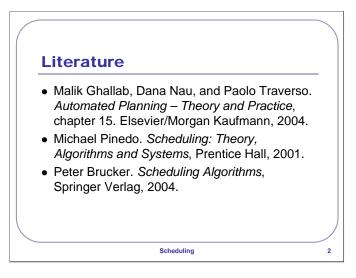


Scheduling

•Planning with Actions that Require Resources

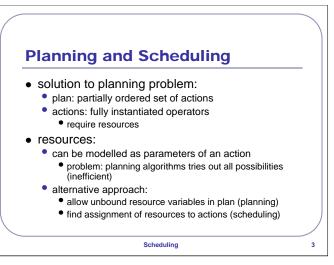


Literature

•Malik Ghallab, Dana Nau, and Paolo Traverso. *Automated Planning – Theory and Practice*, chapter 15. Elsevier/Morgan Kaufmann, 2004.

•Michael Pinedo. Scheduling: Theory, Algorithms and Systems, Prentice Hall, 2001.

•Peter Brucker. Scheduling Algorithms, Springer Verlag, 2004.



Planning and Scheduling

•solution to planning problem:

•plan: partially ordered set of actions

•actions: fully instantiated operators

•require resources

•resources:

•can be modelled as parameters of an action

 problem: planning algorithms tries out all possibilities (inefficient)

•alternative approach:

•allow unbound resource variables in plan (planning)

•planning focuses on causal reasoning (what to do)

•find assignment of resources to actions (scheduling)

•scheduling: resource and time allocation (how and when to do it)

•planning before scheduling (not optimal approach)

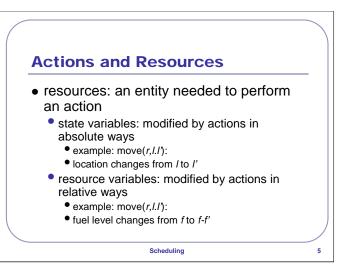


Overview

•Scheduling Problems and Schedules

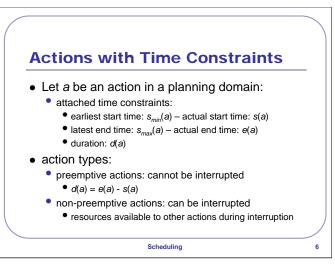
•now: an overview of different types of scheduling problems

Searching for Schedules



Actions and Resources

•resources: an entity needed to perform an action
•state variables: modified by actions in absolute ways
•example: move(*r*,*l*.*l'*):
•location changes from *I* to *I'*•resource variables: modified by actions in relative ways
•example: move(*r*,*l*.*l'*):
•fuel level changes from *f* to *f*-*f'*



Actions with Time Constraints

•Let *a* be an action in a planning domain:

•attached time constraints:

```
•earliest start time: s<sub>min</sub>(a) – actual start time: s(a)
```

```
•latest end time: s_{max}(a) – actual end time: e(a)
```

```
•duration: d(a)
```

•action types:

•preemptive actions: cannot be interrupted

 $\bullet d(a) = e(a) - s(a)$

•non-preemptive actions: can be interrupted

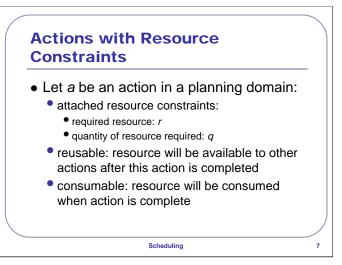
•resources available to other actions during interruption

•cost: interruption usually has cost associated

•further constraints: examples:

•action must be performed at night

•interruptions must be at least 30 minutes long



Actions with Resource Constraints

•Let *a* be an action in a planning domain:

•attached resource constraints:

•required resource: r

•quantity of resource required: q

•reusable: resource will be available to other actions after this action is completed

•tools, machines, HD space, helicopters, docks

•consumable: resource will be consumed when action is complete

•petrol, electricity, CPU time, credit

•time is usually treated differently, as a special case

Reusable Resources
 resource availability: total capacity: Q_r current level at time t z_r(t) resource requirements: require(a,r,q): action a requires q units of resource r while is active resource profile:
$Q_r q_1 q_2$
a_1 : require(a_1, r, q_1) a_2 : require(a_2, r, q_2)
Scheduling

Reusable Resources

•resource availability:

•total capacity: Q_r

```
•current level at time t: z_r(t)
```

•resource requirements:

•require(*a*,*r*,*q*): action *a* requires *q* units of resource *r* while it is active

•resource profile:

•[figure]

•actions are overlapping (temporally)

•profile shows availability of resource to other actions

•returns to full capacity when all actions are completed

Consumable Resources • resource availability: • total reservoir at c_i : Q_r • current level at time t . $z_i(t)$ • resource consumption/production: • consume(a,r,q): action a requires q units of resource r • produce(a,r,q): action a produces q units of resource r • resource profile: $Q_r = \frac{z_r}{q_1}$ q_2 q_3	
$\begin{array}{c} a_2: \text{ consume}(a_2, r, q_2) \\ a_3: \text{ produce}(a_3, r, q_3) \end{array}$	
Scheduling	9

Consumable Resources

resource availability:

•total reservoir at t_0 : Q_r

•current level at time $t: z_r(t)$

•resource consumption/production:

•consume(*a*,*r*,*q*): action *a* requires *q* units of resource *r*•produce(*a*,*r*,*q*): action *a* produces *q* units of resource *r*•resource profile:

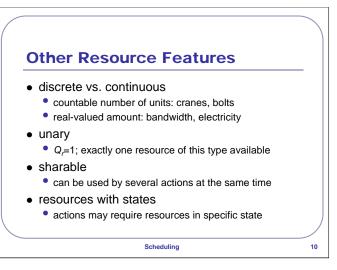
•[figure]

•actions are overlapping (temporally)

•profile shows availability of resource to other actions

availability at end usually different from beginning

•resource profile as step function: usually not accurate



Other Resource Features

discrete vs. continuous

•countable number of units: cranes, bolts

•real-valued amount: bandwidth, electricity

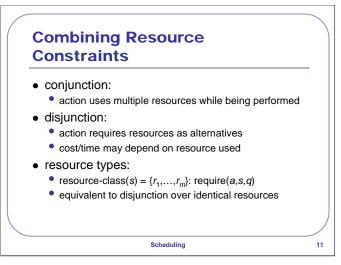
•unary

•*Q_r*=1; exactly one resource of this type available •sharable

can be used by several actions at the same time
resources with states

•actions may require resources in specific state

•example: freezer with temperature setting



Combining Resource Constraints

•conjunction:

•action uses multiple resources while being performed•disjunction:

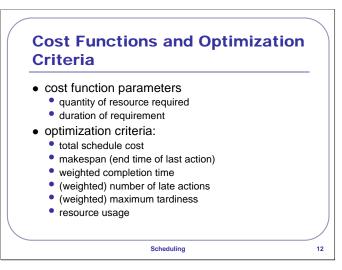
action requires resources as alternatives

cost/time may depend on resource used

•resource types:

```
•resource-class(s) = \{r_1, \dots, r_m\}: require(a,s,q)
```

•equivalent to disjunction over identical resources



Cost Functions and Optimization Criteria •cost function parameters

quantity of resource required

duration of requirement

•optimization criteria:

total schedule cost

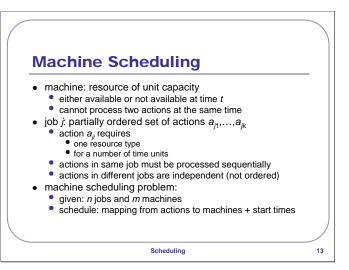
•makespan (end time of last action)

•weighted completion time

•(weighted) number of late actions

•(weighted) maximum tardiness

•resource usage



Machine Scheduling

•class of problems

•machine: resource of unit capacity

•either available or not available at time t

•cannot process two actions at the same time

•job *j*: partially ordered set of actions a_{j_1}, \dots, a_{j_k}

•in general, jobs can have different numbers of activities

•action a_{ii} requires

•one resource type

•for a number of time units

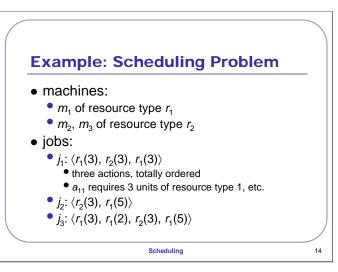
•actions in same job must be processed sequentially

•even if they are only partially ordered: object that is being worked on

•actions in different jobs are independent (not ordered)•machine scheduling problem:

•given: *n* jobs and *m* machines

•schedule: mapping from actions to machines + start times



Example: Scheduling Problem •machines:

• m_1 of resource type r_1

• m_2 , m_3 of resource type r_2

•jobs:

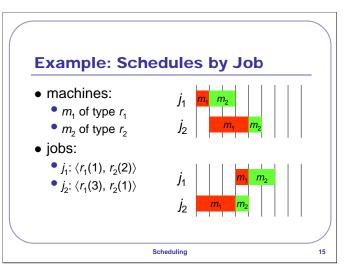
•*j*₁: $\langle r_1(3), r_2(3), r_1(3) \rangle$

•three actions, totally ordered

• a_{11} requires 3 units of resource type 1, etc.

•*j*₂: $\langle r_2(3), r_1(5) \rangle$

• j_3 : $\langle r_1(3), r_1(2), r_2(3), r_1(5) \rangle$



Example: Schedules by Job •machines:

• m_1 of type r_1

• m_2 of type r_2

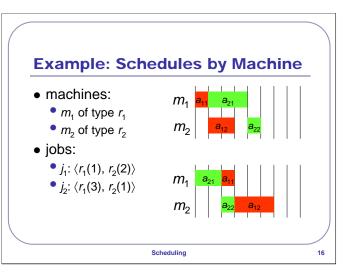
•jobs:

• j_1 : $\langle r_1(1), r_2(2) \rangle$ • j_2 : $\langle r_1(3), r_2(1) \rangle$

·J₂. \/₁(3), /

•[figures]

•schedules showing machines assigned to actions in jobs



Example: Schedules by Machine •machines:

• m_1 of type r_1

• m_2 of type r_2

•jobs:

• j_1 : $\langle r_1(1), r_2(2) \rangle$ • j_2 : $\langle r_1(3), r_2(1) \rangle$

•[figures]

•schedules showing actions assigned to machines



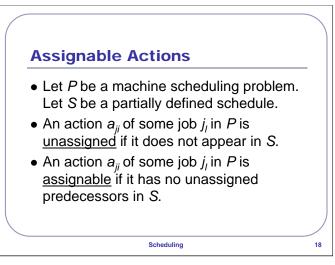
Overview

•Scheduling Problems and Schedules

•just done: an overview of different types of scheduling problems

Searching for Schedules

now: search algorithms that generate schedules



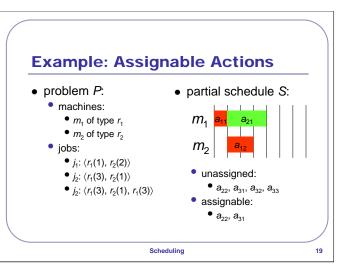
Assignable Actions

•Let *P* be a machine scheduling problem. Let *S* be a partially defined schedule.

•An action a_{ji} of some job j_l in *P* is <u>unassigned</u> if it does not appear in *S*.

•An action a_{ji} of some job j_l in *P* is <u>assignable</u> if it has no unassigned predecessors in *S*.

•all predecessors in schedule; action is ready to be executed



Example: Assignable Actions

•problem P:

•machines:

• m_1 of type r_1

• m_2 of type r_2

•jobs:

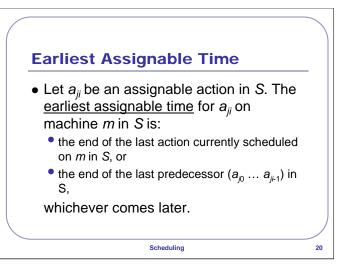
•
$$j_1$$
: $\langle r_1(1), r_2(2) \rangle$
• j_2 : $\langle r_1(3), r_2(1) \rangle$
• j_2 : $\langle r_1(3), r_2(1), r_1(3) \rangle$

•[figure]

•unassigned:

•a₂₂, a₃₁, a₃₂, a₃₃ •assignable:

•*a*₂₂, *a*₃₁



Earliest Assignable Time

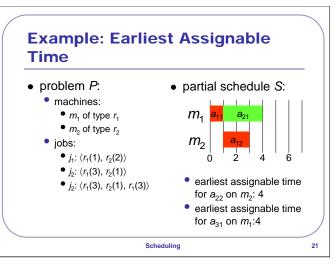
•Let a_{ji} be an assignable action in S. The <u>earliest assignable</u> <u>time</u> for a_{ji} on machine *m* in S is:

•the end of the last action currently scheduled on *m* in S, or

•the end of the last predecessor $(a_{j0} \dots a_{ji-1})$ in S,

•whichever comes later.

•note: assignment not necessarily optimal!



Example: Earliest Assignable Time

•problem P:

•machines:

• m_1 of type r_1

• m_2 of type r_2

•jobs:

•
$$j_1$$
: $\langle r_1(1), r_2(2) \rangle$
• j_2 : $\langle r_1(3), r_2(1) \rangle$
• j_2 : $\langle r_1(3), r_2(1), r_1(3) \rangle$

•[figure]

•earliest assignable time for a_{22} on m_2 : 4 •earliest assignable time for a_{31} on m_1 :4



- Heuristic Search
- heuristicScheduler(P,S)
- assignables ← P.getAssignables(S)
- if assignables.isEmpty() then return S
- action ← assignables.selectOne()
- machines ← P.getMachines(action)
- machine ← machines.selectOne()
- *time* ← S.getEarliestAssignableTime(*action*, *machine*)
- S ← S + assign(*action*, *machine*, *time*)
- return heuristicScheduler(P,S)



Using Local Search

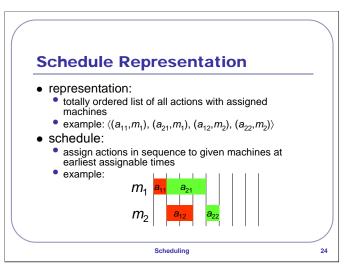
•issues:

•representing schedules

•generating a random initial schedule

•generating neighbours

•evaluating neighbours (schedules)



Schedule Representation

•representation:

•totally ordered list of all actions with assigned machines

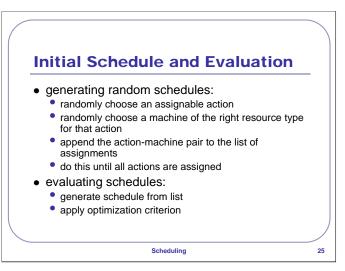
•example: $\langle (a_{11}, m_1), (a_{21}, m_1), (a_{12}, m_2), (a_{22}, m_2) \rangle$

•schedule:

•assign actions in sequence to given machines at earliest assignable times

•example:

•[figure]



Initial Schedule and Evaluation

•generating random schedules:

•randomly choose an assignable action

•randomly choose a machine of the right resource type for that action

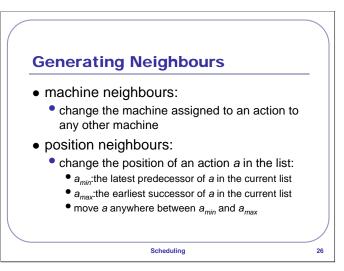
•append the action-machine pair to the list of assignments

•do this until all actions are assigned

•evaluating schedules:

•generate schedule from list

apply optimization criterion



Generating Neighbours

•machine neighbours:

•change the machine assigned to an action to any other machine

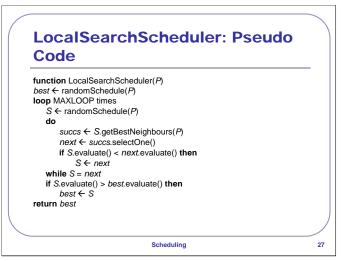
•position neighbours:

•change the position of an action *a* in the list:

•*a_{min}*:the latest predecessor of *a* in the current list

•*a_{max}*:the earliest successor of *a* in the current list

•move *a* anywhere between *a_{min}* and *a_{max}*



•LocalSearchScheduler: Pseudo Code

•function LocalSearchScheduler(P)

best ← randomSchedule(P)

•will contain best schedule found

loop MAXLOOP times

•S ← randomSchedule(P)

•best schedule for local search

•do

•succs ← S.getBestNeighbours(P)

•returns set of neighbours with highest value for evaluation function

next ← succs.selectOne()

•randomly select a (best) neighbour

•if S.evaluate() < next.evaluate() then

•S ← next

•remember best local neighbour

•while S = *next*

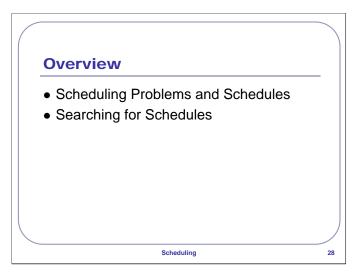
•stop local search when no uphill move possible

•if S.evaluate() > best.evaluate() then

•best ← S

•remember best overall

•return best



Overview

•Scheduling Problems and Schedules

Searching for Schedules

➡just done: search algorithms that generate schedules