## Software Project Scheduling Problem

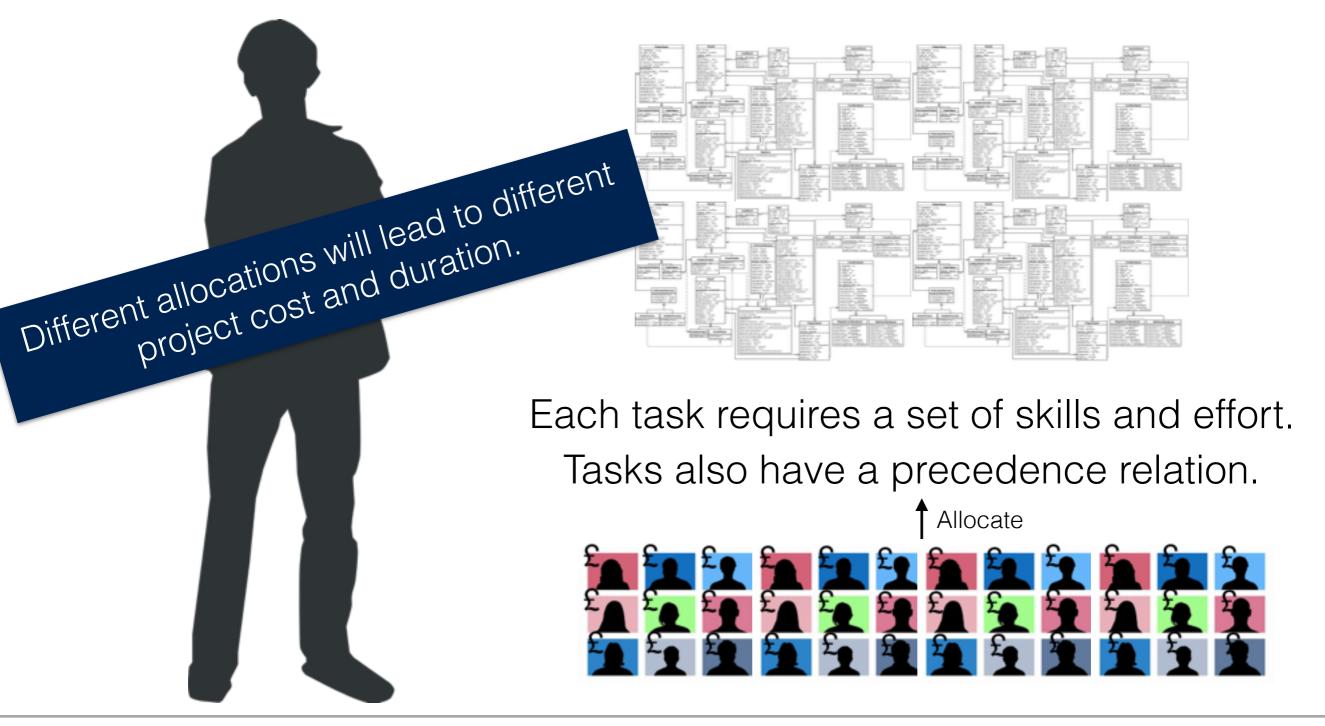
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Nature-inspired Optimisation Lecture

### Outline

- What is the Software Project Scheduling Problem (SPSP)?
- Why are automated optimisation methods important for the SPSP?
- How to solve the SPSP?

### Software Project Scheduling Problem (SPSP)



### Software Project Scheduling Problem (SPSP)

SPSP: find a good allocation of employees to tasks in a software project so as to minimise its **cost** and **completion time**.

It is very difficult to optimally assign employees to tasks manually.

 The space of possible allocations can be enormous.

We can use optimisation algorithms (e.g., EAs) to solve the SPSP!

## Advantages of Optimisation Algorithms for the SPSP

- Insight into how to optimise objectives -- they may find solutions that no human has thought of.
- Speed up the task of allocating employees to tasks.
- Help software manager to find solutions that satisfy all constraints.
  - Team must have skills to perform a task.
  - No overwork is allowed [video].

## Formulating the SPSP

#### Setting: assume we are given

- *n* employees *e*<sub>1</sub>, . . . , *e*<sub>n</sub> with salaries and sets of skills;
- m tasks  $t_1, \ldots, t_m$  with efforts and sets of required skills;
- a task precedence graph (TPG).

#### Problem: allocate employees to tasks so as to:

- minimise cost (total salaries paid) and
- minimise duration (completion time).

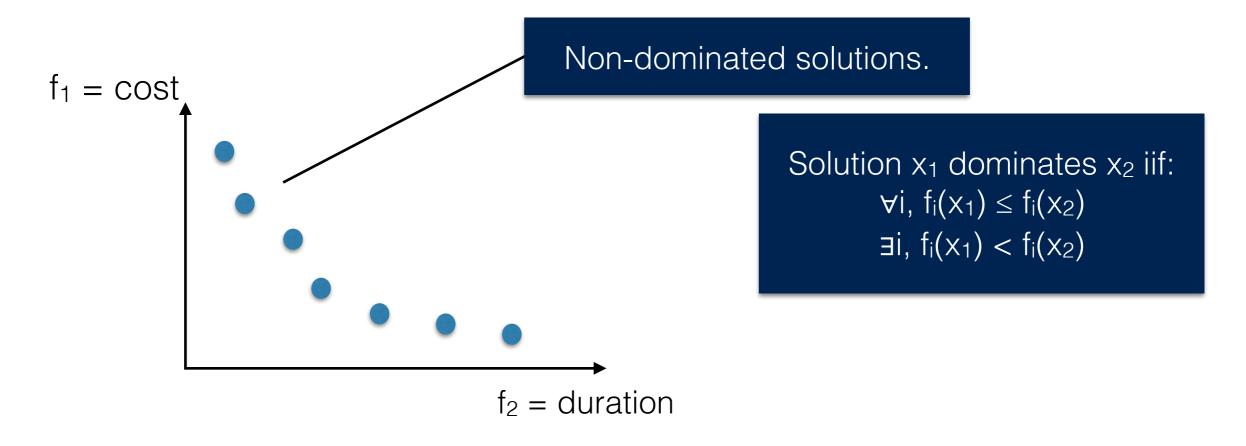
#### Constraints:

- team must have required skills and
- no overwork.

## Solving the SPSP

What type of evolutionary algorithm would be adequate?

 Multi-objective Evolutionary Algorithms, e.g., NSGA-II.



#### NSGA-II

- ullet Step 1: For a population of size M, create a group of M offspring using the desired crossover and mutation operators
- ullet Step 2: The offspring and their parent solutions are combined into a group of size 2M
- Step 3: Selecting the fittest M individuals from this group as parents for the next generation by:
  - Step 3.1: Nondominated Sorting: similar to NSGA, e.g., to identify all non-dominated fronts and sort them
  - Step 3.2: Crowding distance sorting: removes the "most crowded" individuals, e.g., those individual with small D<sub>i</sub> from this final front, in order to make it fit into the group of M parents.

[Video: <a href="https://youtu.be/sEEiGM9em8s">https://youtu.be/sEEiGM9em8s</a>]

<sup>\*</sup> From Lecture 15.

# Designing an Evolutionary Algorithm

- Representation / encoding;
- mutation and crossover;
- fitness / objectives evaluation;
- how to deal with constraints.

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### Representation

Dedication: percentage of time an employee spends on a task, respecting a certain granularity *k*.

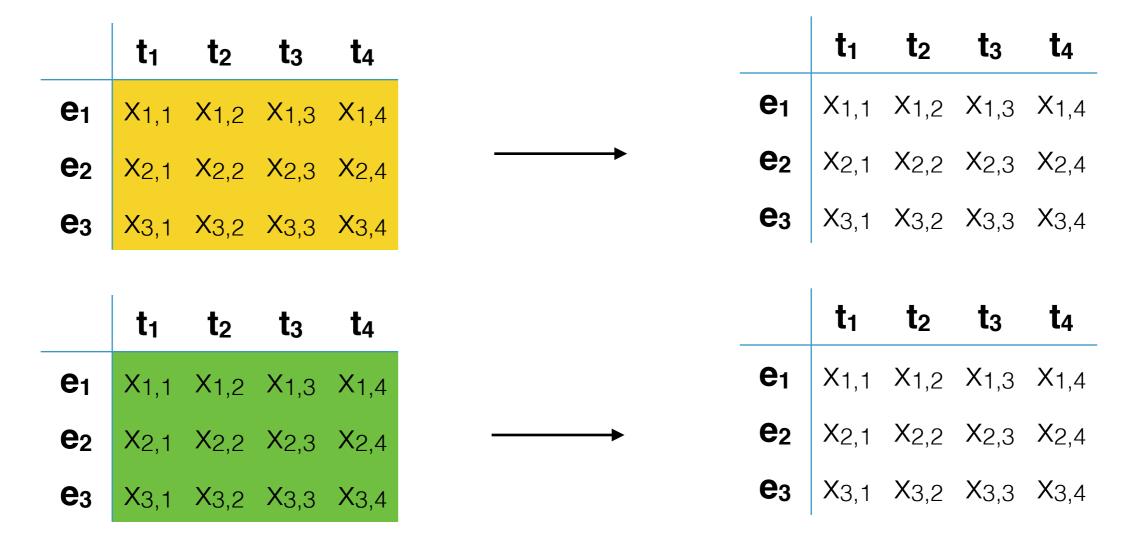
	t <sub>1</sub>	t <sub>2</sub>	 t <sub>m</sub>
e <sub>1</sub>	X1,1	X1,2	 X1,m
<b>e</b> <sub>2</sub>	X2,1	X2,2	 X1,m X2,m  Xn,m
•••			 •••
<b>e</b> n	X <sub>n,1</sub>	X <sub>n,2</sub>	 X <sub>n,m</sub>

Employees can divide their attention among tasks.

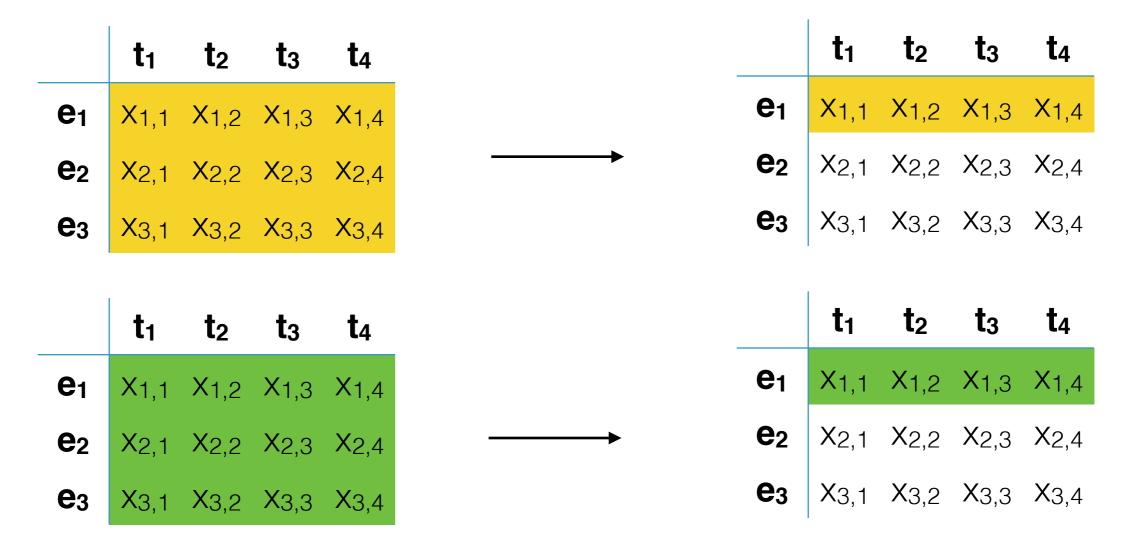
# Designing an Evolutionary Algorithm

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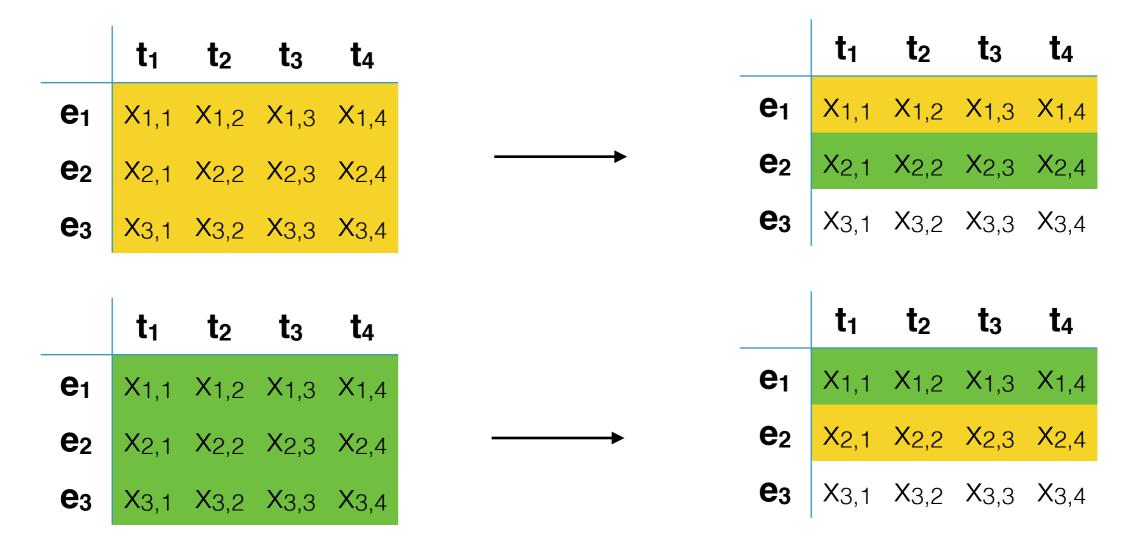
- Mutation of  $x_{i,j}$  picks a new dedication uniformly at random from  $\{0/k, 1/k, ..., k/k\} \setminus x_{i,j}$ .
- Crossover: exchange rows



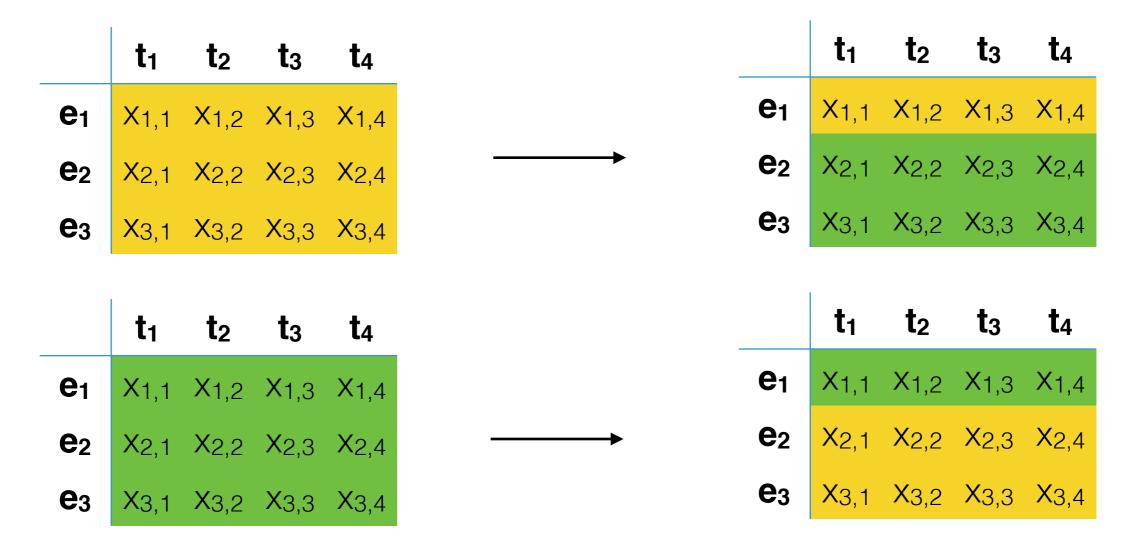
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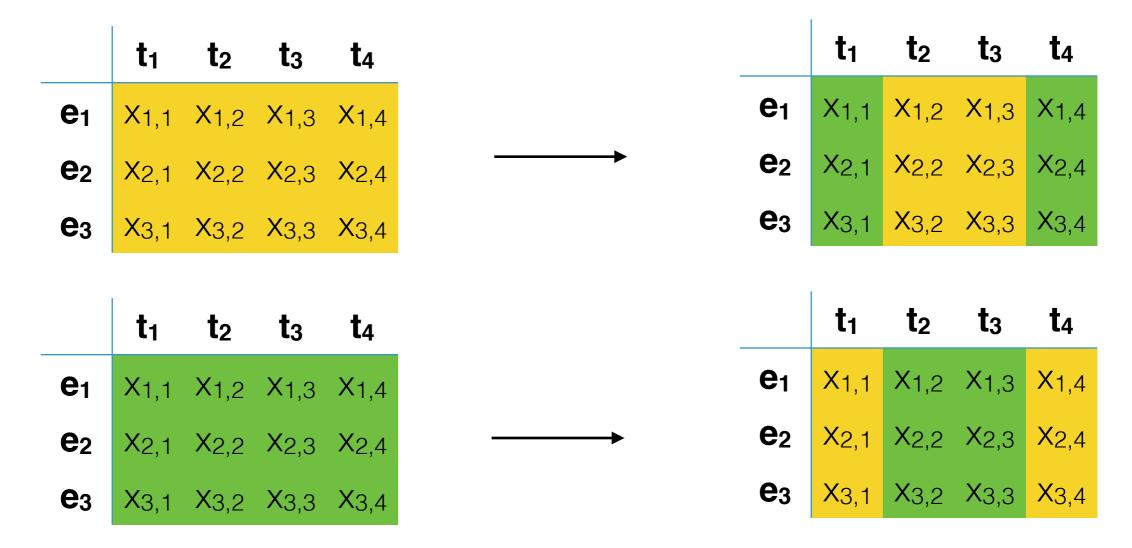
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- Crossover: exchange rows or columns.



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## Formulating the SPSP

#### Setting: assume we are given

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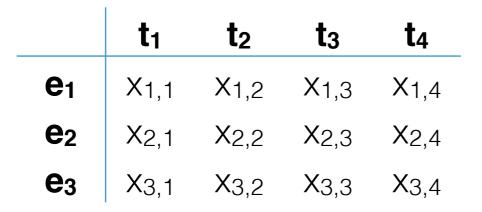
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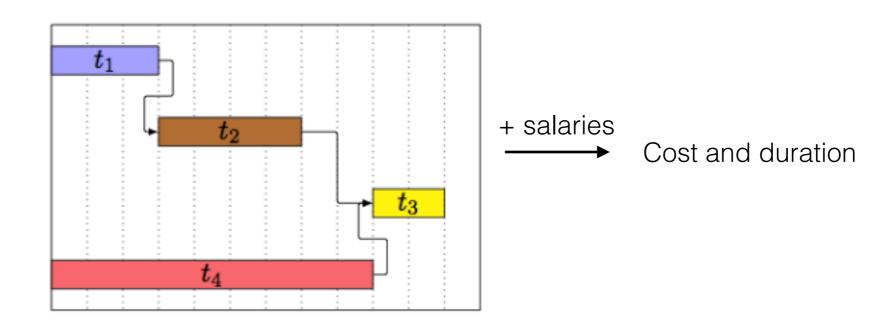
#### Constraints:

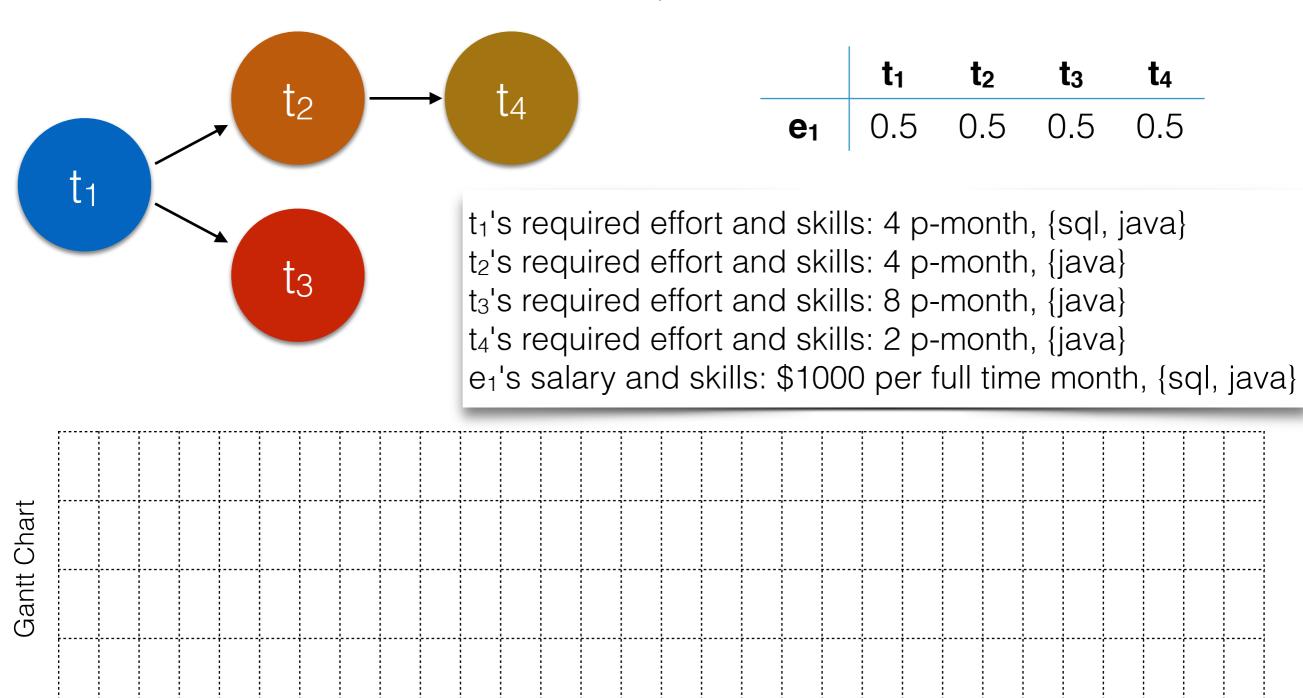
- team must have required skills and
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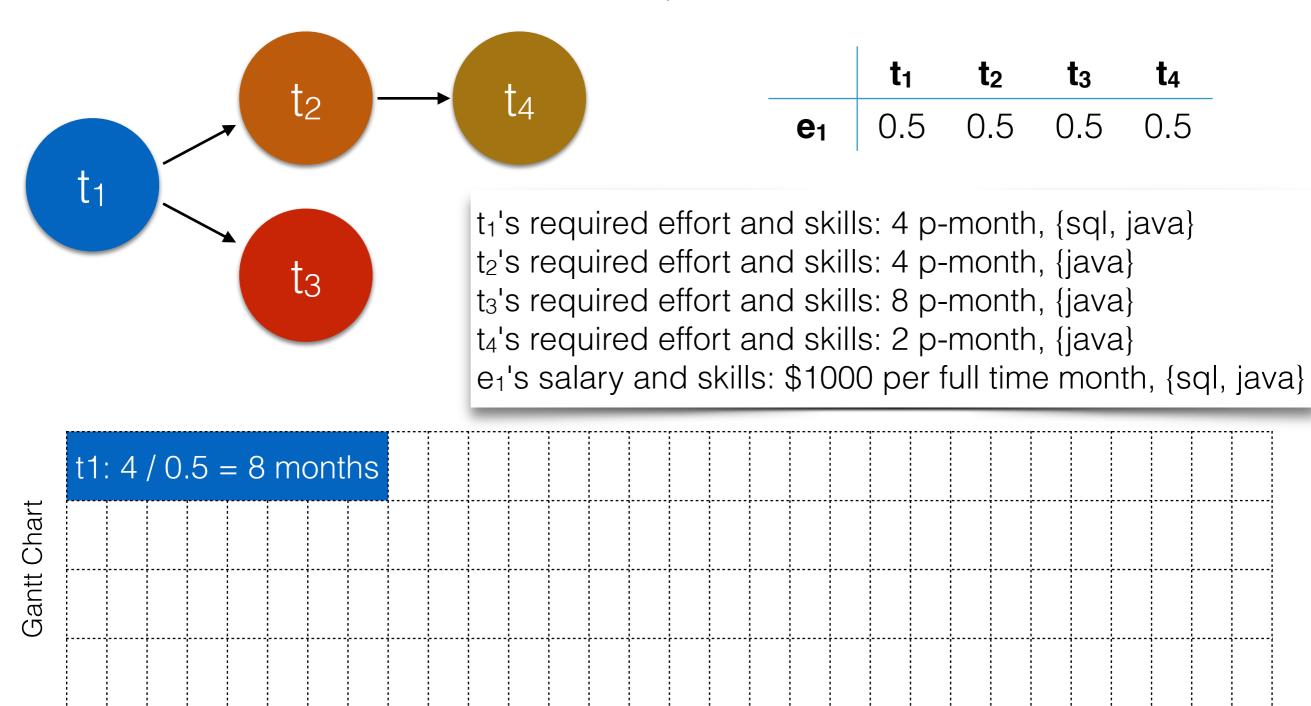
### Evaluating a Solution



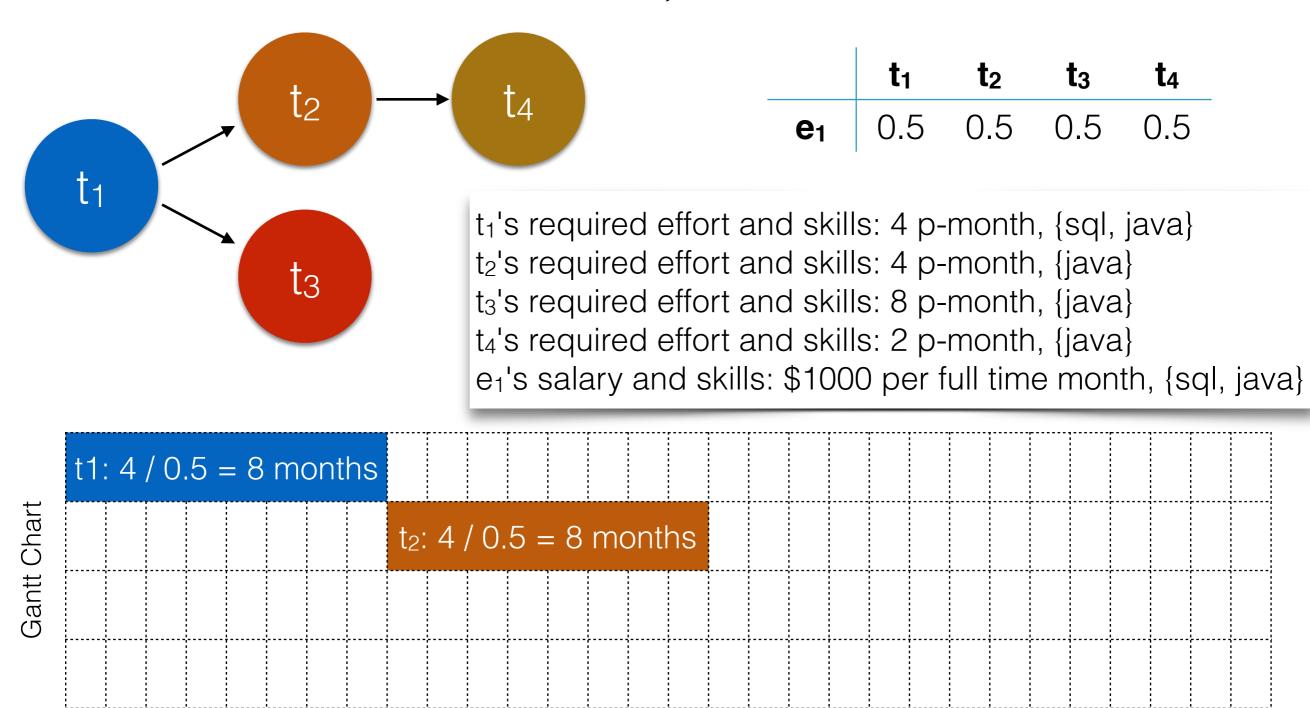
+ TPG, tasks required efforts

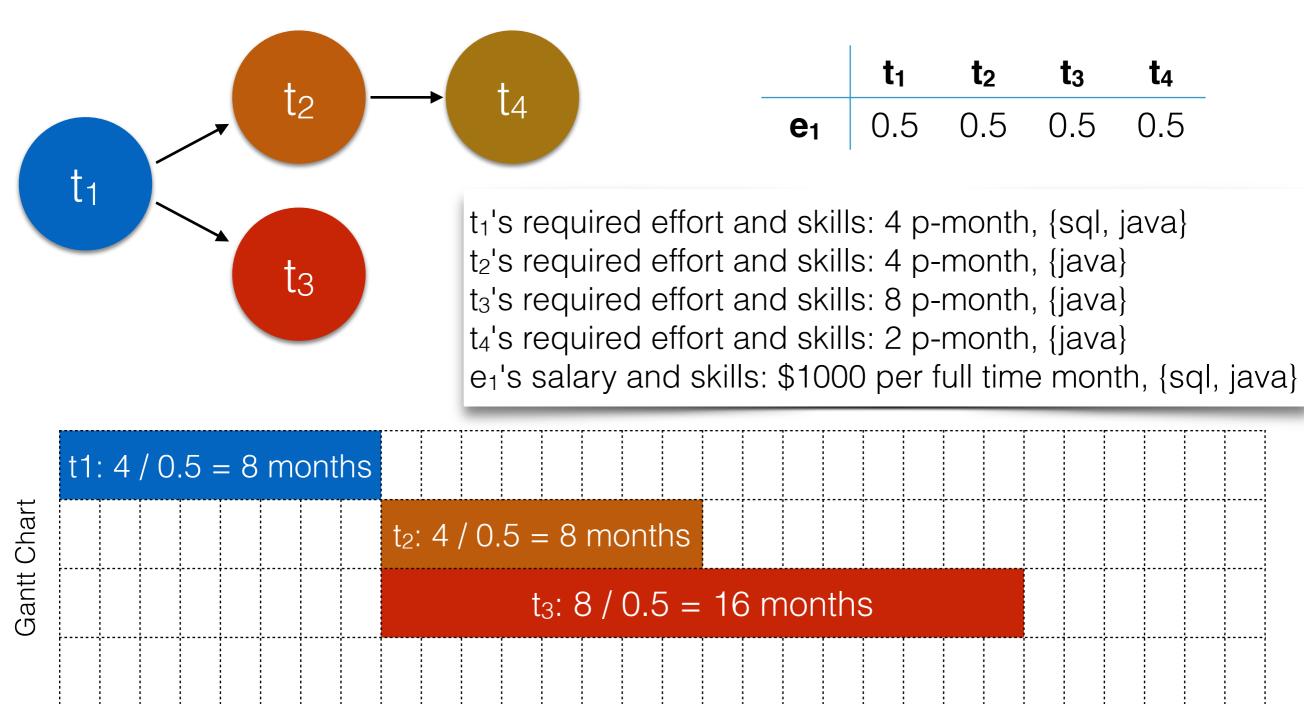






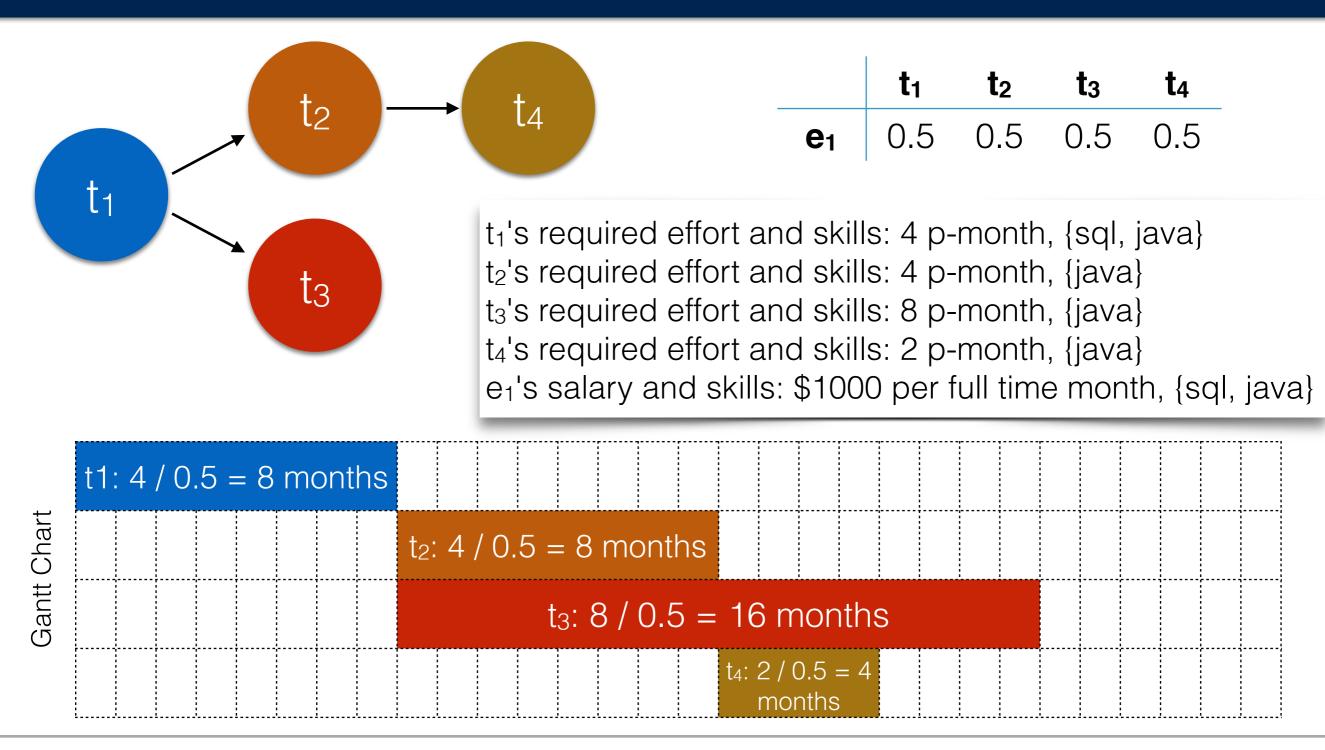
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#### What is the completion time of the project?

#### And the cost?



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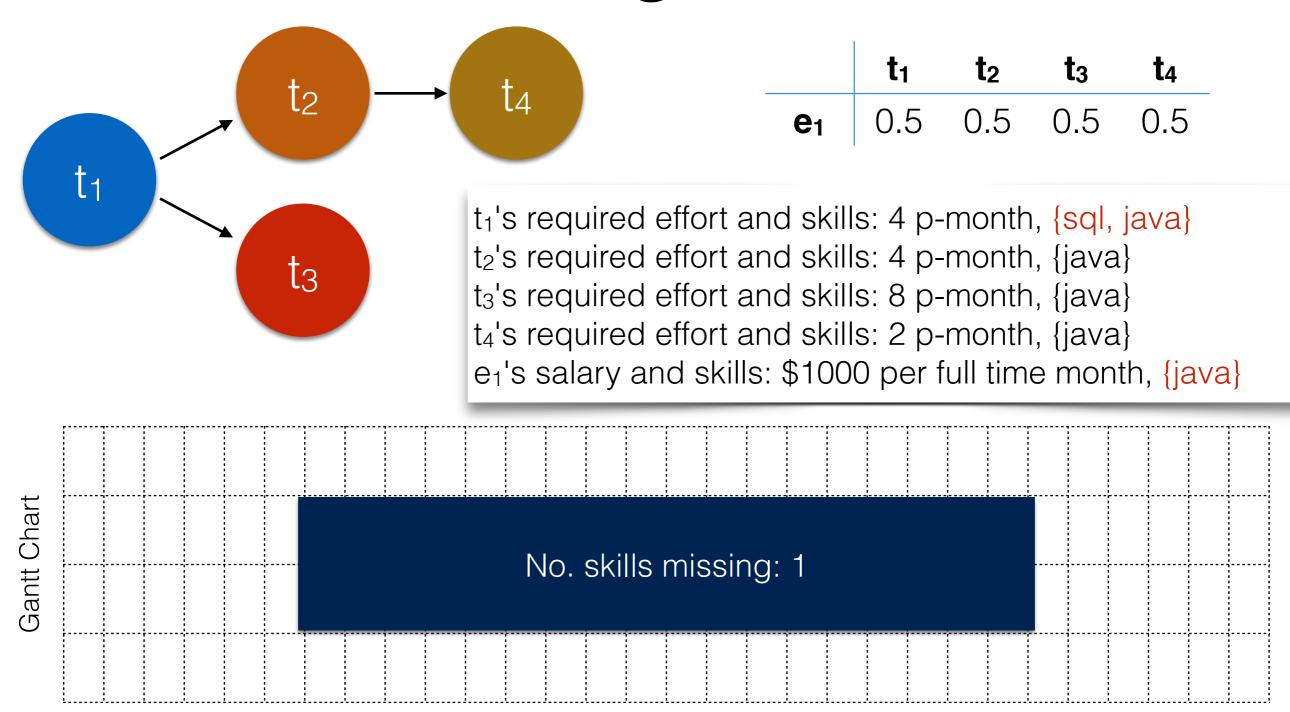
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#### Constraints:

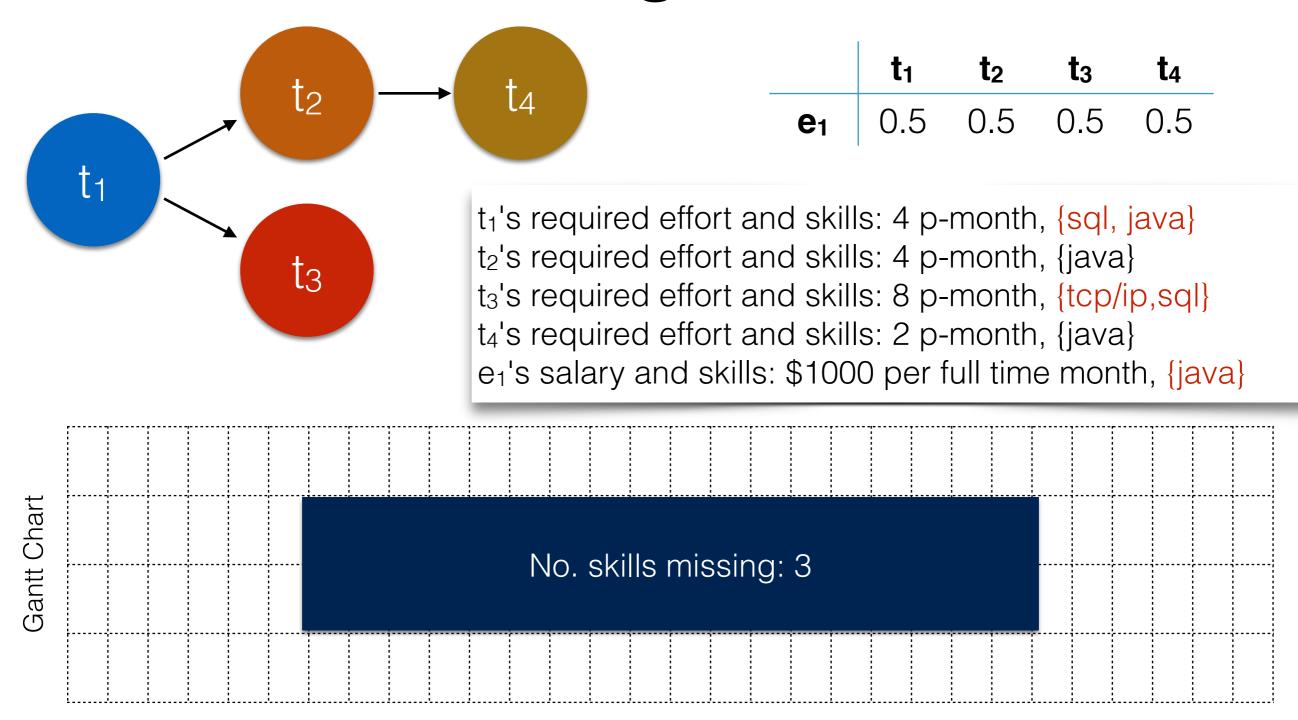
- team must have required skills and
- no overwork.

### Infeasible Schedule --Missing Skills



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### Infeasible Schedule --Missing Skills



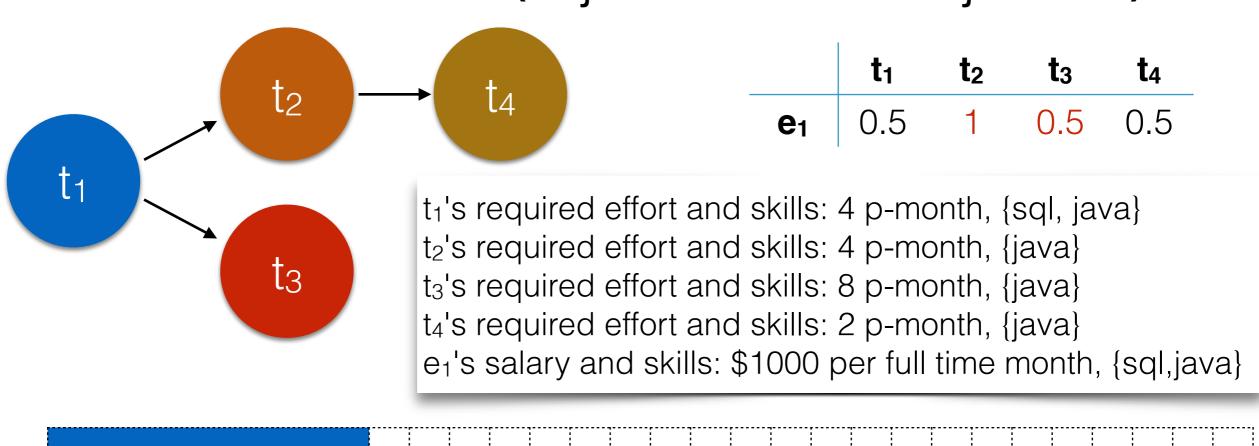
#### Overwork

• There is overwork at time  $\tau$  if, for a given employee  $e_i$ , the total dedication of  $e_i$  to tasks at time  $\tau$  is:

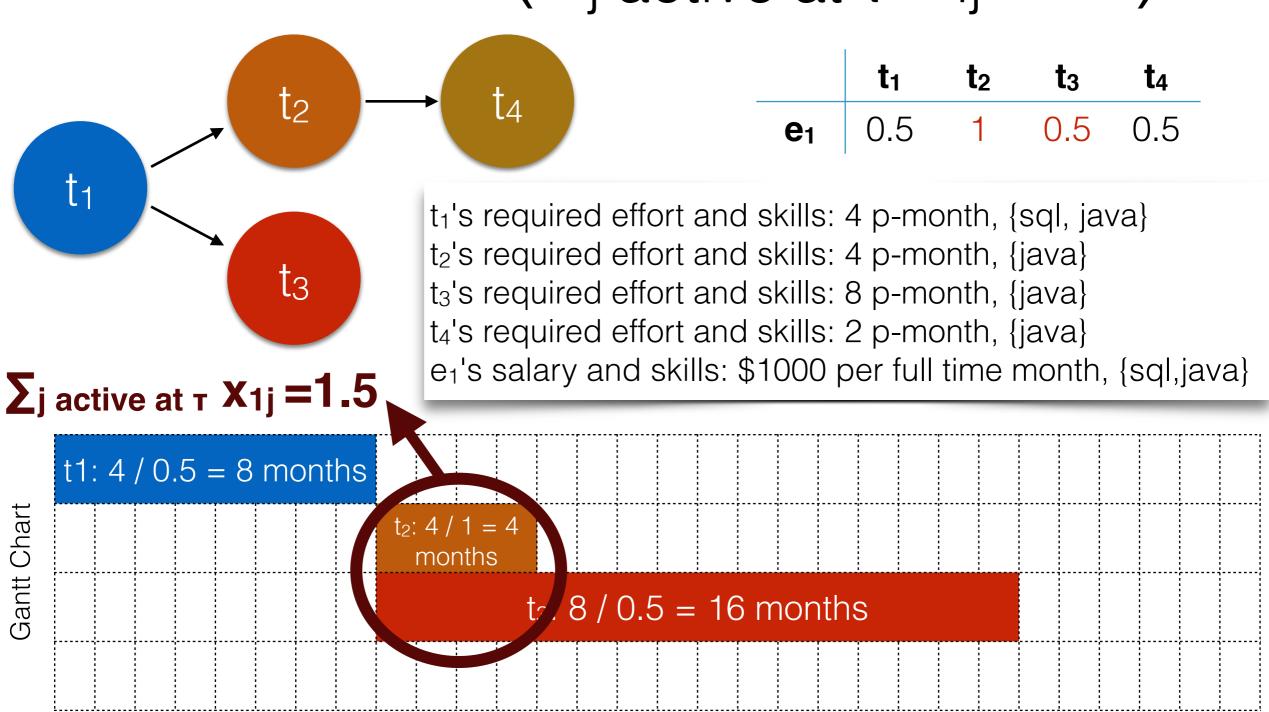
$$\sum_{j \text{ active at } \tau} X_{ij} > 1$$

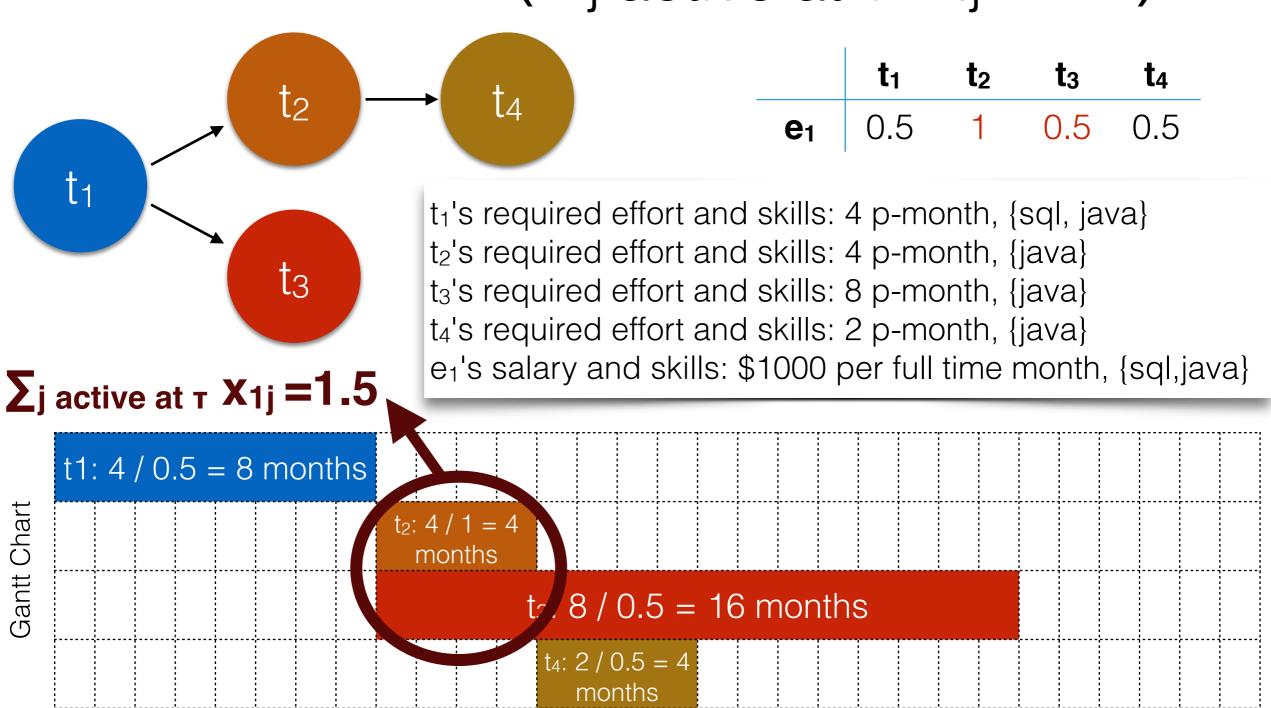
• Overwork for employee  $e_i$  at time  $\tau =$ 

```
max(0, \sum_{j \text{ active at } \tau} x_{ij} - 1)
```





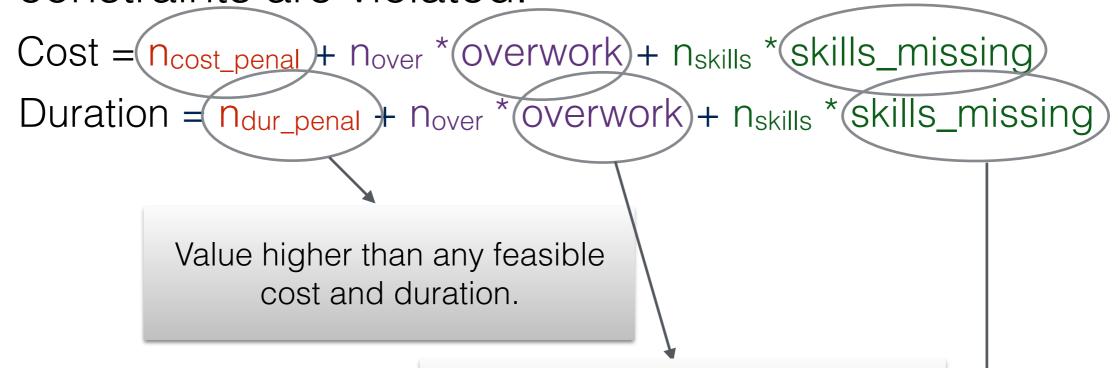




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## How to Deal with the Constraints?

 Option 1: assign very high cost and duration if constraints are violated.



Total amount of overwork time.

Total number of skills missing in the teams.

## How to Deal with the Constraints?

 Option 1: assign very high cost and duration if constraints are violated.

```
Cost = n_{cost\_penal} + n_{over} * overwork + n_{skills} * skills_missing

Duration = n_{dur\_penal} + n_{over} * overwork + n_{skills} * skills_missing
```

What are the problems of this solution?

## How to Deal with the Constraints?

 Option 2: normalise dedications to deal with overwork so that total dedication is at most 1.

If employee *i* has overwork at any moment τ

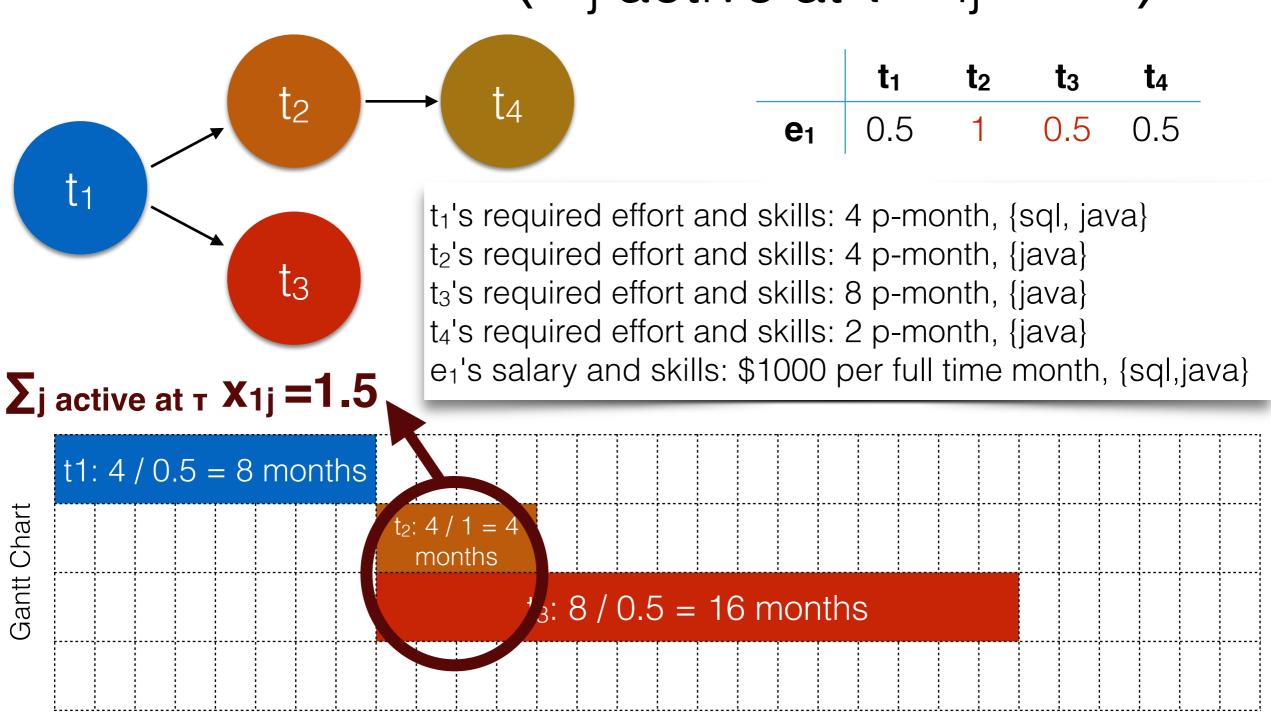
$$d_{ij}(\tau) = x_{ij} / \sum_{j \text{ active at } \tau} x_{ij}$$

else

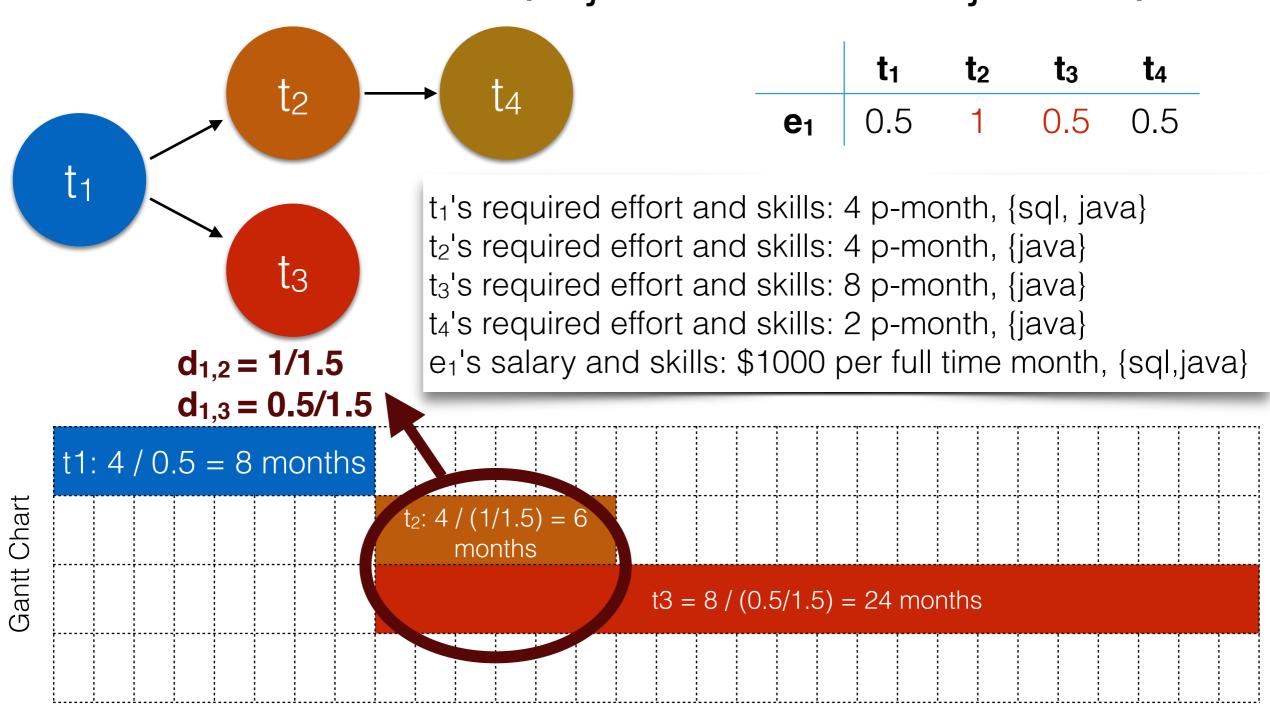
$$d_{ij}(\tau) = x_{ij}$$

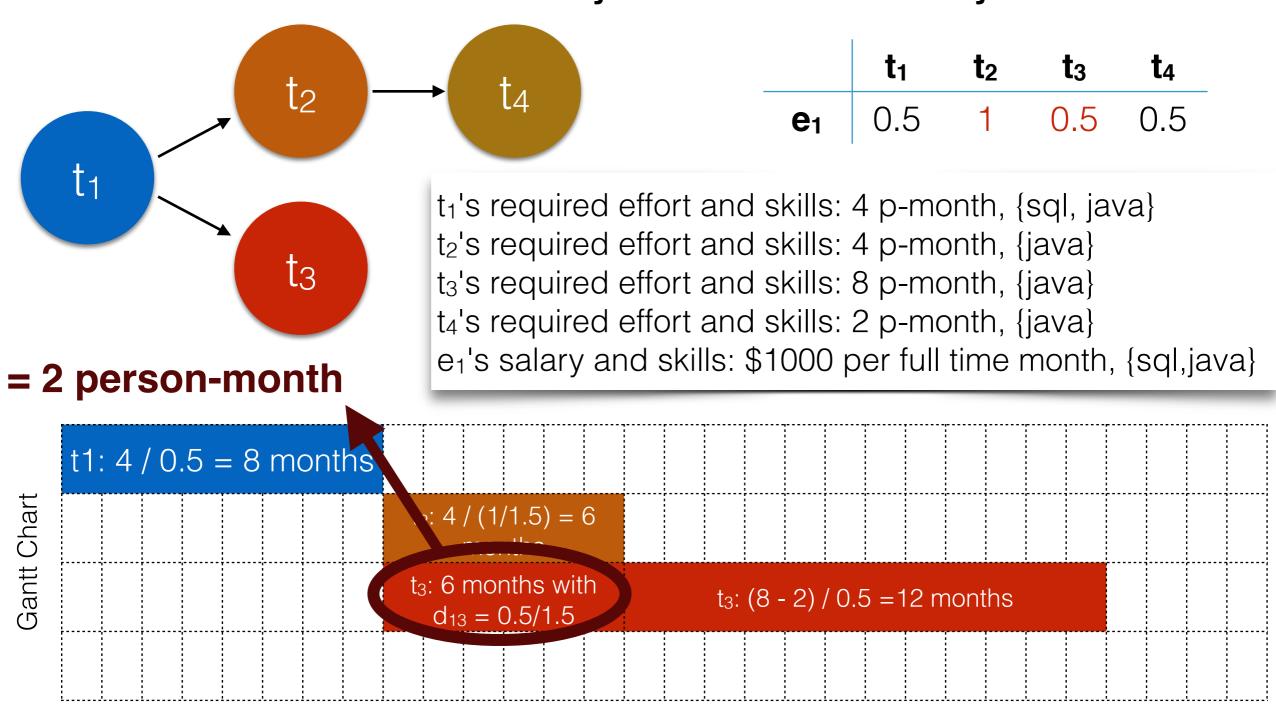
Cost = n<sub>cost\_penal</sub> \* 2 \* skills\_missing

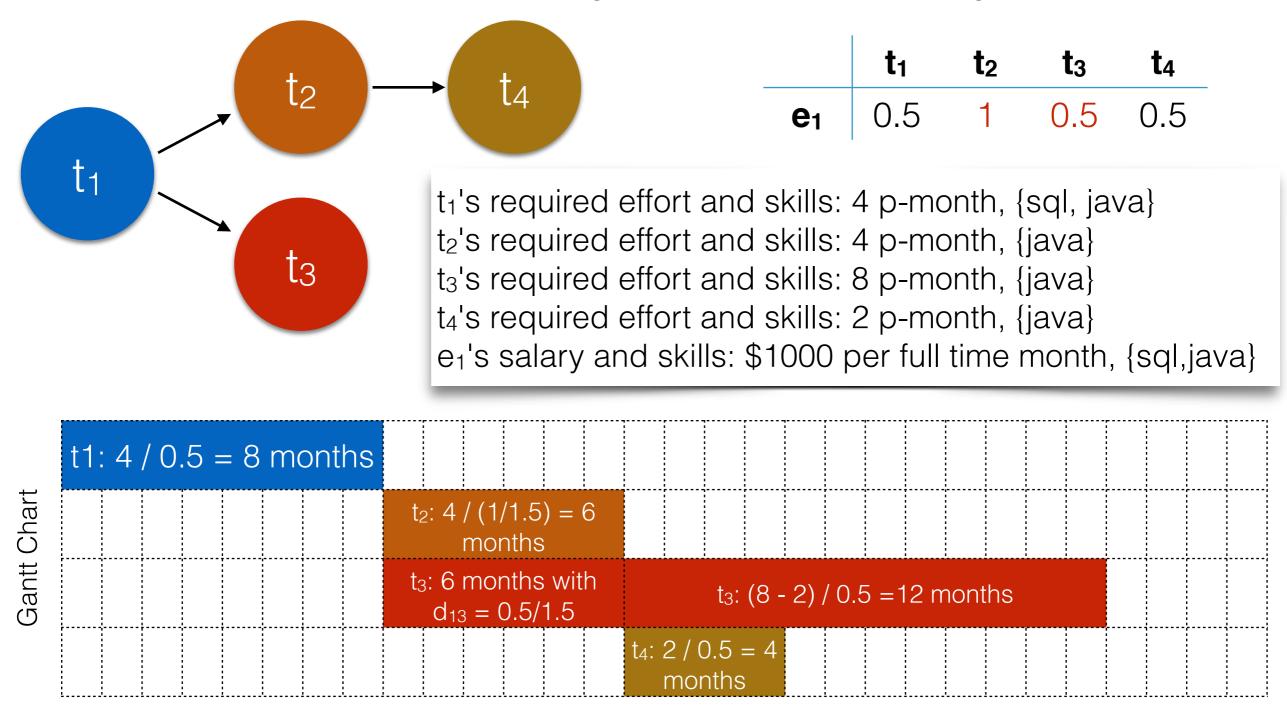
Duration = n<sub>dur\_penal</sub>\* 2 \* skills\_missing



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[Demo -- Ian Watson's final year project (2013)]

### Summary

- What Software Project Scheduling Problem (SPSP) is.
- Why are automated optimisation methods are important for the SPSP.
- How to solve the SPSP:
  - suitable EA;
  - representation;
  - mutation and crossover;
  - objectives;
  - constraints.

### Further Reading

L. Minku, D. Sudholt and X. Yao. "Improved Evolutionary Algorithm Design for the Project Scheduling Problem Based on Runtime Analysis", IEEE Transactions on Software Engineering vol. 40, n. 1, p. 83-102, 2014.

F. Luna, D. Gonzalez-Alvarez, F. Chicano, and M. Vega- Rodriguez, "On the Scalability of Multi-Objective Metaheuristics for the Software Scheduling Problem", Proceedings of the 11th International Conference on Intelligent System Design and Applications, p. 1110-1115, 2011.

E. Alba and F. Chicano, "Software Project Management with GAs", Information Sciences, vol. 177, p. 2380-2401, 2007.