

Cloud Computing Project

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Contents



Project Aim and Requirements



Architecture Overview



Implementation Steps



Monitoring and Testing



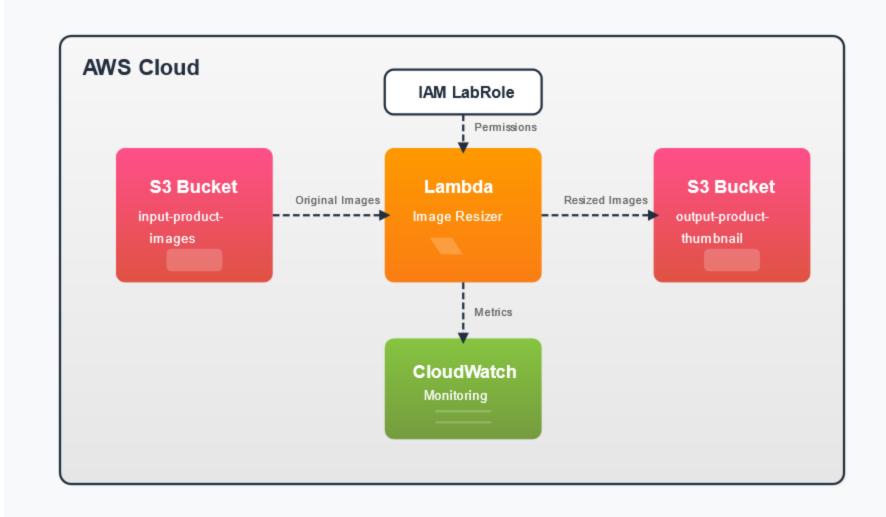
Conclusion

Project Aim and Requirements

- To develop and deploy a scalable image processing application using AWS Lambda
- Primary function: Automatic thumbnail generation from uploaded images
- Key requirements:
 - AWS Lambda service
 - Amazon S3 for image storage
 - CloudWatch for monitoring
 - Predefined IAM role (LabRole)
 - Appropriate trigger configurations

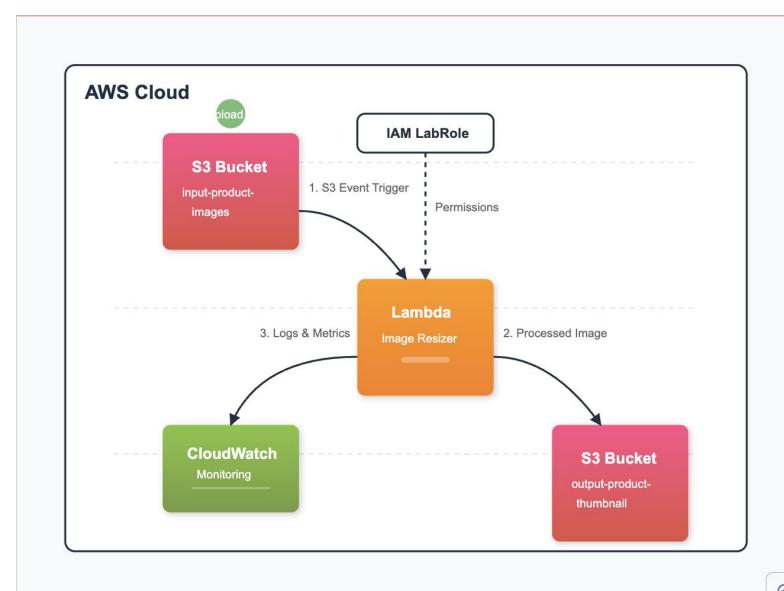
Architecture Overview

The system architecture is designed to efficiently handle image resizing workflows using AWS services. It consists of a source S3 bucket named (input-product-images), which holds the original images, and a destination S3 bucket named (output-product-thumbnail) where the resized images will be stored. Image processing is handled by an AWS Lambda function that works with a set of permissions provided by the LabRole IAM role. This ensures secure and controlled access to the necessary resources. CloudWatch is integrated for monitoring and tracking performance metrics, hence allowing for efficient oversight of the system's operations.



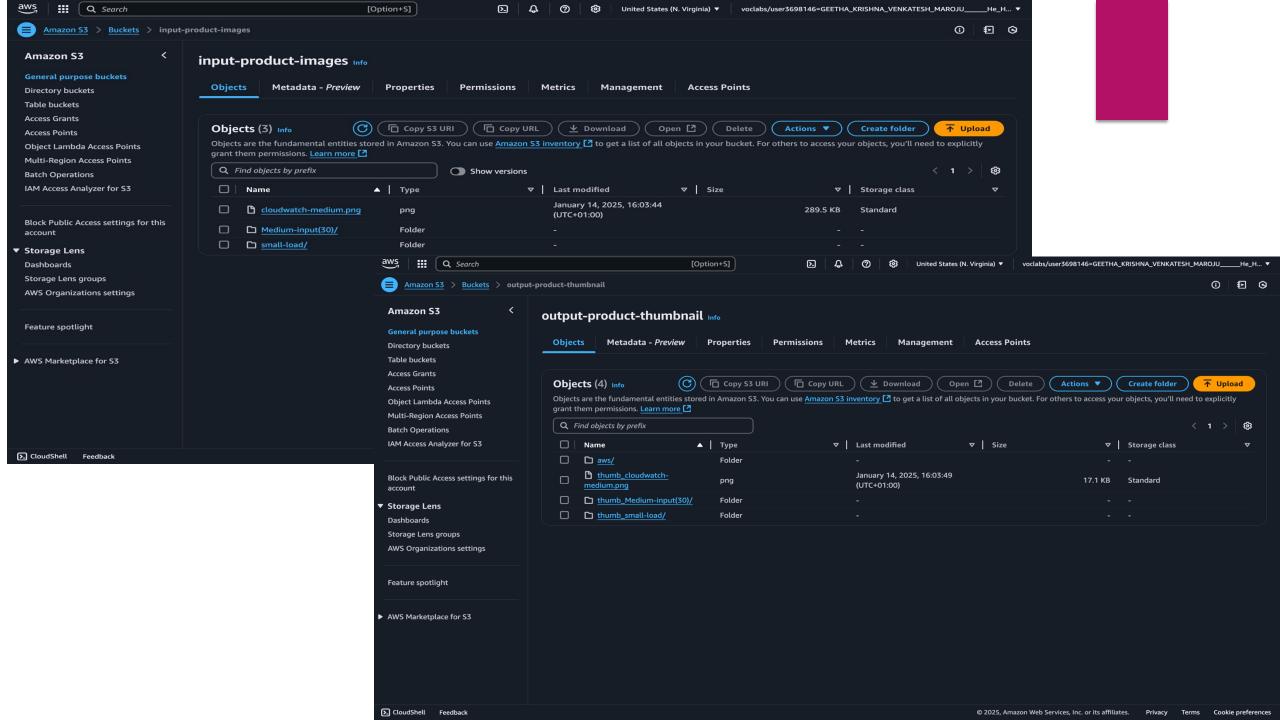
Architecture Overview

▶ This workflow triggers right from uploading an image into the source S3 bucket. The S3 event triggers and calls the Lambda function. The LabRole permissions enable the Lambda function to process the image into the required dimensions and place it into the destination S3 bucket. During this time, CloudWatch would have recorded the performance metrics and logs for system performance.



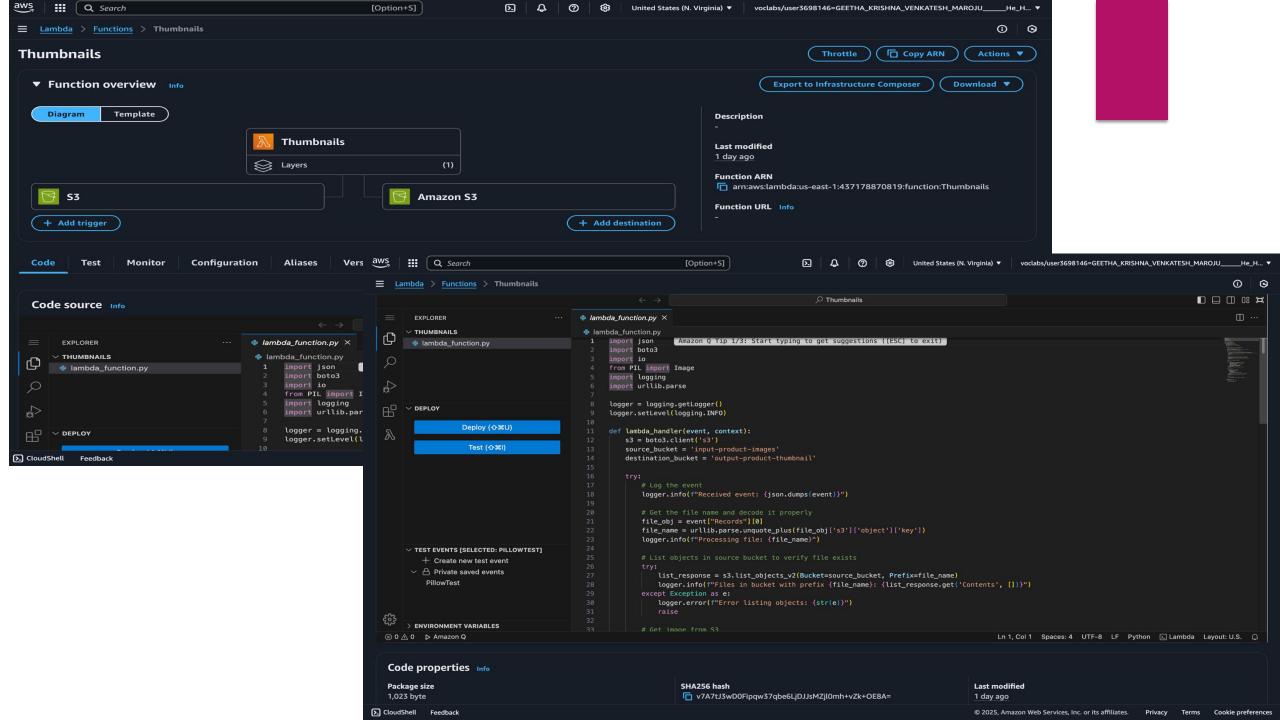
Implementation Steps 1. S3 Bucket Setup

- ➤ Create two S3 buckets-one source bucket is used for uploading the images to be resized and another one used as the target bucket, to store the output/ resized photos. Give it a name in accordance with naming conventions at AWS. In my case I named the source bucket as "input-product-images" and the target bucket is named as "output-product-thumbnail".
- Next, set up stream permissions on both buckets. Set adequate access policies to enable only the actions of the Lambda function to read in the source bucket and write to the destination bucket. Permission control should be highly specific through bucket policies and ACLs.



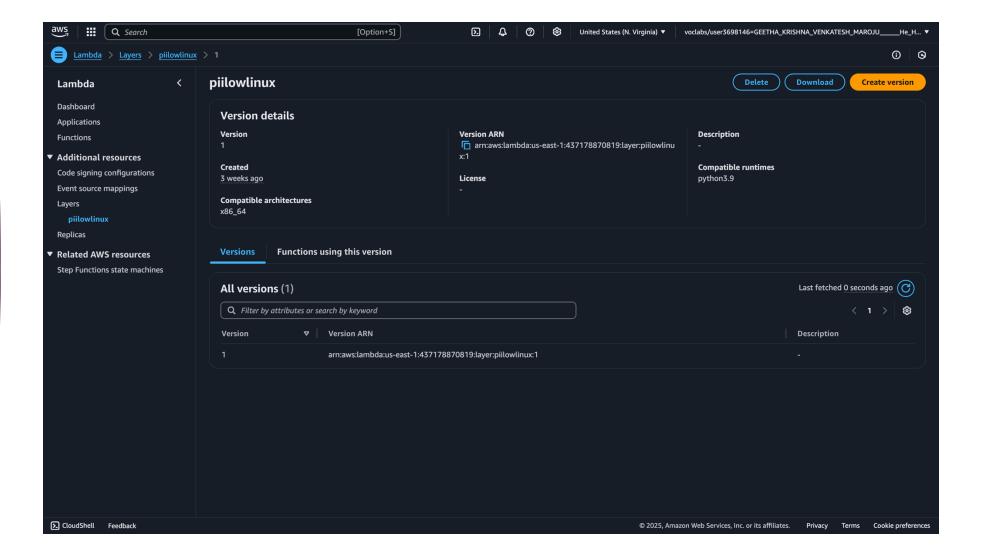
Implementation Steps 2. Lambda Function Creation

▶ The next step is to create a new AWS Lambda function, for this we need to search for lambda in AWS console then after we created the new lambda function named as the "Thumbnails" and it uses the runtime as "Python 3.9" and assigning it the "LabRole" IAM role during setup. This will grant the required permission to work with S3 and resize images using this function. The memory setting and timeout are set to have better performance. As for the image resizing, it can be done in a programming language of choice, Python in this case, which will use its libraries to deal with images. Environment variables are defined for the source and destination bucket names to simplify the configuration.



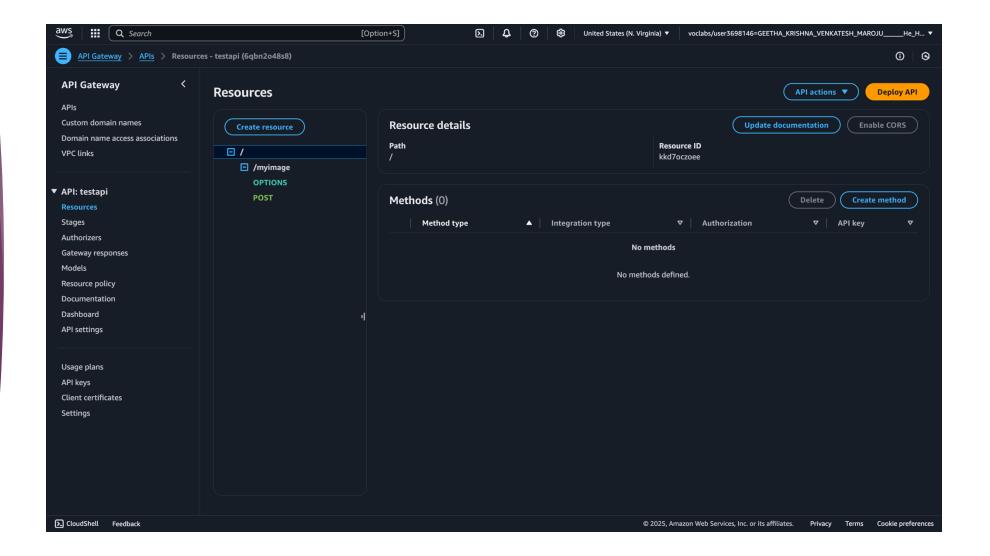
2a. Lambda Layer Creation

In order to enable Python to process images, a Lambda layer was created with the Pillow library-a library not natively available within the default Lambda runtime. First, it was necessary to create a new Lambda layer and upload a ZIP file containing the Pillow package. In that respect, the layer was set to Python 3.9 to ensure smooth integration. The layer was attached to the Lambda function once it had been created with the dependencies required for the resizing and processing of images. This ensured that this function was loaded for the process in full capability.



2b. API Gateway Setup

▶ API Gateway was implemented to expose the Lambda function via HTTP so that users could start an image processing job via a secured endpoint. A new REST API was created and methods like GET and POST were configured to handle the requests of image processing. API Gateway was then integrated with the Lambda function, and appropriate request and response mappings were set up to ensure smooth communication between components. API deployment was finally performed to a targeted stage called 'dev1', through which the flow was triggered. With this setup, interaction with the image processing system via HTTP is solid and secure.



Implementation Steps 3. Trigger Configuration

▶ Set up an S3 event trigger on the source bucket that will call the Lambda function. An event to listen to is "Object Created" for the function to be triggered every time a new image is uploaded. Besides that, file type filters can be further added, say ".jpg" and ".png", to have the trigger narrow down to those relevant image formats. This guarantees that only these files of interest are processed to smoothen the workflow.

Implementation Steps 4. Testing Configuration

▶ Sample image upload to the source bucket will finally test the system. The functionality is tested by verifying that the resized images are populated in the destination bucket. AWS CloudWatch will be observed for logs and metrics to confirm that the Lambda function executes without errors and performance as expected. To this regard, necessary adjustments in the configuration would be done based on the observations so as to have the system up and running properly.

0(0.00%)

66%

443

80%

410

75%

184

90%

2286

95%

220

0.00

98% 99% 99.9% 99.99% 100% # regs

Aggregated

/dev1/myimage

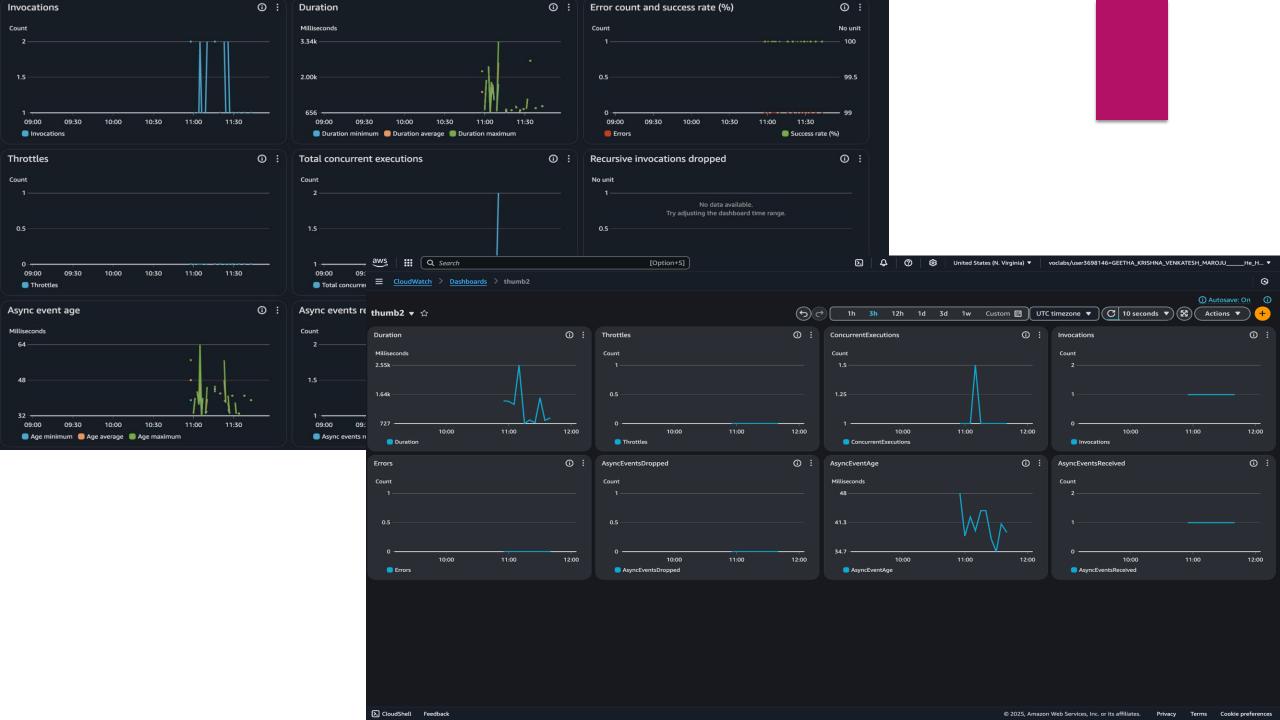
Name

Type

Response time percentiles (approximated)

Monitoring and Testing 1. Performance Monitoring

▶ Performance monitoring of the system is very crucial for reliable operation. AWS CloudWatch provides a set of metrics on the activity of the Lambda function that includes invocation count, execution duration, error count, and memory usage. These will, in turn, enable observation of real-time behavior of the system and pinpoint optimization opportunities. These insights can be used to tune the system to maintain its efficiency and potential bottlenecks. Graphs and screenshots of these metrics are attached herein, to give a clear view of the system performance during testing.

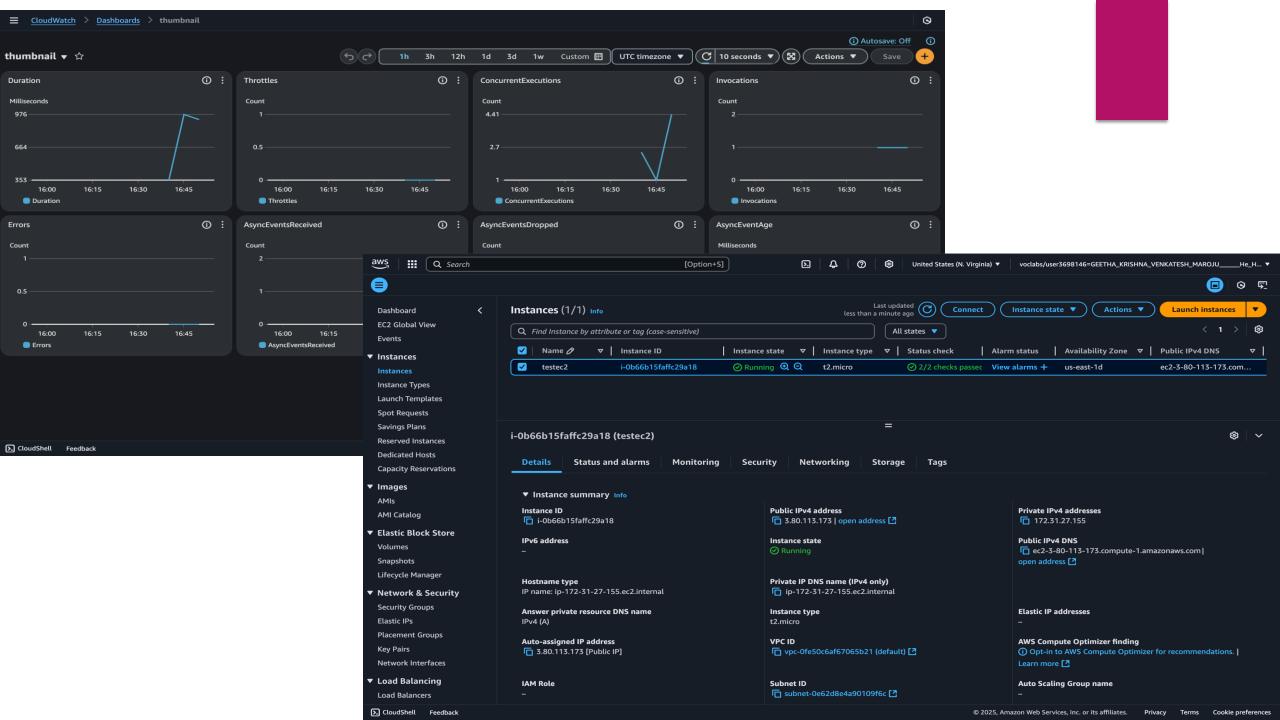


Monitoring and Testing2. Scaling Testing Methods

▶ To test the scalability of the system, several images were manually uploaded to the source bucket to simulate parallel executions. This procedure tested how the Lambda function would handle multiple requests at the same time. Metrics watched in CloudWatch include processing times and success/failure rates to see how the system responds under load. Observations from these tests demonstrate the system's capability to scale efficiently while maintaining performance.

Monitoring and Testing 3. Testing Scenarios

- It involves a number of different test scenarios aimed at proving that the system is operational under various conditions. These included tests for uploading a single image to check the basic functionality, multiple concurrent uploads to check on load handling to evaluate consistent performance of the system. The system was also tested with different formats of images, such as jpg and png.
- For stress testing, Locust was used to simulate high loads on the system. The following command was executed to conduct the stress test: "locust -f test.py --host https://6qbn2o48s8.execute-api.us-east-1.amazonaws.com/dev1/myimage --headless -u 10 -r 2 -t 10s".
- This command initiated 10 simulated users with a spawn rate of 2 users per second for a duration of 10 seconds. The results are updated below screenshots.



Conclusion

- ► The project designed a serverless architecture for image processing using AWS, which helped in implementing an efficient and scalable solution. With AWS Lambda, the system scales automatically to whatever workload is required of it. Further, this is cost-effective and robust since implementation was done leveraging institutional AWS resources.
- It generated thumbnails from uploaded images reliably, proving to be consistent in performance for file types and sizes. This project also provided great experience in the development of serverless applications and further developed skills in cloud architecture, automation, and performance monitoring. These learnings set a very strong foundation for using serverless technologies in future projects.