



GOKARAJU RANGARAJU INSTITUTE OF ENGINEERING AND TECHNOLOGY
Department of Computer Science and Engineering

Mini Project Review-1

AI-Based Cloud Operations Monitoring System

Under the esteemed guidance of

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CONTENTS

1. Problem statement
2. Abstract
3. Introduction
4. Objective and Motivation
5. Literature Survey

PROBLEM STATEMENT

- Cloud systems generate a large amount of monitoring data
- Manual monitoring is time-consuming and inefficient
- Problems are detected only after failures occur
- Multiple alerts create confusion and delay response
- Slow issue resolution increases system downtime
- There is a need for early problem detection

Therefore, some automated solutions are needed to reduce the human intervention .

ABSTRACT

- Continuously collects and monitors cloud system data
- Automatically analyzes data instead of manual checking
- Identifies unusual behavior early
- Predicts possible failures before they occur
- Reduces unnecessary alerts
- Automatically fixes common problems
- Helps reduce downtime and human effort

INTRODUCTION

- Cloud computing is widely used to host applications and services
- Cloud systems generate data like CPU usage, memory usage, and logs
- Monitoring large cloud systems manually is difficult
- Traditional monitoring detects problems only after failure
- This leads to downtime and slow recovery
- This project focuses on predicting problems early
- It also performs basic automatic fixes to reduce downtime

OBJECTIVES

- To design a system that tracks the health of cloud resources.
- To develop a mechanism for identifying warning signs of failures.
- To implement basic automatic actions to improve system availability.

MOTIVATION

- Increasing cloud system failures highlight the need for better monitoring
- Manual handling of cloud issues is slow and inefficient
- Early detection and automation can greatly reduce downtime and effort

LITERATURE SURVEY

| S.No. | Title of The Paper & Year | Authors | Methodology & Metrics | Datasets Used | Observed Shortcomings/ Gaps in The Paper |
|-------|---|---------------------------------|--|--|---|
| 1. | AIOps: Real-Time Analytics for Cloud Operations (2018) | Shubhangi Vashistha, Ravi Kumar | <ul style="list-style-type: none">Analyzed cloud monitoring dataUsed anomaly detection techniquesMetrics: CPU usage, memory usage, response time | <ul style="list-style-type: none">Cloud system monitoring dataLogs collected from cloud servers | <ul style="list-style-type: none">Focuses mainly on detection, not predictionNo automatic incident resolutionLimited discussion on reducing alert noise |

LITERATURE SURVEY

| S.No. | Title of The Paper & Year | Authors | Methodology & Metrics | Datasets Used | Observed Shortcomings/ Gaps in The Paper |
|-------|---|---------------------|---|--|--|
| 2 | Predictive Analytics for Incident Management in Cloud Systems (2020) | A. Sharma, P. Gupta | <ul style="list-style-type: none"> Machine learning models for incident prediction Historical incident analysis Metrics: accuracy, prediction rate, downtime | <ul style="list-style-type: none"> Historical cloud incident data System performance logs | <ul style="list-style-type: none"> Requires large historical datasets Does not include automated remediation High dependency on manual intervention |
| 3 | Anomaly Detection for Cloud Monitoring Systems (2019) | L. Chen, M. Zhang | <ul style="list-style-type: none"> Analyzed system performance data Used statistical and pattern-based anomaly detection Metrics: CPU usage, memory usage, network delay | <ul style="list-style-type: none"> Cloud server performance datasets Monitoring data collected over time | <ul style="list-style-type: none"> Detects issues only after anomalies occur No prediction of future failures Does not provide automatic corrective actions |