SOLVING THE QUALITY-BASED SOFTWARE-SELECTION AND HARDWARE-MAPPING PROBLEM WITH ACO
There are some requested components in each benchmark.

The solution model should contain some assignments:
- one assignment for one requested component.
Each component has some implementations.
For each assignment, one of the implementation of the component, should be selected. We select it randomly.

The solution model should contain some assignments:
- one assignment for one requested component.

There are some requested components in each benchmark.
Each implementation has a resource requirement and some component requirements.
There are some requested components in each benchmark. The solution model should contain some assignments:
- one assignment for one requested component.

For each assignment, one of the implementation of the component, should be selected. We select it randomly.

We select suitable resources according to the ACO algorithm.
THE GOAL:

- The constraint is that:
  - At most one implementation is deployed on each resource. So a resource could not be used in two assignments.

- For each assignment we have to find suitable component and resource mappings such that, all required property clauses of the implementation and all request constraints are fulfilled.

- The final goal is to achieve a valid solution with minimum objective that is constructed in minimum time.
ACO ALGORITHM OVERVIEW

- “Ant Colony Optimization (ACO) studies artificial systems that take inspiration from the behavior of real ant colonies and which are used to solve discrete optimization problems.”
  
  Source: ACO website, [http://iridia.ulb.ac.be/~mdorigo/ACO/about.html](http://iridia.ulb.ac.be/~mdorigo/ACO/about.html)

- Real ants use stigmergy. How?
  - PHEROMONES!!!

- Social behavior of ants is driven by trail pheromone
  - Mark the path for others
  - Deposited while walking between food sources and nest
ACO ALGORITHM OVERVIEW

**Probability of path selection:**

\[ P_{ij} = \frac{(\tau_{ij})^\alpha (\eta_{ij})^\beta}{\sum_j (\tau_{ij})^\alpha (\eta_{ij})^\beta} \]

**Deposit Pheromone and Evaporation:**

\[ \tau_{ij} = \tau_{ij} + \frac{Q}{\text{objective of node } ij} \]

\[ \tau_{ij} = (1 - \rho) \times \tau_{ij} \]

**Heuristic Information:**

\[ \eta_{ij} = \frac{1}{\text{objective of node } ij} \]
ACO Solver Overview

- ACO Solver has two main parts:
  - In part 1 valid software solutions are generated.
  - In part 2, the solutions are passed to a swarm of ants for finding the best valid solutions.

- The population size and iteration size are predefined parameters that determine the termination of parts 1 and 2, respectively.

https://github.com/Ariyanic/TTC18/tree/master/jastadd-mquat-solver-aco
PART 1: GENERATING VALID SOFTWARE SOLUTIONS

Initialize population size;
Initialize \( \tau_0, \alpha, \beta \);
For(int pop=0; i < population size; pop++){
    If time exceeds from maxSolvingTime: break;
    create an empty solution.
    For each requested components, create an empty assignment.
    For each assignment i{
        create a valid component mappings.
        find the possible resources (i.e. the resources that not violated the required property clauses and request constraints) from all of the available resources.
        If an assignment has no possible resources: Ignore the solution.
        Else: set \( \tau_{ij} = \tau_0, \eta_{ij} = \frac{1}{assignment\_objective} \) and \( P_{ij} = \frac{(\tau_{ij})^\alpha(\eta_{ij})^\beta}{\sum_j (\tau_{ij})^\alpha(\eta_{ij})^\beta} \) for each possible resources j
    }
    Save the solutions that all component mappings are valid and all assignments of it has at least one possible resource.
PART 2: FINDING THE BEST VALID SOLUTION

Initialize iteration size;
bestSolution = null;
For(int iteration=0; iteration< iteration size; iteration++){
    If time exceeds from maxSolvingTime: break;
    For each saved solution create an ant, and pass to current solution, possible resources, \( \tau \), \( \eta \) and P to it.
    Run all ants in parallel
    Update \( \tau \) and P to new values
    If (bestSolution is null or antSolution.objective < bestSolution.objective)
        bestSolution = antSolution;
}
Return bestSolution;
RUNNING ANTS

Initialize $\alpha$, $\beta$, $\rho$ and $Q$;

Sort the assignments of the solution based on the number of possible resources

For each assignments $i$ {
    if there is not any possible resource that has not been used by other assignment:
        return null;
    else:
        select a possible resource $j$ using Rolette Wheel mechanism that has not been used by other assignment.
        Assign resource to the assignment $\tau_{ij} = \tau_{ij} + \frac{Q}{assignment.objective}$;
        $\tau_{ij} = (1 - \rho) * \tau_{ij}$; //evaporation $\tau$ for the selected possible resource
        $P_{ij} = \frac{(\tau_{ij})^{\alpha} (\eta_{ij})^{\beta}}{\sum_{j} (\tau_{ij})^{\alpha} (\eta_{ij})^{\beta}}$;
    }

Return the solution;
RESULTS

For larger benchmark ACO Solver meets the time out.

<table>
<thead>
<tr>
<th>Scalability Level</th>
<th>Population Size</th>
<th>Iteration Size</th>
<th>Execution Time (Performance)</th>
<th>Best objective of the valid solution (Quality)</th>
</tr>
</thead>
<tbody>
<tr>
<td>trivial</td>
<td>1</td>
<td>1</td>
<td>8 ms</td>
<td>226823.67</td>
</tr>
<tr>
<td>small</td>
<td>8</td>
<td>1</td>
<td>17 ms</td>
<td>34620.2</td>
</tr>
<tr>
<td>small-many-hw</td>
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<td>5</td>
<td>25 ms</td>
<td>34620.2</td>
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<tr>
<td>small-complex-sw</td>
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<td>1</td>
<td>991 ms</td>
<td>969897.3</td>
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<tr>
<td>medium</td>
<td>50000</td>
<td>100</td>
<td>222961 ms</td>
<td>8658117.329</td>
</tr>
</tbody>
</table>
CONCLUSION

- The ACO algorithm is used to solve the Quality-based Software-Selection and Hardware-Mapping problem.

- The scalability, performance and quality of the proposed solution are tested on different benchmarks.

- The outcome indicates that this approach generates correct results for the evaluated test cases.

- Additionally, better results in terms of performance are given in comparison with the results of the ILP algorithm in the case description.