Benchmarks for basic scheduling problems

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Abstract: In this paper, we propose 260 randomly generated scheduling problems whose size is greater than that of the rare examples published. Such sizes correspond to real dimensions of industrial problems. The types of problems that we propose are: the permutation flow shop, the job shop and the open shop scheduling problems. We restrict ourselves to basic problems: the processing times are fixed, there are neither set-up times nor due dates nor release dates, etc. Then, the objective is the minimization of the makespan.

Keywords: Combinatorial optimization; Scheduling; Benchmarks

Introduction

The types of problems discussed in this paper (permutation flow shop, job shop and open shop scheduling problems) have been widely studied in the literature using exact or heuristic methods, but a common comparison base is missing. We hope that this paper will fill a gap in this domain.

The three-field nomenclature described in Lawler et al. [9] names these problems $F \parallel C_{max}$, $J \parallel C_{max}$ and $O \parallel C_{max}$ respectively. They certainly belong to the most studied ones among scheduling problems. Let us describe them briefly.

There are n jobs that have to be performed on m unrelated machines; in our case, every job consists of m nonpreemptible operations. Every operation of a job uses a different machine during a given time and may wait before being processed.

For the permutation flow shop problem, the operations of every job must be processed on machines $1, \ldots, m$ in this order. Moreover, the processing order of the jobs on the machines is the same for every machine. The problem con-

sists in finding a permutation of the n jobs that minimizes the makespan.

In the case of the job shop problem, any processing order of the jobs on the machines is allowed. For every job, the operations must be processed in a given order on the machines, but this order may differ according to the jobs.

For the open shop problem, every operation is assigned to a given machine but the order of the operations of every job is totally free.

The aim of this paper is to present unsolved problems whose size corresponds to the one of industrial problems. These problems must be easy to generate.

Generating interesting problems

As we do not know any method to solve exactly the problems we want to propose, we have used heuristic methods to get hopefully good solutions of these problems. These heuristic methods are based on taboo search techniques. Taboo search is described very generally in Glover [7] and one can find some of its practical applications to the flow shop sequencing problem in Taillard [13] and Widmer and Hertz [14], and to the job shop scheduling problem in Taillard [12].

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Taboo search is very easy to implement and can provides results that are better than those obtained by any other heuristic method described in [9] or in Applegate and Cook [1] if its parameters are well tuned.

In order to propose problems that are as difficult as possible (the most interesting ones) we have generated many instances of problems that we have 'solved' in a summary way with taboo search. Then, we have chosen the 10 problems that seemed to be the hardest ones and we have solved them once more, allowing our heuristic method to perform a higher number of iteration.

Obviously, the choice of the hardest problems is very subjective. We decided that a problem was interesting if the best makespan we found was far from a lower bound of the makespans and if many attempts to solve the problem (starting from various initial solutions) did not provide the same solution. Such a method enabled us to detect the simplest problems but we may not propose problems that have a local optimum with a large attraction basin.

The problems

The problems we propose are randomly generated with a good random number generator proposed in Bratley [4]. We recall its implementation so that this paper is self contained.

A problem will be entirely defined by the initial value of the seed of the random generator and by the way of generating it. For every type of problem, we give a simple manner of computing a lower bound of the makespan; in particular, this permits to verify the generation of the problem. For every size of problem, we give the total number of instances we have generated (summary resolution), the maximum number of iteration of taboo search that were done (long resolution) and the proportion of problems that were solved up to the lower bound, that is to say optimally. For every type and every size of problem, we give 10 instances. For each instance, we give the initial value of the random generator seed, the best value of the makespan we have found (i.e. an upper bound of the optimal makespan) and a lower bound of all the makespans.

The random number generator

Let us recall the implementation of the linear congruential generator we have used which is based on the recursive formula

$$X_{i+1} = (16\,807X_i) \mod(2^{31}-1)$$

This implementation uses only 32-bit integers and provides a uniformly distribution sequence of numbers between 0 and 1 (not contained):

Step 0. Initial seed and constants:

 $X_0 \ (0 < X_0 < 2^{31} - 1);$ $a = 16\,807, \ b = 127\,773, \ c = 2\,836, \ m = 2^{31} - 1.$ *Step 1.* Modification of the seed:

$$k := \lfloor X_j / b \rfloor,$$

$$X_{i+1} := a(X_i \mod b) - kc,$$

If $X_{i+1} < 0$, then let

$$X_{i+1} \coloneqq X_{i+1} + m$$

Step 2. New value of the seed: X_{i+1} . Current value of the generator: X_{i+1}/m .

Below, we shall denote by U(0, 1) the pseudorandom number that this generator provides. We have 0 < U(0, 1) < 1 for every generated number.

We shall denote by U[a, b] (with a < b, a and b integers) the integer number

$$|a + U(0, 1) \cdot (b - a + 1)|$$

For every random integer generated, we have

$$a \leq U[a, b] \leq b$$

and every integer between a and b has the 'same' probability of being chosen. In order to implement the integer random procedure only with 32-bit integers, the problems have been chosen in such a way that one never has to deal with a seed X such that

$$\left\lfloor a + \frac{X \cdot (b - a + 1)}{m} \right\rfloor \neq \left\lfloor a + \frac{X}{\lfloor m/(b - a + 1) \rfloor} \right\rfloor$$

Flow shop problems

There are in the literature some problems of this type; let us quote for example eight small and simple problems proposed in Carlier [5] and solved exactly in this reference. The flow shop problems are characterized by the processing times d_{ij} of job j on machine i $(1 \le i \le m, 1 \le j \le n)$. We have generated the values of d_{ij} by the following way:

For i = 1 to m

for j = 1 to n

 $d_{ij} := U[1, 99].$

We propose problems with 5, 10 and 20 machines and from 20 to 500 jobs. We compute the lower bound of the makespan as presented below.

Let b_i be the minimum amount of time before machine *i* starts to work and a_i be the minimum amount of time that it remains inactive after its work up to the end of the operations, and let T_i be its total processing time. We have:

$$b_i = \min_j \left(\sum_{k=1}^{i-1} d_{kj} \right),$$

$$a_i = \min_j \left(\sum_{k=i+1}^m d_{kj} \right),$$

$$T_i = \sum_{j=1}^n d_{ij}.$$

Clearly, the optimal makespan C_{\max}^* is greater than or equal to the maximum between the mini-

		_								_
Time seed	UB	LB	Γ	Time seed	UB	LB		Time seed	UB	LB
20 jobs	5 ma	chines	Ī	50 jobs	10 ma	chines		100 jobs	20 m	achines
873654221	1278	1232		1958948863	3037	2907		450926852	6330	5851
379008056	1359	1290		575633267	2911	2821		1462772409	6320	6099
1866992158	1081	1073		655816003	2871	2801		1021685265	6364	6099
216771124	1293	1268		1977864101	3067	2968		83696007	6331	6072
495070989	1235	1198		93805469	3011	2908		508154254	6405	6009
402959317	1195	1180		1803345551	3021	2941		1861070898	6487	6144
1369363414	1239	1226		49612559	3124	3062		26482542	6379	5991
2021925980	1206	1170		1899802599	3048	2959		444956424	6514	6084
573109518	1230	1206		2013025619	2910	2795		2115448041	6386	5979
88325120	1108	1082		578962478	3100	3046		118254244	6534	6298
20 jobs	10 ma	chines		50 jobs	20 ma	chines		200 jobs	10 m	achines
587595453	1582	1448		1539989115	3886	3480		471503978	10872	10816
1401007982	1659	1479		691823909	3733	3424		1215892992	10500	10422
873136276	1496	1407		655816003	3673	3351		135346136	10956	10886
268827376	1377	1308		1315102446	3755	3336	1	1602504050	10893	10794
1634173168	1419	1325		1949668355	3648	3313		160037322	10537	10437
691823909	1397	1290		1923497586	3719	3460		551454346	10347	10255
73807235	1484	1388		1805594913	3730	3427		519485142	10882	10761
1273398721	1538	1363		1861070898	3737	3383		383947510	10754	10663
2065119309	1593	1472		715643788	3772	3457		1968171878	10465	10348
1672900551	1591	1356		464843328	3791	3438		540872513	10727	10616
20 jobs	20 ma	chines		100 jobs	5 ma	chines		200 jobs	20 machines	
479340445	2297	1911		896678084	5493	5437		2013025619	11393	10979
268827376	2099	1711		1179439976	5274	5208		475051709	11445	10947
1958948863	2326	1844		1122278347	5175	5130		914834335	11522	11150
918272953	2223	1810		416756875	5018	4963		810642687	11461	11127
555010963	2291	1899		267829958	5250	5195		1019331795	11427	11132
2010851491	2226	1875		1835213917	5135	5063		2056065863	11368	11085
1519833303	2273	1875		1328833962	5247	5198		1342855162	11536	11194
1748670931	2200	1880		1418570761	5094	5038		1325809384	11544	11126
1923497586	2237	1840		161033112	5448	5385		1988803007	11424	10965
1829909967	2178	1900		304212574	5328	5272		765656702	11548	11122
50 jobs	5 ma	chines		100 jobs	10 ma	achines		500 jobs	20 m	achines
1328042058	2724	2712		1539989115	5776	5759		1368624604	26316	25922
200382020	2836	2808		655816003	5362	5345		450181436	26807	26353
496319842	2621	2596		960914243	5679	5623		1927888393	26626	26320
1203030903	2751	2740		1915696806	5820	5732		1759567256	26642	26424
1730708564	2863	2837		2013025619	5491	5431		606425239	26509	26181
450926852	2829	2793		1168140026	5308	5246		19268348	26654	26401
1303135678	2725	2689		1923497586	5600	5523		1298201670	26575	26300
1273398721	2683	2667		167698528	5640	5556	1	2041736264	26794	26429
587288402	2554	2527		1528387973	5891	5779		379756761	26241	25891
248421594	2782	2776		993794175	5856	5830		28837162	26662	26315

Figure 1. Instances of flowshop problems

Table 1

Flow shop problems

mum amount of time required by the machines and the minimum of time required for each job:

$$LB = \max\left\{ \max_{i} (b_{i} + T_{i} + a_{i}), \max_{j} \left(\sum_{i=1}^{m} d_{ij} \right) \right\}$$

$$\leq C_{\max}^{*}.$$

This lower bound is easy to compute and we conjecture that for fixed m

$$\lim_{n\to\infty}\operatorname{Prob}(C^*_{\max}=\operatorname{LB})=1.$$

For every size of problem we give the following information (Table 1):

- Nb jobs: The number of jobs.
- Nb machines: The number of machines.
- Nb instances: The total number of problems generated.
- LB reached: The proportion of problems for which we found a solution for which the makespan was equal to the lower bound (or equal to the lower bound augmented by 2% for the 500-job 20-machine problems).
- Nb iterations: The maximum number of iterations performed by taboo search (long resolution).
- Nb resolutions: The number of attempts to solve the problem from various initial solutions (long resolution).

Then we give ten instances for every size of problem with the following information (Figure 1):

- Time seed: The initial value of the random generator's seed.
- UB: An upper bound of the optimal makespan (the best value we got).

LB: A lower bound of the makespans.

As the aim is to give an upper bound as good as possible but not a fast solving method, the computation time does not have much importance. However, let us mention that an iteration of taboo search needs about $4 \cdot 10^{-6} n^2 m$ seconds on a 'Silicon Graphics' personal workstation (10 Mips).

Job shop problems

In the literature, we may find instances of small problems in Lawrence [10] and Muth and

1 10 11	shop proof	CIIIS				
Nb jobs	Nb ma- chines	Nb in- stances	LB reached (%)	Nb itera- tions	Nb resolu- tions	
20	5	100	35	104	3	
20	10	100	1	104	3	
20	20	100	0	$2 \cdot 10^{4}$	3	
50	5	70	41	$5 \cdot 10^{3}$	3	
50	10	70	3	104	3	
50	20	70	0	$5 \cdot 10^{4}$	3	
100	5	10000	54	$2 \cdot 10^{3}$	4	
100	10	50	6	$2 \cdot 10^{4}$	3	
100	20	50	0	104	3	
200	10	300	28	$4 \cdot 10^{3}$	3	
200	20	25	0	$4 \cdot 10^{3}$	10	
500	20	100	14 ^a	$4 \cdot 10^{3}$	5	

^a The value reached for this size was less than or equal to 1.02 times the lower bound.

Thompson [11]; most of the optimal values of these problems are given in [1] or Carlier and Pinson [6]. Some very good solution values of Lawrence's problems are also given in [12]. We can consider that problems up to ten machines may be solved satisfactorily with existing methods. This is why we propose problems with 15 and 20 machines and from 15 to 100 jobs.

The processing time d_{ij} of the *i*-th operation of job j $(1 \le i \le m, 1 \le j \le n)$ is obtained as follows:

For
$$j = 1$$
 to n

For i = 1 to m

 $d_{ii} \coloneqq U[1, 99].$

The machine M_{ij} on which the *i*-th operation of job *j* has to be performed is given by the following procedure:

Step 0.

$$M_{ii} := i \ (1 \le i \le m, \ 1 \le j \le n).$$

Table 2 Job shop problems

Nb jobs	Nb ma- chines	Nb in- stances	Lb reached (%)	Nb itera- tions	Nb resolu- tions
15	15	50	0	$7 \cdot 10^{5}$	4
20	15	50	4	106	3
20	20	50	0	107	4
30	15	50	26	$3 \cdot 10^{6}$	4
30	20	50	0	$2 \cdot 10^{6}$	4
50	15	100	78	3.106	4
50	20	26	27	5 · 10 ⁵	4
100	20	100	97	$5 \cdot 10^{5}$	3

Step 1.

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For j = 1 to n

For i = 1 to m

Swap M_{ij} and $M_{U[i,m]j}$.

Let us note the use of another initial seed for the choice of the machines: machine seed.

An instance of a small open shop problem, obtained with the same procedures, is given extensively in Table 4.

The lower bound is computed as previously but b_i , a_i and T_i are defined as follows:

$$b_i = \min_j \left(\sum_{k=1}^{k'-1} d_{kj} \right),$$

	Time seed	Machine seed	UB	LB		Time
1	15 jobs	15 machines				30
	840612802	398197754	1231*	1005		1841414
	1314640371	386720536	1252	953		211695
	1227221349	316176388	1223	1036		796392
	342269428	1806358582	1181	973		53249
	1603221416	1501949241	1234	940		202052
	1357584978	1734077082	1243	1134		52444
1	44531661	1374316395	1228	1103		1569394
	302545136	2092186050	1221	980		146026
	1153780144	1393392374	1289	1020		19832
	73896786	1544979948	1261	940		3807
	20 jobs	15 machines		_		50
	533484900	317419073	1376	1254		1
	1894307698	1474268163	1381	1267		66048
	874340513	509669280	1367	1243		35222
	1124986343	1209573668	1355	1329		119751
	1463788335	529048107	1366	1163		137602
	1056908795	25321885	1371	1211		210663
	195672285	1717580117	1480	1306		176535
	961965583	1353003786	1432	1315		110509
	1610169733	1734469503	1361	1202		90724
	532794656	998486810	1373	1213		201163
	20 jobs	20 machines				50
	1035939303	773961798	1663	1217		849
	5997802	1872541150	1626	1314		199192
	1357503601	722225039	1574	1248		34209
	806159563	1166962073	1660	1284	Ì	163404
	1902815253	1879990068	1598	1256		34170
	1503184031	1850351876	1679	1245		32016
	1032645967	99711329	1704	1403		108969
	229894219	1158117804	1626	1387]	43303
	823349822	108033225	1635	1352	í i	61597
	1297900341	489486403	1614	<u>1277</u>	l	23615
	30 jobs	15 machines				100
	98640593	1981283465	1770	1764		30203
	1839268120	248890888	1853	1774		143764
	573875290	2081512253	1855	1733	l	179247
	1670898570	788294565	1851	1828	1	164727
	1118914567	1074349202	2007	1754	ſ	69648
	178750207	294279708	1844	1777		178556
	1549372605	596993084	1822	1771	l	11780
	798174738	151685779	1714	16/3		163915
	553410952	1329272528	1824	1764		200742
	1661531649	1173386294	1723	1608	J	68276

Figure 2. Instances of job shop problems

with
$$k'$$
 such that $M_{k'i} = i$;

$$a_i = \min_j \left(\sum_{k=k'+1}^m d_{kj} \right),$$

with k' such that $M_{k'j} = i$;

$$T_i = \sum_{j=1}^n d_{k'j},$$

with k' such that $M_{k'j} = i$.

We conjecture again that this bound is tight if $n/m \rightarrow \infty$, because we have always found an optimal schedule if $n/m \ge 6$, considering more than 2000 problems whose size was varying from 20

Time seed	Machine seed	UB	LB
30 jobs	20 machines		
1841414609	1357882888	2064	1850
2116959593	1546338557	1983	1761
796392706	1230864158	1896	1710
532496463	254174057	2031	1820
2020525633	978943053	2032	1785
524444252	185526083	2057	1940
1560304601	487269855	1947	1751
1460267840	1631446539	2005	1770
108374872	1937476577	2013	1758
38071822	1541985579	1973	1678
50 jobs	15 machines		
17271	718030	2760*	2760
660481270	449650254	2756*	2756
352220765	949737011	2717*	2717
1107518780	166840558	2830*	2813
1376020303	483922052	2689	2679
2106639239	955932362	2781	2781
1765352082	1209982549	2943*	2943
1105092880	1349003108	2885*	2885
907248070	919544535	2655*	2655
2011630757	1845447001	2723*	2723
50 jobs	20 machines		
8493988	2738939	2921	2868
1991925010	709517751	3002	2848
342093237	786960785	2835	2755
1634043183	973178279	2775	2697
341706507	286513148	2800	2725
320167954	1411193018	2914	2845
1089696753	298068750	2895	2812
433032965	1589656152	2835	2764
615974477	331205412	3097	3071
236150141	592292984	3075	2995
100 jobs	20 machines		
302034063	1203569070	5464 <u>*</u>	5464
1437643198	1692025209	5181	5181
1792475497	1039908559	5568	5552
1647273132	1012841433	5339	5339
696480901	1689682358	5392	5392
1785569423	1092647459	5342	5342
117806902	739059626	5436	5436
1639154709	1319962509	5394	5394
2007423389	749368241	5358	5358
682761130	262763021	5213	5183

Time seed	Machine seed	UB	LB		Time seed	Machine seed	UB	LB
4 jobs	4 machines				10 jobs	10 machines		
1166510396	164000672	193*	186		1344106948	1868311537	645 ^b	637
1624514147	1076870026	236*	229		425990073	1111853152	588 ^{*b}	588
1116611914	1729673136	271*	262		666128954	1750328066	611 ^b	598
410579806	1453014524	250*	245		442723456	1369177184	577 ^{*b}	577
1036100146	375655500	295*	287		2033800800	1344077538	641 ^b	640
597897640	322140729	189*	185		964467313	1735817385	538 ^{*b}	538
1268670769	556009645	201*	197		1004528509	967002400	623	616
307928077	421384574	217*	212		1667495107	818777384	596 ^b	595
667545295	485515899	261	258		1806968543	1561913259	595 ^{*b}	595
35780816	492238933	2 <u>17</u> *	213	1	938376228	344628625	602 ^b	596
5 jobs	5 machines				15 jobs	15 machines		
527556884	1343124817	300*	295		1561423441	1787167667	937 ^{*b}	937
1046824493	1973406531	262*	255		204120997	213027331	918 ^{*b}	918
1165033492	86711717	323 ^{*a}	321		801158374	1812110433	871 ^{*b}	871
476292817	24463110	310*	306		1502847623	1527847153	934 ^{*b}	934
1181363416	606981348	326 ^{*a}	321		282791231	1855451778	950 ^b	946
897739730	513119113	312*	307		1130361878	849417380	933 ^{*b}	933
577107303	2046387124	303 ^{*a}	298	1	379464508	944419714	891 ^{*b}	891
1714191910	1928475945	300*	292		1760142791	1955448160	893 ^{*b}	893
1813128617	2091141708	353	349		1993140927	179408412	908 ^b	899
808919936	1837537 <u>64</u>	326*	321		1678386613	1567160817	902 ^{*b}	902
7 jobs	7 machines				20 jobs	20 machines		
1840686215	1827454623	438	435		957638	9237185	1155 ^{*b}	1155
1026771938	1312166461	449	443		162587311	1489531109	1244 ^b	1241
609471574	670843185	479	468		965299017	1054695706	1257 ^{*b}	1257
1022295947	398226875	467	463		1158457671	14999999517	1248 ^{*b}	1248
1513073047	1250759651	419	416		1191143707	1530757746	1256	1256
1612211197	95606345	460	451		1826671743	901609771	1209 ^b	1204
435024109	1118234860	435	422		1591533998	1146547719	1294 ^{*b}	1294
1760865440	1099909092	426	424		937297777	92726463	1173 ^b	1169
122574075	10979313	460	458		687896268	1731298717	1289 ^{*b}	1289
248031774	1685251301	400	<u>398</u>		<u>687034842</u>	684013066	<u>1241^{™b}</u>	1241

Figure 3. Instances of open shop problems

jobs, 2 machines to 150 jobs, 15 machines, passing by 500 jobs, 4 machines.

The time needed to perform one iteration of taboo search is about $20 \cdot 10^{-6}nm$ seconds on 'Silicon Graphics' workstation. Table 2 and Figure 2 are analogous to Table 1 and Figure 1, but, for job shop problems, we give in addition ma-

chine seed in Figure 2. Stars in Figure 2 indicate optimum values.

Open shop problems

We do not know instances of such problems in the literature. This is why we give problems of



Figure 4. An optimum schedule for the first problem of Table 4

small size. These problems are obtained using exactly the same procedures as those used for the job shop problems, and the lower bound of the makespans corresponds to the maximum amount of time that a job or a machine requires, i.e.

$$\mathbf{LB} = \max\left\{\max_{i}\left(\sum_{j}d_{ij}\right), \max_{j}\left(\sum_{i,k\mid M_{ik}=j}d_{ik}\right)\right\}.$$

Because one has to choose the order of the operations of a job, one can find very often an optimal schedule, except for the problems in which the number of jobs is about the number of machines. In this case, either an optimal solution is easily reached, or the problem is harder than job shop problems of the same size.

For problems with $n \gg m$, we have observed empirically that the mean complexity of taboo search applied to open shop problems, $O(n^{2.37} \cdot m^{3.69})$, is lower than the complexity of taboo search applied to job shop problems, $O(n^{2.50} \cdot m^{3.81})$.

In Table 4, we describe extensively the first 4-job 4-machine problem we propose, i.e. the

```
function unif(var seed: integer; low, high: integer): integer;
const
   m = 2147483647; a = 16807;
   b =127773:
                     c = 2836;
var
   k٠
                integer:
   value_0_1: double; (* floating point coded on 64 bits *)
begin
   k := \text{seed div } b:
   seed := a * (seed \mod b) - k * c;
   if seed < 0 then seed := seed + m;
   value _0_1 = \text{seed}/m;
   unif := low + trunc(value _ 0_1 * (high - low + 1))
end:
procedure generate_flow_shop( var time_seed:
                                                              integer;
                                   nb_jobs, nb_machines:
                                                              integer;
                                   var d:
                                                              matrix);
(* type matrix = array[1..20, 1..500] of integer; must be declared above *)
var i, j: integer;
begin
   for i := 1 to nb machines do
      for j := 1 to nb_jobs do
          d[i, j] := unif(time_seed, 1, 99)
end:
procedure generate_job_and_open_shop( var time_seed, machine_seed: integer;
                                               nb_jobs, nb_machines:
                                                                                integer:
                                               var d, M:
                                                                                matrix);
var i, j: integer;
   procedure swap(var a, b: integer);
   var temp: integer;
   beain
      temp := a; a := b; b := temp
   end:
begin
   for j := 1 to nb_jobs do
      for i = 1 to nb_machines do
          d[i, j] := unif(time_seed, 1, 99);
   for i = 1 to nb_ jobs do
      for i = 1 to nb_machines do
          M[i, j] \coloneqq i;
   for j := 1 to nb_ jobs do
      for i = 1 to nb_machines do
          swap(M[i, j], M[unif(machine_seed, i, nb_machines), j])
end:
```

Figure 5. Pascal code for the generation of scheduling problems

Table 3 Open shop problems

Nb jobs	Nb ma- chines	Nb in- stances	LB reached (%)	Nb itera- tions	Nb resolu- tions
4	4	50 000	98.5	10 ⁵	5
5	5	45000	99.7	$5 \cdot 10^{5}$	4
7	7	1 000	94	106	5
10	10	300	89	$2 \cdot 10^{6}$	5
15	15	40	52	$3 \cdot 10^{5}$	3
20	20	25	24	$3 \cdot 10^{6}$	3

processing times d_{ij} of the operation *i* of job *j* and its associated machine M_{ij} . We give an optimum schedule of this problem in the Gantt chart of Figure 4.

The time needed by taboo search to perform one iteration is about $23 \cdot 10^{-6}nm$ seconds. Tables 3 and Figure 3 are analogous to Table 2 and Figure 2. Stars in Figure 3 indicate optimum values; a) indicates solution values issued from Kleinau [8] who proves optimality for 4×4 and 5×5 open shop problems; b) indicates solution values issued from Bräsel et al. [3].

Finally, we give in Figure 5 the procedures, written in Pascal language, for generating the scheduling problems described in this paper.

Concluding remarks

We hope that the problems that we propose will constitute a comparison base for future resolution methods.

Everyone may send us his own results about these problems, specifying whether his solutions

Table 4 The first instance of the 4-job 4-machine open shop problem

Opera-	Job	Job j ^b							
tion <i>i</i>	1	2	3	4	1	2	3	4	
1	54	9	38	95	3	4	1	1	
2	34	15	19	34	1	1	2	3	
3	61	89	28	7	4	2	3	2	
4	2	70	87	29	2	3	4	4	

^a Processing times d_{ij} .

^b Machines M_{ii}.

are proved optimal or not, in order to update the best solutions known. The data included in Figures 1 to 3 are available via e-mail in the OR-Library created by Beasley [2].

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