USING MACHINE LEARNING FOR GREENING THE TRAIN TICKET APPLICATION

Carbon-aware autoscalers for microservices in the cloud Abhinandan Roul, Ishara Madhavi, Mahira Joytu, Phuc Tran







INTRODUCTION

Cloud-based microservices are increasing energy demand in data centers, making sustainability a critical challenge in modern computing.

As digital infrastructure expands, we need smarter, more sustainable autoscaling solutions that optimize resource use and align with carbonaware computing to reduce environmental impact without compromising performance.

OBJECTIVES

A smart autoscaling policy that improves resource utilization, optimizing for reduction in carbon emissions.







Energy Consumption Inefficient scaling leads to higher energy usage

Carbon Intensity No consideration about carbon footprint

Response Time High response times impact user experience

User Count Over Time

To have a reasonable trade-off between sustainability and performance.

DATASET

User Count Variation Over Time

PIPELINE



Coverage Analysis

JACOCO Java Code Coverage



Tracing



Train Ticket Application Deployment

Carbon Intensity Data













Scaling Decision













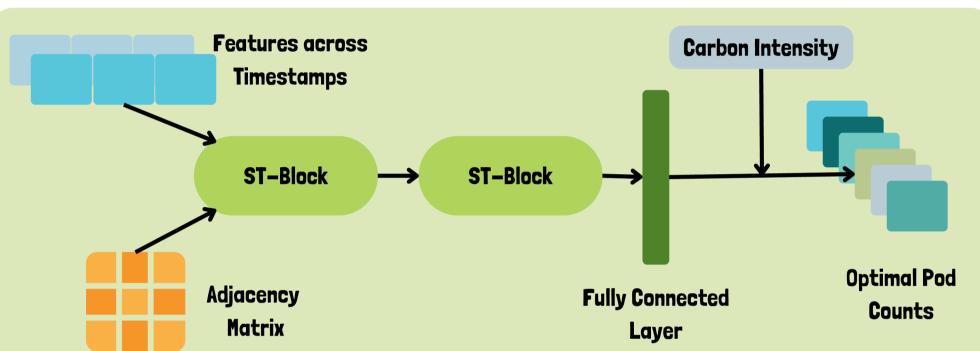


Model N





METHODOLOGY



For Workload Forecasting

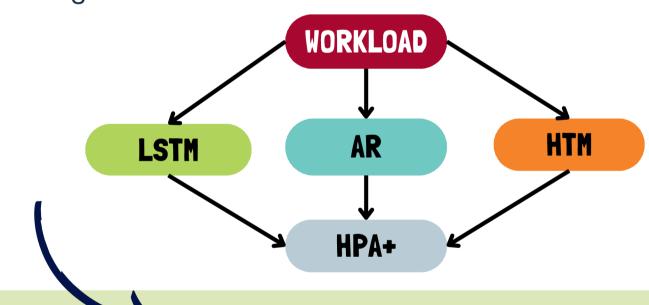
MODELS

Neural Network (STGCN)

For Scaling Decision

Spatio-Temporal Graph Convolution

Use timeseries forecasting models (LSTM, AR, HTM) to predict incoming web requests and make proactive scaling decisions.



TRADE-OFF ANALYSIS \bigcirc

Formulate trade-off policies that prioritize either response time or carbon emissions, depending on situational demands.

ENERGY USAGE OF THE MODELS 🗲

Analyze the energy overhead associated with both the training and inference processes.

- 63 microservices written in many languages: Java, Go, Python.
- Peak user count: 100 users.
- Duration: ~ 3 hours.
- Slight increases, slight decreases, sharp increases, sharp decreases, and fluctuations.

99% Percentile of Response Time Over Time 25000 HPA STGCN Finland STGCN Poland g 20000 <u>i</u> 15000 10000 5000 2000 Time offset (s)

CARBON FOOTPRINT Poland Finland Carbon Intensity 660 g CO2/kWh 128 g CO2/kWh HPA **8.1** g **41.81** g Carbon emissions **STGCN 3.31** g **12.52** g

Energy usage varies depending on carbon intensity, scaling up when energy is

cleaner and down when it's more carbon-intensive.

CHALLENGES

- Balancing sustainability and performance.
- Managing a complex microservices architecture.
- Handling workload variability.
- Ensuring accurate workload forecasting.
- Identifying useful metrics.

