

Configuration Manual

MSc Research Project Cloud Computing

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Configuration Manual

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1 Introduction

This configuration guide explains the specification of the entire configuration and deployment of the intelligent multi-cloud CI/CD system built up in this research. You then build it across GCP, Azure and AWS using GitHub Actions, Kubernetes, Docker and Azure Machine Learning to deploy a Djangoweb app. The arrangement allows for AI based cloud selection as well as Kubernetes rollback on performance failures. This guide serves practitioners and researchers so they can reproduce, replicate, and extendthe system.

2 System Configuration Setup

2.1 Software Requirements

Tool/Platform	Version / Notes
OS	Ubuntu 22.04 / Windows 11 / macOS Ventura
Python	3.10
Docker	24.x
GitHub CLI	Latest version
Terraform	1.4.x
Google Cloud SDK	With gke-gcloud-auth-plugin
Azure CLI	Latest version
AWS CLI	v2
VS Code (Recommended)	With Docker, Python & Kubernetes extensions
kubectl	v1.26+
Azure Machine Learning SDK	azureml-core, azure-identity, joblib, etc.

Table 1: Tools and Platforms Used

2.2 Hardware Requirements

• **Processor:** Intel i5/i7 or Apple M1

• RAM: Minimum 8 GB (16 GB recommended)

• Disk Space: Approximately 20 GB (for images, logs, data)

• Network: Stable broadband internet connection

3 Environmental Setup

3.1 GitHub Repository Setup

- Github Repository Setup: git clone https://github.com/sahil0480/NCI_Research_Project.git
- Setup a Python environment: python -m venv .venv
 source .venv/bin/activate
 pip install -r requirements.txt
- Install Docker and enable Buildx: docker buildx install



Figure 1: Requirement.txt

4 Cloud & Credential Setup

4.1 GitHub Secrets Configuration

All secrets are stored under GitHub repo \rightarrow Settings \rightarrow Secrets \rightarrow Actions.

Secret Name	Purpose
GCP_SA_KEY	GCP Service Account JSON
AZURE_KEY	Azure Client ID, Secret, Tenant ID
AZURE_STORAGE_KEY	Azure Blob Storage Key
AWS_KEY	AWS IAM User Key (in JSON format)
AWS_ACCESS_KEY_ID	AWS Key ID (if separate)
AWS_SECRET_ACCESS_KEY	AWS Secret (if separate)

Table 2: Secrets and Their Purpose in Multi-Cloud Deployment



Figure 2: Github Secret Keys

5 Infrastructure Setup

5.1 Terraform Scripts

The Terraform configuration for this project has been structured into three cloud environments AWS, Azure, and GCP each in seperate folder and deployment scripts. This ensures modular and cloud-specific infrastructure provisioning.

Terraform AWS Infrastructure

terraform/aws/ is the directory used to provision an Amazon EKS (Elastic Kubernetes Service) cluster on AWS. It contains the following key files and directories:

- **provider.tf** Specifies the AWS provider configuration including access credentials and region.
- main.tf Contains the core infrastructure logic to create a VPC, subnets, IAM roles, and the EKS cluster itself.
- aws-auth.yaml A configuration file for granting access to EKS worker nodes via kubectl.
- **outputs.tf** Defines output variables such as the EKS cluster name and kubeconfig for access.
- variables.tf Declares input variables like region and cluster name to enable parameterized deployments.
- modules/ An optional directory for reusable and modular Terraform code components.
- .tfstate files Automatically generated files that track the current state of infrastructure resources for consistent and idempotent deployment.

Terraform Azure Infrastructure

terraform/azure/ provisions an AKS (Azure Kubernetes Service) cluster. Key components include:

- providers.tf Configures the Azure provider with subscription ID and tenant ID.
- main.tf Provisions the AKS cluster and its associated node pools.
- **terraform.tfvars** Contains values for variables such as location, node count, and resource group.
- **outputs.tf** Defines output values such as kubeconfig credentials for accessing the AKS cluster.
- variables.tf Declares input variables like region, resource group, and node count for customization.

Terraform GCP Infrastructure

terraform/gcp/ provisions a GKE (Google Kubernetes Engine) cluster. The folder includes the following:

- **provider.tf** Configures the Google Cloud provider using credentials and project information.
- main.tf Creates the GKE cluster, including node pools and necessary networking resources.
- **terraform.tfvars** Stores values for variables such as GCP project ID, zone, and cluster name.
- **outputs.tf** Outputs important details such as the cluster endpoint and kubeconfig for access.
- variables.tf Accepts input parameters like region, project name, and desired cluster name.



Figure 3: Terraform Folder Structure

6 Application Setup - Containerization with Docker

6.1 Dockerfile

Docker for containerization (for consistency across dev, testing, production) for the Django RMSapplication. If we lookat the Dockerfile located at the root of the project, we see that it starts from an official Python 3.9 slim image. It also sets some environment variables to boost Python runtime and sets /appas the working dir. It installs some essential system dependencies (gcc, libpq-dev, curl) that are required to make the database work and also it will generate therun environment. Python dependencies are installed from requirements. txt and copy over the entire project files into the container. Prior to launching the application, Django's collect static command is run, collecting all static files and preparing for serving in a production-like environment. The container listens on port 8000 and runs the app through Gunicorn, a high-performance WSGI server, bound to all interfaces.

```
dockerfile 1 X  docker-composeyml

dockerfile 2 ...

# Use official Python image as base

FROM python:3.9-slim

# Set environment variables

ENV PYTHONDONTWRITERYTECODE-1

ENV PYTHONDONTWRITERYTECODE-1

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```

Figure 4: Docker File Code Snippet

6.2 Dockerfile-Compose

Docker Compose (docker-compose. yml) serves as the orchestrator of this multi-container environment, bringing together the Django web application and PostgreSQL database. The web service constructs the docker image based on the defined Dockerfile, it mounts the local project directory into the docker container and uses named volumes topersist static and media files. It is reading configuration variables from the. env file,runs collectstatic and starts the app with Gunicorn The db service pulls an official PostgreSQL 15 image and sets up a database from some environment variables including the database name, user, and password. We attach volumes to persist data, static,and media. It allows for smooth local development, and can easily be moved to run on the cloud as part of Kubernetes and CI/CD.

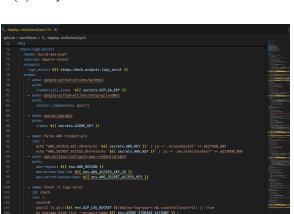
Figure 5: Docker-Compose File Code Snippet

7 CI/CD Pipeline Setup

On each commit to the master branch, itautomatically triggers the CI/CD pipeline. It startswith building and pushing Docker images to all three prominent cloud providers' container registries, namely Google Cloud Artifact Registry, Azure Container Registry (ACR), and Amazon Elastic Container Registry (ECR). After that, it will deploy the app on Kubernetes clusters that hosted on GKE, AKS, and EKS. Once deployed, the pipeline collects the deployment metrics and logs from each cloud environment, ultimately storing them in their appropriate storage services: Google Cloud Storage (GCS), Azure Blob Storage, and AWS S3. These logs are then aggregated into a single dataset, which spurs an Almodel training job to initiate inside Azure Machine Learning. The intelligent deployment to the best cloud provider is done on the basis of the model prediction. In case thedeployment fails it triggers an automatic Kubernetes rollback mechanism to bring back the last stable version.



(a) Step 1: Initialize and Define Variables



(c) Step 3: Multi-Cloud Log Check

(e) Step 5: Log Aggregation

(b) Step 2: Azure & AWS Login

```
Control of the Contro
```

(d) Step 4: Initial Deployment (GCP & Azure)

(f) Step 6: AI-Based Cloud Selection & Final Deploy

Figure 6: GitHub Actions Workflow for Multi-Cloud Deployment

8 AI Model Training via Azure ML

Training and deployment decisions are executed automatically within the pipeline.

• ML models trained: Logistic Regression, Random Forest, XGBoost, KNN

- Dataset: Merged deployment logs from all cloud providers
- Platform: Azure ML Studio with Python SDK

```
print_model.py x

scripts > @ train_model.py

isinport sys
isingort
isingo
```

Figure 7: AI Model Training Code Snippet

9 Rollback Strategy

- Triggered using: kubectl rollout undo deployment/django-app
- Activated if:
 - Logs show sub-threshold metrics
 - AI model predicts poor performance post-deployment

10 Dataset Summary

- Total Records: 9000
- Columns: cloud, start_time, end_time, duration_seconds, cpu_utilization, memory_usage, deployment_size, cost_estimate, status_code, retry_count, infra_load_score, region, status
- Cloud Platforms: GCP, Azure, AWS
- Duration Range (seconds): 30 to 600
- Status Types: success

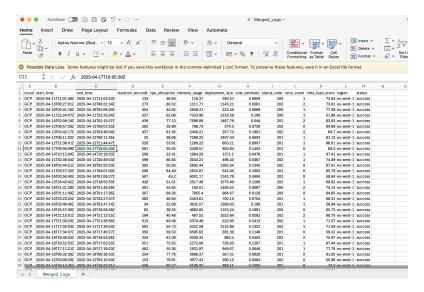


Figure 8: Dataset Snippet

11 Output & Prediction

- Final cloud decision stored in best_cloud.txt.
- Automated deployment to predicted cloud provider (GCP/AWS/Azure).

12 Troubleshooting

Issue	Solution
Deployment failed	Check log output in CI job, validate credentials
Rollback not triggered	Verify thresholds & log parsing script
AI not training	Check Azure ML authentication or dataset format

Table 3: Common Issues and Their Solutions