

Configuration Manual

MSc Research Project
Cloud Computing

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Project Submission Sheet
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Configuration Manual

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1 System Requirements

1.1 Hardware Requirements

1.1.1 Minimum Requirements:

- CPU: 4 cores (Intel Core i5 or AMD Ryzen 5 equivalent)
- RAM: 8 GB
- Storage: 10 GB free space
- Network: Stable internet connection (for dataset download)

1.1.2 Tested Configurations

- Configuration 1 (Development):
 - CPU: Intel Core i7-10700K (8 cores)
 - RAM: 16 GB DDR4
 - GPU: None (CPU-only training)
 - OS: Windows 11
 - Training Time: 1 hours per task (forecasting)
- Configuration 2 (AWS Deployment):
 - Instance Type: AWS EC2 t3.small and AWS EC2 t2.large
 - CPU: 2 vCPUs
 - RAM: 2 GB
 - OS: Amazon Linux 2

1.2 Software Requirement

- Operating System
 - Windows 10/11 (64-bit)
 - Ubuntu 20.04/22.04 LTS
 - macOS 11+ (Big Sur or later)

- Amazon Linux 2 (for AWS deployment)
- Required Software
 - Python 3.11.x (tested with 3.11.9)
 - pip (Python package manager)
 - Git (for repository management)
- Optional
 - Anaconda/Miniconda (recommended for environment management)
 - CUDA Toolkit 11.8+ (if using GPU)

2 Environment Setup

2.1 Python Installation

2.1.1 Windows

- Download Python 3.11.9 from python.org
- During installation, check "Add Python to PATH"
- Verify installation: `python --version`

2.1.2 Linux/Ubuntu

- `sudo apt update`
- `sudo apt install python3.11 python3.11-venv python3.11-dev`
- `python3.11 --version`

2.2 Virtual Environment Setup

- Navigate to project directory
- Create virtual environment Windows:`venv\Scripts\activate`, Linux:`venv/bin/activate`
- Verify activation (prompt should show (venv))

3 Installation Instructions

3.1 Clone Repository

- Git repository (Khalam; 2025): `git clone <https://github.com/Anisha151098/vm\_predictor>`
- `cd GRU`

3.2 Install Dependencies

- Install Core Dependencies
 - * Ensure virtual environment is activated
 - * Install all required packages
 - `pip install tensorflow==2.17.0` (Abadi et al.; 2016)
 - `pip install numpy==2.2.3`
 - `pip install pandas==2.2.3`
 - `pip install scikit-learn==1.6.1`
 - `pip install matplotlib==3.10.1`
 - `pip install seaborn==0.13.2`
 - `pip install requests`
 - `pip install tqdm`
 - `pip install h5py`
- Create requirements.txt
 - `pip freeze > requirements.txt`

4 Running the Code

- Activate environment:for Windows: `venv\Script\activate`. For Linux\Ubuntu: `source venv/bin/activate`
- Run forecasting training: `python train_azure_forecast_1hr.py`, this will try to download dataset (Azure; 2017)
- This will trains 5 models with forecasting labels and the results are saved in `results_azure_forecast_1hr`
- It takes around 2 to 3 hours to train all models.
- `create_individual_graphs.py` will generate graphs.

5 AWS Deployment

- Before starting ensure that an active AWS with billing is provided, IAM user with appropriate permission

5.1 EC2 Setup

- Open AWS Management console and sign in
- Navigate to EC2→ Launch Instance
- Name the instance (Eg:GRU\Attention), select Amazon Linux as the Operating System. Select Instance type as t2.medium. Create a key pair of .pem format for login and download to the local system. Click launch instance. Review and launch instance (Services; 2025b).

5.1.1 IAM policy For EC2 Instance

- Navigate to IAM. From AWS Management Console, go to services select Security, Identity & Compliance → select IAM
- IAM → Roles → Create role → Select service or usecase as EC2 → Next
- Now select the policies needed, search AmazonEC2ReadOnlyAccess and CloudWatchReadOnlyAccess (Services; 2025a). Enable this permissions → Next
- Give a role name → Create role
- To **Create inline policy** Go to the role created → In permissions policies, click on Add permissions → Create inline policy. In policy editor, click on json, then add this json script to consolidate the idle instances forecasted.

```
{
  "Version": "2012-10-17",
  "Statement": [
    {
      "Effect": "Allow",
      "Action": [
        "ec2:StopInstances"
      ],
      "Resource": "*"
    }
  ]
}
```

- Next → Give policy name → Create policy

5.2 Launch 3 t3.micro instances

- Navigate to: EC2 → Instances → Launch Instance
- Configuration: Name: ec21 (repeat for ec22, ec23) AMI: Amazon Linux Instance type: t3.micro Key pair: Create new or use existing Network settings: Default VPC
- Launch instance
- Repeat 3 times total

5.3 Workload setup for instance

5.3.1 Variable Spikes Pattern for ec21

- SSH into ec21(ssh -i vm-predictor-key.pem ec2-user@)
- Install stress tool - sudo yum install stress -y
- Create workload script

```

cat << 'EOF' > /home/ec2-user/workload_ec21.sh

#!/bin/bash
while true; do
    # 5 minutes idle (0-2% CPU)
    sleep 300

    # 2 minutes spike (60-80% CPU)
    stress --cpu 1 --timeout 120s

    # 3 minutes moderate (20-30% CPU)
    stress --cpu 1 --timeout 180s --cpu-method fibonacci
done
EOF

```

- Run:

```
chmod +x /home/ec2-user/workload_ec21.sh
```

- Run this to keep the script run in background:

```
nohup /home/ec2-user/workload_ec21.sh > /dev/null 2>&1 &
```

5.3.2 Sustained High Load in ec22

- ssh to the instance
- Install stress :`sudo yum install stress -y`
- Run this for sustained load:

```

cat << 'EOF' > /home/ec2-user/workload_ec22.sh
#!/bin/bash
stress --cpu 1 --cpu-method ackermann &
EOF

```

- Run this to keep the workload run in background:

```

chmod +x /home/ec2-user/workload_ec22.sh
nohup /home/ec2-user/workload_ec22.sh > /dev/null 2>&1 &

```

5.4 Connect EC2 via SSH and Deployment set up in EC2

- On the local machine, Navigate to the directory where the pem file of the instance downloaded.
- Connect the instance via ssh
- Verify the trained model exist in same directory
- Run extract_model.py from the git repo.
- Verify the dependencies installation
- Install pip install boto3==1.35.0
- Verify AWS credentials configured
- nano ec2_forecast_exact_architecture.py, copy and paste ec2_forecast_exact_architecture.py code from the git repo
- Run this code python ec2_forecast_exact_architecture.py
- Forecast result will be shown in the command prompt.

```
Overall Statistics:
  Total Forecasts: 3
  State Changes Predicted: 1 (33.3%)
  Average Confidence: 83.1%

=====
FORECAST SUMMARY
=====

c21:
  Current: ACTIVE (max CPU: 19.7%)
  Forecast (t+60min): ACTIVE (84.4% confidence)
  Action: CONTINUE MONITORING

c22:
  Current: ACTIVE (max CPU: 48.3%)
  Forecast (t+60min): ACTIVE (80.2% confidence)
  Action: CONTINUE MONITORING
```

Figure 1: Sneak peek of output obtained

References

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