

# Optimization of Energy Efficiency and Execution Time in Cloud Computing

Using MILP, Simulated Annealing, Genetic Algorithms, and Multi-Objective Goal Programming

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August 14, 2024

# Introduction

- ▶ As cloud computing grows, so does energy consumption in data centers, leading to increased operational costs and CO2 emissions.
- ▶ This project explores how optimization techniques can enhance energy efficiency and reduce execution time in cloud environments.

# Problem Statement

**Objective:** To minimize energy consumption and execution time in a cloud computing environment.

**Key Features:**

- ▶ CPU usage
- ▶ Memory usage
- ▶ Network traffic
- ▶ Power consumption
- ▶ Number of executed instructions
- ▶ Execution time
- ▶ Task type, priority, and status

# Mixed Integer Linear Programming (MILP) - Part 1

**Objective:** Minimize the total cost of energy consumption and execution time.

**Decision Variables:**

- ▶  $x_{it}$ : Binary variable indicating whether task  $i$  is scheduled to run at time  $t$ .
- ▶  $r_{ij}$ : Continuous variable representing the amount of resource  $j$  allocated to task  $i$ .

**Parameters:**

- ▶  $R_j^{\max}$ : Maximum available amount of resource  $j$ .
- ▶  $\alpha, \beta$ : Weights assigned to energy consumption and execution time.

# Mixed Integer Linear Programming (MILP) - Part 2

## Objective Function:

Minimize  $Z = \alpha \times \text{Energy Consumption} + \beta \times \text{Execution Time}$

## Subject to:

$$\sum_i r_{ij} \times x_{it} \leq R_j^{\max}, \quad \forall j, t$$

$$\sum_t x_{it} = 1, \quad \forall i$$

$$x_{it} \in \{0, 1\}, \quad r_{ij} \geq 0$$

# Simulated Annealing (SA)

**Objective:** Iteratively minimize energy consumption and execution time by exploring different configurations.

**Parameters:**

- ▶  $S$ : Current solution.
- ▶  $T$ : Temperature parameter controlling acceptance of worse solutions.
- ▶  $\Delta E$ : Change in cost between current and new solution.

**Steps:**

- ▶ Start with an initial solution  $S$  and initial temperature  $T$ .
- ▶ For each iteration, generate a new solution  $S'$  in the neighborhood of  $S$ .
- ▶ Compute the change in cost  $\Delta E = E(S') - E(S)$ .
- ▶ If  $\Delta E < 0$ , accept  $S'$ . Else, accept  $S'$  with probability  $\exp(-\Delta E/T)$ .
- ▶ Decrease the temperature  $T$  and repeat until convergence.

# Genetic Algorithms (GA)

**Objective:** Evolve task allocation strategies to minimize energy consumption and optimize execution time.

**Parameters:**

- ▶  $f(\text{chromosome})$ : Fitness function evaluating energy consumption and execution time.
- ▶  $w_1, w_2$ : Weights for energy consumption and execution time in the fitness function.

**Steps:**

- ▶ Initialize a population of chromosomes (task allocations).
- ▶ Evaluate fitness:  $f(\text{chromosome}) = w_1 \times \text{Energy Consumption} + w_2 \times \text{Execution Time}$ .
- ▶ Selection: Choose the fittest chromosomes.
- ▶ Crossover: Create offspring by combining pairs of chromosomes.
- ▶ Mutation: Introduce small changes to chromosomes.
- ▶ Repeat until convergence or maximum iterations.

# Multi-Objective Goal Programming (MOGP)

**Objective:** Balance multiple goals like minimizing energy consumption and execution time.

**Parameters:**

- ▶  $G_1$ : Goal to minimize energy consumption.
- ▶  $G_2$ : Goal to minimize execution time.
- ▶  $d^+, d^-$ : Deviation variables for each goal.
- ▶  $w_1, w_2$ : Weights for deviations in the objective function.

**Formulation:**

- ▶ Define Goals:  $G_1$  - Minimize Energy,  $G_2$  - Minimize Execution Time.
- ▶ Formulate deviation variables ( $d^+$  and  $d^-$ ) for each goal.
- ▶ Minimize the weighted sum of deviations:  
Minimize  $Z = w_1 \times d_1^+ + w_2 \times d_2^-$ .
- ▶ Subject to: Goal constraints, resource constraints.



# Implementation Strategy

- ▶ Implement the problem using R, with focus on data preprocessing, algorithm application, and evaluation metrics.
- ▶ **Data Preparation:** Extract and normalize key features from the dataset.
- ▶ **Algorithm Application:** Apply MILP, SA, GA, and MOGP to optimize the objectives.

# Expected Outcomes

- ▶ Anticipated results include improved energy efficiency, reduced execution time, and insights into trade-offs between objectives.
- ▶ Evaluation will be based on how well the optimization techniques balance energy consumption and execution time.