COMPUTING NEAR-OPTIMAL SCHEDULES JOB SHOP SCHEDULING BY LOCAL SEARCH

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Computing near-optimal schedules

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For many NP-hard optimization problems there are polynomial-time algorithms for finding solutions that are provably quite close to the optimum. For others certain performance guarantees are unlikely to be attained, in the sense that if there is such a good algorithm, then P = NP. We survey a number of positive and negative results on computing near-optimal solutions for machine scheduling problems, with an emphasis on multiprocessor scheduling and shop scheduling.

Job shop scheduling by local search

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The job shop scheduling problem is one of the most difficult problem types in combinatorial optimization. Even relatively small instances of the problem are hard to solve to optimality. Recently some progress has been made in finding good approximations of the optimum by a variety of local search techniques, such as iterative improvement, simulated annealing, tabu search, variable-depth search, and genetic algorithms. We survey this work.

SHORT SCHEDULES

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LESLIE HALL
HAN HOOGEVEEN
COR HURKENS
ALEXANDER RINNOOY KAN
SERGEY SEVAST'JANOV
DAVID SHMOYS
ÉVA TARDOS
BART VELTMAN
DAVID WILLIAMSON

BIN PACKING

n ITEMS OF SIZE $a_1, a_2, ..., a_n > 0$ BINS OF CAPACITY b > 0PACK ALL ITEMS IN MINIMUM # BINS

∃ PACKING IN 2 BINS ? IS NO-COMPLETE [KARP 1972]

POLYNOMIAL ALGORITHM A

WITH \forall INSTANCE I: $\frac{A(I)}{OPT(I)} < \frac{3}{2}$ UNLESS $\mathcal{P} = \mathcal{NP}$

SUPPOSE 3A

• OPT(I) $\leq 2 \Rightarrow A(I) < \frac{3}{2} OPT(I) \leq 3 \Rightarrow A(I) \leq 2$ • OPT(I) $\geq 3 \Rightarrow A(I) \geq 3$

 \Rightarrow A ANSWERS 2-BIN QUESTION IN POLYNOMIAL TIME $\Rightarrow \mathcal{S} = \mathcal{N}\mathcal{S}$

 $FFD(I) \leq \frac{11}{9} OPT(I) + 4 [JOHNSON 1976]$

+ 3

[BAKER 1985]

[YUE 1990]

COMBINATORIAL OPTIMIZATION PROBLEM

- CLASS OF INSTANCES
- POLYNOMIAL ALGORITHM:
 INSTANCE I & OBJECT F
 → IS F A FEASIBLE SOLUTION FOR I?
- POLYNOMIAL ALGORITHM:
 INSTANCE I & FEASIBLE SOLUTION F
 → NONNEGATIVE & INTEGRAL VALUE OF F
- INSTANCE I

 ?> FEASIBLE SOLUTION OF MINIMUM VALUE OPT (I)

ALGORITHM A:

INSTANCE I -> FEASIBLE SOLUTION OF VALUE A(I)

PERFORMANCE RATIO OF A:

$$R(A) = \inf\{r \mid r \geqslant 1, \forall I : \frac{A(I)}{OPT(I)} \le r\}$$

A IS POLYNOMIAL Y-APPROXIMATION ALGORITHM IF

- $R(A) \leq r$
- · A RUNS IN TIME POLYNOMIAL IN II

{A_r} IS POLYNOMIAL APPROXIMATION SCHEME IF \$\forall r > 1

- $R(A_r) \leq r$
- · A, RUNS IN TIME POLYNOMIAL IN II

{Ar} IS FULLY FOLYNOMIAL APPROXIMATION SCHEME IF \$7>1

- · R(An) < r
- · AT RUNS IN TIME POLYNOMIAL IN [I] & T-1

IMPOSSIBILITY THEOREM

IF JCEN

∃ FEASIBLE SOLUTION OF VALUE ≤ c?

THEN

 $\frac{1}{2}$ POLYNOMIAL ALGORITHM A WITH R(A) < $\frac{c+1}{c}$ UNLESS $\mathcal{S} = \mathcal{N} \mathcal{S}$

SUPPOSE 3A

• OPT(I) $\leq c \Rightarrow A(I) < \frac{c+1}{c} OPT(I) \leq c+1 \Rightarrow A(I) \leq c$

• OPT(I) ≥ C+1 ⇒

 $A(I) \geqslant c$

→ A ANSWERS ≤ C-QUESTION IN POLYNOMIAL TIME

⇒ 8=N8

$$! \le c \implies \notin \frac{c+1}{c}$$

BIN PACKING

$$! \le 2 [KARP 1972] \implies \neq \frac{3}{2}$$

SYMMETRIC TSP

GRAPH G = (V, E), $d: E \rightarrow \mathbb{N} \cup \{0\}$ FIND HAMILTON CYCLE OF MINIMUM TOTAL WEIGHT $! \leq 0 \text{ [KARP 1972]} \implies \not r \text{ FOR ANY } r > 1$

MULTIPROCESSOR SCHEDULING

m IDENTICAL MACHINES My, ..., Mm n INDEPENDENT JOBS J, ..., Jn PROCESSING J; REQUIRES TIME P; E IN

SCHEDULE = ASSIGNMENT OF EACH JOB TO A MACHINE LENGTH OF SCHEDULE = max; ∑_{Ji→M}; Pj

FIND SCHEDULE OF MINIMUM LENGTH

$$R(LS) = 2 - \frac{1}{m}$$

[GRAHAM 1966]

$$R(LPT) = \frac{4}{3} - \frac{1}{3m}$$

[GRAHAM 1969]

PAS

[HOCHBAUM & SHMOYS 1987]

NO FPAS UNLESS P=WP [GAREY & JOHNSON 1978]

- DEPENDENT JOBS PRECEDENCE RELATION ON JOB SET
- NONIDENTICAL MACHINES PROCESSING J; ON M; REQUIRES TIME pij

PRECEDENCE-CONSTRAINED SCHEDULING

$$R(LS) = 2 - \frac{1}{m}$$

[GRAHAM 1966]

$$! \le 3 \implies \cancel{4} \stackrel{4}{3}$$

[L& RK 1978]

CLIQUE

GRAPH G= (V, E), k & N

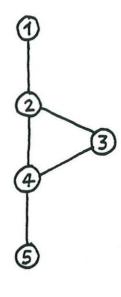
DOES G HAVE A COMPLETE SUBGRAPH ON & VERTICES?

CLIQUE ∝ ≤3

G:

M LARGE ENOUGH

3m UNIT-TIME JOBS



1-	1 2
2	→ 23
3<	\$24
4	34
5	\$45
↑ Y) [E]

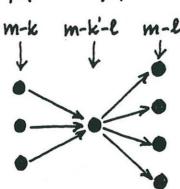
Mm		6	•
•	•	5	•
	•	1	•
	4	34	•
M ₂	3	24	45
M ₁	2	23	12

k=3

 $\ell = \binom{k}{2}$

k' = |V| - k

 $\ell' = |E| - \ell$



∃ CLIQUE OF SIZE k \Leftrightarrow

∃ SCHEDULE OF LENGTH ≤ 3

PRECEDENCE - CONSTRAINED SCHEDULING WITH COMMUNICATION DELAYS

m IDENTICAL MACHINES

n UNIT-TIME JOBS

IF J; → Jk AND J; & Jk ON DIFFERENT MACHINES
THEN UNIT-TIME DELAY BETWEEN J; & Jk

m RESTRICTED

m UNRESTRICTED

* < 3

* ≤ 5

! ≤4 ⇒ ₹=

! ≤ 6 ⇒ ₹ 76

[HOOGEVEEN, L, VELTMAN, 1992]

R(GREEDY) = 3

 $R(LP) = \frac{4}{3}$

[RAYWARD-SMITH]

[MUNIER, KÖNIG]

R_{JOB DUPLICATION} (LS) = 2

[PAPADIMITRIOU, YANNAKAKIS]

FOOTNOTE

m RESTRICTED	GENERAL	TREE	m=2
DELAYS O	! ≤ 3	*	*
DELAYS 1	! < 4	!	?

SCHEDULING UNRELATED MACHINES

$$R(LP+M) = 2$$

[L, SHMOYS, & TARDOS 1990]

$$! \le 3 \Rightarrow 4\frac{4}{3}$$

$$! \le 2 \Rightarrow 4\frac{3}{2}$$

UNRELATED MACHINES

POTTS' LP-BASED ALGORITHM

(1) SOLVE LP

INPUT : Pij = PROCESSING TIME OF JOB j ON MACHINE i

OUTPUT: xi = FRACTION OF JOB j ASSIGNED TO MACHINE i

Z < OPT (= MINIMUM SCHEDULE LENGTH)

MINIMIZE Z

SUBJECT TO $\sum_i \varkappa_{ij} = 1$, j=1,...,n $\sum_j p_{ij} \varkappa_{ij} \leq \mathbf{Z}, \quad i=1,...,m$ $\varkappa_{ij} \geqslant 0$, i=1,...,m, j=1,...,n

-> AT MOST m-1 SPLIT JOBS

- (2) ASSIGN UNSPLIT JOBS BY ROUNDING DOWN LP-OUTPUT:

 PUT FRACTION [xij] OF JOB j ON MACHINE i

 → SCHEDULE OF LENGTH ≤ Z ≤ OPT
- (3) ASSIGN SPLIT JOBS OPTIMALLY BY COMPLETE ENUMERATION:

 CHECK O(mm-1) POSSIBILITIES

-> SCHEDULE OF LENGTH & OPT

R(LP) = 2

RUNNING TIME POLYNCMIAL FOR FIXED M

3DM

≪ 3

3n ELEMENTS → 3n JOBS

- 1 1 1
- 2 2 2
- 3 3 3
- 4 4 4

- 000
- 2 2 2
- 3 3 3
- **(4)**

Pij= {1 IF j E TRIPLE i 3 OTHERWISE

m TRIPLES

-> m MACHINES & m-n DUMMY JOBS

(1111111111)

 $p_{ij} = 3$

- - 1 2 3
 - 141
 - 2 1 3
 - 2 2 2
 - 2 3 4
- ► **② ⊕ ③**
- **▶** ③ ② ①
 - 3 3 4
 - 4 1 2
- **● ④ ③ ④**
 - 4 4 1

- 000
- CHINE INS
- SHIEHIYD.
- CHININED,
- WIIIIII)
- WIIIIII R
- ② ④ ③
- 330
- SHISHIRS,
- CHILLIAN.
- **(4)**
- CHATTED,

3 n TRIPLES

⇔ 3 SCHEDULE

CONTAINING

OF LENGTH < 3

3n ELEMENTS

3DM

oc ≤2

3n ELEMENTS 2n JOBS

4 4

m MACHINES & m-n DUMMY JOBS m TRIPLES

- (2 FOR i OF TYPE I

- 3

2 2

- 4

2

- 2 FOR i OF TYPE 4

- **▶ ④ ③**

3 n TRIPLES

⇔ 3 SCHEDULE

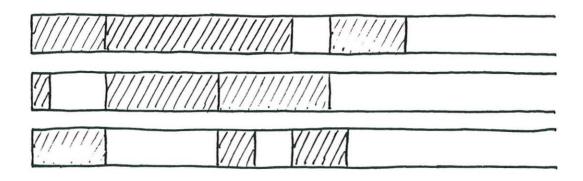
CONTAINING

OF LENGTH & 2

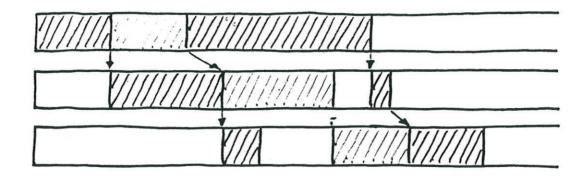
3n ELEMENTS

SHOP SCHEDULING

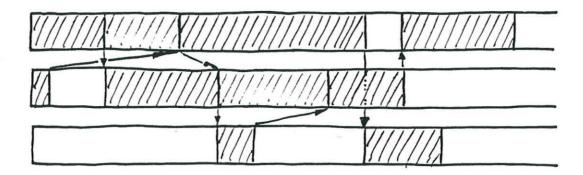
OPEN SHOP



FLIW CHOP



JOB SHOP



- . INTEGRAL PROCESSING TIMES
- · MINIMIZE SCHEDULE LENGTH

OC OPEN SHOP

FC GIVEN AN FLOW SHOP INSTANCE & A CEN,

JC JOB SHOP

DOES THERE EXIST A SCHEDULE OF LENGTH & c?

* 02 * F2 * J2 [FOLKLORE]

* O3 * F3 * J3

[WILLIAMSON.H.H.H.L.S]

IMPOSSIBILITIES (UNLESS P=WP):

- POLYNOMIAL ALGORITHM A WITH R(A) < €
- · POLYNOMIAL APPROXIMATION SCHEME

POSSIBILITIES

O: R(GREEDY) = 2

[RACSMÁNY]

 $J: R(\cdots) = O(\log^2(m \cdot m_{max}))$

[SHMOYS.STEIN.WEIN 1991]

F, m:FIXED: PAS

[HALL 1995]

O, m.FIXED: PAS

[SEVAST'JANOV . WOEGINGER 1996]

JOB SHOP SCHEDULING

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- LIMIT TO APPROXIMABILITY
 Williamson. Hall. Hoogeveen. Hurkens. L. Sevast janov. Shmoys
- + APPROXIMATION BY LOCAL SEARCH Vaessens · Aarts · L

SET OF MACHINES

EACH MACHINE . IS AVAILABLE AT TIME O

• CAN HANDLE ≤ 1 JOB AT A TIME

EACH JOB

. IS A CHAIN OF OPERATIONS

EACH OPERATION REQUIRES UNINTERRUPTED PROCESSING

ON GIVEN MACHINE

FOR GIVEN AMOUNT OF TIME

EXAMPLE: JOB A: 8 -> 3 -> 4

JOB B: $1 \rightarrow 6 \rightarrow 2 \rightarrow 10$

JOB C: 7 → 5 → 9

SCHEDULE = ALLOCATION OF OPERATIONS

TO TIME INTERVALS ON MACHINES

EXAMPLE:	MA	B	C				
	M B	A		С			ennemente se Arribas (
	ML		ВА				
	м				C	В	
	ò						45

OPTIMAL SCHEDULE = SCHEDULE OF MINIMUM LENGTH

DISJUNCTIVE PROGRAM

SET M OF MACHINES

SET J OF JOBS

SET O OF OPERATIONS

OPERATION i & O . BELONGS TO JOB Jie J

· REQUIRES PROCESSING ON MACHINE MIEM

· DURING TIME PIEIN

BINARY RELATION -> ON O

· DECOMPOSING O INTO CHAINS (= JOBS)

FIND STARTING TIMES S; (ieO)

· MINIMIZING

· LENGTH: max iso {5; +p;}

SUBJECT TO

· AVAILABILITY: Si ≥0

 $(i \in \mathcal{O})$

· PRECEDENCE: Si-Si≥ pi

 $(i,j\in\mathcal{O},\ i\rightarrow j)$

• CAPACITY: $S_j - S_i \ge p_i \ V \ S_i - S_j \ge p_j \ (i,j \in O, \ M_i = M_j)$

DISJUNCTIVE GRAPH G = (O, A, E)

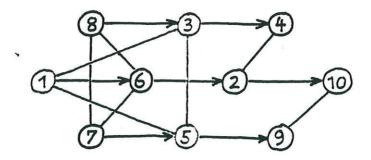
· VERTEX SET O

• ARC SET $A = \{(i,j)|i \rightarrow j\}$... DIRECTED!

• EDGE SET E = {{i,j}|Mi=Mj} ... UNDIRECTED!

• WEIGHT Pi FOR EACH i & O

EXAMPLE:



CAPACITY CONFLICT: EDGE {i,j} & E

SCHEDULING DECISION: REPLACE {i,j} BY (i,j) OR (j,i)

DISJUNCTIVE GRAPH G = (O, A, E)

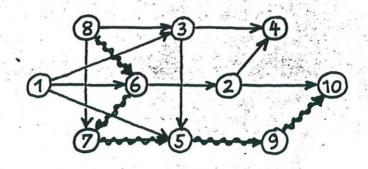
VERTEX SET O

ARC SET $A = \{(i,j)|i \rightarrow j\}$... DIRECTED!

EDGE SET $E = \{\{i,j\} | M_i = M_j\}$... UNDIRECTED!

WEIGHT P: FOR EACH i & O

EXAMPLE:



EDGE {i,j} EE CAPACITY CONFLICT:

SCHEDULING DECISION: REPLACE {i,j} BY (i,j) OR (j,i)

FEASIBLE SCHEDULE:

ORIENTATION E OF E SUCH THAT

DIGRAPH G= (O, AUE) IS ACYCLIC

LENGTH OF SCHEDULE:

LONGEST PATH LENGTH IN G

PROBLEM: FIND ORIENTATION E OF E

THAT MINIMIZES LONGEST PATH LENGTH IN G

POLYNOMIAL

EXPONENTIAL

ALGORITHMS

ALGORITHMS

OPTIMAL

FAST

ENUMERATION:

SOLUTIONS

OPTIMIZATION

DP, B&B

APPROXIMATE

FAST

LOCAL

SOLUTIONS

APPROXIMATION

SEARCH

COMPUTATIONAL COMPLEXITY

POLYNOMIA	AL	TI	ME
-----------	----	----	----

NP-HARD

2 MACHINES

≤2 OPERATIONS/JOB

2 MACHINES

< 3 OPERATIONS/JOB

3 MACHINES

≤ 2 OPERATIONS/JOB

2 MACHINES

ALL pi = 1

2 MACHINES

ALL pi ∈ {1,2}

3 MACHINES

ALL Pi = 1

2 JOBS

3 JOBS

LENGTH < 3

LENGTH ≤ 4

J4 FIND SCHEDULE OF LENGTH & 4

NP-HARD

JE FIND SCHEDULE OF LENGTH < 5 x OPTIMUM NP-HARD

J4 ∝ J = :

SUPPOSE A IS POLYNOMIAL ALGORITHM FOR JE

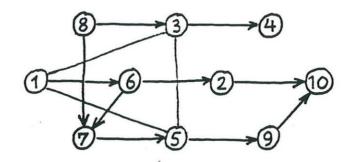
TAKE ANY INSTANCE I

• OPT(1) $\geqslant 5 \Rightarrow$ A(1) $\geqslant 5$ • OPT(1) $\leqslant 4 \Rightarrow$ A(1) $< \frac{5}{4} \cdot OPT(1) \leqslant 5 \Rightarrow$ A(1) $\leqslant 4$

→ A IS POLYNOMIAL ALGORITHM FOR J+

BRANCH & BOUND

NODE: SOME E'CE HAS ORIENTATION E'



LOWER BOUNDS

- LBO . IGNORE EDGES IN E-E'
 - . COMPUTE LONGEST PATH LENGTH IN (O, AUE')
- LB1 . CHOOSE M*
 - . IGNORE EDGES IN E-E' NOT ON M*
 - · SOLVE 1-MACHINE PROBLEM WITH FOR EACH i ON M*:
 - . HEAD Ti = LONGEST PATH LENGTH UP TO i
 - . BODY P: = PROCESSING TIME
 - . TAIL q' = LONGEST PATH LENGTH FROM i

STRENGTHEN LB1

• PRECEDENCE CONSTRAINTS: 1-

• PRECEDENCE DELAYS:

- ADJUSTED HEADS & TAILS: rj = max (8+7, 1+6+7) → rj = 8+6.
- · L3k, k>1

SURROGATE DUALITY BOUNDS

POLYHEDRAL BOUNDS

LOWER BOUND VALUES

FOR 10×10 INSTANCE [Fischer, Thompson, 1963]

OPTIMUM 930

[Carlier, Pinson, 1989]

LB1

808

[McMahon, Florian, 1975]

LB1 WITH ADJUSTED HEADS & TAILS [Cartier, Pinson, 1994] 868

LB5

907

[Lageweg, 1984]

SURROGATE DUALITY : [Fisher, Lageweg, L, Rinnowy Kan, 1983

813 >1 HR

POLYHEDRAL BOUNDS [Applegate, Cook, 1991]

• CUTS 1 823 5 SEC

• CUTS 2 824 5 MIN

• CUTS 3 827 > 2 HR

BRANCHING RULES

- (a) GENERATE "ACTIVE SCHEDULES"
- (4) ORIENT CRUCIAL EDGE
- (c) APPLY 'BLOCK APPROACH'

IMPLEMENTATIONS

- . UB: APPROXIMATION ALGORITHM, E.G., SHIFTING BOTTLENECK
- . LB: PREEMPTIVE LB1
- · BRANCHING: (4) OR (c)
- . ELIMINATION CRITERIA: MANY!

RESULTS FOR 10×10 INSTANCE

- 22021 NODES, 300 MIN [Carlier, Pinson, 1989]
- 16055 NODES, 6 MIN [Applegate, Cook, 1991]
- · 4242 NODES, 19 MIN [Brucker, Jurisch, Sievers, 1992]
- 37 NODES, 8 MIN [Carlier, Pinson, 1994]

MAJOR ISSUE

• FIND BETTER (LP-BASED?) LOWER BOUNDS

APPROXIMATION ALGORITHMS: CONSTRUCTION

PRIORITY RULE

· SCHEDULE OPERATIONS ACCORDING TO SOME PRIORITY FUNCTIO

ACTIVE SCHEDULES

DEFINITION: A SCHEDULE IS ACTIVE'
IF MOVING BACK ONE OPERATION WILL DELAY ANOTHER ONE.

THEOREM: AT LEAST ONE OPTIMAL SCHEDULE IS ACTIVE.

CONSTRUCTION :

- LET S; = EARLIEST POSSIBLE STARTING TIME OF i (i & 0)

 LET O' = SET OF UNSCHEDULED OPERATIONS
- DETERMINE $j \in O'$ SUCH THAT $S_{j}+p_{j} = \min_{i \in O'} \{S_{i}+p_{i}\}$
- SELECT OPERATION FROM SET

 {i|ie0', Mi=Mj, Si < Sj+pj}

 ACCORDING TO SOME PRIORITY FUNCTION

LOCAL SEARCH

F : SET OF FEASIBLE SOLUTIONS 2

c(x) & R : COST OF x

N(x) CF : NEIGHBORHOOD OF &

ITERATIVE IMPROVEMENT

- · GENERATE X & F
- · AS LONG AS ∃y ∈N(x) WITH c(y) < c(x), DO x ← y
- . IS LOCALLY OPTIMAL WITH RESPECT TO N

NEW VARIANTS

- . SIMULATED ANNEALING
- · THRESHOLD ACCEPTANCE
- · TABU SEARCH
- VARIABLE-DEPTH SEARCH
- . GENETIC ALGORITHMS

. . . .

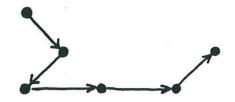
HYBRID VARIETIES

LOCAL SEARCH COMBINED WITH

- · CONSTRUCTIVE RULE
- · OTHER LOCAL SEARCH METHOD
- . BACKTRACKING SCHEME

SMALL-CHANGE NEIGHBORHOODS

S SWAP: REVERSE MACHINE ARC ON LONGEST PATH



- . SWAP ELSEWHERE WILL NOT GIVE IMPROVEMENT
- . SWAP ON LONGEST PATH WILL NOT CREATE CYCLE
- · VSCHEPULE 3 PATH TO OPTIMAL SCHEDULE

S+ SWAP CRITICAL ARC O AND ARCS -t & 1



- J JUMP: REINSERT OPERATION ON LONGEST PATH ELSEWHERE
- P PERMUTE 3 OR MORE OPERATIONS ON LONGEST PATH

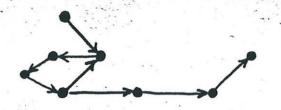
BIG - CHANGE NEIGHBORHOODS

R1 RESCHEDULE 1 MACHINE ... BY COMPUTING LB1

Rt RESCHEDULE & MACHINES ... BY JOB SHOP SCHEDULING

SMALL-CHANGE NEIGHBORHOODS

S SWAP: REVERSE MACHINE ARC ON LONGEST PATH



- . SWAP ELSEWHERE WILL NOT GIVE IMPROVEMENT
- . SWAP ON LONGEST PATH WILL NOT CREATE CYCLE
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S+ SWAP CRITICAL ARC O AND ARCS -t & 1



J JUMP: REINSERT OPERATION ON LONGEST PATH ELSEWHERE

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BIG - CHANGE NEIGHBORHOODS

R1 RESCHEDULE 1 MACHINE ... BY COMPUTING LB1

Rt RESCHEDULE & MACHINES ... BY JOB SHOP SCHEDULING

SHIFTING BOTTLENECK

[Adams, Balas, Zawack, 1988]

FOR k - 1 TO m DO:

· CONSTRUCTION :

SCHEDULE MACHINE MAXIMIZING LB1 AMONG UNSCHEDULED MACHINES

• ITERATIVE IMPROVEMENT USING R1:

REOPTIMIZE PARTIAL SCHEDULE BY RESCHEDULING 1 MACHINE AT A TIME

VARIATIONS

- SB + DELAYS

 [Balas, L, Vazacopoulos, 1995]

 SCHEDULE MACHINE MAXIMIZING LB1 INCORPORATING PRECEDENCE DELAYS
- SB + GUIDED LOCAL SEARCH [Balas, Vazacopoulos, 1994]

 REOPTIMIZE BY YARIABLE-DEPTH SEARCH USING JUMPS

HYBRIDS

- PARTIAL ENUMERATION [Adams, Balas, Zawack, 1988]

 SUB-BRANCH & SUPER-BOUND, APPLYING SB IN EACH NODE
- BOTTLE-t [Applegate, Cook, 1991]

 FOR k > m-t: BRANCH, CUNSIDERING EACH MACHINE IN TURN
- · SHUFFLE [Applegate, Cook, 1991]
 - . CONSTRUCT SCHEPULE USING BOTTLE-5
 - · ITERATIVELY IMPROVE IT USING RE

RESULTS

SB> {SB. DELAYS, PE, BOTTLE-4,5,6} > SHUPFLE >> SB. GLS

- II ITERATIVE IMPROVEMENT [Aarts, vlaarhoven, L, Ulder, 1994] ACCEPT y IF C(y) C(z) < 0
- TA THRESHOLD ACCEPTING [Aarts, vlaarhoven, L, Ulder, 1994]

 ACCEPT y IF c(y)-c(x) < t WITH to
- SA SIMULATED ANNEALING [Vlaarhoven, Aarts, L, 1992]

 ACCEPT y IF c(y)-c(x) <0 OR WITH PROBABILITY \$ 0

SA!! BI-LEVEL VARIANT

[Matsuo, Suh, Sullivan, 1983]

- · ACCEPT Y AS IN SA.
- OTHERWISE, SUBJECT y TO !! TO OBTAIN Z ACCEPT Z AS IN !!

RESULTS

. TIME EQUIVALENT

11s > 11s+ >> TAs >> {PE, SAs, SAs+} > SHUFFLE > SA!! S+

. TIME OF NO CONCERN

SHUFFLE >> SAs(∞)

CONCLUSIONS

- · RANDOMIZATION HELPS
- . EXPLOITING PROBLEM STRUCTURE HELPS

TABOO SEARCH

- SELECT BEST NEIGHBOR

 (UNLESS IT IS FORBIDDEN (UNLESS IT IS GOOD ENOUGH))
- MAINTAIN FORBIDDEN SET
 (OFTEN DEFINED IN TERMS OF FORBIDDEN MOVES)

TS_S SWAP CRITICAL ARC [Barnes, Chambers, 1995]

TS_P PERMUTE MORE CRITICAL OPERATIONS [Pell'Amico, Trubian, 1993]

TS_SBT BACKTRACKING TO REJECTED MOVES [Nowicki, Smutnicki, 1995]

 $TS_S > SA_S(\infty) > TS_P > TS_SBT$

VARIABLE-DEPTH SEARCH

- X → CHAIN OF SOLUTIONS, MAKING SMALL GREEDY MOVES
- · y & BEST SOLUTION IN CHAIN
- GLS, GUIDED LOCAL SEARCH [Balas, Vazacopoulos, 1994]
 - . TREES INSTEAD OF CHAINS, BOUNDED BY HEURISTIC RULES
 - · MOVES ARE JUMPS
- IGLS ITERATED GLS

 SB.GLS → [GLS ON m-1 MACHINES →

 GLS ON M MACHINES][∞]
- RGLS-K REITERATED GLS

 IGLS → [GLS ON m-Vm MACHINES →

 SB TO ADD Vm MACHINES →

 IGLS ON m MACHINES]^k

GLS > IGLS > TSBT > RGLS-5; TSBT IS FASTER

GENETIC ALGORITHMS

- · CHOOSE POPULATION OF SOLUTIONS, SPLIT IN PAIRS
- . FOR EACH PAIR, GENERATE TWO HYPERNEIGHBORS
- . REDUCE POPULATION TO ORIGINAL SIZE

VARIETY OF HYPERNEIGHBORHOODS ALLOWING USE OF

- . STRING REPRESENTATIONS
- . MUTATIONS
- · CROSSOVERS

BILEVEL VARIANTS
SUBJECTING HYPERNEIGHBURS TO, E.G., II

RESULTS

- . OFTEN POOR
- . AT BEST IN RANGE [SB, SHUFFLE]

CONSTRAINT SATISFACTION

SOLVE FEASIBILITY PROBLEM BY TREE SEARCH USING

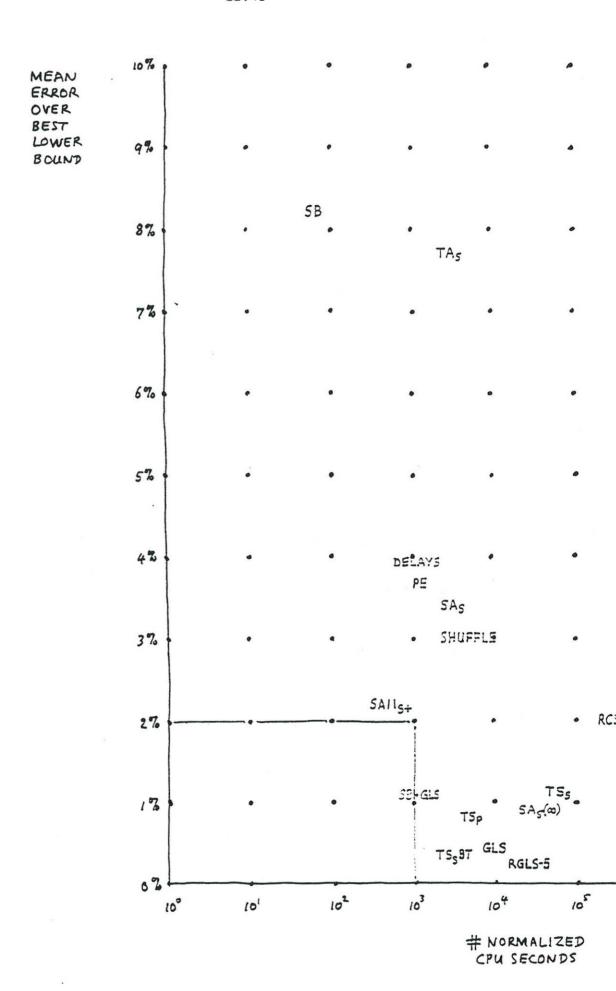
- . [BRANCH] VARIABLE AND VALUE SELECTION
- · [BOUND] CONSISTENCY ENFORCING
- COMPLETE ENUMERATION TRADITION OF LOGIC PROGRAMMING
- -> COMBINATION WITH TECHNIQUES FROM MATH. PROGRAMMING

RESULTS

RANDOMIZED RCS IS OK BUT SLOW

NEURAL NETWORKS

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DISCUSSION

Rapporteur: Martin Beet

Lecture One

Referring to the TSP problem mentioned in the talk, dr Arne Andersson inquired about the alleged success of the neural network approach compared to the use of linear programming or other techniques. Professor Lenstra was somewhat sceptical about this, and replied that in his opinion this kind of combinatorial problem did not lend itself to effective solution by neural nets. In his view many researchers were uncertain about the great power and wideranging applicability of linear programming.

Professor Nievergelt inquired about scheduling research concerning the problem of disturbed schedules, as might be encountered in transport management; for example how to handle a delayed train or airline flight while causing as little disturbance to the remaining schedule as possible. Professor Lenstra stated that this robustness of a schedule seemed to be highly dependent on the initial schedule, but that most researchers in scheduling had concentrated on static problems. Mr Ainsworth supported this view. Both agreed on the importance and the demand for techniques to solve these problems.

Lecture Two

In answering a query from the audience, Professor Lenstra stated that the comparison of the running time of different local-search algorithms relied solely on empirical findings, as a unifying theory was not available.

Professor Mehlhorn was curious as to why the linear programming approach was not also used for job shop scheduling (JSS). Professor Lenstra explained that in his experience the LP approach with relaxations did not provide satisfactory results, due to the difficulty of dealing with the disjunction in the formulation of the precedence constraints. He added that local-search algorithms were a valid approach, since there seemed to be little demand for algorithms to solve large-scale JSS problems.