Mathematical Tools for Engineering and Management

Lecture 13

25 Jan 2012





- ▷ Models, Data and Algorithms
- ▷ Linear Optimization
- Mathematical Background: Polyhedra, Simplex-Algorithm
- Sensitivity Analysis; (Mixed) Integer Programming
- ▷ MIP Modelling

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- ▷ MIP Modelling: More Examples; Branch & Bound
- > Cutting Planes; Combinatorial Optimization: Examples, Graphs, Algorithms
- ▷ TSP-Heuristics
- ▷ Network Flows
- Shortest Path Problem
- Complexity Theory
- Nonlinear Optimization
- ▷ Scheduling
- ▷ Lot Sizing & Intro to Multiobjective Optimization (Feb 01)
- ▷ Summary (Feb 08)
- ▷ Oral exam (Feb 15)



























Job 1: Book 200 pages, 500 copies 3h printing time

Jobs

Job 2: Book 60 pages, 2500 copies 4h printing time

Job 3: Thesis 170 pages, 10 copies 1h printing time













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▷ Determine an optimal order for the jobs to be processed...











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 - ➡ ...if jobs have to be finished at a given time













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 - ➡ ...if some jobs are more important than others













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- > Determine an optimal order for the jobs to be processed...
 - ➡ ...if jobs have to be finished at a given time
 - ...if some jobs are more important than others
 - ➡ ...if there is more than one machine (identical or different machines)





 \triangleright

Jobs

















$\triangleright~~{\sim}1500$ compute nodes with ${\sim}13000$ cores











- ho \sim 1500 compute nodes with \sim 13000 cores
- ▷ Schedule computation jobs...
 - ...consisting of thousands of parallel processes
 - ...according to their release times













▷ Schedule (Gantt chart):







Machine			
		time	
GPE	·····		

B



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- $\bullet \quad \textbf{Completion time} \quad C_j := s_j + p_j$



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$$\bullet \quad \textbf{Completion time} \quad C_j := s_j + p_j$$

• Average completion time for n jobs:

$$\frac{1}{n}\sum_{j=1}C_j$$

-

n





 \triangleright For fixed number *n* of jobs: minimize sum of completion times











Machine



14



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 \triangleright For fixed number *n* of jobs: minimize sum of completion times







$$\Rightarrow \sum_{j=1} C_j = 16 + 26 + 32 + 54 + 66 + 80 + 95 + 115 + 129 = 613$$









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Idea: Schedule jobs in order of non-decreasing processing time!







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▷ Example schedule:

0

For fixed number n of jobs: minimize sum of completion times

10

↑

16

6

↑

32

↑

26

12

44

22

14

↑

80

↑

66

15

↑

95

 \triangleright

Machine

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n

j=1

 $\sum C_j$

20

 \uparrow

115

129



For fixed number n of jobs: minimize sum of completion times

n






\triangleleft	Minimize makespan													
\triangleright	Latest completion time \Rightarrow minimize makespan $\max_{j=1,,n} C_j$													
\triangleright	Example schedule:													
Ma	hine 16 10 6 22 12 14 15 20 14													





\triangleleft	Minimize makespan													
\triangleright	Latest com	pletion	time ➡	minimiz	e mak	espan	$\max_{j=1,\dots,n}$	C_{j}						
\triangleright	Example so	chedule:					-							
M	achine 16	1	0 6	22	12	1	4	15	20	14				
	↑ 0	↑ 16	↑ ↑ 26 32		↑ 54	↑ 66	↑ 80	↑ 95		↑ 115	↑ 129			





\triangleleft	Minimize makespan													
	⊳ Lat	est com	pletion	time	⇒ r	ninimiz	e ma	kespan	$\max_{i=1,\ldots}$	$\sum_n C_j$				
	⊳ Exa	ample sc	hedule	:					3)	,				
	Machine	16		10 6		22	1	.2	14	15	20		14	
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➡ makespan: 129





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depends only on the input (processing times), not on the schedule itself





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• Start time of job j cannot be before its release date







 \blacksquare Start time of job j cannot be before its release date





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 \triangleright Jobs can have: a release date r_j



- Start time of job j cannot be before its release date
 - → Constraint: $s_j \ge r_j$















▷ Minimize makespan







- ▷ Minimize makespan
- Optimal algorithm: schedule jobs in the order of non-decreasing release dates





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- ...have both due dates and release dates





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 - minimize maximum lateness : schedule jobs in order of non-decreasing due dates
- ...have both due dates and release dates
 - → more complicated (\mathcal{NP} -hard)
- ...be allowed to be interrupted (possibly at additional cost/time)
 - ➡ easier if no cost/time involved, harder otherwise
- ...consume resources
 - ➡ Resource-constrained scheduling





▷ Single Machine Scheduling: only one machine available





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- ➡ With release dates, greedy strategy only works for makespan minimization
- ▷ Summary if no interruptions and resources are involved:

		Objective										
	$\sum C_j$	$\max C_j$	lateness									
no release dates	non-decreasing process times	trivial	non-decreasing due dates									
with release dates	\mathcal{NP} -hard	non-decreasing release dates	\mathcal{NP} -hard									





⊳ Example:

job j	1	2	3	4	5	6	7	8	9	10	11	12	13	14
process time p_j	5	6	9	12	7	12	10	6	10	9	7	8	7	5





▷ Example:

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➡ Order by non-decreasing process times:




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➡ Order by non-decreasing process times:







⊳ Example: v	vith p	reced	ence	const	raint	S	P	roject	sche	duling				
job j	1	2	3	4	5	6	7	8	9	10	11	12	13	14
process time p_j	5	6	9	12	7	12	10	6	10	9	7	8	7	5
preceded by	-	1	1	2	3	3	4	5,6	5,6	7	8,9	10,11	11	12,13
Machine 1	14 2	8	5	11 1	.3 1	.2	3	10	7	9	4	6		
\uparrow													\rightarrow	
0												ma	$\operatorname{x} C_j$	





▷ Example: with precedence constraints ➡ Project scheduling														
job j	1	2	3	4	5	6	7	8	9	10	11	12	13	14
process time p_j	5	6	9	12	7	12	10	6	10	9	7	8	7	5
preceded by	-	1	1	2	3	3	4	5,6	5,6	7	8,9	10,11	11	12,13





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B



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Schedule is infeasible!

▷ Feasible schedule:



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preceded by		1	1	2	3	3	4	5,6	5,6	7	8,9	10,11	11	12,13
1	14 2	8	5	11 1	3 1	2	2	10	7	Q	Д	6		



- ➡ Schedule is infeasible!
- ▷ Feasible schedule:































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▷ Greedy strategy: schedule an arbitrary job next with already fulfilled precedences







▷ Greedy strategy: schedule an arbitrary job next with already fulfilled precedences

Polynomial runtime







- ▷ Greedy strategy: schedule an arbitrary job next with already fulfilled precedences
 - ➡ Polynomial runtime ➡ Efficient algorithm!







> Suppose there are arbitrarily many machines available





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- ▷ Example: Project scheduling on construction site







- > Suppose there are arbitrarily many machines available
 - All jobs with fulfilled precedences can be carried out immediately and parallely
- ▷ Example: Project scheduling on construction site
 - Different tasks done by different contractors:
 Concrete builder, stonemasonry, house painter, glazier, ...
 - can provide as many workers as necessary to carry out each task













S



































































































- ▷ Forward procedure: compute earliest possible completion times for all jobs
 - ➡ Makespan is the maximal earliest possible completion time computed







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polynomial

▷ A critical path is a chain of critical jobs, starting at time 0 and ending at the makespan







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•••••

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▷ Project scheduling: minimize makespan with single machine





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 - Efficiently solvable (by greedy algorithm)





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- ▷ Minimize sum of completion times





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 - Critical Path Method





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- ▷ Except: unlimited number of machines
 - Critical Path Method
- ▷ Summary:

	Objective	
	$\sum C_j$	$\max C_j$
single machine	$\mathcal{NP} ext{-hard}$	polynomial (greedy algorithm)
\ge 2 machines	$\mathcal{NP} ext{-hard}$	$\mathcal{NP} ext{-hard}$
unlimited machines	?	polynomial (critical path method)













▷ Minimize sum of completion times







▷ Minimize sum of completion times









▷ Minimize sum of completion times



$$\sum_{j=1}^{n} C_j = 6 + 20 + 36$$

= 10 + 24 + 44
= 12 + 28 + 50
= 230





▷ Jobs can be carried out on one of 3 identical machines



▷ Minimize sum of completion times



$$\sum_{j=1}^{n} C_j = 6 + 20 + 36$$

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= 230

➡ Optimal: schedule by non-decreasing processing times, on earliest available machine





▷ Minimize sum of completion times: polynomial (greedy algorithm)





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- ▷ Minimize sum of completion times: polynomial (greedy algorithm)
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- ▷ Variants: types of machines
 - Identical machines





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 - Uniform machines: machines differ by a fixed speed factor





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 - Uniform machines: machines differ by a fixed speed factor
 - Unrelated machines: processing times differ for every job on each machine





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- \triangleright All the other additional features: weights, release dates, precedence constraints, ...





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- > Multi-operation models : job has to be processed sequentially on multiple machines
 - **Open shop** : order in which jobs pass through machines is unimportant
 - Flow shop : each job has the same machine order (A, B, ...)





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- \triangleright All the other additional features: weights, release dates, precedence constraints, ...
- > Multi-operation models : job has to be processed sequentially on multiple machines
 - **Open shop**: order in which jobs pass through machines is unimportant
 - Flow shop : each job has the same machine order (A, B, ...)
 - Job shop : each job can have a different machine order




- ▷ Minimize sum of completion times: polynomial (greedy algorithm)
- \triangleright Minimize makespan: \mathcal{NP} -hard
- ▷ Variants: types of machines
 - Identical machines
 - Uniform machines: machines differ by a fixed speed factor
 - Unrelated machines: processing times differ for every job on each machine
- \triangleright All the other additional features: weights, release dates, precedence constraints, ...
- > Multi-operation models : job has to be processed sequentially on multiple machines
 - **Open shop**: order in which jobs pass through machines is unimportant
 - Flow shop : each job has the same machine order (A, B, ...)
 - Job shop : each job can have a different machine order
 - ➡ Makespan minimization for job shop scheduling can also be solved using networks



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- ▷ Models, Data and Algorithms
- ▷ Linear Optimization
- Mathematical Background: Polyhedra, Simplex-Algorithm
- Sensitivity Analysis; (Mixed) Integer Programming
- ▷ MIP Modelling

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- ▷ MIP Modelling: More Examples; Branch & Bound
- > Cutting Planes; Combinatorial Optimization: Examples, Graphs, Algorithms
- ▷ TSP-Heuristics
- ▷ Network Flows
- Shortest Path Problem
- Complexity Theory
- Nonlinear Optimization
- \triangleright Scheduling
- ▷ Lot Sizing & Intro to Multiobjective Optimization (Feb 01)
- ▷ Summary (Feb 08)
- ▷ Oral exam (Feb 15)



