

# Swarm Intelligence

## Extensions of Ant System

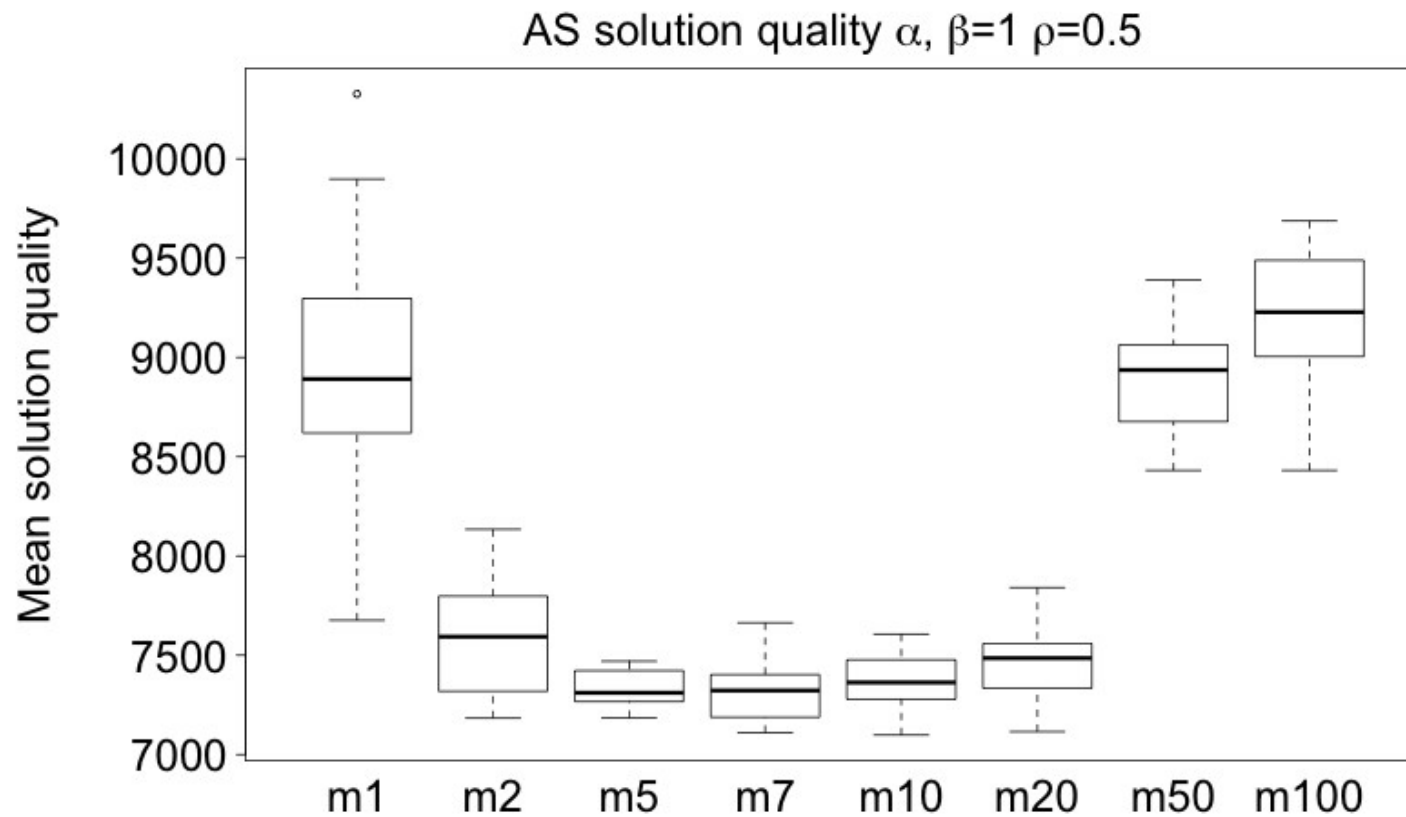
Christian Camacho Villalón  
[christian.camacho.villalon@ulb.ac.be](mailto:christian.camacho.villalon@ulb.ac.be)  
Federico Pagnozzi  
[federico.pagnozzi@ulb.ac.be](mailto:federico.pagnozzi@ulb.ac.be)

IRIDIA – Université Libre de Bruxelles (ULB)  
Bruxelles, Belgium

# Outline

1. Implementation exercise
2. Review of AS
3. MAX-MIN Ant System (MMAS)
4. Ant Colony System
5. Elitist Ant System
6. Rank-based Ant System
7. Best-worst Ant System
8. ACOTSP

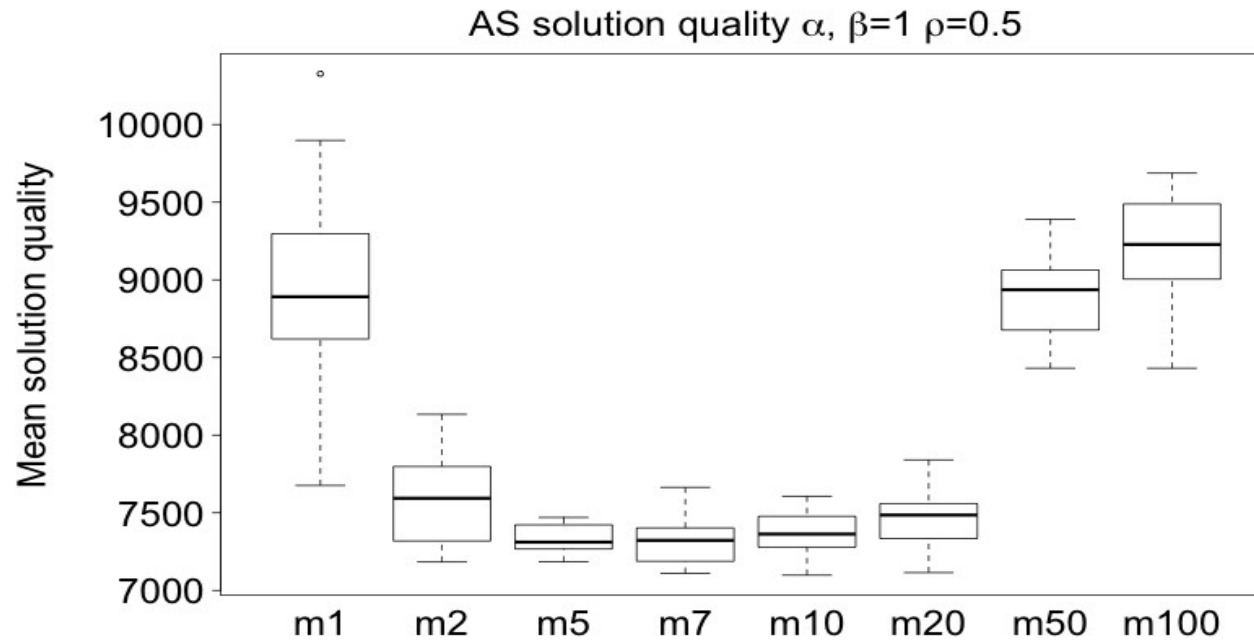
# Implementation exercise 1



# Compare results statistical tests

- Is there a **statistically significant** difference between the solution quality generated by the different algorithms?
- **Null hypothesis**: The statement to be tested.
  - Example: For the Wilcoxon signed-rank test, the null hypothesis is that ‘the median of the differences is zero’
- The **significance level (  $\alpha$  )** determines the maximum allowable probability of incorrectly rejecting the null hypothesis.
- The null hypothesis is rejected if this p-value is smaller than the previously chosen significance level.

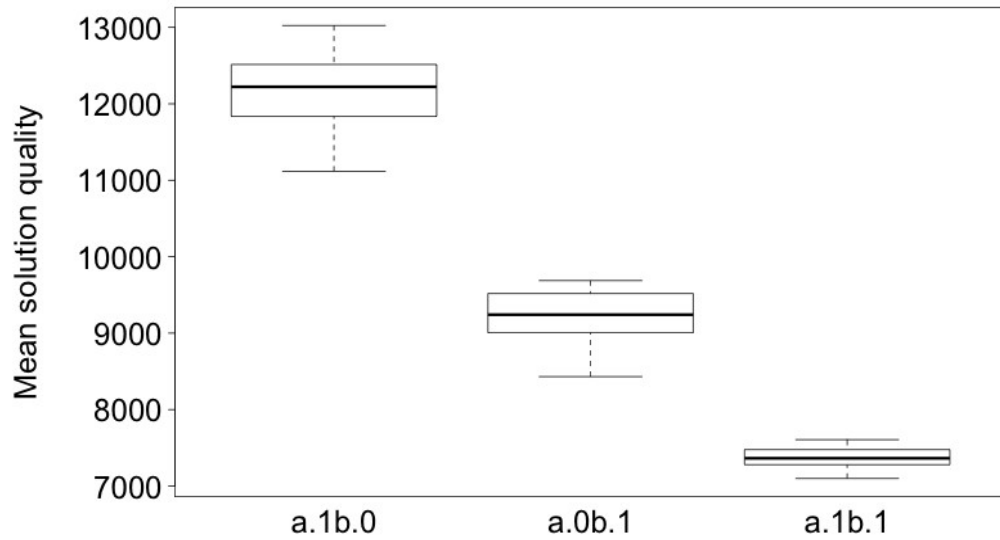
# Implementation exercise 1



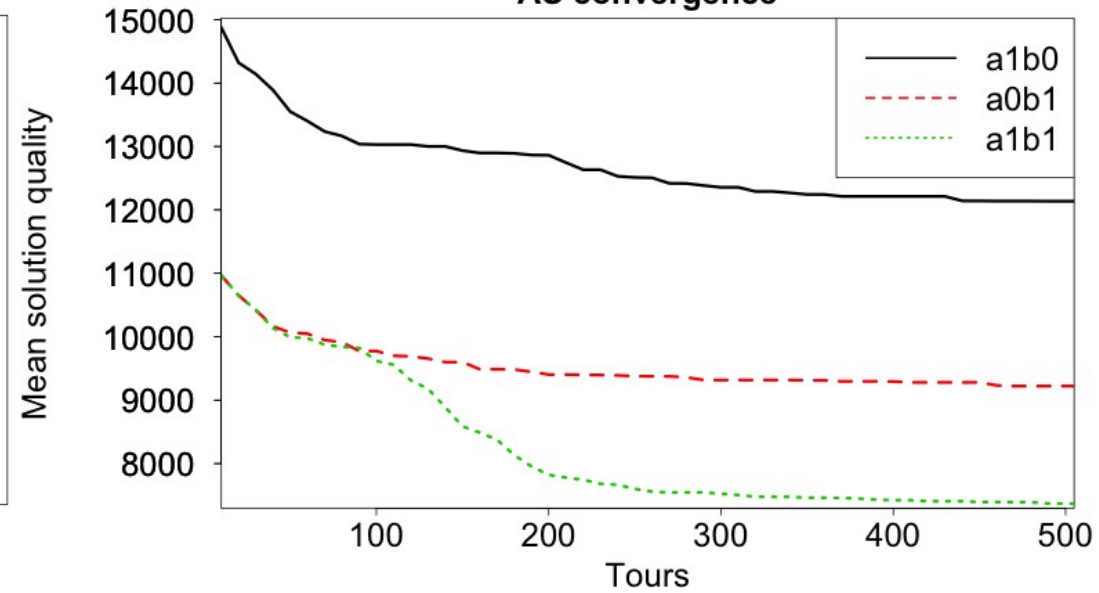
- Wilcoxon test p-values: Corrected (Bonferroni)
  - m1 vs. m2: **1.907e-06** **0.00001**
  - m2 vs. m5: **0.003654** **0.011**
  - m5 vs. m7: **0.6676** **1.000**

# Implementation exercise 1

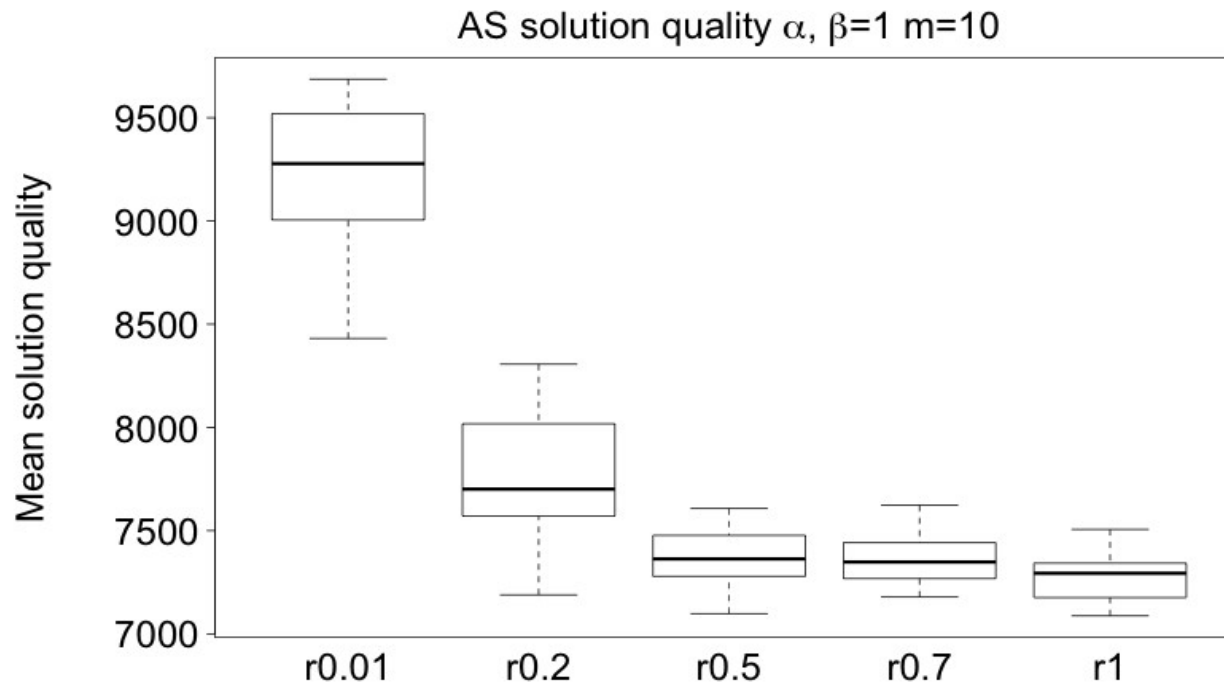
AS solution quality  $m=10$   $\rho=0.5$



AS convergence



# Implementation exercise 1

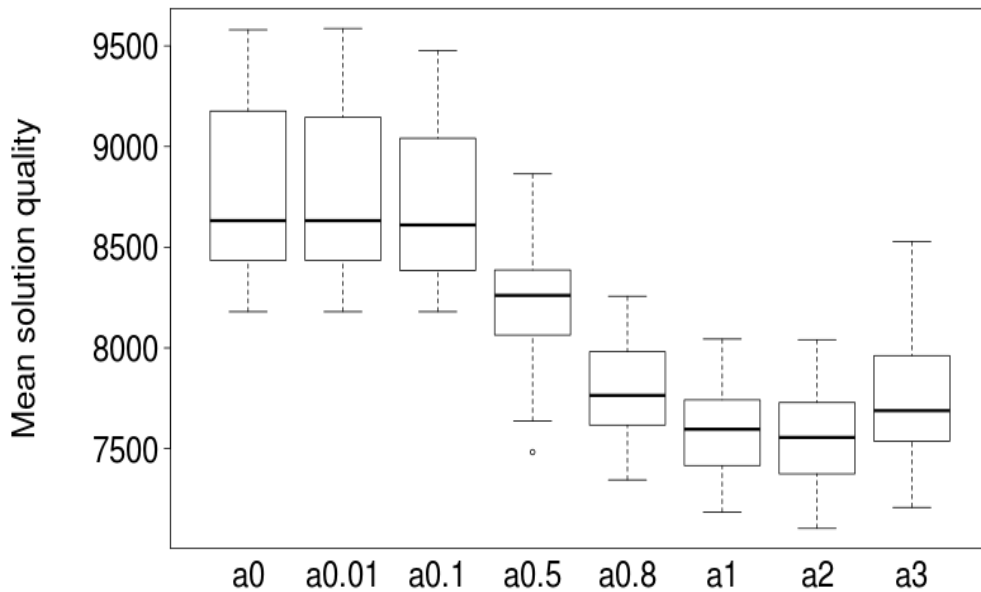


Wilcoxon test p-values				
	0.01	0.2	0.5	1
0.01	-	4.67E-009	4.97E-009	4.67E-009
0.2	-	-	1	1
0.5	-	-	-	1
1	-	-	-	-

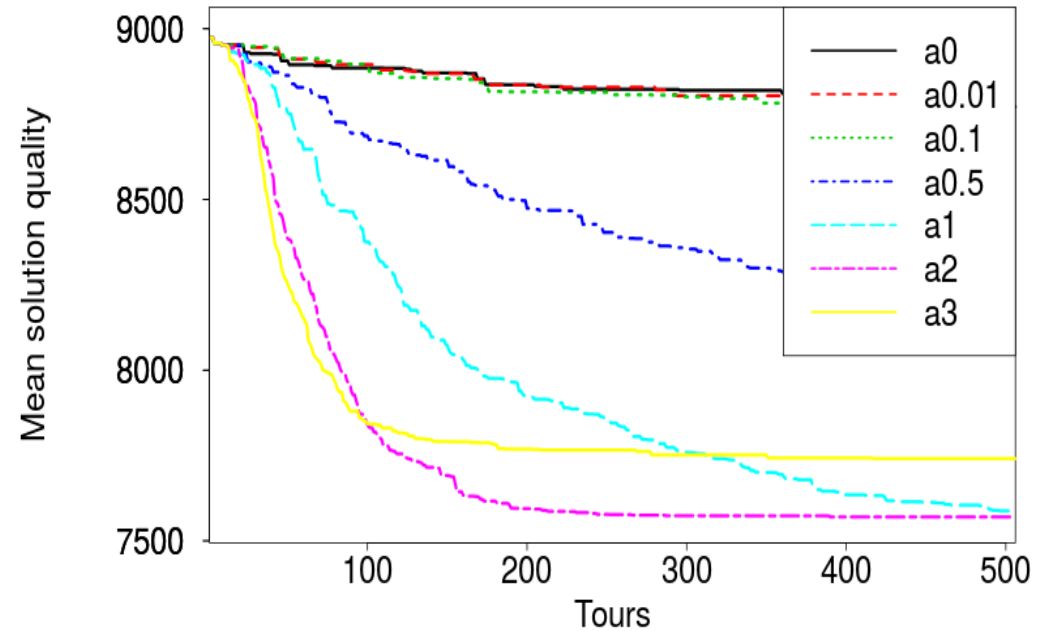
# Implementation exercise 1

- More examples of parameter analysis:

AS solution quality  $\alpha$ ,  $\beta=1$   $\rho=0.2$   $m=10$



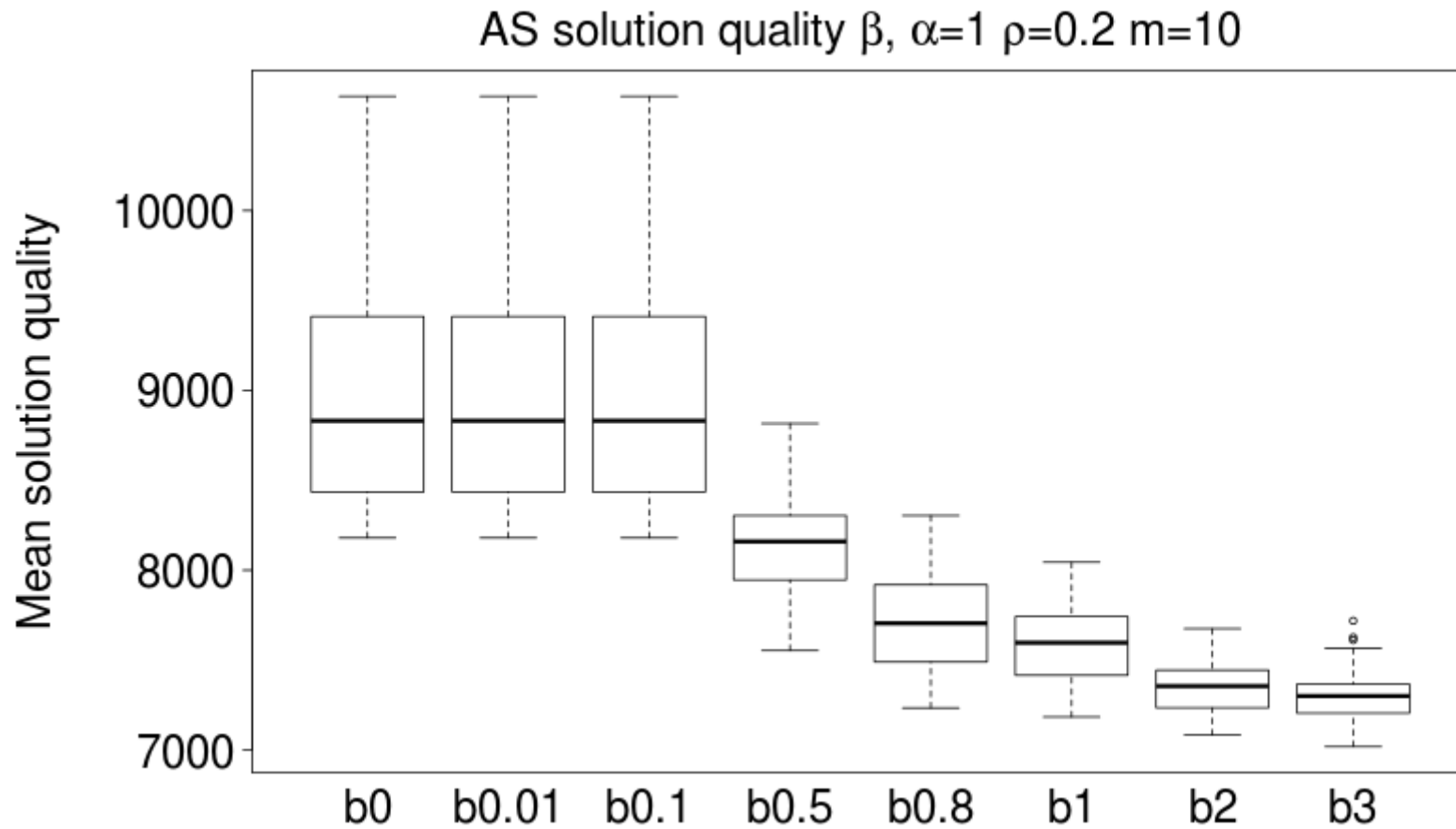
Sample acotsp





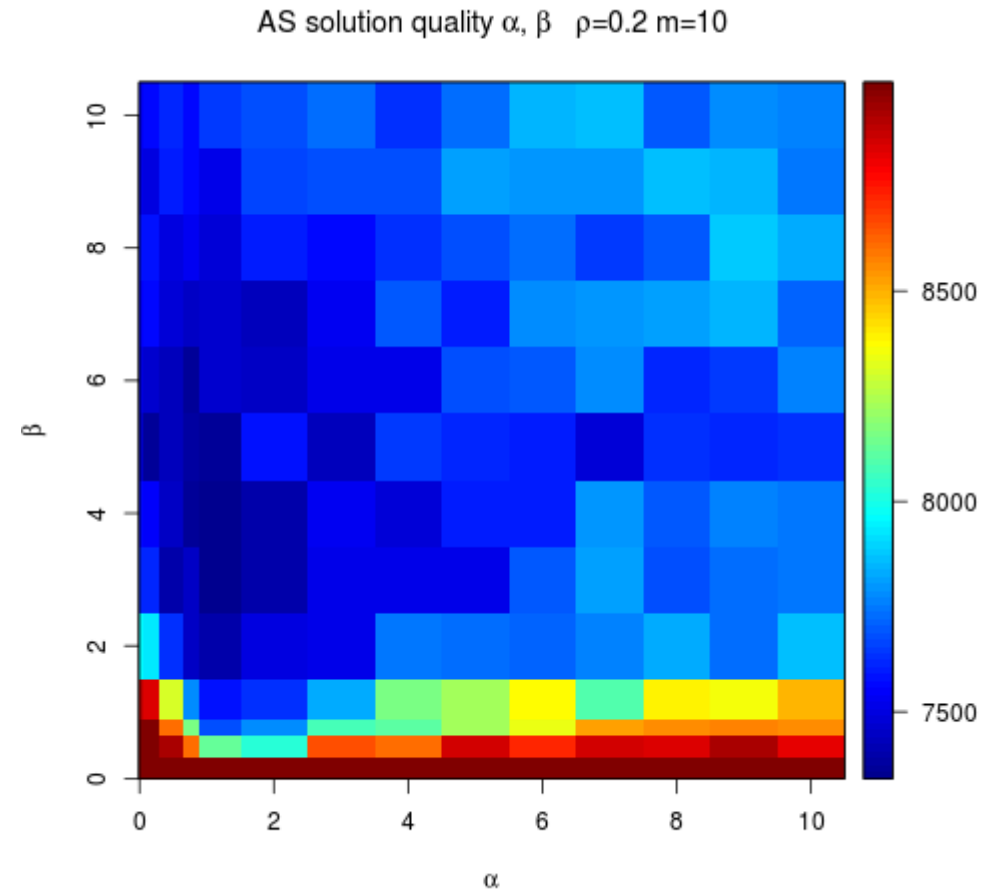
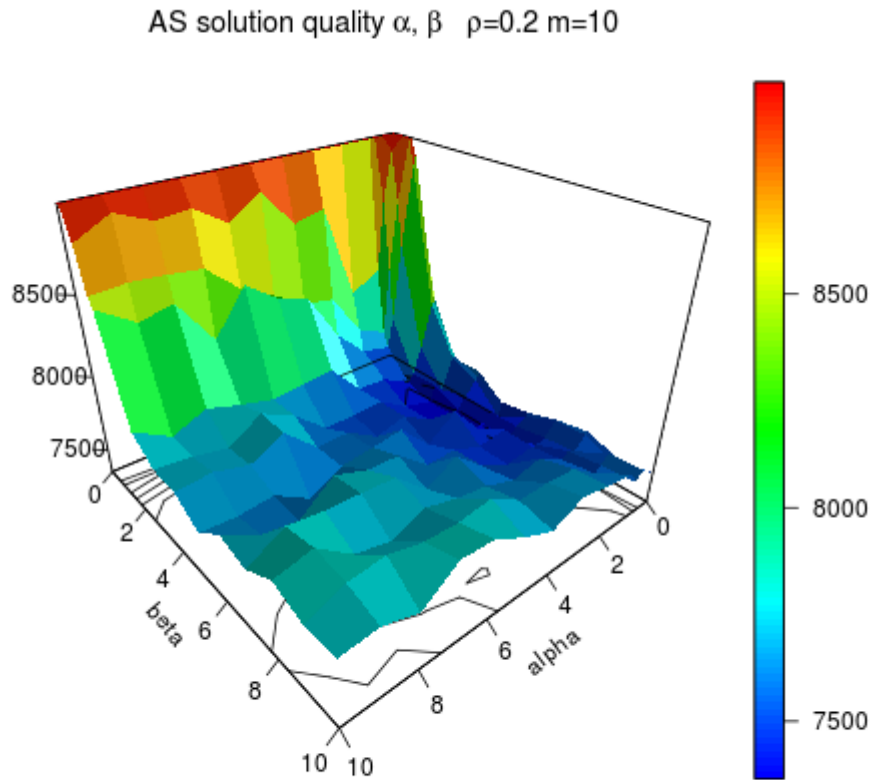
# Implementation exercise 1

- More examples of parameter analysis:



# Implementation exercise 1

- We can also analyse interactions:



# Extensions of AS

- MAX-MIN Ant System (MMAS)
  - Only iteration best or best-so-far ants update pheromone
  - Pheromone trails have explicit upper and lower limits
  - Pheromone trails initialized to upper limit
  - Pheromone trails are re-initialized when stagnated
- Ant Colony System (ACS)
  - Pheromone is updated also while building the solution
  - Only iteration best or best-so-far ants update pheromone

# MAX-MIN Ant System (MMAS)

- Only iteration best or best-so-far ants update pheromone

$$\tau_{ij}(t) = (1 - \rho) \cdot \tau_{ij}(t-1) + \Delta \tau_{ij}^{best}$$

$$\Delta \tau_{ij}^{best} = \frac{1}{L_{best}}, \text{ if } \text{arc}(i, j) \in \text{best tour}$$

- $L_{best}$ : length of the shortest tour found

# MAX-MIN Ant System (MMAS)

- Pheromone trail values are subject to bounds

$$\tau_{min} \leq \tau_{ij} \leq \tau_{max}$$

$$\tau_{max} = \frac{1}{\rho \cdot L^{opt}}$$

$$\tau'_{max} = \frac{1}{\rho \cdot L^{bs}}$$

$$\tau_{min} = \frac{\tau_{max}}{a}$$

$$\tau'_{min} = \frac{\tau_{max}}{2 \cdot n}$$

$$\tau_0 = \infty$$

# MAX-MIN Ant System (MMAS)

- Pheromone trails are re-initialized:
  - When the algorithm converges
  - When no improving solution has been generated for a certain number of consecutive iterations

# Ant Colony System (ACS)

- Three main ideas:
  - Different state transition rule
  - Different global pheromone update rule
  - New local pheromone update rule
- Goal is: better control on exploration/exploitation

# Ant Colony System (ACS)

- State transition (pseudo-random proportional) rule, which is biased towards:

- exploitation with probability  $q_0$

$$j = \operatorname{argmax}_{j \in N_i^k} (\tau_{ij} \cdot \eta_{ij}^\beta) \quad \text{if } q \leq q_0$$

- exploration with probability  $1 - q_0$

$j$  is chosen according to the usual proportional transition rule.



# Ant Colony System (ACS)

- **Local update rule** (to introduce diversification): while building a solution, each ant updates pheromone on visited edges

$$\tau_{ij} = (1 - \rho) \cdot \tau_{ij} + \rho \cdot \tau_0$$

# Ant Colony System (ACS)

- **Global update rule:** pheromone updated **only on edges of the best tour** found so far

$$\tau_{ij} = (1 - \rho) \cdot \tau_{ij} + \rho \cdot \Delta \tau_{ij}^{bs} \quad \forall (i, j) \in T^{bs}$$

$$\Delta \tau_{ij}^{bs} = \frac{1}{L^{bs}}$$

# Ant Colony System for TSP

## Simple pseudo code

```
1  While !termination()  
2    For each ant Do  
3      select random initial starting city  
4      While tour is not complete  
5        select next city using state transition rule  
6        apply local pheromone update rule  
7      EndWhile  
8    EndFor  
9    Apply global pheromone update rule  
10 EndWhile
```

# Elitist Ant System

- Elitism refers to favour best individuals to guide the search. → **intensification**
- After each iteration the **global best ant** deposit pheromone along with the others.
- Introduce a new parameter **e** that controls the contribution of the global best ant to the pheromone update.

$$\tau_{ij}(t) = (1 - \rho) \cdot \tau_{ij}(t-1) + \sum_{k=1}^m \Delta \tau_{ij}^k + e \Delta \tau_{ij}^{bs}$$

# Rank-based Ant System

- A number of the best ants are allowed to update pheromone.
- All the ants are **ranked** regarding their tour quality and the best  $\omega-1$  are selected.
- They deposit pheromone according to their rank. So the best ones contribute more.
- Parameter  $\omega$  controls the **number of ants** allowed to deposit pheromone (usually 25%) and also controls the **amount of pheromone** contributed by each ant.
- The global best ant deposit pheromone with the others.

$$\tau_{ij}(t) = (1 - \rho) \cdot \tau_{ij}(t-1) + \sum_{r=1}^{\omega-1} (\omega - r) \Delta \tau_{ij}^r + \omega \Delta \tau_{ij}^{bs}$$

# Best-worst Ant System

- Transition rule and pheromone evaporation as in Ant System
- Pheromone update after each iteration:
  - The global best ant contributes positively to the pheromone update
  - The worst ant contributes negatively to the pheromone update (additional evaporation)
    - This is only applied in the edges present in the worst ant and absent in the global best ant.
- Pheromone trails **mutation** → diversification
- Restart of the search when stagnation ( $\tau_0$ )

# ACOTSP

- ACOTSP developed by Thomas Stutzle, provides the implementation of a set of ACO algorithms to solve TSP.
- Which algorithms are implemented?
  - Ant System
  - Elitist Ant System
  - Max-min Ant System
  - Rank based Ant System
  - Best-worst Ant System
  - Ant Colony System

# ACOTSP

## Options: Algorithms

- How to specify the algorithm?
  - **--as** : *Ant System*
  - **--eas** : *Elitist Ant System*
  - **--ras** : *Rank-based version of Ant System*
  - **--mmas** : *MAX-MIN ant system*
  - **--bwas** : *Best-worst ant system*
  - **--acs** : *Ant colony system*
- Look for other parameters using **./acotsp --help**
- Related parameters:
  - q0**: prob. of best choice in tour construction (ACS)
  - elitistants**: number of elitist ants (MMAS)
  - rasranks**: number of ranks in rank-based Ant System (RAS)



# ACOTSP

## Options: Other

- Other general parameters
  - tries**: number of independent trials (runs)
  - tours**: number of steps in each trial (max tours evaluated per trial)
  - time**: maximum time for each trial (seconds)
  - seed**: seed for the random number generator
  - optimum**: to stop if tour better or equal optimum is found
  - ants**: number of ants
  - nnants**: nearest neighbours in tour construction
    - To use of candidate list to construct solutions
  - alpha**: alpha (influence of pheromone trails)
  - beta**: beta (influence of heuristic information)
  - rho**: rho (pheromone trail evaporation)
  - localsearch**: 0: no local search 1: 2-opt 2: 2.5-opt 3: 3-opt

# ACOTSP

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# ACOTSP

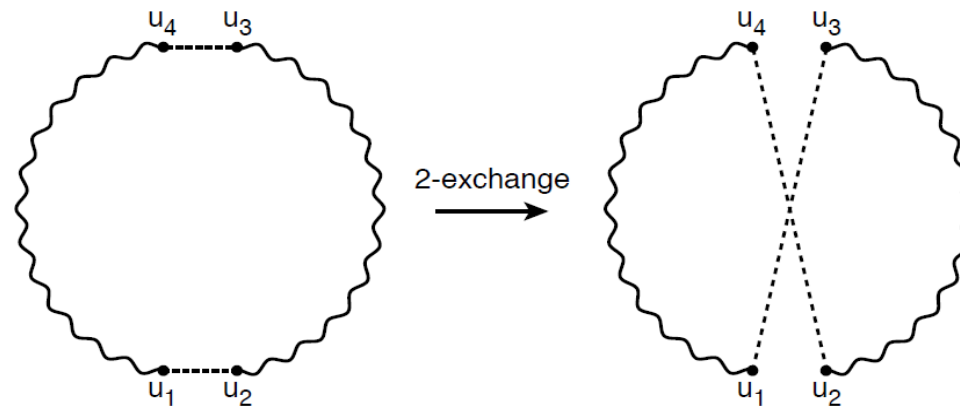
## Options: Local search

- **Local search** starts from a solution already constructed and moves through the search space from one neighbour to other.
- ACOTSP offers the possibility to apply a local search procedure to improve the tours found.
- The options are:
  - 2-opt
  - 2.5-opt
  - 3-opt

# ACOTSP

## Options: Local search

- 2-opt
  - Heuristic: Select two edges and exchange them (2-exchange)
  - Repeat this process for all the edges combinations looking for improvement



- 3-opt follows the same idea using 3 edges, also 2-opt moves are evaluated.
- 2.5-opt: Evaluates the insertion of a node coming from edge (A-B) between the nodes of other edge (C-D). Ex. A-C-B-D