# TMLL

# Automated Performance Diagnosis through ML-Driven Trace Analysis

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This should be 05, but I don't have access to change it anymore :( (the day is coincidentally correct)

## What is Trace Compass?

#### **Trace Compass** is a tool for solving **performance** and **reliability** issues by **analyzing** system **traces**, providing userfriendly **views**, **graphs**, and **metrics**.

## What is Trace Server?

#### An **open-source tool** for **analyzing logs** and **traces**, enabling users to interact via **client-side wrappers** in **Python** and **TypeScript**, and providing **various analyses** and **outputs** without requiring direct use of Trace Compass.

#### Think of it as the core of Trace Compass, but without the UI

## What areas can be investigated with TS?

#### Identifying performance bottlenecks

- CPU usage, disk I/O view, memory usage, scheduler and system call latency
- Diagnosing synchronization issues
  - Futex contention and IRQ latency
- Analyzing system and application behavior
  - Events table, flame chart call stack, thread and resources statuses
- Optimizing memory usage
  - Memory usage, memory latency
- Detailed latency analysis
  - Futex contentions, IRQ, schedulers, system calls, call stack

#### What is the procedure right now?



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#### **Issues and Downsides**

#### **Trace Server**

- User should know How to Get
- User should know What to Get

#### AI/ML

- User should know What to Use
- User should know How to Use
- User should know How to Analyze

#### What are we proposing?



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#### **TMLL Overview**



#### **TMLL Architecture**



Viewer (e.g., Jupyter Notebooks, VSCode Trace Extension, Theia Trace Viewer, etc.), Exporter (e.g., SVG, PNG, etc.), or Trace Compass Bookmark

#### **TMLL Features and Modules**

#### **Anomaly Detection**

- Identify unusual patterns in system
- Provide potential regions of interest

#### **Resource Optimization**

- Optimize system resource allocation
- Analyze which areas can be optimized

#### **Predictive Maintenance**

- Predict and prevent system failures
- Done in real-time analysis scenarios

#### **Performance Trend Analysis**

- Understand long-term performance trends
- Provide statistical insights on how system evolves

## Scenario 1 Something weird is happening, which is out of our expectations...

### **Anomaly Detection**



## **Anomaly Detection (cont'd)**



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## Anomaly Detection (cont'd)



Anomaly Detection for "combined (pca)" using "zscore" method

Potential Anomaly Period: 2024-04-24 02:13:**11.796**169984 to 2024-04-24 02:13:**11.890**216704

# <u>Scenario 2</u>

# How are my system components interacting with each other?

## **Correlation Analysis**



## **Correlation Analysis (Lag Analysis)**



## Scenario 3

## No significant internal/external factors have changed, yet the system's behavior has altered at an unknown point in time...

### **Change Point Detection**



### Change Point Detection (cont'd)



# <u>Scenario 4</u>

# Are there significant idle times in my system components? I want to save up some money...

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#### **Idle Resource Detection**



MEMORY Utilization: Memory Usage (2) (Idle 52.9% of time, Avg: 508.06MB, Peak: 651.20MB, Pattern: Moderate variation)





DISK Utilization: Disk I/O View (Idle 76.9% of time, Avg: 57.34MB/s, Peak: 1015.70MB/s, Pattern: Highly variable)

### **Idle Resource Detection (CPU Scheduling)**



# <u>Scenario 5</u>

# I'm going on vacation. Should I be concerned about any upcoming workloads while I'm away?

### Capacity Planning (i.e., Forecasting)



### **Scenario X** I want to analyze features $c_1, c_2, ..., c_n$ to perform tasks $t_1, t_2, ..., t_n$ . What can I do?

#### TMLL is highly extensible. Since the data is now in Python Pandas Data Frames, you can essentially do anything you want! Just checkout the documentations!

#### **Data Density Adjustments**



#### How to Use?

```
main.py
          from tmll.tmll_client import TMLLClient
          from tmll.ml.modules.anomaly_detection.anomaly_detection_module import AnomalyDetection
          # Initialize the TMLL client
          client = TMLLClient()
          # Create an experiment from trace files
          experiment = client.create_experiment(traces=[
                   "path": "/path/to/trace/file",
                                                  Which trace(s)?
           ], experiment_name="My Experiment")
                                                           Which output(s)?
          # Run anomaly detection
          outputs = experiment.find_outputs(keyword=["cpu usage", "memory usage"]
                                                                                   type=["xy"])
  Which
          ad = AnomalyDetection[client, experiment, outputs]
module?
          anomalies = ad.tind_anomalies(method="iforest", zscore_threshold=3.5)
          ad.plot_anomalies(anomalies)
```

