

# Maintenance Scheduling System

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# Agenda

- ▶ Introduction to Maintenance Scheduling
- ▶ Architecture of a Scheduling System
- ▶ Possible Contributions to Operation Research

## Mathematical Notation: Sets

$a \in$

$A_{b,c}^m(t, x, y)$

- ▶ a: set element
- ▶ A: set itself
- ▶ b: set element from set B
- ▶ c: set element from set C
- ▶ m: model formulation m
- ▶ t: time
- ▶ x: value of decision variable from a different model
- ▶ y: value of decision variable from a different model

## Mathematical Notation: Parameters

*name\_of\_parameter*<sub>a,b</sub>(*t*, *x*, *y*)

- ▶ parameters are functions of set elements and input parameters
- ▶ *a*: set element from A
- ▶ *b*: set element from B
- ▶ *t*: time
- ▶ *x*: value of decision variable from another model
- ▶ *y*: value of decision variable from another model

## Mathematical Notation: Variables

$$x_{a,b}^m(t)$$

- ▶ variables are functions of set elements, specified model, and time
- ▶ x: decision variable
- ▶ a: set element from A
- ▶ b: set element from B
- ▶ m: specifying the model
- ▶ t: time
- ▶ Notice: decision variables cannot depend on other decision variables as it would make them belong to the same model.

# Strategic

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**Meta variables:**

$$s \in S \quad (1)$$

$$\alpha(\tau) \quad (2)$$

$$\tau \in [0, \infty] \quad (3)$$

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**Minimize:**

$$\begin{aligned} & + \sum_{o \in O(\tau, \alpha(\tau))} \sum_{d \in D(\tau)} \text{tactical\_value}_{do}(\tau) \cdot \beta_{do}(\tau) \\ & + \sum_{r \in R(\tau)} \sum_{d \in D(\tau)} \text{tactical\_penalty} \cdot \mu_{rd}(\tau) \end{aligned} \quad (4)$$

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**Subject to:**

$$\sum_{o \in O(\tau, \alpha(\tau))} \text{work}_o(\tau) \cdot \beta_{do}(\tau) \leq \Psi_{drt}(\tau) + \mu_{rd}(\tau) \quad \forall d \in D(\tau) \quad \forall r \in R(\tau) \quad (5)$$

$$\sum_{r \in R(\tau)} \Psi_{drt}(\tau) \leq \text{tactical\_resource}_{dr}(\tau) \quad \forall d \in D(\tau) \quad \forall t \in T(\tau) \quad (6)$$

$$\sum_{d \in D(\tau)} \Psi_{drt}(\tau) \leq \text{technician\_skills}_{rt}(\tau) \quad \forall r \in R(\tau) \quad \forall t \in T(\tau) \quad (7)$$

$$\beta_{do}(\tau) \leq \text{number}_o(\tau) \cdot \text{operating\_time}_o \cdot \sigma_{do}(\tau) \quad \forall d \in D(\tau) \quad \forall o \in O(\tau, \alpha(\tau)) \quad (8)$$

$$\sum_{d = \text{latest\_finish}_o(\tau)} \sigma_{do}(\tau) = \text{duration}_o(\tau) \quad \forall o \in O(\tau, \alpha(\tau)) \quad (9)$$

$$\sum_{d^* \in D_{\text{duration}_o(\tau)}(\tau)} \sigma_{d^*o}(\tau) = \text{duration}_o(\tau) \cdot \eta_{do}(\tau) \quad \forall o \in O(\tau, \alpha(\tau)) \quad \forall d \in D(\tau) \quad (10)$$

$$\sum_{o \in O(\tau, \alpha(\tau))} \eta_{do}(\tau) = 1, \quad \forall d \in D(\tau) \quad (11)$$

$$\sum_{d \in D(\tau)} d \cdot \sigma_{do1}(\tau) + \Delta_o(\tau) = \sum_{d \in D(\tau)} d \cdot \sigma_{do2}(\tau) \quad \forall (o1, o2) \in \text{finish\_start}_{o1, o2} \quad (12)$$

$$\sum_{d \in D(\tau)} d \cdot \sigma_{do1}(\tau) = \sum_{d \in D(\tau)} d \cdot \sigma_{do2}(\tau) \quad \forall (o1, o2) \in \text{start\_start}_{o1, o2} \quad (13)$$

$$\beta_{do}(\tau) \in \mathbb{R} \quad \forall d \in D(\tau) \quad \forall o \in O(\tau, \alpha(\tau)) \quad (14)$$

$$\mu_{rd}(\tau) \in \mathbb{R} \quad \forall r \in R(\tau) \quad \forall d \in D(\tau) \quad (15)$$

$$\sigma_{do}(\tau) \in \{0, 1\} \quad \forall d \in D(\tau) \quad \forall o \in O(\tau, \alpha(\tau)) \quad (16)$$

$$\eta_{do}(\tau) \in \{0, 1\} \quad \forall d \in D(\tau) \quad \forall o \in O(\tau, \alpha(\tau)) \quad (17)$$

$$\Delta_o(\tau) \in \{0, 1\} \quad \forall o \in O(\tau, \alpha(\tau))$$

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# Supervisor

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**Meta variables:**

$$z \in Z \quad (18)$$

$$\alpha(\tau) \quad (19)$$

$$\theta(\tau) \quad (20)$$

$$\tau \in [0, \infty] \quad (21)$$

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**Maximize:**

$$\sum_{a \in A(\tau, \alpha(\tau))} \sum_{t \in T(\tau)} \text{supervisor\_value}_{at}(\tau, \lambda_t(\tau), \Lambda_t(\tau)) \cdot \gamma_{at}(\tau) \quad (22)$$

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**Subject to:**

$$\sum_{a \in A_o(\tau, \alpha(\tau))} \rho_a(\tau) = \text{work}_o(\tau) \quad \forall o \in O(\tau, \alpha(\tau)) \quad (23)$$

$$\sum_{t \in T(\tau)} \sum_{a \in A_o(\tau, \alpha(\tau))} \gamma_{at}(\tau) = \phi_o(\tau) \cdot \text{number}_o(\tau) \quad \forall o \in O(\tau, \alpha(\tau)) \quad (24)$$

$$\sum_{o \in O_w(\tau, \alpha(\tau))} \phi_o(\tau) = |O_w(\tau, \alpha(\tau))| \cdot \Phi_w(\tau) \quad \forall w \in W(\tau, \alpha(\tau)) \quad (25)$$

$$\sum_{a \in A_o(\tau, \alpha(\tau))} \gamma_{at}(\tau) \leq 1 \quad \forall o \in O(\tau, \alpha(\tau)) \quad \forall t \in T(\tau) \quad (26)$$

$$\gamma_{at}(\tau) \leq \text{feasible}_{at}(\theta(\tau)) \quad \forall a \in A_o(\tau, \beta) \quad (\tau) \forall o \in O(\tau, \alpha(\tau)) \quad \forall t \in T(\tau) \quad (27)$$

$$\gamma_{at}(\tau) \in \{0, 1\} \quad \forall o \in O(\tau, \alpha(\tau)) \quad \forall t \in T(\tau) \quad (28)$$

$$\phi_o(\tau) \in \{0, 1\} \quad \forall o \in O(\tau, \alpha(\tau)) \quad (29)$$

$$\Phi_w(\tau) \in \{0, 1\} \quad \forall w \in W(\tau, \alpha(\tau)) \quad (30)$$

$$\rho_a(\tau) \in [\text{lower\_activity\_work}_a(\tau), \text{work}_a(\tau)] \quad \forall a \in A(\tau, \alpha(\tau)) \quad (31)$$

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# Operational

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**Meta variables:**

$$t \in \mathcal{T}(\tau) \quad (32)$$

$$\alpha(\tau) \quad (33)$$

$$\gamma(\tau) \quad (34)$$

$$\tau \in [0, \infty] \quad (35)$$

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**Maximize:**

$$\sum_{a \in A(\tau, \gamma_t(\tau))} \sum_{k \in K(\gamma(\tau))} \delta_{ak}(\tau) \quad (36)$$

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**Subject to:**

$$\sum_{k \in K(\gamma(\tau))} \delta_{ak}(\tau) \cdot \pi_{ak}(\tau) = \text{activity\_work}_a(\tau, \rho(\tau)) \cdot \theta_a(\tau) \quad \forall a \in A(\tau, \gamma_t(\tau)) \quad (37)$$

$$\lambda_{a21}(\tau) \geq \Lambda_{a1, \text{last}(a1)}(\tau) + \text{preparation}_{a1, a2} \quad \forall a1 \in A(\tau, \gamma_t(\tau)) \quad \forall a2 \in A(\tau, \gamma_t(\tau)) \quad (38)$$

$$\lambda_{ak}(\tau) \geq \Lambda_{ak-1}(\tau) - \text{constraint\_limit} \cdot (2 - \pi_{ak}(\tau) + \pi_{ak-1}(\tau)) \\ \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad (39)$$

$$\delta_{ak}(\tau) = \Lambda_{ak}(\tau) - \lambda_{ak}(\tau) \quad \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad (40)$$

$$\lambda_{ak}(\tau) \geq \text{event}_{ie} + \text{duration}_{ie} - \text{constraint\_limit} \cdot (1 - \omega_{akie}(\tau)) \\ \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad \forall i \in I(\tau) \quad \forall e \in E(\tau) \quad (41)$$

$$\Lambda_{ak}(\tau) \leq \text{event}_{ie} + \text{constraint\_limit} \cdot \omega_{akie}(\tau) \\ \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad \forall i \in I(\tau) \quad \forall e \in E(\tau) \quad (42)$$

$$\lambda_{a1}(\tau) \geq \text{time\_window\_start}_a(\beta(\tau)) \quad \forall a \in A(\tau, \gamma_t(\tau)) \quad (43)$$

$$\Lambda_{a, \text{last}(a)}(\tau) \leq \text{time\_window\_finish}_a(\beta(\tau)) \quad \forall a \in A(\tau, \gamma_t(\tau)) \quad (44)$$

$$\pi_{ak}(\tau) \in \{0, 1\} \quad \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad (45)$$

$$\lambda_{ak}(\tau) \in [\text{availability\_start}(\tau), \text{availability\_finish}(\tau)] \\ \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad (46)$$

$$\Lambda_{ak}(\tau) \in [\text{availability\_start}(\tau), \text{availability\_finish}(\tau)] \\ \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad (47)$$

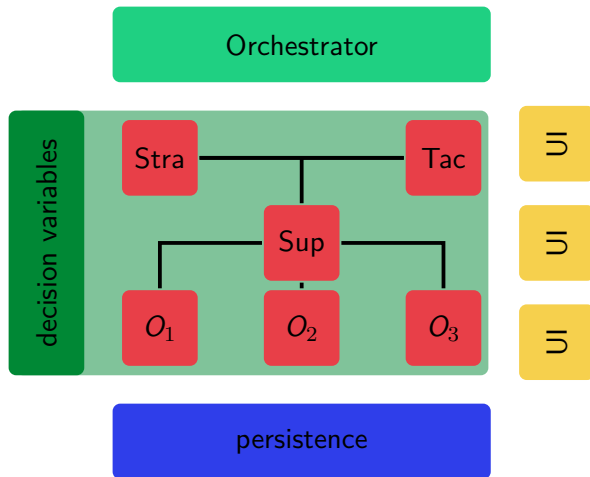
$$\delta_{ak}(\tau) \in [0, \text{work}_{a\_to\_o(a)}(\tau)] \quad \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad (48)$$

$$\omega_{akie}(\tau) \in \{0, 1\} \quad \forall a \in A(\tau, \gamma_t(\tau)) \quad \forall k \in K(\gamma(\tau)) \quad \forall i \in I(\tau) \quad \forall e \in E(\tau) \quad (49)$$

$$\theta_a(\tau) \in \{0, 1\} \quad \forall a \in A(\tau, \gamma_t(\tau)) \quad (50)$$

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# Architecture



Managing metahe

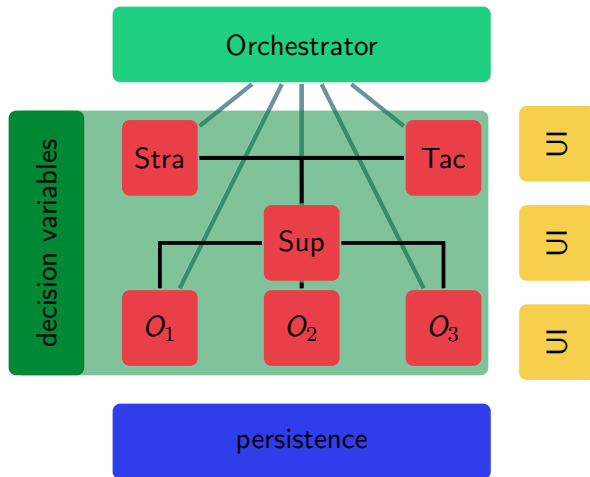
Solution sharing (

Actor-based Optim

Data storage (Me

Message passing (

# Architecture



Managing metahe

Solution sharing (

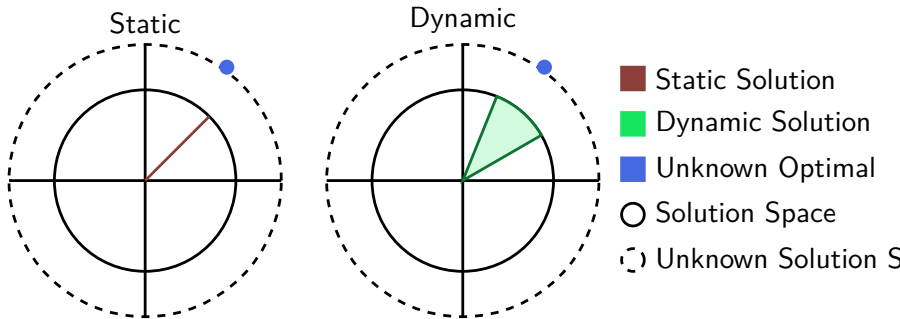
Actor-based Optim

Data storage (Me

Message passing (

1. Reactive Versus Static Constraints
2. Dynamic
3. Business
4. Technical
5. Academic

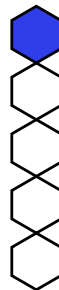
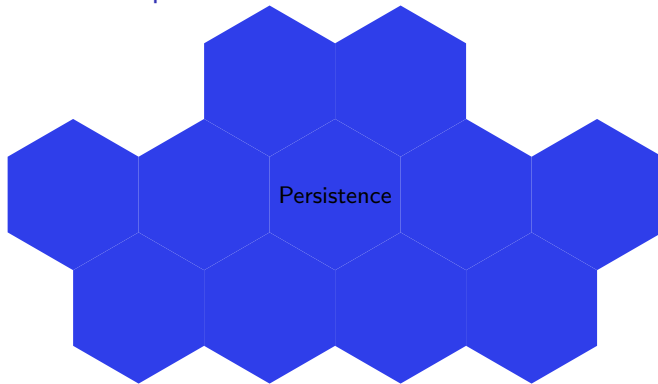
## Dynamic versus Static Models



- ▶ Mathematical models guide direction but does not provide direct solutions.
- ▶ Static solutions are rarely fully executable.
- ▶ Dynamic models are less constrained and ensure a contained optimal solution.
- ▶ Remember: The real optimal solution is ever knowable at time  $= 0$



## Model setup



Persistence

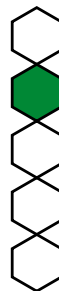
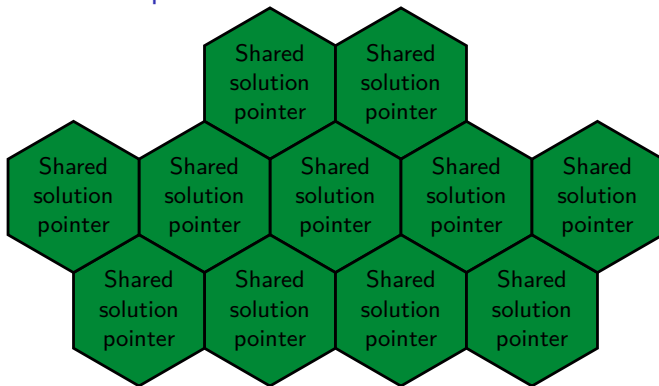
Atomic Point

Metaheuristi

Orchestratio

User interfac

## Model setup



Persistence

Atomic Point

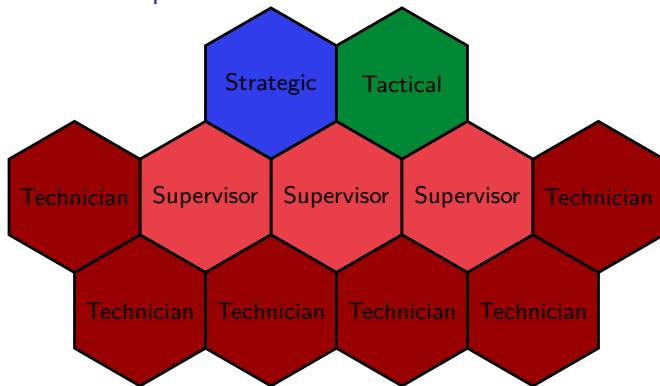
Metaheuristic

Orchestration

User interface



## Model setup



Persistence

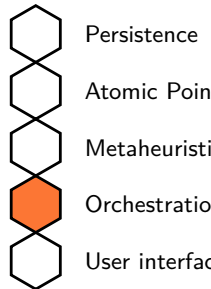
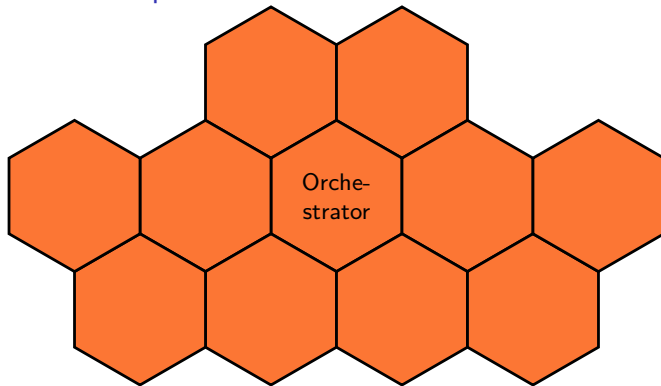
Atomic Point

Metaheuristi

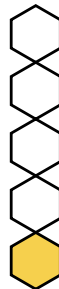
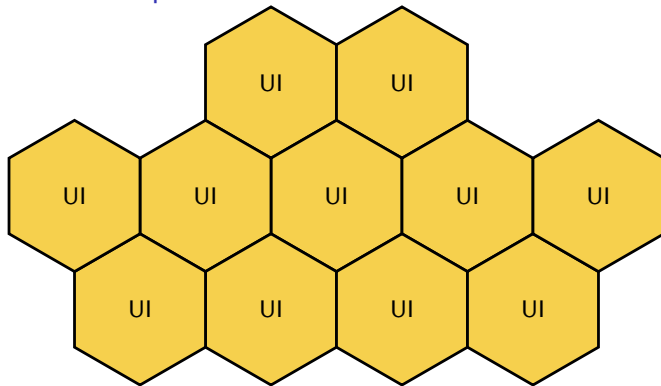
Orchestratio

User interfacer

## Model setup



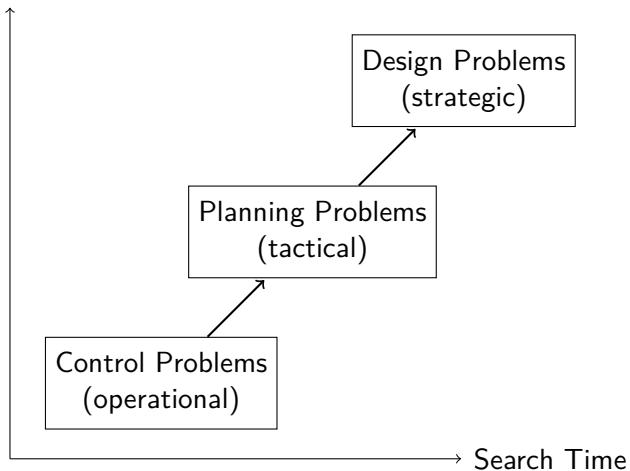
## Model setup



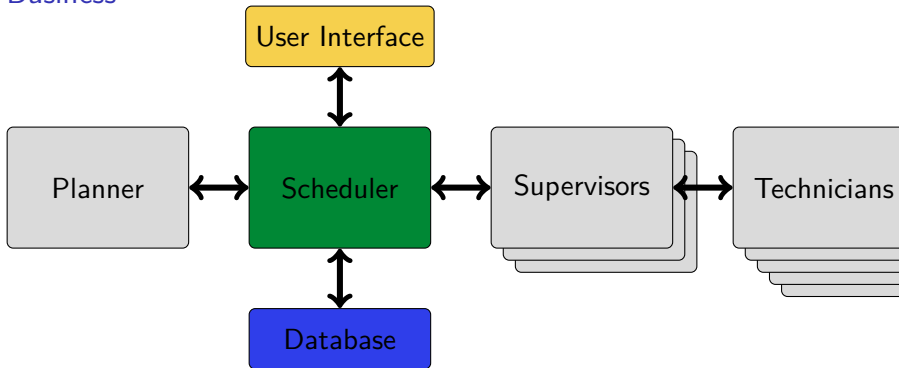
- Persistence
- Atomic Point
- Metaheuristi
- Orchestratio
- User interface

# The Story of Zaabalawi

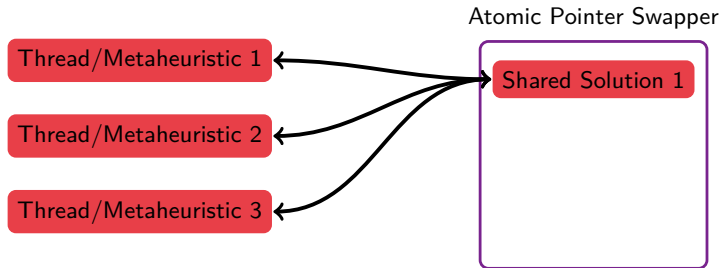
Quality of Solutions



## Business

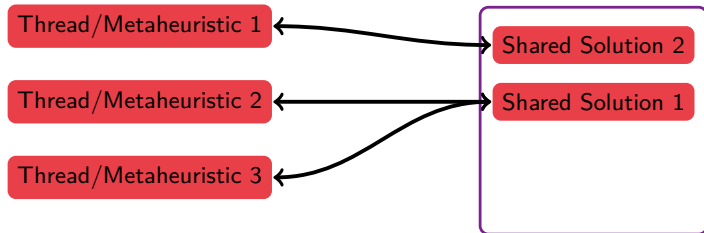


## Atomic Pointer Swapping



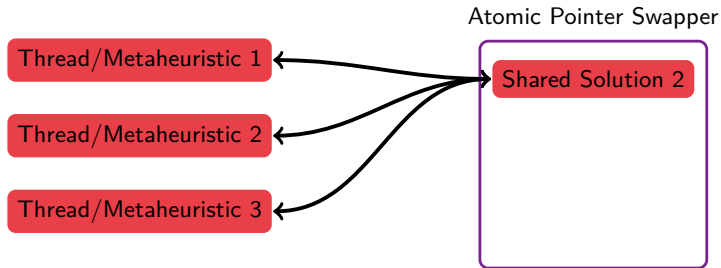
Thread one finds a better solution and swaps it in

Atomic Pointer Swapper

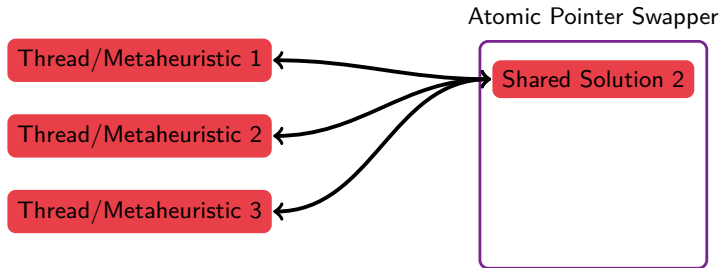




Thread two and three loads the new Shared Solution at the top of their optimization loop



Shared Solution 1 is dropped from memory when is it no longer referenced by any threads



Include the many different ways that metaheuristics can be understood.

- ▶ Chinese literature.

## Decision Problems versus Process Problems.

- ▶ Chinese literature.

## Ordinator Architecture

