

# Configuration Manual

MSc Research Project  
Cloud Computing

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# Configuration Manual

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## 1 Requirements

The implementation of Enhanced Zero Trust Security Model (EZTSM) required various tools and programs. This configuration manual shows the step by step process for setting up the environment and deployment to achieve the desired result. The setup is divided into two major categories (1) AWS environment setup (2) Anomaly Detection model training setup.

### 1.1 AWS environment setup

In this research, the standard AWS account and a user with required access has been created. The user has both programmatic and console access. All the AWS services were created through console only. Steps for creating required AWS services and integrating that between services are mentioned below.

### Amazon Cognito

1. In the Cognito dashboard create user pool, select Single Page Application (SPA) under application type because this type won't create any client secret which we don't require.
2. Under configure option, for sign in attributes "Email" and "Username" has been provided and for signup attributes "email" and "name" were given.
3. This will create a User Pool and App Client shown in Figure 1 and 2.
4. An user is created with the required attributes, with email verified, shown in Figure 3.

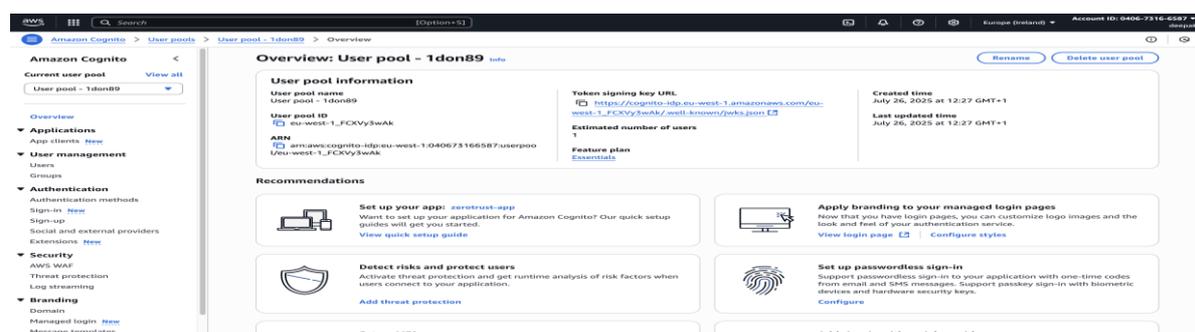


Figure 1: User pool dashboard

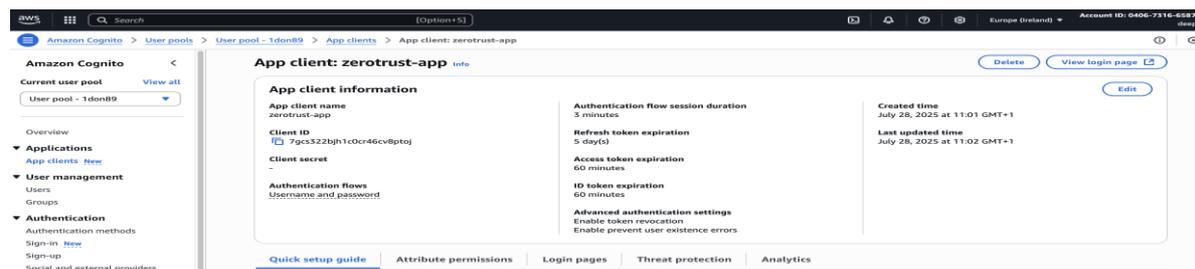


Figure 2: App client page

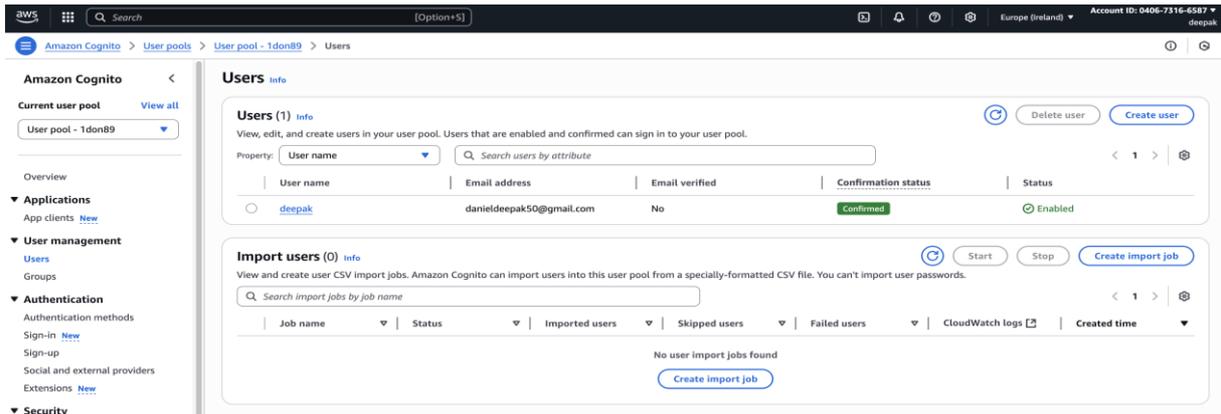


Figure 3: Users dashboard

## AWS Lambda

1. In this project two lambda functions are created, one act as Gatekeeper “gatekeeper-lambda” and another as business logic lambda “real\_lambda”.
2. In the lambda creation page shown in figure 4 mention function name, runtime is selected as python 3.9, default execution role is “LambdaExecutionRole-ZeroTrust” which has necessary permissions for the lambda to communication to other services.
3. After gatekeeper-lambda function created shown in figure 5, the python code “gatekeeper\_lambda.py” is zipped and uploaded to the function, mention the compatible python package as python 3.9.
4. Similarly created the business logic lambda “real\_lambda” shown in figure 6, here the logic is created simply to return 200 status code once the lambda is invoked.
5. The required dependencies for the gatekeeper-lambda is attached to the function in the form of layer “crypt-layer” , so the layer is created shown in figure 7 with the zip file contained required dependencies installed.
6. Both lambda functions are created with required dependencies and deployed to be accessed.

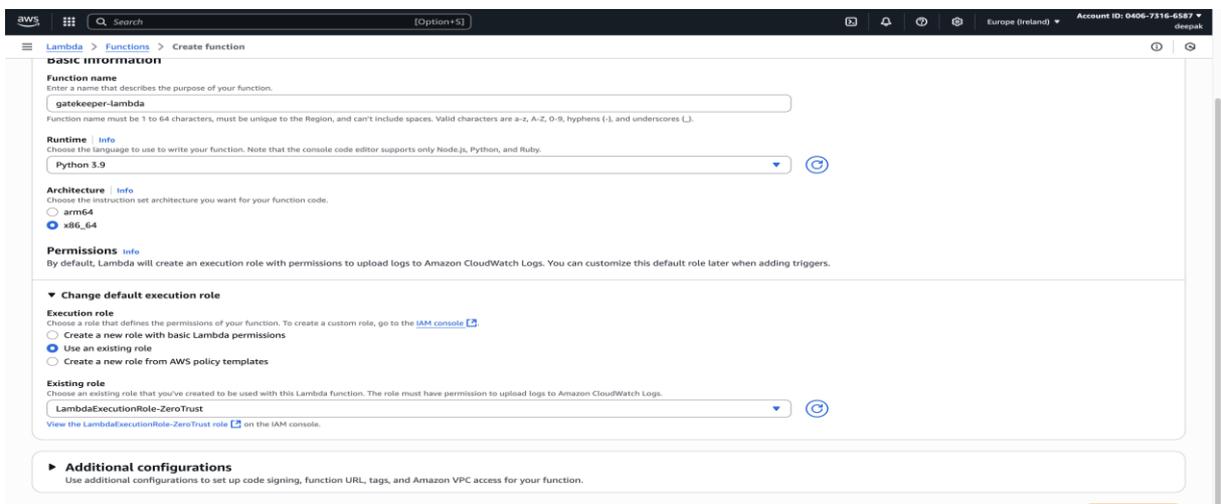


Figure 4: Lambda creation page

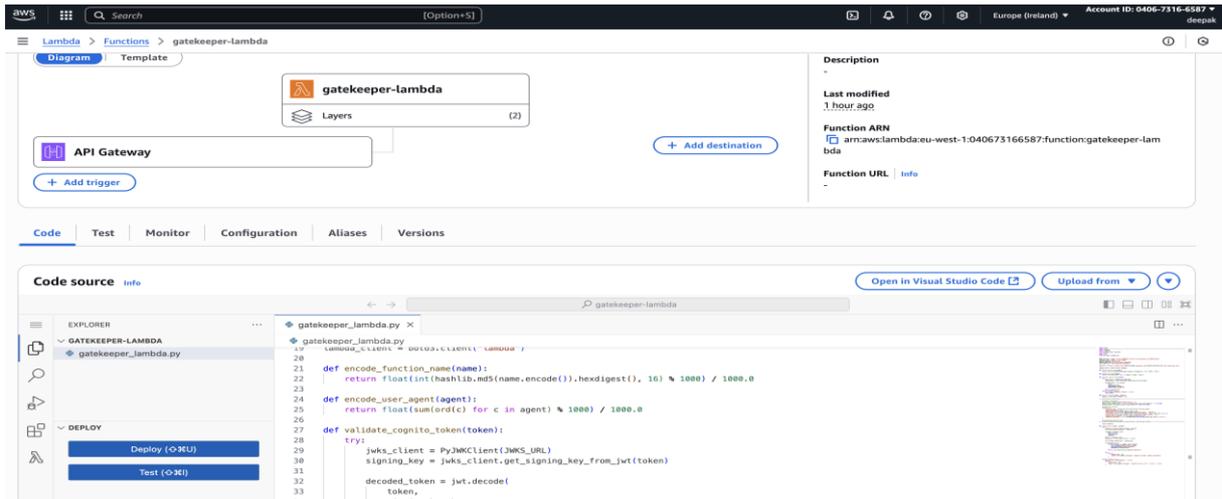


Figure 5: Gatekeeper lambda

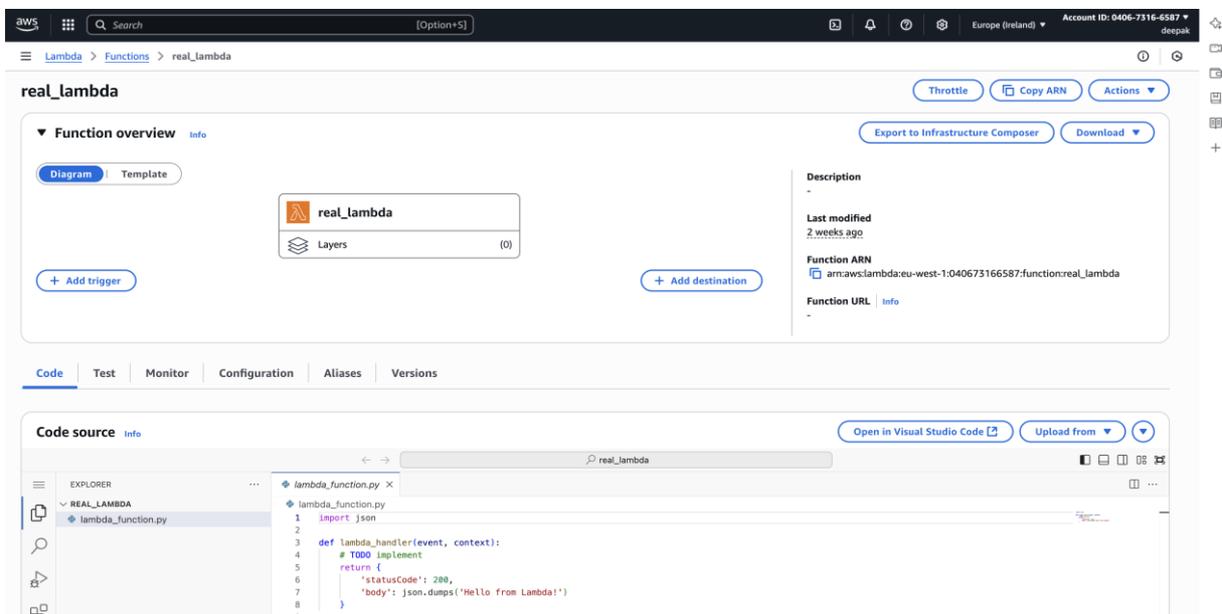


Figure 6: Business logic lambda

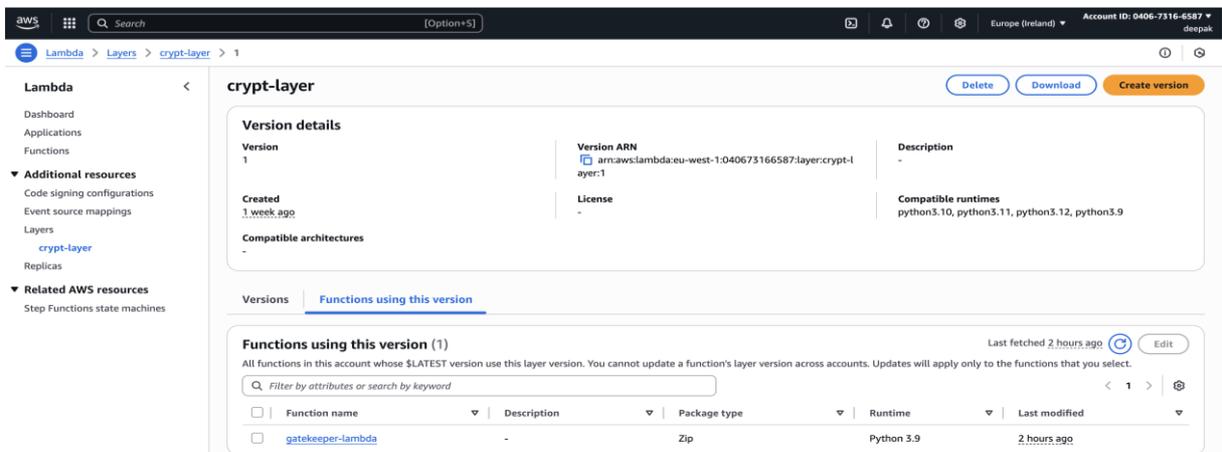


Figure 7: Lambda layers dashboard

## API Gateway

1. In this project, REST API is created under API gateway for exposing the gatekeeper-lambda, in the API creation page provide the name for the API and create.
2. Once API is created, under the API create a resource with the path “/” and resource name “secure”.
3. Under the created resource, the method needs to be created. Here POST method is required so select POST method and provide the gatekeeper Lambda ARN for integration since the API invokes the lambda function.
4. Once method is created, the authorization needs to be attached, under method request attach “CognitoAuth” since we are using Cognito for user authentication. The figure 8 shows the all the configuration of API.

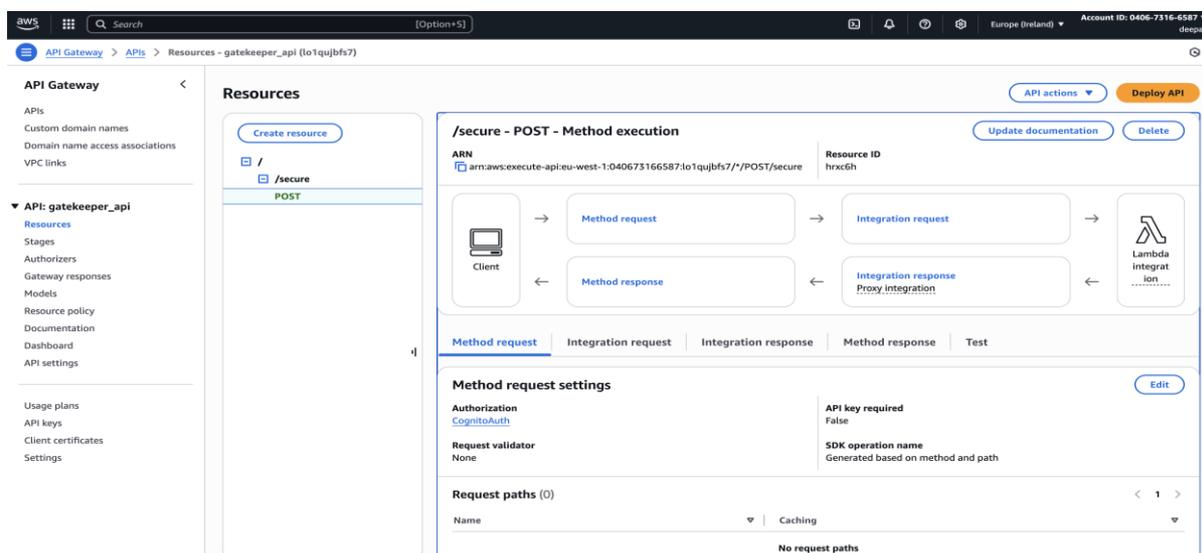


Figure 8: API gateway resource page

## Docker and AWS Elastic Container Registry

1. The trained machine learning model (model.joblib) is containerize as Docker image, first a directory is created “ml-fargate” inside that placed Dockerfile, app.py, model.joblib and requirement.txt.
2. Once all the files placed inside the directory “ml-fargate” containerize as a docker image using below commands.  
`$docker build -t ml-fargate .`
3. The docker image can be run and tested locally using below command. Docker needs to be installed in the platform as prerequisites.  
`$docker run -p 8080:8080 ml-fargate`
4. Now, an ECR repository created using below AWS CLI command. AWS CLI should be installed in the platform as prerequisites.  
`$ aws ecr create-repository --repository-name ml-fargate`
5. Once the repository is created, docker needs to be authenticated to access ECR repository, using the below AWS CLI command.

```
$ aws ecr get-login-password --region eu-west-1 | \
docker login --username AWS --password-stdin 040673166587.dkr.ecr.eu-west-1.amazonaws.com
```

6. Now, tag the docker image as “latest” and push to the repository using below command.

```
$ docker tag ml-fargate 040673166587.dkr.ecr.eu-west-1.amazonaws.com/ml-fargate:latest
```

```
$ docker push 040673166587.dkr.ecr.eu-west-1.amazonaws.com/ml-fargate:latest
```

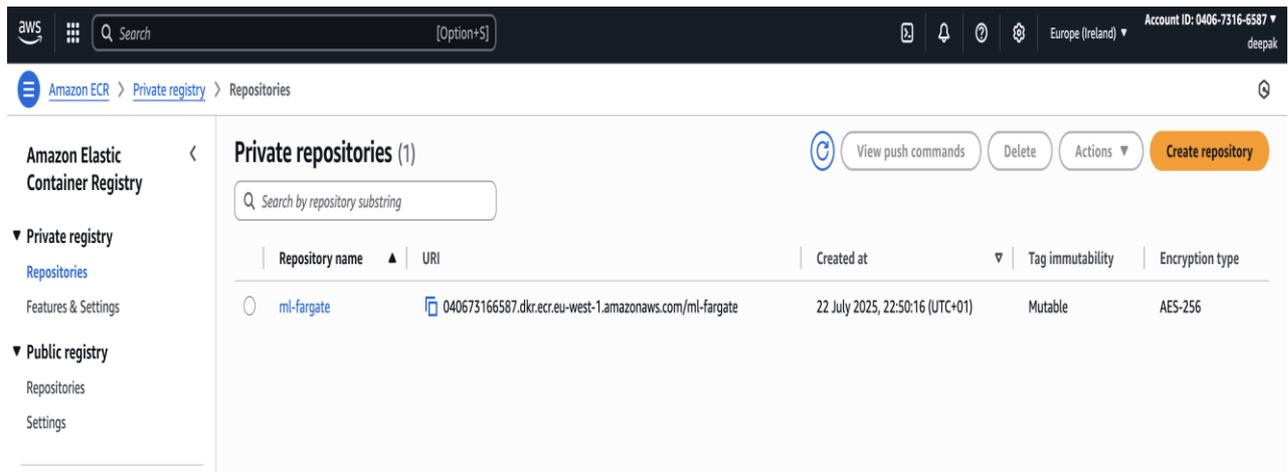


Figure 9: AWS ECR repository

## Amazon Elastic Container Service

1. To deploy the containerized docker image and to manage the docker container, AWS ECS is used. First, a cluster is created with Fargate (serverless) and with the cluster name “ml-fargate-cluster”, shown in figure 10.
2. Once cluster is created, a task definition is required. Under task definition create new, launch type as fargate, name as “ml-task-def” and choose role “ecsTaskExecutionRole” with necessary permission, shown in figure 11.
3. Under container definition, add container then select the docker image “040673166587.dkr.ecr.eu-west-1.amazonaws.com/ml-fargate:latest”, port mapping as “8080” click add and then create.
4. Now create a service under cluster, choose launch type as fargate, then choose “ml-task-def”, then give the service name “ml-api-service” and number of task as “1”.
5. Now under networking section, provide the necessary network configuration, select the vpc and two subnets as required. Select a security group which has inbound rule for port “80” and “8080” open.
6. Enable ALB (Application Load Balancer), create new then give listener port as “8080” then create a target group with name “ml-api” and port “8080”.
7. Finally deploy the service and check the load balancer page for DNS, then test.
8. The DNS will be available in ALB page “ml-lb-1330515347.eu-west-1.elb.amazonaws.com”.

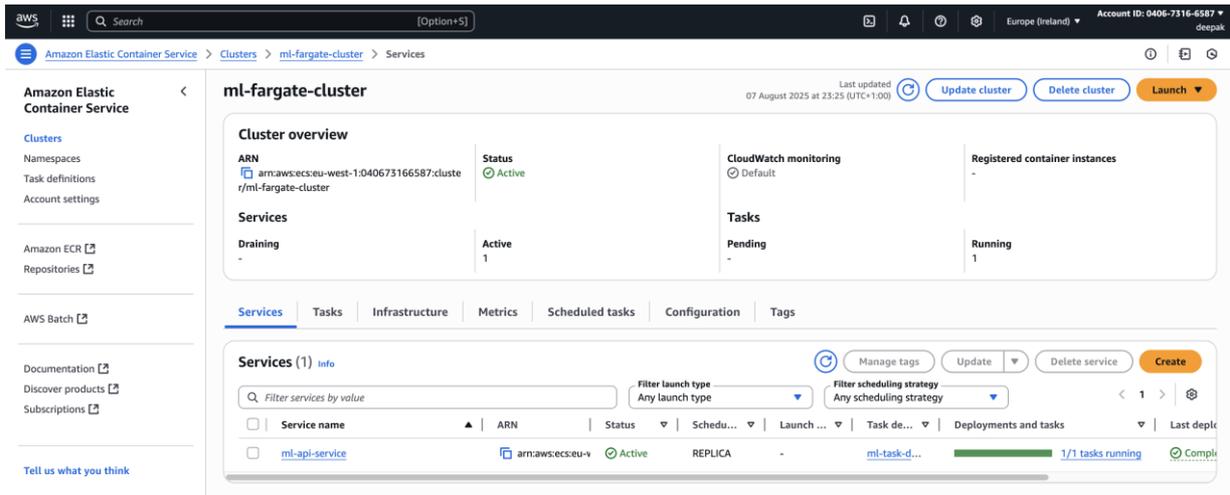


Figure 10: AWS ECS cluster page

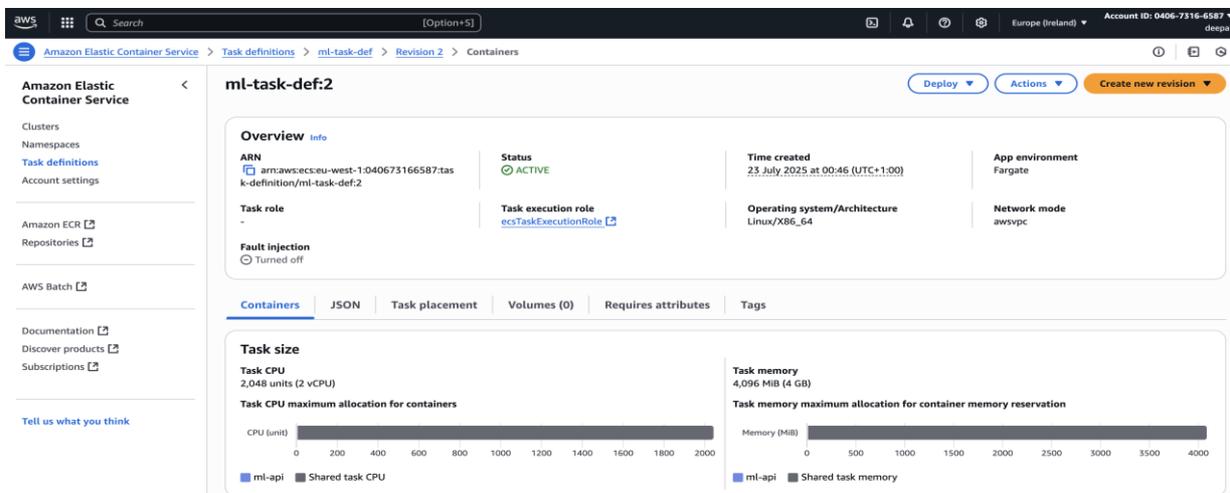


Figure 11: Task definition

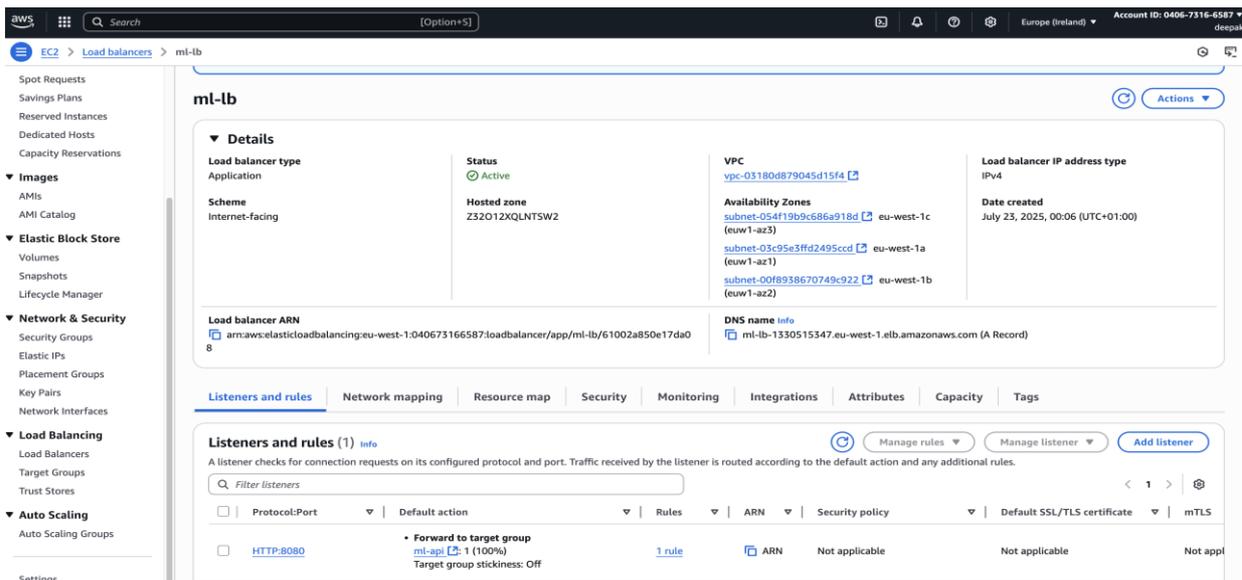


Figure 12: ALB dashboard

## 1.2 Anomaly Detection model training setup

This section covers the environment setup required for machine learning model training and storing the trained model.

### AWS Sagemaker Jupyter notebook

1. To perform the machine learning model training, a jupyter notebook is required. To create a notebook instance go to AWS Sagemaker AI service, then create a notebook instance, provide the instance name, instance type and platform, also provide the IAM role with necessary permission.
2. Once the instance is created, open the jupyter notebook and upload the synthetically generated dataset “lambda\_dataset.csv” and run the data preprocessing code to preprocess and clean the code for training, this will produce “features.csv” and “labels.csv”
3. Now train the model with training code and save the model as “model.joblib”, before training the sklearn library version should be “1.2.1” it can be installed by below command.

```
$pip install scikit-learn==1.2.1
```

4. Now the trained model is stored in the jupyter notebook instance and download for deployment.

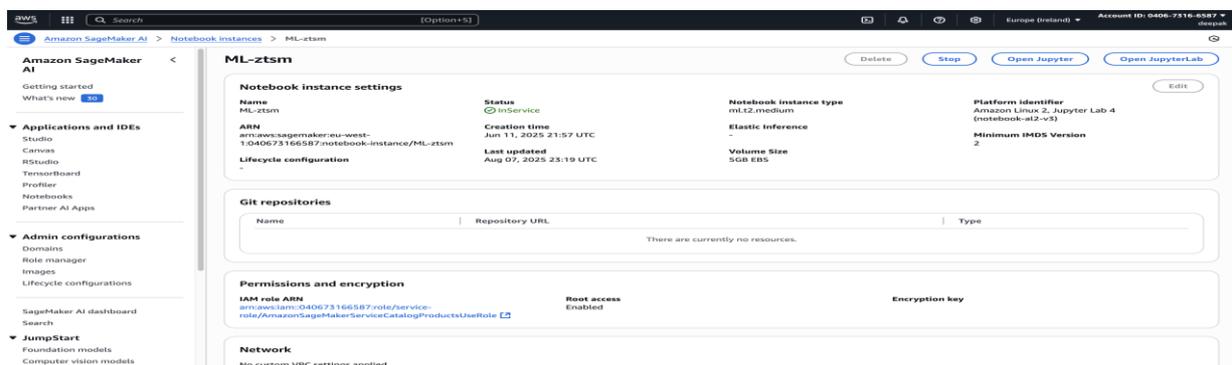


Figure 13: AWS Sagemaker instance detail

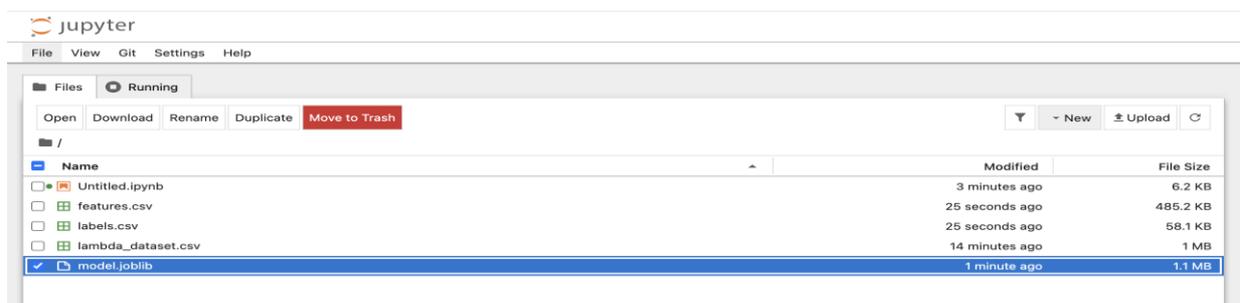


Figure 14: Jupyter notebook instance

## 2 Testing

After the entire setup is done, the setup can be tested using below commands either in the local linux machine or ec2 instance.

1. Below command stores the value in variables

```
$ USER_POOL_CLIENT_ID="<userpool_id>"
$ USERNAME="<username>"
$ PASSWORD="<password>"
$ API_URL="<deployed api endpoint url>/secure"
```

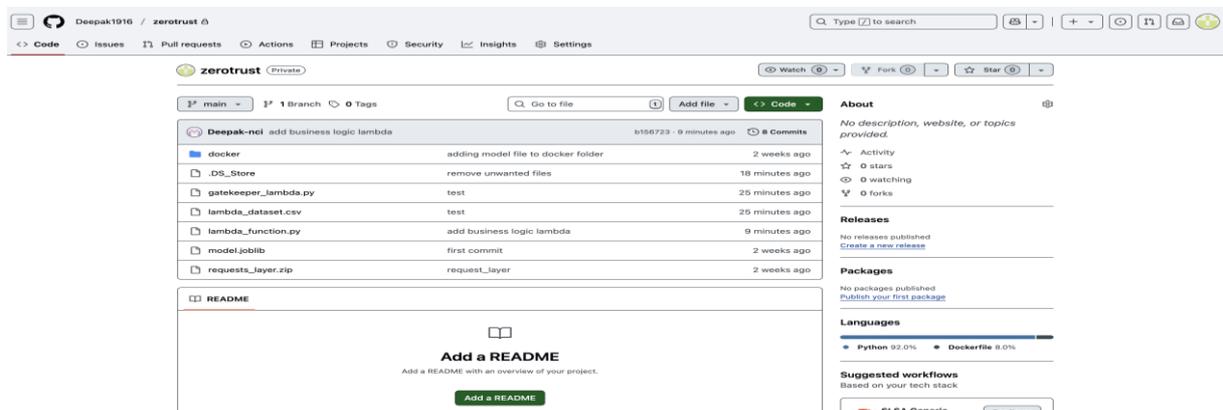
2. This command gets the cognito token.

```
$ TOKEN=$(aws cognito-idp initiate-auth \
--auth-flow USER_PASSWORD_AUTH \
--client-id $USER_POOL_CLIENT_ID \
--auth-parameters
USERNAME=$USERNAME,PASSWORD=$PASSWORD \
--query 'AuthenticationResult.IdToken' \
--output text)
```

3. This command calls the gatekeeper lambda to validate the token and to call the model to predict.

```
$ curl -X POST "$API_URL" \
-H "Authorization: Bearer $TOKEN" \
-H "User-Agent: test-client" \
-H "Content-Type: application/json" \
-d '{"message": "hello"}'
```

## 3 GitHub Repository



The screenshot shows a GitHub repository page for 'zerotrust' by user 'Deepak1016'. The repository is private and has 1 branch (main) and 0 tags. The file list includes:

File Name	Commit Message	Commit Time
docker	adding model file to docker folder	2 weeks ago
.DS_Store	remove unwanted files	18 minutes ago
gatekeeper_lambda.py	test	25 minutes ago
lambda_dataset.csv	test	25 minutes ago
lambda_function.py	add business logic lambda	9 minutes ago
model.joblib	first commit	2 weeks ago
requests_layer.zip	request_layer	2 weeks ago

The README section is currently empty, with a prompt to 'Add a README'. The right sidebar shows repository statistics: No description, website, or topics provided; 0 stars, 0 watching, 0 forks; No releases published; No packages published; Languages: Python 92.0%, Dockerfile 8.0%; Suggested workflows: SLSA Generic.

GitHub Repository URL: <https://github.com/Deepak1916/zerotrust.git>

## References

Amazon Web Services. (2024). *Amazon API Gateway developer guide* [Documentation]. <https://docs.aws.amazon.com/apigateway/latest/developerguide/welcome.html>

Amazon Web Services. (2024). *Amazon Cognito developer guide* [Documentation]. <https://docs.aws.amazon.com/cognito/latest/developerguide/what-is-amazon-cognito.html>

Amazon Web Services. (2024). *AWS Lambda developer guide* [Documentation]. <https://docs.aws.amazon.com/lambda/latest/dg/welcome.html>

Amazon Web Services. (2024). *Amazon Elastic Container Service documentation* [Documentation]. <https://docs.aws.amazon.com/AmazonECS/latest/developerguide/Welcome.html>

Amazon Web Services. (2024). *Amazon Elastic Container Registry documentation* [Documentation]. <https://docs.aws.amazon.com/AmazonECR/latest/userguide/what-is-ecr.html>

Amazon Web Services. (2024). *Amazon SageMaker developer guide* [Documentation]. <https://docs.aws.amazon.com/sagemaker/latest/dg/whatis.html>