

VI

A Branch-Cut-and-Price Algorithm

As mentioned in the former chapters, since the column generation algorithm solves the linear relaxation of the set partitioning formulation, another technique is needed in order to obtain an integer solution. In this chapter, we use Branch-and-Bound, a widely known technique, which enumerates all possible solutions through a search tree, discarding the branches with solutions greater than a known upper bound for the instance. This technique, when used together with column generation and cut separation is called *Branch-Cut-and-Price* (BCP). We devise a Branch-Cut-and-Price algorithm for the CARP using the column generation algorithm described in Chapter V and exactly separating the odd edge cutset and capacity cuts as described in Chapter IV.

VI.1 The Branch-and-Bound

The BCP algorithm starts by solving the relaxation problem defined at root node of the search tree. This operation consists in executing the column generation algorithm as described in Chapter V, which includes the exact separation of cuts. If the solution is integral, i.e., all variables are integers, the algorithm stops and returns this solution. Otherwise, this node is inserted into a list and the main loop of the BCP begins. At each iteration, the algorithm chooses from the list the next node to be opened. This choice of a node can be implemented in different ways, each one reflecting in a different search through the tree. For instance, if the list is a stack, the search will be a depth-first search. In our implementation, a priority queue is used, always returning the node with the lowest value. This approach is usually called best-first search and, each time a new node is chosen from the list, its value is considered the new current lower bound. This can be assured by the fact that there is no node with a lower value remaining in the list.

The opening of a node consists in selecting a fractional variable of its solution and creating two branches, following a defined branching rule, which will be explained further. Then, from each branch a new node is generated,

which will be solved in a similar manner as the root node. However, differently from the root node, the value is tested against the known upper bound and, if it is greater or equal than the bound, it is discarded. This operation is usually known as pruning the node of the search tree. On the other hand, if the value is less than the upper bound, the node is considered as a valid node. Furthermore, if the solution of a valid node is integral, its value is then considered as the new upper bound and there is no way to branch from this node anymore. When this is not the case and the solution of a valid node still contains any fractional variable, the node is inserted into the list.

It is important to notice that, at a given iteration, if the chosen node has a value that differs less than one unit from the known upper bound, the BCP algorithm stops and the solution which generated the upper bound is then returned as the overall best solution.

(a) Branching Rules

The most intuitive branching rule is, for a given variable λ_r from the set partitioning formulation with a fractional value $\bar{\lambda}_r \in [0, 1]$, create two branches, the first one with $\lambda_r = 0$ and other with $\lambda_r = 1$. However, this cannot be done due to the column generation algorithm. It is easy to see that if the variable associated with a route r is fixed to zero, eventually the pricing subproblem will find this route again on a subsequent iteration. This route will be returned to the restricted master which will consider it as a new route and will create a new variable λ_r for it.

There are some ways to cope with this difficulty. But instead of that, we prefer to branch on the deadheaded edges variables z_e from the one-index formulation. By the time it is necessary to branch, the values of the z_e variables are obtained using the relation $\sum_{r \in \Omega} b_r^e \lambda_r = z_e, \forall e \in E$, which is analogous to the one created for the two-index formulation, shown in (III.20).

The drawback of this solution is that when an integer solution is found, it is integer just on z_e variables and it may not be integer on the λ_r variables. This integer solution has the same properties of one from the one-index formulation, it may have a value less than the optimal solution, which would also be infeasible. However, it does not prevent the BCP algorithm from returning good solutions, as will be shown in the results.

The selection of the branching variable can follow different rules. The simplest rule is to search for the variable whose value in the solution \bar{z}_e is closer to 0.5 and then create two branches, one with $z_e \leq \lfloor \bar{z}_e \rfloor$ and other with $z_e \geq \lceil \bar{z}_e \rceil$. Mapping these inequalities to λ_r variables results in the following two constraints.

$$\sum_{r \in \Omega} b_r^e \lambda_r \leq \lfloor \bar{z}_e \rfloor \quad (\text{VI.1})$$

$$\sum_{r \in \Omega} b_r^e \lambda_r \geq \lceil \bar{z}_e \rceil . \quad (\text{VI.2})$$

Instead of creating these two constraints for each new branch generated during the BCP algorithm, one bound constraint for each deadheaded edge $e \in E$ can be inserted *a priori* in the restricted master as shown below.

$$lb_e \leq \sum_{r \in \Omega} b_r^e \lambda_r \leq ub_e \quad \forall e \in E . \quad (\text{VI.3})$$

With the introduction of the bounds constraints, the pricing subproblem does not change, but the reduced cost of an edge, shown in equations (V.2) and (V.3), must consider the dual values associated with these constraints. Therefore, given ρ_e , the dual variable associated with constraints (VI.3), this value must be subtracted from the reduced cost of the associated edge e for both equations.

In this work, the branching rule described above is used. However, there are some different branching rules which could be used. For instance, instead of choosing the variable closer to 0.5, the rule could be to choose the variable closer to any other value between 0 and 1, exclusive. Furthermore, a set of edges could be chosen. One way to do this is to select a vertex set S with $\sum_{e \in \delta(S)} z_e$ fractional and create the associated branches.

VI.2 Improving the Search

(a) Using Strong Branching

Strong Branching is a technique used in the attempt to find better lower bounds faster. Instead of choosing the best fractional variable w.r.t. the branching rule, it uses a candidate set of variables and tests which of them give the best progress before actually perform any branching. This test is done by solving the linear relaxations – in this case the column generation – of the two nodes which would be created from each candidate. If the candidate set is the full set of fractional variables and, for each candidate, both linear relaxations are solved to optimality, this strategy is called *full strong branching*. A benefit of this strategy is to always choose the locally best fractional variable to branch on. However, the effort needed for solving each node of the search tree is high. Alternatively, most strong branchings in literature tries to estimate quickly which variable is a good one to branch on.

One possibility to speed up the strong branching is to evaluate only a subset of the candidate set. In this work, we create the candidate set by choosing the best 5 fractional variables w.r.t. the branching rule. For each candidate s , an estimate on the values of both the linear relaxations is obtained by running the column generation using just the heuristic pricing described on Section V.1(c) and without separating any cut. Let $left_s$ and $right_s$ be these values. In order to choose which candidate is the one to branch, an evaluation criterion is used. The most common one is to select the one with largest $\min\{left_s, right_s\}$ and, in the case of a tie, the one with largest $\max\{left_s, right_s\}$. We use a more complex criterion, as follows, which usually gives better results than this simple one.

$$\sigma(s) = \alpha \min(left_s, right_s) + (1 - \alpha) \max(left_s, right_s) \quad (\text{VI.4})$$

This function can be calibrated using different values for the parameter $\alpha \in [0, 1]$. During our experiments, we found $\alpha = 3/4$ to be a good choice. Therefore, we branch on the candidate with largest $\sigma(s)$.

As mentioned, during the strong branching evaluation the linear relaxations are not solved to optimality. This approach is called strong branching on pseudo-costs, since we do not know the optimal value of the linear relaxations. This was originally proposed for problems which do not use column generation. In this case, the pseudo-costs are obtained performing only a few iterations of the linear programming solver and these values are used to choose the branching variable. Anyway, in both cases, when the branching variable is finally chosen, the optimal values must be computed. This requires our algorithm to do a complete execution of the column generation for both new nodes of the search tree.

(b) Using Reduced Cost Fixing

Reduced Cost Fixing is a well-known technique which is used to estimate the limits of a variable using its current reduced cost and the known lower and upper bounds. We refer the reader to the 1998 book of Wolsey [79] where a detailed description of its applications is given. The benefit of using this technique is the possible reduction in the number of nodes during the branch-and-bound search if some variable fixings are done at each iteration of the algorithm.

Given a variable z_e with reduced cost \bar{c}_e and solution value $\tilde{z}_e = lb_e$ equals to its current lower bound, it is correct to state that if the value of this variable is incremented by 1, the value of the current solution is incremented

by \bar{c}_e . Therefore, knowing the value of the current solution Z^* and an upper bound UB for the solution, it can be concluded that an upper bound for this variable is as follows.

$$ub'_e = lb_e + \left\lfloor \frac{UB - Z^*}{\bar{c}_e} \right\rfloor \quad (\text{VI.5})$$

The upper bound for this variable is updated in case $ub'_e < ub_e$. Analogously, given a variable z_e with solution value $\tilde{z}_e = ub_e$ equals to its current upper bound and a known lower bound LB for the solution, it can also be concluded that a lower bound for the variable is as follows.

$$lb'_e = ub_e - \left\lceil \frac{UB - Z^*}{|\bar{c}_e|} \right\rceil \quad (\text{VI.6})$$

Notice that these bounds can only be computed for variables with non-negative reduced costs $\bar{c}_e \neq 0$, otherwise it would result in a division by zero.

VI.3 Experimental Results

The computational experiments were done using the same configuration as Chapters IV and V. The BCP algorithm was tested on all CARP datasets described on section IV.4, except for the *egl-large* dataset. As mentioned before, the instances of this dataset are prohibitively large to column generation.

The results are shown in Table VI.1. Columns **Ins**, **LB** and **UB** show the name and the known lower and upper bounds for each instance. Following these columns, for each group of columns **NG=X**, where $\Delta(N_i) = X$, columns **Value**, **Routes**, **Cuts**, **Nodes** and **Time** show the solution cost, the number of routes found by the column generation, the number of cuts found by the separation algorithms, the number of nodes opened by the branch-and-bound and the overall time of computation in seconds, respectively. For each instance, a time limit of three hours was used. Furthermore, when a solution cost is better than the known lower bound, its value is highlighted in boldface, and when it is an optimal solution for the instance, it is underlined.

Notice that even with the small time limit used and for small values of $\Delta(N_i)$, the algorithm was able to improve several known lower bounds and it was also able to find some new optimal values for the beullens dataset.

Table VI.1: Branch-Cut-and-Price Results for the CARP

Ins	LB	UB	NG=8					NG=16					NG=32				
			Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time
kshs1	14661	14661	<u>14661.0</u>	116	5	1	0.110	<u>14661.0</u>	92	5	1	0.109	<u>14661.0</u>	92	5	1	0.094
kshs2	9863	9863	<u>9863.0</u>	124	8	1	0.109	<u>9863.0</u>	88	8	1	0.109	<u>9863.0</u>	88	8	1	0.078
kshs3	9320	9320	<u>9320.0</u>	167	4	1	0.141	<u>9320.0</u>	93	4	1	0.125	<u>9320.0</u>	93	4	1	0.125
kshs4	11498	11498	<u>11498.0</u>	134	5	1	0.156	<u>11498.0</u>	99	5	1	0.141	<u>11498.0</u>	99	5	1	0.188
kshs5	10957	10957	<u>10957.0</u>	110	4	1	0.156	<u>10957.0</u>	112	4	1	0.219	<u>10957.0</u>	112	4	1	0.203
kshs6	10197	10197	<u>10197.0</u>	234	4	1	0.172	<u>10197.0</u>	123	4	1	0.109	<u>10197.0</u>	123	4	1	0.188
gdb1	316	316	<u>316.0</u>	259	12	1	0.031	<u>316.0</u>	235	12	1	0.016	<u>316.0</u>	201	12	1	0.031
gdb2	339	339	<u>339.0</u>	260	8	1	0.031	<u>339.0</u>	303	8	1	0.031	<u>339.0</u>	290	8	1	0.031
gdb3	275	275	<u>275.0</u>	217	5	1	0.016	<u>275.0</u>	200	5	1	0.015	<u>275.0</u>	191	5	1	0.016
gdb4	287	287	<u>287.0</u>	126	6	1	0.000	<u>287.0</u>	86	6	1	0.000	<u>287.0</u>	86	6	1	0.016
gdb5	377	377	<u>377.0</u>	228	8	1	0.032	<u>377.0</u>	200	8	1	0.015	<u>377.0</u>	197	8	1	0.031
gdb6	298	298	<u>298.0</u>	158	5	1	0.015	<u>298.0</u>	160	4	1	0.016	<u>298.0</u>	143	4	1	0.016
gdb7	325	325	<u>325.0</u>	219	10	1	0.016	<u>325.0</u>	147	10	1	0.016	<u>325.0</u>	192	10	1	0.015
gdb8	348	348	<u>347.0</u>	944	28	13	1.688	<u>347.0</u>	828	24	11	1.266	<u>347.0</u>	885	25	13	1.672
gdb9	303	303	<u>303.0</u>	813	18	7	1.469	<u>303.0</u>	734	17	3	0.844	<u>303.0</u>	857	19	3	0.922
gdb10	275	275	<u>275.0</u>	245	12	1	0.032	<u>275.0</u>	228	12	1	0.031	<u>275.0</u>	237	13	1	0.047
gdb11	395	395	<u>395.0</u>	1105	13	1	0.953	<u>395.0</u>	925	13	1	0.782	<u>395.0</u>	1105	13	1	0.968
gdb12	458	458	<u>454.0</u>	253	11	5	0.203	<u>455.0</u>	236	11	3	0.156	<u>455.0</u>	202	11	3	0.141
gdb13	536	536	<u>536.0</u>	412	5	1	0.172	<u>536.0</u>	400	5	1	0.172	<u>536.0</u>	291	5	1	0.125
gdb14	100	100	<u>100.0</u>	151	1	1	0.016	<u>100.0</u>	165	1	1	0.031	<u>100.0</u>	135	1	1	0.015
gdb15	58	58	<u>58.0</u>	127	1	3	0.062	<u>58.0</u>	174	1	3	0.109	<u>58.0</u>	220	1	5	0.157
gdb16	127	127	<u>127.0</u>	378	6	3	0.125	<u>127.0</u>	322	6	3	0.110	<u>127.0</u>	228	7	3	0.093
gdb17	91	91	<u>91.0</u>	134	7	1	0.047	<u>91.0</u>	134	7	1	0.093	<u>91.0</u>	74	7	1	0.032
gdb18	164	164	<u>164.0</u>	565	1	1	0.219	<u>164.0</u>	685	1	1	0.297	<u>164.0</u>	385	1	1	0.156
gdb19	55	55	<u>55.0</u>	55	4	1	0.016	<u>55.0</u>	60	4	1	0.015	<u>55.0</u>	60	4	1	0.000
gdb20	121	121	<u>121.0</u>	265	9	1	0.047	<u>121.0</u>	205	10	1	0.063	<u>121.0</u>	187	10	1	0.047

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Table VI.1: *Continued from previous page*

Ins	LB	UB	NG=8					NG=16					NG=32				
			Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time
gdb21	156	156	<u>156.0</u>	454	7	3	0.219	<u>156.0</u>	354	6	3	0.281	<u>156.0</u>	314	7	3	0.172
gdb22	200	200	<u>200.0</u>	452	7	11	0.875	<u>200.0</u>	472	7	9	0.890	<u>200.0</u>	412	6	11	1.015
gdb23	233	233	<u>233.0</u>	412	1	17	1.484	<u>233.0</u>	432	1	7	0.703	<u>233.0</u>	352	1	9	0.859
1A	173	173	<u>173.0</u>	3067	20	3	8.516	<u>173.0</u>	3192	20	3	10.812	<u>173.0</u>	2754	20	3	10.750
1B	173	173	<u>173.0</u>	1762	20	1	2.985	<u>173.0</u>	1711	20	3	4.094	<u>173.0</u>	1809	20	3	4.641
1C	245	245	<u>242.0</u>	1094	28	31	5.750	<u>242.0</u>	696	18	5	1.484	<u>242.0</u>	713	18	5	1.672
2A	227	227	<u>227.0</u>	2107	12	1	4.063	<u>227.0</u>	2138	11	1	5.063	<u>227.0</u>	1250	11	1	6.109
2B	259	259	<u>258.0</u>	2662	25	5	7.515	<u>258.0</u>	2522	25	3	16.141	<u>258.0</u>	1708	25	1	8.219
2C	457	457	<u>457.0</u>	406	16	1	0.328	<u>457.0</u>	398	16	1	0.297	<u>457.0</u>	345	16	1	0.265
3A	81	81	<u>81.0</u>	2589	25	3	4.282	<u>81.0</u>	2867	23	3	5.969	<u>81.0</u>	2340	24	3	5.938
3B	87	87	<u>87.0</u>	2591	35	7	5.344	<u>87.0</u>	2755	36	7	5.797	<u>87.0</u>	2670	40	9	9.594
3C	138	138	<u>136.0</u>	764	13	9	0.750	<u>136.0</u>	792	15	9	0.953	<u>136.0</u>	775	17	9	0.921
4A	400	400	<u>400.0</u>	9283	37	3	110.140	<u>400.0</u>	10281	37	3	141.953	<u>400.0</u>	12046	37	3	247.156
4B	412	412	<u>412.0</u>	4576	34	3	29.515	<u>412.0</u>	5996	36	3	49.375	<u>412.0</u>	6596	34	3	64.203
4C	428	428	<u>428.0</u>	4297	33	3	26.390	<u>428.0</u>	4771	33	3	30.844	<u>428.0</u>	3846	33	3	28.406
4D	530	530	<u>527.0</u>	5302	75	85	149.984	<u>528.0</u>	10580	102	371	1224.310	<u>528.0</u>	6431	78	95	303.688
5A	423	423	<u>423.0</u>	4487	25	3	31.297	<u>423.0</u>	5547	26	3	42.781	<u>423.0</u>	5327	26	3	44.687
5B	446	446	<u>446.0</u>	4901	73	11	60.890	<u>446.0</u>	6615	125	11	102.406	<u>446.0</u>	7771	74	11	129.906
5C	474	474	<u>471.0</u>	8552	119	25	196.500	<u>471.0</u>	6127	127	13	97.890	<u>471.0</u>	8893	125	19	200.469
5D	577	577	<u>573.0</u>	4734	166	27	61.219	<u>573.0</u>	2564	110	11	20.532	<u>573.0</u>	2414	111	9	21.547
6A	223	223	<u>223.0</u>	3475	21	3	13.797	<u>223.0</u>	3849	21	3	16.907	<u>223.0</u>	3458	22	3	14.969
6B	233	233	<u>231.0</u>	6120	86	37	73.422	<u>231.0</u>	6068	125	19	75.266	<u>231.0</u>	2961	79	7	18.718
6C	317	317	<u>313.0</u>	1198	48	25	8.797	<u>313.0</u>	1623	46	111	45.422	<u>313.0</u>	1344	46	57	23.328
7A	279	279	<u>279.0</u>	4507	23	1	23.235	<u>279.0</u>	4019	21	1	24.547	<u>279.0</u>	5376	21	1	40.532
7B	283	283	<u>283.0</u>	3616	21	1	14.156	<u>283.0</u>	3576	21	1	17.563	<u>283.0</u>	3815	21	1	19.469
7C	334	334	<u>330.0</u>	6057	135	71	114.078	<u>330.0</u>	4987	133	47	96.735	<u>330.0</u>	4125	109	37	78.453
8A	386	386	<u>386.0</u>	3167	23	1	15.563	<u>386.0</u>	3127	23	1	16.531	<u>386.0</u>	3567	23	1	21.515

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Table VI.1: *Continued from previous page*

Ins	LB	UB	NG=8					NG=16					NG=32				
			Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time
8B	395	395	<u>395.0</u>	2496	24	1	9.609	<u>395.0</u>	3116	21	1	13.140	<u>395.0</u>	3216	22	1	15.000
8C	521	521	518.0	2715	71	23	23.735	518.0	2506	68	23	23.796	518.0	2108	60	15	15.750
9A	323	323	<u>323.0</u>	9967	54	3	173.109	<u>323.0</u>	11886	54	3	263.953	<u>323.0</u>	11746	54	3	303.500
9B	326	326	<u>326.0</u>	9956	49	3	149.875	<u>326.0</u>	11336	49	3	228.532	<u>326.0</u>	16736	50	3	701.781
9C	332	332	<u>332.0</u>	6025	44	5	64.593	<u>332.0</u>	7765	44	5	107.969	<u>332.0</u>	6485	45	3	77.156
9D	391	391	386.6	24915	753	545	<i>3h</i>	387.0	22472	599	429	6646.190	387.0	15283	438	201	2414.980
10A	428	428	<u>428.0</u>	10749	44	3	241.000	<u>428.0</u>	16333	44	3	640.140	<u>428.0</u>	21067	44	3	1516.860
10B	436	436	<u>436.0</u>	8756	59	5	183.438	<u>436.0</u>	9116	59	5	236.032	<u>436.0</u>	10176	58	5	304.297
10C	446	446	<u>446.0</u>	8563	60	9	209.859	<u>446.0</u>	8285	65	7	176.375	<u>446.0</u>	9752	63	7	245.609
10D	526	526	525.0	8806	132	33	260.328	525.0	7110	99	21	140.906	525.0	9474	154	31	303.156
C01	4105	4150	4125.0	33153	470	471	6974.130	4125.0	20467	345	265	2994.160	4125.0	18158	328	209	2917.990
C02	3135	3135	<u>3135.0</u>	2081	32	1	3.453	<u>3135.0</u>	2108	32	3	5.281	<u>3135.0</u>	2040	32	3	6.016
C03	2575	2575	2565.0	5145	113	23	48.016	2565.0	4668	125	27	52.515	2565.0	3938	134	17	45.219
C04	3478	3510	3495.0	7310	76	11	55.188	3500.0	8775	82	21	127.375	3500.0	7333	81	13	125.000
C05	5365	5365	5335.0	5022	90	27	64.078	5335.0	4266	73	23	46.547	5340.0	6215	119	73	179.031
C06	2535	2535	2525.0	6446	76	69	89.735	2525.0	4156	84	9	44.969	2525.0	4543	80	9	47.047
C07	4075	4075	4050.0	5224	106	37	54.485	4050.0	4479	94	31	46.485	4050.0	4690	99	37	55.015
C08	4090	4090	4060.0	4004	155	31	53.953	4060.0	4513	196	33	92.125	4060.0	3767	125	27	68.766
C09	5233	5260	5240.0	32727	495	305	6933.890	5240.0	19023	390	137	2180.630	5240.0	14748	335	87	1172.170
C10	4700	4700	4670.0	4057	176	59	87.563	4685.0	5753	241	119	278.281	4685.0	5022	221	125	285.094
C11	4583	4635	4580.0	10966	317	25	345.438	4585.0	11582	339	37	569.312	4585.0	9918	288	25	569.234
C12	4209	4240	4200.0	18606	267	271	2175.080	4200.0	18858	243	267	2093.520	4200.0	15971	248	189	1587.270
C13	2955	2955	2940.0	4229	65	17	29.578	2940.0	4129	61	15	27.734	2940.0	4171	63	19	39.500
C14	4030	4030	4010.0	3639	64	31	28.922	4010.0	4222	65	45	63.578	4010.0	2725	59	27	34.125
C15	4912	4940	4910.0	28469	163	209	2734.550	4920.0	43642	178	561	<i>3h</i>	4913.3	33902	151	307	<i>3h</i>
C16	1475	1475	1470.0	2201	25	1	3.594	1470.0	2065	28	3	6.735	1470.0	1256	25	1	2935.440
C17	3555	3555	3550.0	1635	49	3	3.218	3550.0	1421	48	3	3.485	3550.0	1205	48	3	3.797

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Table VI.1: *Continued from previous page*

Ins	LB	UB	NG=8					NG=16					NG=32				
			Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time
C18	5577	5620	5564.5	39731	273	150	3h	5567.6	33027	179	85	3h	5568.6	34423	197	85	3h
C19	3096	3115	3115.0	16707	290	147	1134.000	3115.0	9861	201	65	1142.050	3115.0	5337	120	25	1725.640
C20	2120	2120	<u>2120.0</u>	4452	59	3	11.641	<u>2120.0</u>	5022	53	5	29.703	<u>2120.0</u>	4197	38	5	18.953
C21	3960	3970	3960.0	6976	52	7	45.562	3960.0	6005	50	7	70.171	3960.0	5532	53	5	259.312
C22	2245	2245	<u>2245.0</u>	3137	36	3	5.109	<u>2245.0</u>	2244	36	5	5.484	<u>2245.0</u>	1673	36	1	3.828
C23	4032	4085	4070.0	38026	383	241	9081.970	4067.0	20228	242	60	3h	4056.1	8877	179	3	3h
C24	3384	3400	3385.0	18883	372	51	1042.030	3385.0	8924	139	15	144.703	3385.0	8856	173	25	255.703
C25	2310	2310	<u>2310.0</u>	1899	29	3	2.797	<u>2310.0</u>	1584	28	1	1.593	<u>2310.0</u>	1343	29	3	2.484
D01	3215	3215	<u>3215.0</u>	8421	54	1	66.453	<u>3215.0</u>	10840	53	1	118.704	<u>3215.0</u>	10856	61	1	135.250
D02	2520	2520	<u>2520.0</u>	3618	25	1	9.250	<u>2520.0</u>	4276	25	1	13.344	<u>2520.0</u>	4744	25	1	17.953
D03	2065	2065	<u>2065.0</u>	6715	31	3	30.704	<u>2065.0</u>	6929	30	3	33.984	<u>2065.0</u>	6413	31	3	39.532
D04	2785	2785	<u>2785.0</u>	12240	54	3	156.766	<u>2785.0</u>	12136	52	3	171.468	<u>2785.0</u>	12839	51	3	233.437
D05	3935	3935	<u>3935.0</u>	5850	49	3	35.188	<u>3935.0</u>	6540	52	3	41.734	<u>3935.0</u>	5870	49	3	43.031
D06	2125	2125	<u>2125.0</u>	5901	30	3	23.484	<u>2125.0</u>	6564	30	3	30.078	<u>2125.0</u>	7326	30	3	45.938
D07	3115	3115	3075.0	14750	137	53	600.516	3090.0	17065	179	57	1515.880	3090.0	14226	104	45	3h
D08	2995	3045	3020.0	17349	103	39	495.625	3020.0	16325	89	33	579.000	3020.0	14640	89	27	688.891
D09	4120	4120	<u>4120.0</u>	12842	50	3	172.812	<u>4120.0</u>	15510	50	3	319.016	<u>4120.0</u>	20382	49	3	756.891
D10	3340	3340	3335.0	5304	33	3	34.000	3335.0	5148	33	3	27.875	3335.0	4451	33	3	1132.190
D11	3745	3745	<u>3745.0</u>	18943	78	7	531.953	<u>3745.0</u>	20367	80	7	604.890	<u>3745.0</u>	19543	80	7	752.625
D12	3310	3310	<u>3310.0</u>	9794	49	3	88.734	<u>3310.0</u>	10748	49	3	123.860	<u>3310.0</u>	11227	49	3	151.266
D13	2535	2535	<u>2535.0</u>	4279	37	1	12.968	<u>2535.0</u>	4769	36	1	17.032	<u>2535.0</u>	3461	36	1	12.437
D14	3272	3280	3280.0	10278	59	11	125.859	3280.0	8839	82	9	120.266	—	—	—	—	—
D15	3990	3990	<u>3990.0</u>	19842	55	3	412.141	<u>3990.0</u>	21235	54	3	550.359	<u>3990.0</u>	24021	54	1	656.812
D16	1060	1060	<u>1060.0</u>	5622	19	1	20.500	<u>1060.0</u>	6662	16	1	131.390	—	—	—	—	—
D17	2620	2620	<u>2620.0</u>	3816	27	3	10.172	<u>2620.0</u>	3464	26	5	13.469	<u>2620.0</u>	2244	26	3	7.907
D18	4165	4165	<u>4165.0</u>	19955	67	3	534.844	<u>4165.0</u>	20805	66	3	680.625	<u>4165.0</u>	21352	67	3	850.312
D19	2393	2400	2385.0	20774	87	13	370.360	—	—	—	—	—	—	—	—	—	—

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Table VI.1: *Continued from previous page*

Ins	LB	UB	NG=8					NG=16					NG=32				
			Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time
D20	1870	1870	<u>1870.0</u>	7745	33	3	32.594	<u>1870.0</u>	9467	34	3	76.281	<u>1870.0</u>	8547	34	3	66.500
D21	2985	3050	3010.0	48420	116	67	3607.050	3010.0	31992	88	49	2229.050	2995.3	21954	60	9	3h
D22	1865	1865	<u>1865.0</u>	7724	32	3	26.594	<u>1865.0</u>	9213	32	1	43.844	<u>1865.0</u>	5867	32	1	79.453
D23	3114	3130	3120.0	33900	116	9	2355.740	3116.8	23785	110	3	3h	—	—	—	—	—
D24	2676	2710	2705.0	61167	218	41	8401.740	2697.5	61419	201	21	3h	—	—	—	—	—
D25	1815	1815	<u>1815.0</u>	3501	24	1	6.625	<u>1815.0</u>	3841	58	1	10.406	<u>1815.0</u>	3376	24	1	355.859
E01	4885	4910	4870.0	18143	413	439	2719.260	4870.0	15571	321	293	1823.880	4870.0	12827	311	195	1822.480
E02	3990	3990	3975.0	4301	100	25	37.766	3975.0	3950	79	17	31.516	3975.0	3334	65	21	34.922
E03	2015	2015	<u>2015.0</u>	3603	34	3	9.718	<u>2015.0</u>	3372	30	1	5.546	<u>2015.0</u>	2839	30	1	6.312
E04	4155	4155	4145.0	12606	307	31	273.859	4145.0	8560	196	41	1502.480	4145.0	8569	252	29	619.984
E05	4585	4585	4580.0	3110	85	5	15.297	4580.0	3001	75	5	13.828	<u>4585.0</u>	4153	106	17	46.687
E06	2055	2055	<u>2055.0</u>	1668	25	1	2.391	<u>2055.0</u>	1704	25	1	2.125	<u>2055.0</u>	1315	25	1	1.687
E07	4155	4155	4080.0	2961	145	39	42.547	4080.0	2494	130	31	44.953	4080.0	1851	92	13	22.937
E08	4710	4710	4700.0	5454	114	21	56.375	<u>4710.0</u>	5609	129	27	60.016	<u>4710.0</u>	5662	163	23	65.468
E09	5780	5820	5775.0	11761	194	39	492.093	5775.0	9151	173	21	244.031	5780.0	18003	277	213	3092.750
E10	3605	3605	<u>3605.0</u>	1411	58	1	2.078	<u>3605.0</u>	1486	58	1	3.313	<u>3605.0</u>	1098	58	1	1.969
E11	4637	4655	4645.0	10111	204	25	305.047	4645.0	9216	137	17	223.203	4650.0	14520	219	43	1687.860
E12	4180	4180	4165.0	10031	183	197	516.750	4165.0	8475	158	127	324.703	4165.0	8195	151	115	374.657
E13	3345	3345	3330.0	2686	63	5	10.203	3330.0	2519	63	5	11.422	3330.0	2279	63	5	179.797
E14	4115	4115	4105.0	7817	192	179	262.625	4105.0	6092	124	129	190.187	4105.0	5463	139	137	293.500
E15	4189	4205	4197.6	47931	177	369	3h	4200.0	41161	198	273	10450.800	4197.7	17654	125	51	3h
E16	3755	3775	<u>3775.0</u>	10561	124	53	176.422	<u>3775.0</u>	5599	123	39	118.672	<u>3775.0</u>	6866	115	45	801.266
E17	2740	2740	<u>2740.0</u>	1684	38	3	2.422	<u>2740.0</u>	1565	41	5	4.531	<u>2740.0</u>	1542	39	3	3.297
E18	3825	3835	3825.0	11053	96	13	173.219	3830.0	19493	232	43	1215.640	3830.0	16008	320	43	3447.670
E19	3222	3235	3230.0	10584	144	31	240.453	3230.0	8908	108	23	184.312	3230.0	7447	121	19	279.844
E20	2802	2825	2810.0	6629	110	21	71.672	2810.0	6644	126	23	87.172	2810.0	4840	73	15	53.235
E21	3728	3730	<u>3730.0</u>	8623	67	21	121.235	<u>3730.0</u>	8976	79	17	143.515	<u>3730.0</u>	7732	79	17	213.984

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Table VI.1: *Continued from previous page*

Ins	LB	UB	NG=8					NG=16					NG=32				
			Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time
E22	2470	2470	2470.0	3118	63	11	23.922	2470.0	2554	61	9	41.359	2470.0	2614	56	7	116.750
E23	3686	3710	3705.0	33482	564	143	5479.630	3705.0	27254	443	91	4330.250	3696.2	8536	178	13	3h
E24	4001	4020	4010.0	15192	199	71	926.172	4010.0	11096	155	21	266.094	4015.0	13671	191	47	699.579
E25	1615	1615	<u>1615.0</u>	956	20	1	0.578	<u>1615.0</u>	1155	21	1	1.063	<u>1615.0</u>	662	20	1	1.047
F01	4040	4040	<u>4040.0</u>	11375	93	3	210.640	<u>4040.0</u>	13103	92	3	242.062	<u>4040.0</u>	15593	94	3	451.547
F02	3300	3300	<u>3300.0</u>	6067	49	1	26.218	<u>3300.0</u>	5381	49	1	24.281	<u>3300.0</u>	7296	49	1	52.469
F03	1665	1665	<u>1665.0</u>	6982	36	1	21.141	<u>1665.0</u>	7249	36	1	25.219	<u>1665.0</u>	4442	39	1	48.594
F04	3476	3485	3485.0	27270	137	27	1471.310	3485.0	31918	109	11	3h	3485.0	17243	93	5	3h
F05	3605	3605	<u>3605.0</u>	4154	65	1	17.594	<u>3605.0</u>	5334	65	3	32.110	<u>3605.0</u>	5431	66	3	39.609
F06	1875	1875	<u>1875.0</u>	3149	28	1	6.500	<u>1875.0</u>	3353	28	1	7.875	<u>1875.0</u>	2812	28	1	7.875
F07	3335	3335	<u>3335.0</u>	5511	72	5	39.125	<u>3335.0</u>	4262	59	3	17.468	<u>3335.0</u>	3445	59	3	13.484
F08	3695	3705	3705.0	8869	103	21	163.062	3705.0	10759	79	27	249.140	3705.0	10168	84	19	271.016
F09	4730	4730	<u>4730.0</u>	12822	88	5	230.609	<u>4730.0</u>	17582	88	5	541.109	<u>4730.0</u>	18022	87	5	629.421
F10	2925	2925	<u>2925.0</u>	4130	44	3	15.016	<u>2925.0</u>	4788	44	3	20.218	<u>2925.0</u>	3695	44	3	16.828
F11	3835	3835	<u>3835.0</u>	14947	85	5	263.625	<u>3835.0</u>	19243	206	5	523.250	<u>3835.0</u>	21815	84	5	818.938
F12	3390	3395	<u>3390.0</u>	17590	92	19	436.734	<u>3390.0</u>	18161	94	19	513.000	<u>3390.0</u>	13644	99	13	409.156
F13	2855	2855	<u>2855.0</u>	3355	48	3	10.953	<u>2855.0</u>	3950	48	3	17.110	<u>2855.0</u>	3632	48	3	14.422
F14	3330	3330	<u>3330.0</u>	4443	78	3	23.687	<u>3330.0</u>	5078	50	1	231.485	<u>3330.0</u>	5333	50	1	2019.550
F15	3560	3560	<u>3560.0</u>	24625	74	9	924.421	<u>3560.0</u>	35989	70	13	2394.890	<u>3560.0</u>	29466	70	13	1497.230
F16	2725	2725	<u>2725.0</u>	7484	38	3	30.328	<u>2725.0</u>	7537	37	1	36.406	<u>2725.0</u>	8646	37	1	3384.970
F17	2055	2055	<u>2055.0</u>	2981	31	1	5.875	<u>2055.0</u>	3206	31	1	8.453	<u>2055.0</u>	1497	31	1	3.125
F18	3063	3075	3065.0	19142	74	3	329.297	3065.0	21413	66	3	2540.670	—	—	—	—	—
F19	2500	2525	<u>2500.0</u>	22195	148	11	595.969	—	—	—	—	—	—	—	—	—	—
F20	2445	2445	<u>2445.0</u>	9775	54	3	74.265	<u>2445.0</u>	10775	54	3	93.703	<u>2445.0</u>	10455	54	3	106.688
F21	2930	2930	<u>2930.0</u>	9647	51	1	65.797	<u>2930.0</u>	11328	51	1	110.875	<u>2930.0</u>	11865	51	1	138.672
F22	2075	2075	<u>2075.0</u>	5902	50	3	21.703	<u>2075.0</u>	7474	56	3	91.063	<u>2075.0</u>	4756	49	1	322.672
F23	2994	3005	3005.0	34350	213	10	3h	3005.0	34409	194	5	5923.860	—	—	—	—	—

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Table VI.1: *Continued from previous page*

Ins	LB	UB	NG=8					NG=16					NG=32				
			Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time	Value	Routes	Cuts	Nodes	Time
F24	3210	3210	<u>3210.0</u>	13914	88	3	215.125	<u>3210.0</u>	18096	102	3	2245.720	<u>3210.0</u>	22045	88	1	9429.840
F25	1390	1390	<u>1390.0</u>	2023	22	1	2.594	<u>1390.0</u>	2092	22	1	3.984	<u>1390.0</u>	736	22	1	3.750
e1-A	3548	3548	<u>3548.0</u>	3116	55	1	26.344	<u>3548.0</u>	2898	55	1	23.312	<u>3548.0</u>	2623	55	1	18.735
e1-B	4498	4498	4496.0	7402	94	131	449.797	4496.0	6400	86	85	344.562	4496.0	5498	87	89	549.750
e1-C	5595	5595	5556.0	2847	70	49	147.890	5556.0	2010	70	19	59.078	5556.0	1627	70	19	62.375
e2-A	5018	5018	5012.0	6531	88	5	160.000	<u>5018.0</u>	5912	94	7	655.797	<u>5018.0</u>	5472	90	7	5886.940
e2-B	6305	6317	6299.0	5529	93	49	340.343	6299.0	4615	89	21	275.172	6301.0	5326	105	53	1349.550
e2-C	8335	8335	8302.0	4882	99	281	810.688	8308.0	5156	123	439	1290.770	8308.0	4569	139	349	1209.700
e3-A	5898	5898	<u>5898.0</u>	7930	81	5	227.359	<u>5898.0</u>	8096	76	5	322.360	<u>5898.0</u>	9244	75	5	1002.020
e3-B	7711	7775	7715.0	15787	178	749	7309.340	7722.0	15523	210	831	10135.300	7722.0	10861	156	289	7833.480
e3-C	10244	10292	10207.0	4787	131	267	1088.140	10214.0	5750	137	407	1914.380	10214.0	5047	111	291	1538.280
e4-A	6408	6444	6393.0	11099	223	23	1060.330	6393.0	12182	163	17	6083.190	—	—	—	—	—
e4-B	8935	8961	8897.0	7552	110	229	1962.030	8900.0	7543	120	203	2289.560	8912.0	9889	218	689	3h
e4-C	11493	11550	11490.0	6519	238	901	5085.380	11495.0	6754	192	1189	7936.920	11495.0	6072	202	723	5176.520
s1-A	5018	5018	<u>5018.0</u>	6904	143	3	176.984	<u>5018.0</u>	10945	143	3	377.640	<u>5018.0</u>	8825	143	3	417.734
s1-B	6388	6388	<u>6388.0</u>	7400	166	35	465.281	<u>6388.0</u>	6510	160	21	369.219	<u>6388.0</u>	7626	182	29	712.125
s1-C	8518	8518	8511.0	5612	183	169	792.062	8511.0	4223	158	91	520.344	8511.0	4598	172	63	409.750
s2-A	9825	9884	9812.7	23671	767	102	3h	9816.8	25049	529	79	3h	9811.3	20759	300	45	3h
s2-B	13017	13100	12986.8	13353	634	357	3h	12991.3	12292	591	295	3h	12993.0	11357	488	263	3h
s2-C	16425	16425	16376.2	8257	397	725	3h	16380.0	8608	317	591	3h	16382.3	8381	327	601	3h
s3-A	10145	10220	10153.2	23485	368	93	3h	10152.6	14615	226	25	3h	10149.7	10637	224	3	3h
s3-B	13648	13682	13633.1	14969	588	275	3h	13635.7	11574	512	237	3h	13636.6	11146	469	193	3h
s3-C	17188	17188	17126.2	9987	614	531	3h	17134.1	8422	466	453	3h	17136.8	7842	396	473	3h
s4-A	12143	12268	12141.8	22904	387	118	3h	12144.5	18010	398	51	3h	12143.1	13142	183	15	3h
s4-B	16098	16283	16085.9	11926	789	229	3h	16099.1	10087	685	135	3h	16098.6	9591	520	125	3h
s4-C	20430	20481	20395.5	6455	438	256	3h	20407.3	6681	349	307	3h	20409.6	6278	339	309	3h