Introduction to Artificial Intelligence



COMP307 Planning and Scheduling 3: Dynamic Scheduling

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Outline

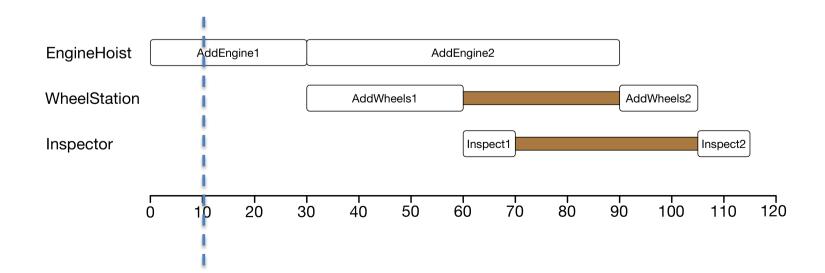
- Dynamic Scheduling
- Dispatching Rules
 - Generating schedules by rules
- Designing Dispatching Rules
 - Terminal set
 - Function set
 - Fitness function

Dynamic Scheduling

- In static scheduling, it is assumed that all the information is known in advance and do not change over time
- In real life, usually not the case (dynamic environment)
 - The plan today won't work tomorrow

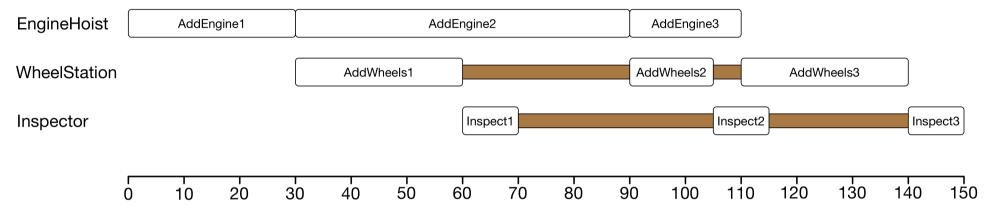
Dynamic Scheduling

- Manufacture two cars
 - 2 jobs known in advance
 - Already made a plan: makespan = 115
 - A new job arrives at time 10
 - Job({AddEngine3 ≺ AddWheels3 ≺ Inspect3})
 - Operation(AddEngine3, ProcTime: 20, Use: EngineHoist)
 - Operation(AddWheels3, ProcTime: 30, Use: WheelStation)
 - Operation(Inspect3, ProcTime: 10, Use: Inspector)



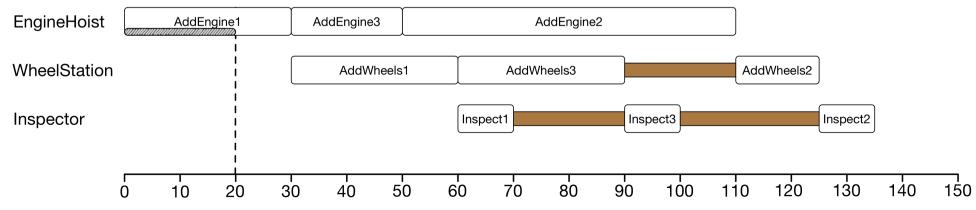
Dynamic Rescheduling

- Simply append to the end of the current schedule
 - Makespan = 150

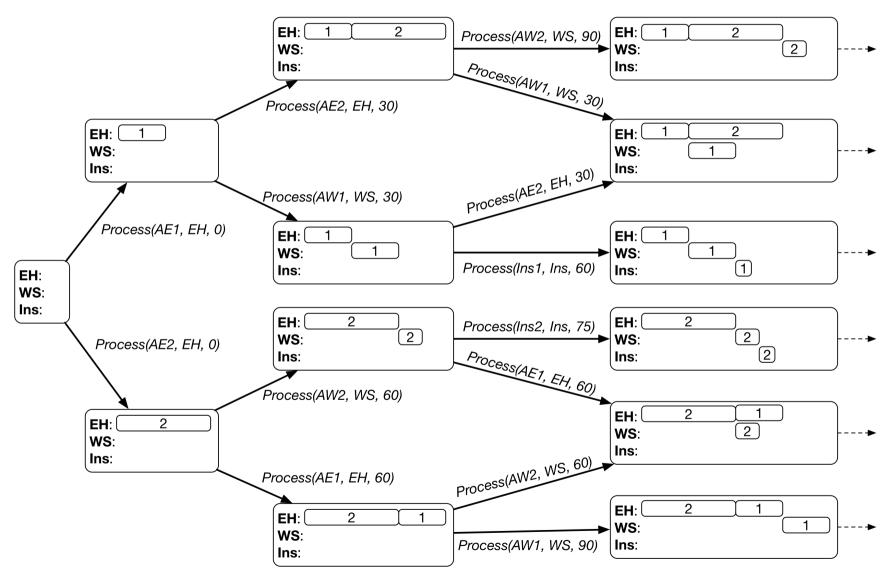


- Re-optimise the unexecuted schedule
 - Makespan = 135, but can be SLOW

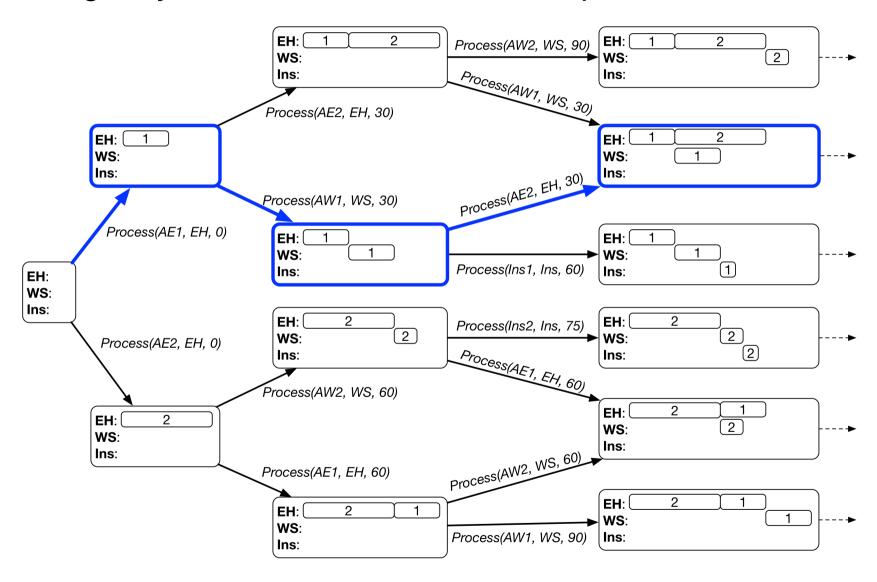
10 jobs, 5 machines, 6.3×10^{32} solutions



Forward search: expand all branches, time consuming



Intelligently select one branch at each point?



- Dispatching Rule: a rule to select one action in each state
 - Considering ONLY the earliest applicable actions (non-delay)
 - Assigning a priority to each earliest action by a priority function
 - Selecting the action with the highest priority
- An example: Shortest Processing Time (SPT)
 - Always select the shortest processing
 - Priority of Process(o, m, t) is 0-ProcTime(o)

• Which one is selected?

Action	Priority
Process(AddEngine2, EngineHoist, 30)	-60
Process(Inspect1, Inspector, 60)	-10

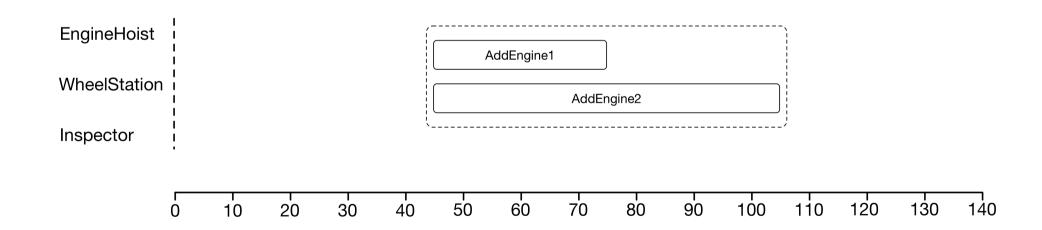
• Which one is selected?

Action	Priority
Process(AddEngine1, EngineHoist, 0)	-30
Process(AddEngine2, EngineHoist, 0)	-60

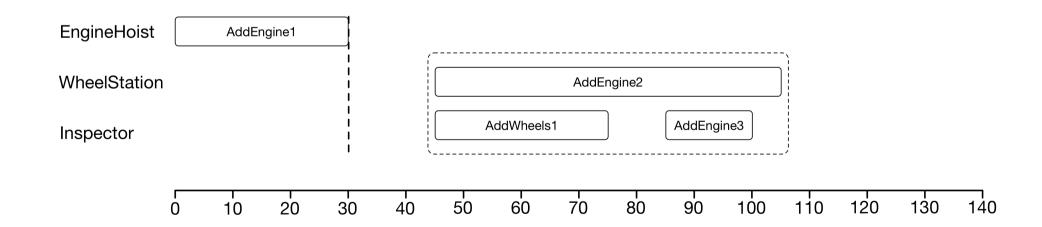
Generate a Schedule by Dispatching Rule

- Step 1: Initialise state
 - empty schedule, all operations unprocessed, time = 0, machine idle time = 0, first operation ready time = α other operation ready time = ∞
- Step 2: Find the earliest applicable actions;
- Step 3: Select the next action by the dispatching rule
- Step 4: Add the selected action into the schedule, update the state
- Step 5: If all operations are processed, stop. Otherwise, go to step 2.

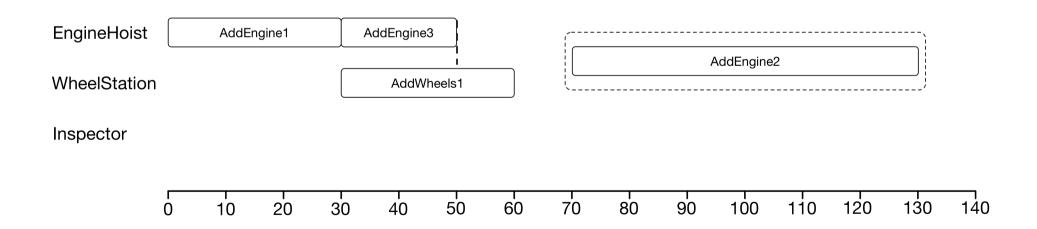
	Arrive	ProcTime		
		AddEngine	AddWheels	Inspect
Job 1	0	30	30	10
Job 2	0	60	15	10
Job 3	10	20	30	10



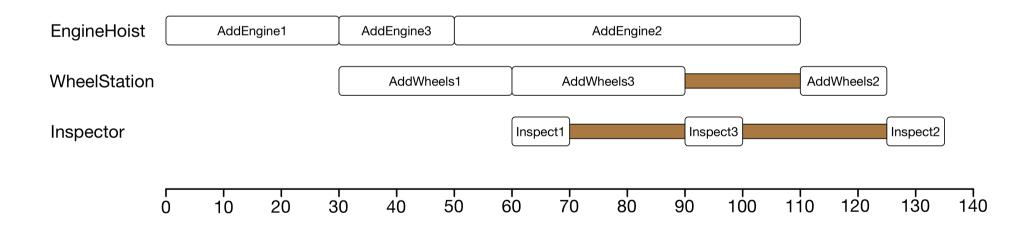
	Arrive	ProcTime		
		AddEngine	AddWheels	Inspect
Job 1	0	30	30	10
Job 2	0	60	15	10
Job 3	10	20	30	10



	Arrive	ProcTime		
		AddEngine	AddWheels	Inspect
Job 1	0	30	30	10
Job 2	0	60	15	10
Job 3	10	20	30	10



	Arrive	ProcTime		
		AddEngine	AddWheels	Inspect
Job 1	0	30	30	10
Job 2	0	60	15	10
Job 3	10	20	30	10



Advantage of Dispatching Rule

- Can be apply at ANY time point to change the remaining schedule
 - Initial state = current state
 - But only need at critical time point (a machine becomes idle, an operation becomes ready)
- Select ONLY the next action to be taken, NO need to generate the entire remaining schedule
- Very quick in real time, can handle dynamic environment very well
 - At each time point, complexity = #unprocessed ops * O(priority)

Design of Dispatching Rule

Intuition

- First-Come-First-Serve (Minimum Waiting Time)
- Shortest Processing Time
- Earliest Due Date
- Maximum Work Remaining

— ...

Look-Ahead

- Work waiting on the next machine
- Processing time of the next operation

Composite rules

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- -(PT+WINQ)
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– -(2PT+WINQ+NPT)

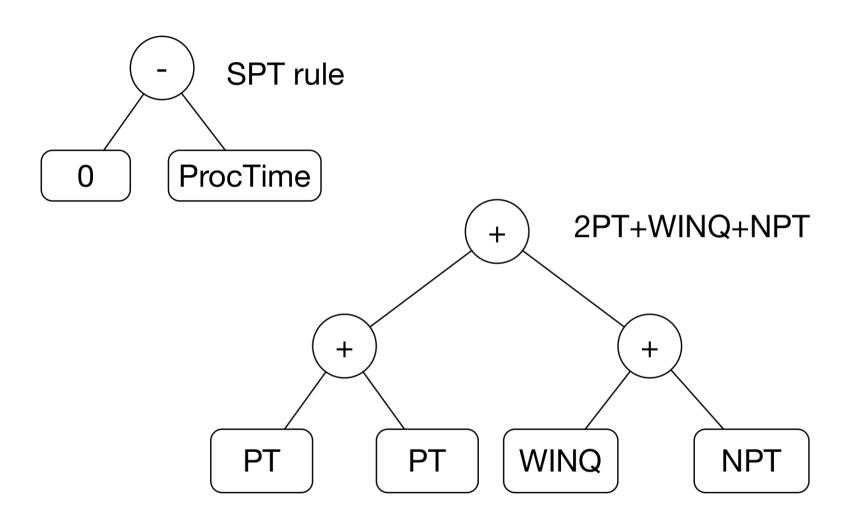
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Design of Dispatching Rule

- Different scenarios need different rules
 - Ford car manufacturing factory in summer season
 - Samsung mobile production lines in spring season
- Hard to design effective rule for a particular given scenario
- Use Genetic Programming (GP) to learn/train dispatching rule based on historical data/simulation

Learning Dispatching Rule with GP

Goal: find the best priority function (GP trees)



Learning Dispatching Rule with GP

- Terminal set: features/attributes of the state and the considered Process(o, m, t)
 - Processing time of o
 - Processing time of o's next operation
 - Total processing time of all the subsequent operations after o (work remaining)
 - Constant coefficients
 - **—** ...
- Function set
 - {+, -, x, /}
 - {max, min}
 - **—** ...
- Fitness: average makespan (or any other objective) of the generated schedules for a set of training instances

Summary

- Simple (re-)search cannot handle dynamic scheduling
- Dispatching rule
- Generate a schedule by a dispatching rule
- Learning dispatching rules by GP