#### Introduction to Artificial Intelligence



#### **COMP307**

#### **Planning and Scheduling 3:**

#### **Dynamic Scheduling**



## Outline

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



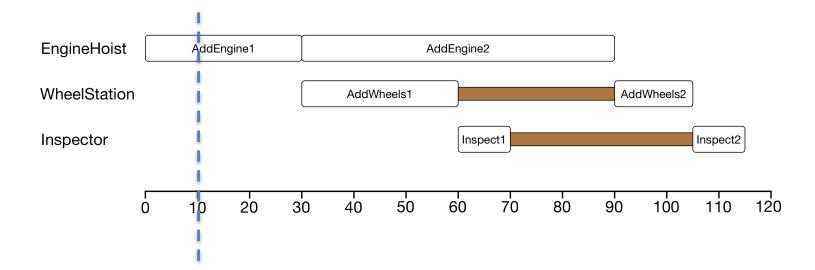
# **Dynamic Scheduling**

- In static scheduling, it is assumed that all the information is  $\bullet$ 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**

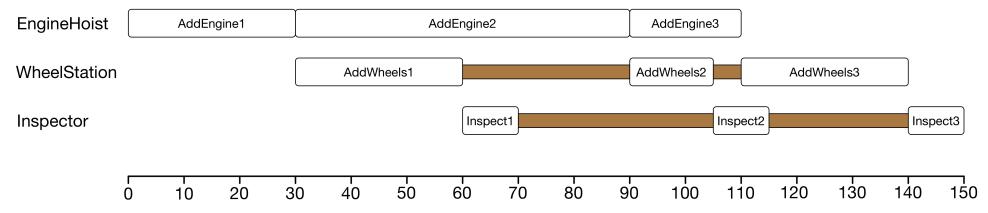
- Manufacture two cars
  - 2 jobs known in advance
  - Already made a plan: makespan = 115
  - A new job arrives at time 10
  - Job({AddEngine3 < AddWheels3 < Inspect3})</p>
  - Operation(AddEngine3, ProcTime: 20, Use: EngineHoist)
  - Operation(AddWheels3, ProcTime: 30, Use: WheelStation)
  - Operation(Inspect3,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





## **Dynamic Rescheduling**

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

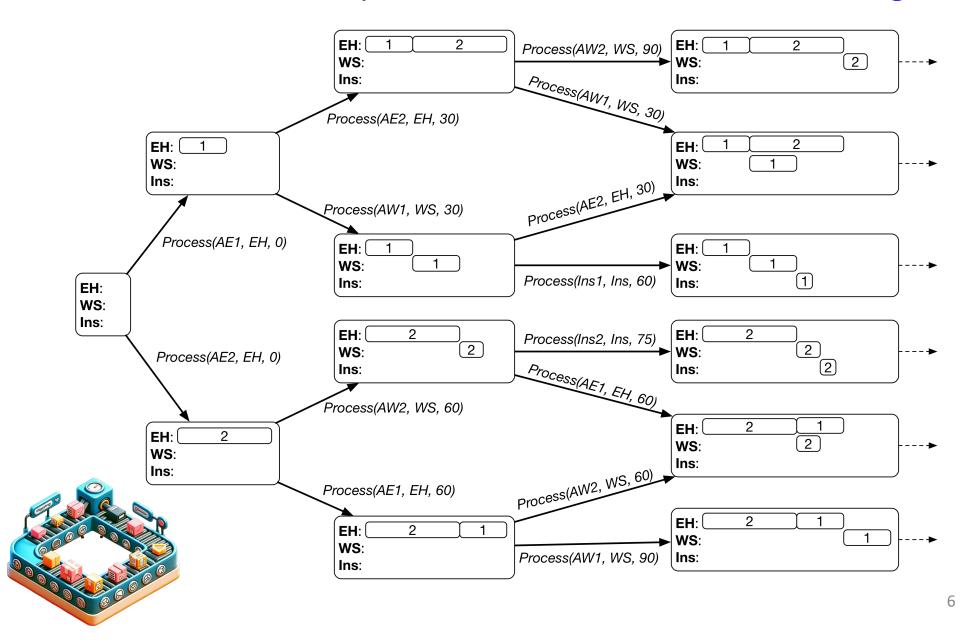


Re-optimise the unexecuted schedule
Makespan = 135, but can be SLOW

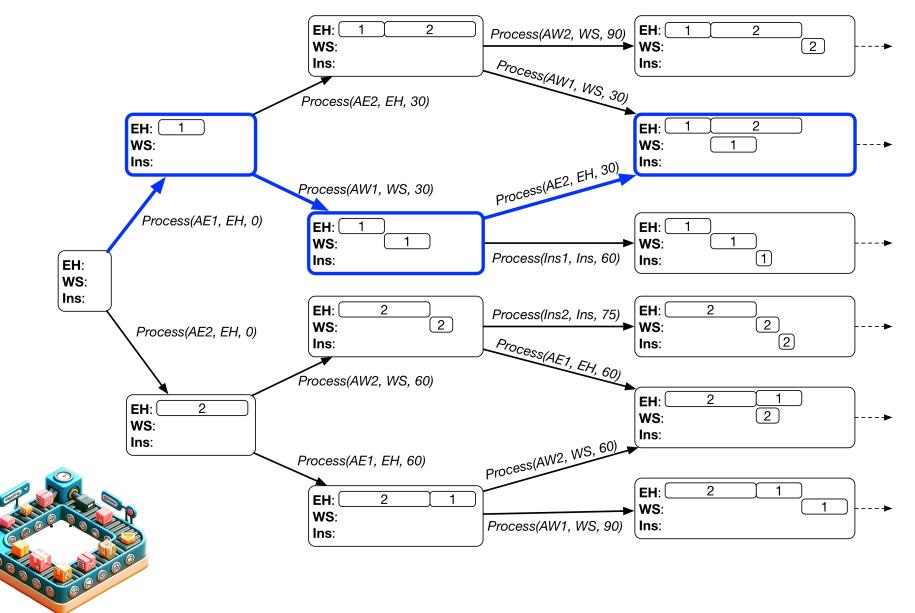
10 jobs, 5 machines, 6.  $3 \times 10^{32}$  solutions



• 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 time
  - Priority of Process(o, m, t) is -ProcTime(o)



• Which one is selected?

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

• Which one is selected?

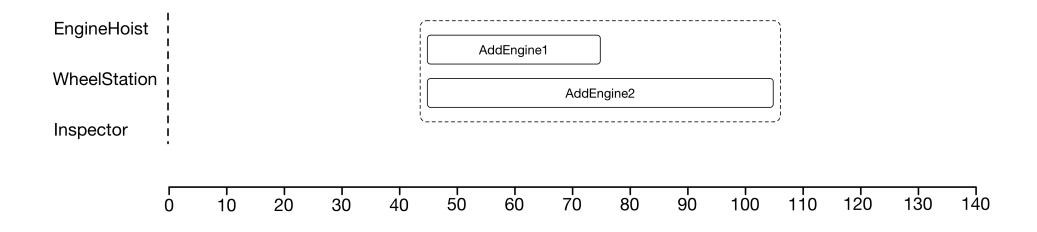
Action	Priority
Process(AddEngine1, EngineHoist, 0)	-30
Process(AddEngine2, EngineHoist, 0)	-60

## Generate a Schedule by Dispatching Rule

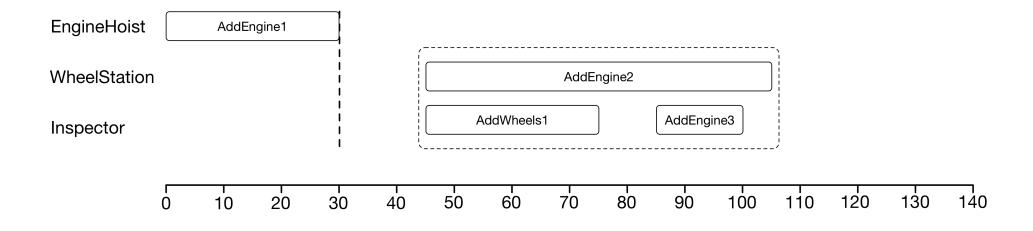
- Step 1: Initialize state
  - empty schedule, all operations unprocessed, time = 0, machine idle time = 0, first operation ready time = arrival time, other operation ready time = ∞
- Step 2: Find the earliest applicable actions;
- Step 3: Select the next action by the dispatching rule
- Step 4: Add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



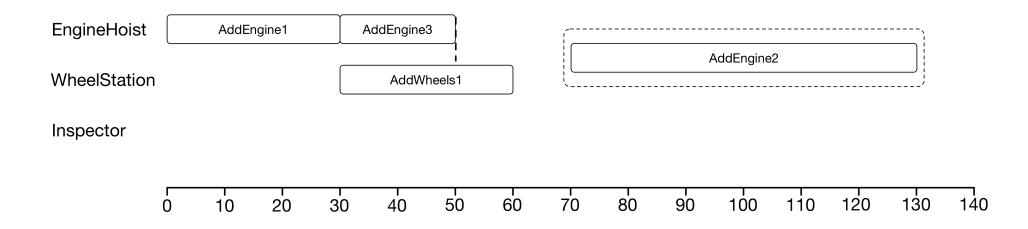
	Arrive	ProcTime		
		AddEngine	AddWheels	Inspect
Job 1	0	30	30	10
Job 2	0	60	15	10
Job 3	10	20	30	10



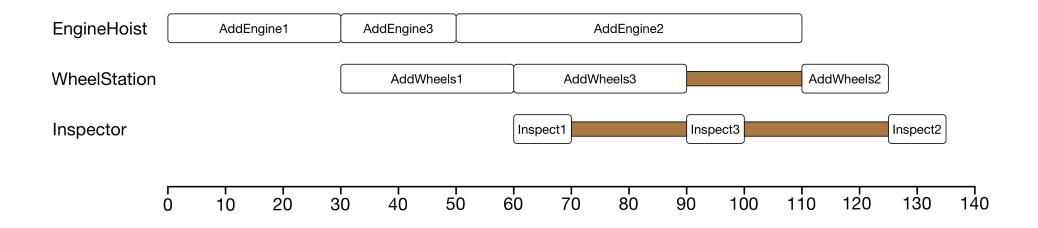
	Arrive	ProcTime		
		AddEngine	AddWheels	Inspect
Job 1	0	30	30	10
Job 2	0	60	15	10
Job 3	10	20	30	10



	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



## Advantages of Dispatching Rule

- Can be apply at ANY time point to change the remaining  $\bullet$ 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
  - -(PT+WINQ)
  - -(2PT+WINQ+NPT)

- ...



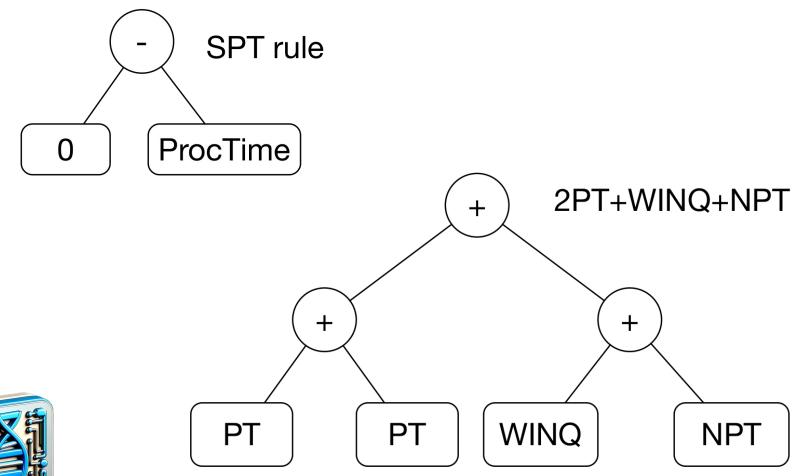
# **Design of Dispatching Rule**

- 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 any particular scheduling 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

