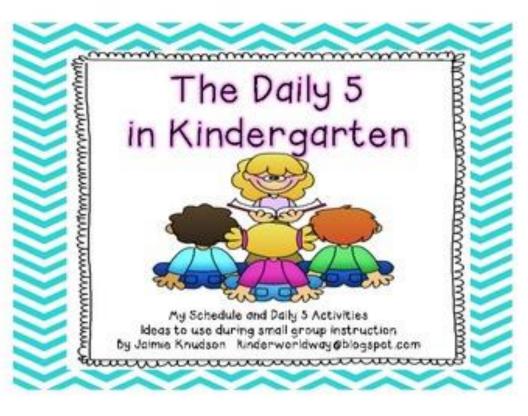
Problem 1

Stojanova A, Kocaleva M, Bikov D, K Lazarova L, Stojkovikj N, Miteva M, Zlatanovska B, Zdravevska S, Ljubenovska M

Problem description

- Kindergartens and day care centers face with a problem when the employees should be organized throughout the working day, or only partially included in a certain period of the day.
- ▶ It is necessary to make a schedule of employees in the kindergartens or in a day care center for children.
- Our aim
 - make an optimization model to deal with the large number of children and employees.
- This problem can be extended to problem for healthcare services for elderly people.





Problem description

















Bathing

Toileting



Dressing



Walking or moving around











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General model proposed by Marta Ferreira (doctoral thesis-2013)

- The general model was developed for the staff scheduling problem of an organization that works continuously, 24 hours a day.
- The day is divided in n_s working shifts. The model considers a set of n_T teams of homogeneous (single skilled and full-time) employees, that must be assigned to either a work or a break shift, in each of the n_p planning period days. Daily shift demand levels must be satisfied, meaning that the model must guarantee a required number of teams working in each shift on each day.
- ▶ $d \in \{1, ..., n_D\}$, day;
- ightharpoonup $t \in \{1, \ldots, n_T\}$, team;
- ▶ $s \in \{1, ..., n_s\}$, working shift;
- ▶ $s' \in \{1, ..., 2 \times n_s\}$, extended shift.

This model can be used for healthcare services to elderly people.

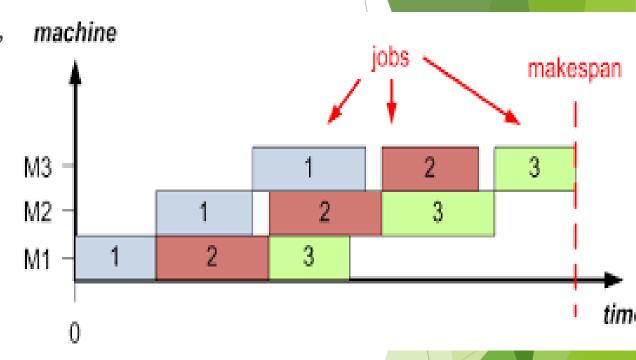
Mathematical models for scheduling problem

- Base Model (FSMP)
- Single machine model
- Parallel machine model
- Flow shop model
- Job shop model
- Pure job shop model

Proposed models

- ► Job shop scheduling model
 - ▶ Optimization of time needed for completing all tasks
 - Agent based approach with AnyLogic
- ► Genetic algorithm
 - ▶ Used alone or in combination with JSS

- This model can be described by a set of n jobs J_j , where $1 \le j \le n$, and each job has to be processed on a set of m machines M_r , where $1 \le r \le m$.
- ► Each job has a sequence of machines that must be processed.
- The processing of job J_j on machine M_r is called the operation O_{jr} . Operation O_{jr} requires the exclusive use of M_r for an uninterrupted duration p_{jr} , where p_{jr} is its processing time.
- A schedule is a set of completion times for each operation MS_{jr}, where 1≤j≤n and 1≤r≤m that satisfies those constraints.
- The time required to complete all the jobs is called the makespan MS.
- The scheduling objective is makespan minimization, which means to minimize the completion time of the last operation of any job.



In our problem we have two sets

- \triangleright E = { E1, E_m} where E are employees (machine of the model)
- \vdash K = { K1, K_m} where K are kids or elderlies (jobs of the model)
- ► The aim is to minimize the time to perform all tasks (all needs of the kids or all needs of elderlies in a shortest possible time)
- For the assignment problem of kids or elderlies to employs we use MIP (mixed integer programming) model.
- ▶ The MIP formulation is often used to model the classical deterministic JSSP

- MIP model
 - **Parameters:**
- r_{ilk} with value 1, if elderly (kid) i requires task l from employ k, and 0 otherwise.
- p_{ik} is servicing time in which an elderly (kid) i has to be serviced from employ k.
 - Decision variables:
- \triangleright s_{ik} is start time of servicing an elderly (kid) i by employ k.
- \mathbf{y}_{ijk} has a value 1 when an elderly j precedes elderly i for caregiver k.

MIP model

$$\sum_{k=1}^{m} r_{imk} (s_{ik} + p_{ik}) \le MS \qquad i = 1, 2, \dots, n$$
 (1)

$$\sum_{k=1}^{m} r_{ilk} (s_{ik} + p_{ik}) - \sum_{k=1}^{m} r_{i,l+1,k} s_{ik} \le 0,$$

$$i=1,2,\ldots,m;$$
 (2)

$$l = 1, 2, \dots, m - 1;$$

$$K(1-y_{ijk})+s_{jk}-s_{ik} \ge p_{ik}, \tag{3}$$

$$k = 1, 2, \dots m; 1 \le i < j \le n$$

$$Ky_{ijk} + s_{ik} - s_{jk} \ge p_{jk}, \quad k = 1, 2, \dots m; 1 \le i < j \le n$$
 (4)

Constraint 1 gives the lower bound for the function MS.

Constraint 2 ensures that the starting time of servicing an elderly i with task l+1 is not earlier than its finish time in its predecessor, task l.

Constraints 3 and 4 ensure that only one elderly is served from employer at any given time. The parameter K is a large number, sometimes taken as the sum of all processing times.

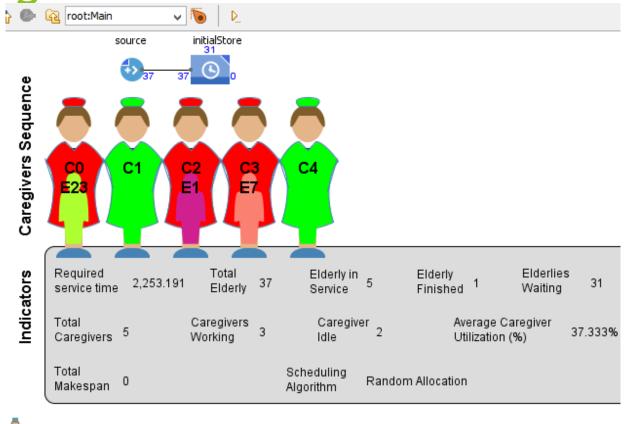
Agent based approach of the problem

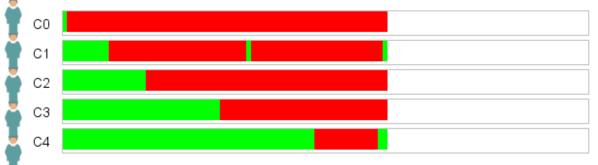
- Agent-based models (ABMs) consist of a set of elements (agents) characterized by some attributes, which interact each other through the definition of appropriate rules in a given environment.
- Our agent based approach of the problem is consist of two types of agents.
 - kids (elderly people or patients)
 - employees (caregivers).

They communicate among each other in a manner that employees can provide different services to elderlies or kids.

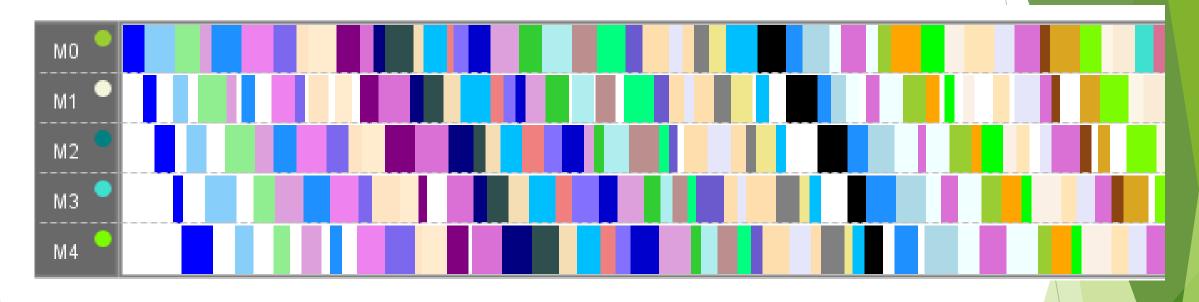
This approach is applied in combination with discrete event simulations

AnyLogic Simulation





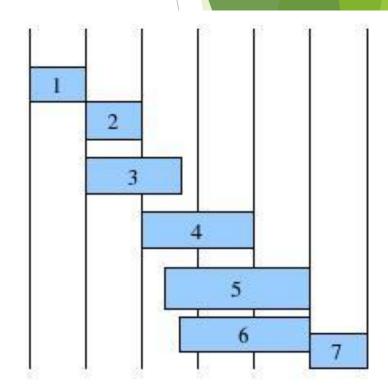
AnyLogic Simulation



| | J0 | J1 | J2 | J3 | J4 | J5 | J6 | J7 | J8 | J9 | J10 | J11 | J12 | J13 | J14 | J15 | J16 | J17 |
|----|--------|--------|--------|--------|--------|--------|--------|--------|-----------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| M0 | 19.098 | 16.558 | 12.847 | 7.771 | 8.264 | 14.787 | 19.985 | 19.436 | 8.07 | 14.562 | 19.807 | 14.257 | 17.745 | 9.069 | 16.457 | 8.697 | 10.537 | 14.66 |
| M1 | 19.208 | 14.898 | 16.16 | 5.16 | 11.48 | 19.765 | 14.456 | 8.142 | 8.824 | 5.131 | 7.299 | 5.422 | 10.307 | 15.947 | 11.416 | 8.727 | 11.958 | 6.657 |
| M2 | 19.056 | 7.351 | 7.13 | 7.416 | 8.497 | 8.101 | 18.644 | 7.59 | 16.669 | 14.468 | 15.659 | 13.872 | 13.979 | 18.301 | 19.988 | 15.603 | 7.659 | 16.619 |
| M3 | 10.958 | 10.673 | 12.226 | 7.671 | 18.349 | 10.62 | 12.615 | 13.234 | 8.375 | 8.12 | 17.36 | 6.095 | 9.381 | 5.656 | 5.929 | 10.066 | 15.142 | 9.597 |
| M4 | 10.213 | 7.096 | 13.168 | 13.106 | 5.575 | 11.95 | 12.372 | 13.332 | 19.747 | 18.205 | 6.855 | 6.111 | 18.147 | 11.866 | 13.795 | 14.11 | 15.734 | 18.209 |

Approach with Genetic algorithm

- Genetic algorithm can be used in combination with JSS
- To apply a genetic algorithm to a scheduling problem we must first represent it as a chromosome.
 - Implements each schedule as a chromosome/individual in a population of schedules.
 - ► Each schedule is evaluated with a fitness function
 - Schedules with greater fitness function values are allowed to "mate" with other schedules via crossover.
 - Mutation provides for diversity in the population.
 - ► The crossover and mutation operations generate new populations of schedules.
 - New generations are created until a schedule is formed that is deemed acceptable.
 - ▶ Generally speaking, a schedule is deemed acceptable if its fitness function value is high enough.



Future work

- ► Try to find application of General model and adapt it for our problem.
- To optimize our proposed approach and to adapt to other constraints needed.
- ► To extend proposed model with genetic algorithms.