE :

Smart Image Cataloguing System using AWS Lambda & Recognition

Description

Tired of tagging every single image by hand? We get it — it's time-consuming and honestly, pretty boring. In this project, we'll show you how to build a smart, serverless system that does all the hard work for you. Using AWS Lambda and Amazon Rekognition, your images will be automatically analyzed the moment they're uploaded to an S3 bucket. Whether it's detecting objects, faces, or even text, the system adds tags for you — making everything super easy to organize and search. It's fast, scalable, and completely hands-free — ideal for any app that deals with loads of images.



In the era of digital transformation, the volume of image data generated by individuals and organizations has grown exponentially. Efficiently organizing and retrieving these images has become a critical challenge. Traditional image cataloguing systems, which rely heavily on manual tagging and classification, are time-consuming and prone to human error. To address these challenges, smart image cataloguing systems leverage the power of cloud computing and artificial intelligence.

Amazon Web Services (AWS) provides a robust suite of tools that enable the development of intelligent, serverless image processing pipelines. In particular, AWS Lambda offers scalable and cost-effective compute resources without the need to manage servers, while Amazon Rekognition provides advanced image and video analysis capabilities powered by machine learning. By integrating these services, we can build a smart image cataloguing system that automatically analyzes, tags, and organizes images in real-time as they are uploaded to cloud storage.



This approach not only reduces manual workload but also improves the accuracy and speed of image classification. The system can be further enhanced to support features such as facial recognition, object detection, and content moderation, making it highly versatile for use cases across industries like media, e-commerce, security, and digital asset management.



USE CASE SCENARIO

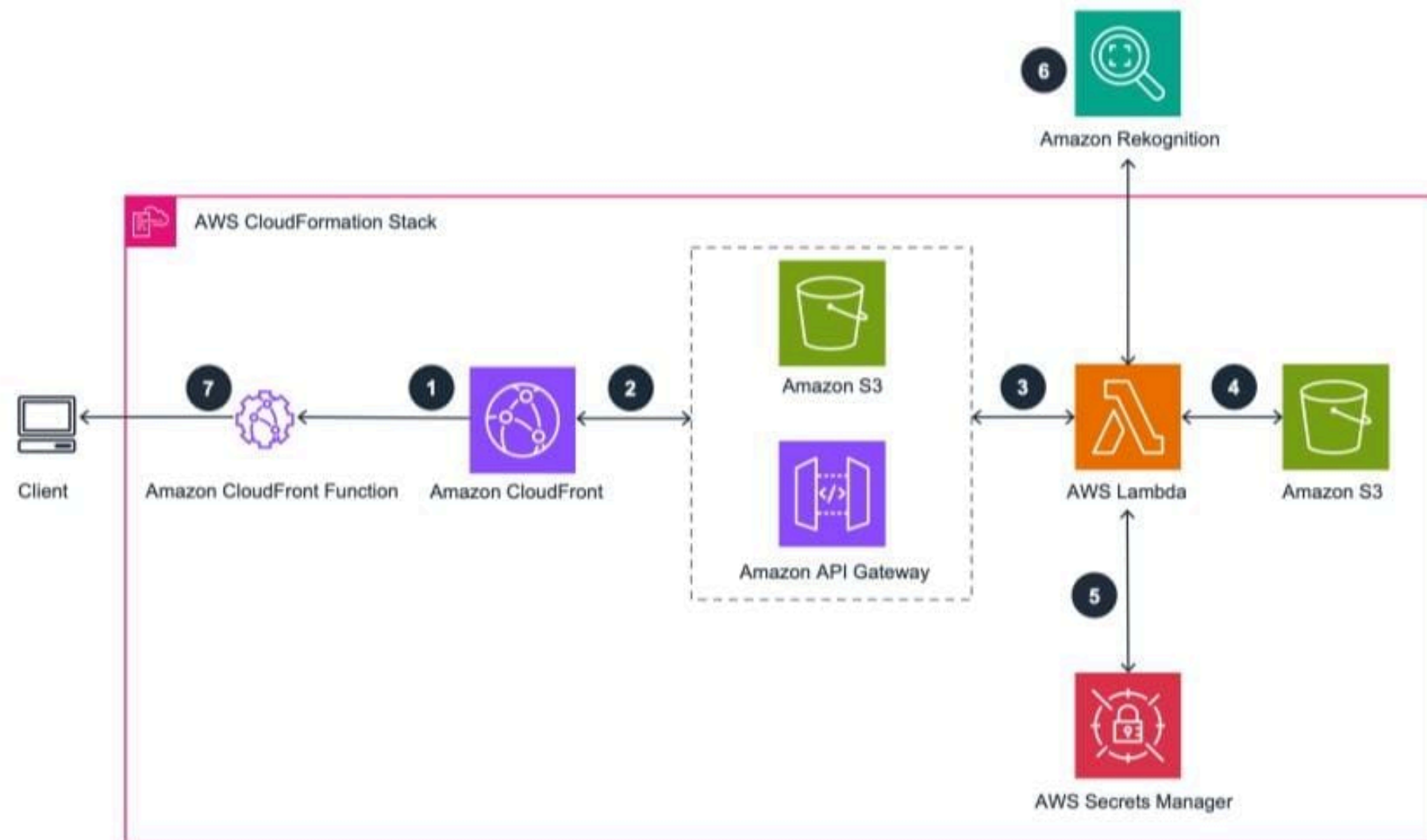
Let's consider a media company that receives thousands of images daily from photographers around the world. These images need to be categorized, tagged with relevant metadata (e.g., people, objects, scenes), and stored for easy retrieval in their content management system (CMS). Doing this manually would be labor-intensive and inconsistent. By using a smart image cataloguing system, the company can automate this process, ensuring faster turnaround and more accurate metadata tagging.



SYSTEM ARCHITECTURE

HERE IS A HIGH-LEVEL VIEW OF THE SYSTEM ARCHITECTURE:

COMPONENTS:





1. AMAZON S3: IMAGE STORAGE. USERS UPLOAD IMAGES HERE.

EXPLANATION OF THE CODE:

- IMPORTS: THE CODE IMPORTS NECESSARY MODULES FROM THE AWS SDK AND NODE.JS FILE SYSTEM
- S3 CLIENT: AN S3 CLIENT IS CREATED TO INTERACT WITH THE S3 . SERVICE
- UPLOAD FUNCTION: THE UPLOADIMAGETOS3 FUNCTION READS AN IMAGE FILE FROM THE LOCAL FILE SYSTEM AND UPLOADS IT TO THE SPECIFIED S3 BUCKET.
- PARAMETERS: THE FUNCTION TAKES THE BUCKET NAME AND FILE PATH AS PARAMETERS.
- ERROR HANDLING: IT INCLUDES ERROR HANDLING TO CATCH ANY ISSUES DURING THE UPLOAD PROCESS.

```
const { S3Client, PutObjectCommand } = require("@aws-sdk/client-s3");
const fs = require("fs");
const path = require("path");

const s3Client = new S3Client({ region: "us-east-1" });

async function uploadImageToS3(bucketName, filePath) {
  try {
    const fileStream = fs.createReadStream(filePath);
    const fileName = path.basename(filePath);

    const uploadParams = {
      Bucket: bucketName,
      Key: fileName,
      Body: fileStream,
      ContentType: "image/jpeg"
    };

    const data = await s3Client.send(new PutObjectCommand(uploadParams));
    console.log(`Image uploaded successfully. ${data.$metadata.httpStatusCode}`);
  } catch (err) {
    console.error("Error uploading image: ", err);
  }
}

const bucketName = "your-s3-bucket-name";
const filePath = "path/to/your/image.jpg";

uploadImageToS3(bucketName, filePath);
```




2. AMAZON S3 TRIGGER: WHEN A NEW IMAGE IS UPLOADED, IT TRIGGERS AN AWS LAMBDA FUNCTION.

WHAT THIS LAMBDA FUNCTION DOES :-

THIS LAMBDA FUNCTION IS AUTOMATICALLY TRIGGERED WHENEVER A NEW IMAGE IS UPLOADED TO A SPECIFIC AMAZON S3 BUCKET. ONCE TRIGGERED, IT PERFORMS THE FOLLOWING ACTIONS:

- IDENTIFIES THE UPLOADED IMAGE USING EVENT DATA FROM S3
- SENDS THE IMAGE TO AMAZON REKOGNITION, AWS'S IMAGE ANALYSIS SERVICE
- DETECTS OBJECTS AND SCENES IN THE IMAGE (E.G., "DOG", "CAR", "TREE")
- LOGS THE RESULTS WITH CONFIDENCE SCORES TO CLOUDWATCH FOR VISIBILITY AND DEBUGGING

```
import boto3
import json
import os

rekognition = boto3.client('rekognition')
s3 = boto3.client('s3')

def lambda_handler(event, context):
    # Get the S3 object details from the event
    bucket_name = event['Records'][0]['s3']['bucket']['name']
    object_key = event['Records'][0]['s3']['object']['key']

    print(f"Processing image: s3://{bucket_name}/{object_key}")

    # Call Amazon Rekognition to detect labels in the image
    response = rekognition.detect_labels(
        Image={'S3Object': {'Bucket': bucket_name, 'Name':
object_key}},
        MaxLabels=10,
        MinConfidence=75
    )

    # Log detected labels
    labels = response['Labels']
    print("Detected labels:")
    for label in labels:
        print(f"{label['Name']} - {label['Confidence']:.2f}%")

    # Optionally, you can store results in DynamoDB or S3

    return {
        'statusCode': 200,
        'body': json.dumps('Image processed successfully.')
    }
```



3. AWS LAMBDA: A SERVERLESS FUNCTION THAT PROCESSES THE IMAGE.

EXPLANATION OF THE LAMBDA FUNCTION :-

- TRIGGERED BY S3: RUNS AUTOMATICALLY WHEN A NEW IMAGE IS UPLOADED TO AN S3 BUCKET.
- CONNECTS TO REKOGNITION: SENDS THE IMAGE TO AMAZON REKOGNITION FOR ANALYSIS.
- DETECTS LABELS: IDENTIFIES OBJECTS, SCENES, OR CONCEPTS IN THE IMAGE (E.G., “DOG”, “LAPTOP”, “SKY”).
- LOGS CONFIDENCE SCORES: DISPLAYS HOW CONFIDENT REKOGNITION IS ABOUT EACH LABEL.
- RETURNS A RESPONSE: SENDS BACK A CLEAN JSON RESPONSE WITH DETECTED LABELS.
- HANDLES ERRORS: IF ANYTHING FAILS, IT RETURNS A PROPER ERROR MESSAGE INSTEAD OF CRASHING.

```
import boto3
import json

# Initialize the Rekognition client
rekognition = boto3.client('rekognition')

def lambda_handler(event, context):
    # Get S3 bucket and object key from the event
    bucket = event['Records'][0]['s3']['bucket']['name']
    image_key = event['Records'][0]['s3']['object']['key']

    print(f"Processing image: {image_key} from bucket: {bucket}")

    try:
        # Call Rekognition to detect labels in the image
        response = rekognition.detect_labels(
            Image={'S3Object': {'Bucket': bucket, 'Name': image_key}},
            MaxLabels=10,
            MinConfidence=70
        )

        # Format and log the results
        labels = response['Labels']
        for label in labels:
            print(f"Label: {label['Name']} - Confidence: {label['Confidence']:.2f}%")
        return {
            'statusCode': 200,
            'body': json.dumps({
                'message': 'Image processed successfully',
                'labels': [label['Name'] for label in labels]
            })
        }
    except Exception as e:
        print(f"Error processing image: {str(e)}")
        return {
            'statusCode': 500,
            'body': json.dumps({'error': str(e)})
        }
```




4. AMAZON REKOGNITION: ANALYZES THE IMAGE TO DETECT OBJECTS, SCENES, FACES, AND TEXT.

EXPLANATION:

- IMPORT BOTO3: AWS SDK FOR PYTHON.
- CREATE CLIENT: CONNECTS TO AMAZON REKOGNITION.
- READ IMAGE: LOAD THE IMAGE AS BYTES.
- DETECT_LABELS(): SENDS THE IMAGE TO REKOGNITION TO FIND OBJECTS/SCENES.
- PRINT RESULTS: DISPLAYS EACH DETECTED LABEL WITH CONFIDENCE SCORE.

```
import boto3

# Create Rekognition client
rekognition = boto3.client('rekognition')

# Load the image
image_file = 'image.jpg'
with open(image_file, 'rb') as image:
    image_bytes = image.read()

# Detect objects/scenes
labels_response = rekognition.detect_labels(Image={'Bytes': image_bytes}, MaxLabels=10)
print("Detected Labels:")
for label in labels_response['Labels']:
    print(f"- {label['Name']} ({label['Confidence']:.2f}%)")

# Detect faces
faces_response = rekognition.detect_faces(Image={'Bytes': image_bytes}, Attributes=['ALL'])
print("\nDetected Faces:")
for i, face in enumerate(faces_response['FaceDetails'], 1):
    print(f"- Face {i}: {face['Gender']['Value']} (Confidence: {face['Gender']['Confidence']:.2f}%)")

# Detect text
text_response = rekognition.detect_text(Image={'Bytes': image_bytes})
print("\nDetected Text:")
for text in text_response['TextDetections']:
    print(f"- {text['DetectedText']} ({text['Confidence']:.2f}%)")
```



5. AMAZON DYNAMODB / RDS: STORES METADATA RETURNED BY REKOGNITION.

EXPLANATION :-

- **CONNECT TO AMAZON REKOGNITION**
USE BOTO3 TO CREATE A REKOGNITION CLIENT TO ANALYZE IMAGES.
- **CONNECT TO AMAZON RDS (MYSQL)**
USE MYSQL.CONNECTOR TO CONNECT TO YOUR RDS INSTANCE
USING YOUR HOST, USERNAME, PASSWORD, AND DATABASE.
- **READ IMAGE AS BYTES**
LOAD THE IMAGE FILE FROM YOUR LOCAL SYSTEM USING PYTHON'S
OPEN() AND READ().
- **SEND IMAGE TO REKOGNITION**
CALL DETECT_LABELS() TO ANALYZE THE IMAGE AND GET OBJECTS/SCENES
WITH CONFIDENCE SCORES.
- **INSERT DETECTED LABELS INTO RDS TABLE**
LOOP THROUGH THE LABELS RETURNED BY REKOGNITION. FOR EACH LABEL:
GET LABEL NAME AND CONFIDENCE
INSERT A ROW INTO THE IMAGE_LABELS TABLE WITH THE IMAGE NAME, LABEL
NAME, AND CONFIDENCE VALUE.

```
import boto3

# Initialize Rekognition and DynamoDB clients
rekognition = boto3.client('rekognition')
dynamodb = boto3.resource('dynamodb')
table = dynamodb.Table('ImageLabels')

# Image details
image_name = 'image.jpg'

# Read image as bytes
with open(image_name, 'rb') as img_file:
    img_bytes = img_file.read()

# Call Rekognition to detect labels
response = rekognition.detect_labels(Image={'Bytes': img_bytes})

# Prepare labels for storage
labels = {label['Name']: str(round(label['Confidence'], 2)) for label in
response['Labels']}

# Store into DynamoDB
table.put_item(
    Item={
        'ImageName': image_name,
        'Labels': labels
    }
)

print("Metadata stored in DynamoDB successfully.")
```

HOW TO DEPLOY AND RUN ON AWS

STEP 1

- CREATE AN S3 BUCKET (E.G., MY-SMART-IMAGE-BUCKET)
- CREATE A DYNAMODB TABLE
- TABLE NAME: IMAGEMETADATA
- PRIMARY KEY: IMAGE_ID (STRING)

- CREATE IAM ROLE
- GO TO IAM > ROLES AND CREATE A NEW ROLE FOR LAMBDA.
- ATTACH THE FOLLOWING PERMISSIONS:
 1. AMAZONS3READONLYACCESS
 2. AMAZONREKOGNITIONFULLACCESS
 3. AMAZ
 4. AWSLAMBDAASICEXECUTIONROLE

STEP 2

STEP 3

- CREATE A LAMBDA FUNCTION
- GO TO AWS LAMBDA > CREATE FUNCTION
- CHOOSE: AUTHOR FROM SCRATCH
- RUNTIME: PYTHON 3.9 OR 3.10
- SET THE EXECUTION ROLE TO THE ONE CREATED ABOVE.

STEP 4

ADD ENVIRONMENT VARIABLE

- KEY: TABLE_NAME
- VALUE: IMAGEMETADATA

DEPLOY THE CODE

- COPY THE CODE I PROVIDED INTO THE LAMBDA FUNCTION EDITOR.
- CLICK DEPLOY

STEP 5

STEP 6

CONNECT S3 TRIGGER

- GO TO YOUR S3 BUCKET
- SET UP EVENT NOTIFICATION:
- EVENT TYPE: PUT
- DESTINATION: YOUR LAMBDA FUNCTION



PROBLEM :-

MANAGING LARGE-SCALE IMAGE LIBRARIES PRESENTS SIGNIFICANT CHALLENGES, PRIMARILY DUE TO THE LABOR-INTENSIVE NATURE OF TRADITIONAL CATALOGUING PRACTICES, WHICH OFTEN INTRODUCE HUMAN ERROR. CONVENTIONAL METHODS STRUGGLE TO PROVIDE EFFICIENT SEARCHING, FILTERING, AND RETRIEVAL CAPABILITIES DUE TO THEIR RELIANCE ON CUMBERSOME MANUAL TAGGING AND CLASSIFICATION.

TO ADDRESS THESE CHALLENGES, WE ARE DEVELOPING AN ADVANCED SMART IMAGE CATALOGUING SYSTEM THAT LEVERAGES AMAZON REKOGNITION INTEGRATED WITH AWS LAMBDA. THIS ARCHITECTURE ENABLES THE AUTOMATION OF VARIOUS CRITICAL PROCESSES INVOLVED IN IMAGE ANALYSIS, TAGGING, AND STORAGE, OFFERING ORGANIZATIONS A STREAMLINED SOLUTION THAT EXCELS IN KEY AREAS:

AUTOMATED CONTENT RECOGNITION: THE SYSTEM WILL LEVERAGE MACHINE LEARNING TO AUTOMATICALLY IDENTIFY AND CLASSIFY A WIDE RANGE OF COMPONENTS WITHIN IMAGES—SUCH AS OBJECTS, FACES, TEXT, AND COMPLEX SCENES—ENSURING COMPREHENSIVE INDEXING OF VISUAL CONTENT.



DYNAMIC METADATA GENERATION: BY EMPLOYING SOPHISTICATED ALGORITHMS, THE SYSTEM WILL DELIVER ACCURATE TAGS AND RICH METADATA THAT TRULY REFLECT THE UNDERLYING CONTENT OF EACH IMAGE, THEREBY ENHANCING THE SEARCHABILITY AND RELEVANCE OF THE IMAGE REPOSITORY.

EFFICIENT METADATA MANAGEMENT: IMAGE METADATA WILL BE PRESERVED IN A STRUCTURED, SEARCH-OPTIMIZED FORMAT, FACILITATING EASY ACCESS AND RETRIEVAL FOR USERS.

SERVERLESS SCALABILITY: UTILIZING A SERVERLESS ARCHITECTURE WILL MAXIMIZE COST EFFICIENCY WHILE ENSURING ROBUST SCALABILITY, ALLOWING THE SOLUTION TO ADAPT SEAMLESSLY TO INCREASING IMAGE VOLUMES AND FLUCTUATING WORKLOADS.

IMPLEMENTING THIS SOPHISTICATED SYSTEM WILL SIGNIFICANTLY ALLEVIATE THE RELIANCE ON MANUAL PROCESSES, ENHANCE THE ACCURACY AND RICHNESS OF IMAGE METADATA, AND ENABLE RAPID, INTELLIGENT RETRIEVAL OF VISUAL CONTENT. THIS WILL FUNDAMENTALLY TRANSFORM THE WAY ORGANIZATIONS MANAGE AND UTILIZE THEIR IMAGE ASSETS.



SOLUTION :-

LEVERAGING ADVANCED ARTIFICIAL INTELLIGENCE AND CUTTING-EDGE CLOUD TECHNOLOGIES, THE SMART IMAGE CATALOGUING SYSTEM REPRESENTS AN INNOVATIVE SOLUTION THAT EFFECTIVELY AUTOMATES THE ORGANIZATION, TAGGING, AND RETRIEVAL OF IMAGES. BY UTILIZING THE CAPABILITIES OF AMAZON REKOGNITION AND AWS LAMBDA, THIS SYSTEM CONDUCTS THOROUGH ANALYSES OF UPLOADED IMAGES TO IDENTIFY A DIVERSE RANGE OF ELEMENTS, INCLUDING OBJECTS, SCENES, FACES, AND TEXT. THE OUTCOME OF THIS INTELLIGENT ANALYSIS IS THE GENERATION OF COMPREHENSIVE METADATA, WHICH IS THEN SECURELY STORED IN A HIGHLY SEARCHABLE DATABASE.

FOR INDIVIDUALS AND ORGANIZATIONS MANAGING EXTENSIVE COLLECTIONS OF IMAGES, THIS PIONEERING SYSTEM ALLEVIATES THE BURDENSOME REQUIREMENT FOR MANUAL TAGGING. IT ENHANCES SEARCHABILITY, THEREBY FACILITATING RAPID AND EFFICIENT IMAGE RETRIEVAL, WHILE SIGNIFICANTLY SIMPLIFYING THE OVERALL MANAGEMENT OF IMAGE ASSETS. THIS TRANSFORMATIVE METHODOLOGY NOT ONLY STREAMLINES WORKFLOWS BUT ALSO EMPOWERS USERS TO FULLY LEVERAGE THE POTENTIAL OF THEIR VISUAL CONTENT.



CONCLUSION

IN TODAY'S FAST-PACED DIGITAL WORLD, THE ABILITY TO DERIVE MEANINGFUL INSIGHTS FROM VISUAL DATA HAS NEVER BEEN MORE VALUABLE. BY INTEGRATING AMAZON RECOGNITION WITH SCALABLE DATA STORAGE SOLUTIONS LIKE AMAZON DYNAMODB OR AMAZON RDS, WE UNLOCK POWERFUL CAPABILITIES—FROM REAL-TIME OBJECT DETECTION TO STRUCTURED DATA ARCHIVING—THAT CAN TRANSFORM HOW APPLICATIONS INTERACT WITH IMAGES.

THIS COMBINATION NOT ONLY SHOWCASES THE STRENGTH OF AWS SERVICES WORKING TOGETHER BUT ALSO EMPHASIZES THE IMPORTANCE OF BUILDING CLOUD-NATIVE SOLUTIONS THAT ARE AUTOMATED, SCALABLE, AND SECURE. WHETHER YOU'RE BUILDING A SMART PHOTO ORGANIZER, AN IDENTITY VERIFICATION SYSTEM, OR A MONITORING SOLUTION, THIS STACK PROVIDES A ROBUST FOUNDATION TO INNOVATE WITH CONFIDENCE.

ULTIMATELY, THE SYNERGY BETWEEN AI-DRIVEN ANALYSIS AND CLOUD-BASED DATA STORAGE ENABLES DEVELOPERS TO MOVE FROM RAW DATA TO ACTIONABLE INSIGHTS EFFORTLESSLY—BRINGING US ONE STEP CLOSER TO SMARTER, MORE INTUITIVE APPLICATIONS.



RESOURCES / REFERENCES

1. AMAZON REKOGNITION DOCUMENTATION

[HTTPS://DOCS.AWS.AMAZON.COM/REKOGNITION/LATEST/DG/WHAT-IS.HTML](https://docs.aws.amazon.com/rekognition/latest/dg/what-is.html)

2. DETECTING LABELS IN AN IMAGE WITH REKOGNITION

[HTTPS://DOCS.AWS.AMAZON.COM/REKOGNITION/LATEST/DG/LABELS-DETECT-LABELS-IMAGE.HTML](https://docs.aws.amazon.com/rekognition/latest/dg/labels-detect-labels-image.html)

3. AMAZON DYNAMODB DOCUMENTATION

[HTTPS://DOCS.AWS.AMAZON.COM/AMAZONDYNAMODB/LATEST/DEVELOPERGUIDE/INTRODUCTION.HTML](https://docs.aws.amazon.com/amazondynamodb/latest/developerguide/introduction.html)

4. AMAZON RDS DOCUMENTATION

[HTTPS://DOCS.AWS.AMAZON.COM/AMAZONRDS/LATEST/USERGUIDE/WELCOME.HTML](https://docs.aws.amazon.com/amazonrds/latest/userguide/welcome.html)

5. BOTO3 AWS SDK FOR PYTHON

[HTTPS://BOTO3.AMAZONAWS.COM/V1/DOCUMENTATION/API/LATEST/INDEX.HTML](https://boto3.amazonaws.com/v1/documentation/api/latest/index.html)

6. MYSQL CONNECTOR/PYTHON DEVELOPER GUIDE

[HTTPS://DEV.MYSQL.COM/DOC/CONNECTOR-PYTHON/EN/](https://dev.mysql.com/doc/connector-python/en/)



THANK YOU

