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Dynamic Programming Driven Memetic Search for the Steiner Tree Problem with Revenues, Budget and Hop Constraints - Online Supplement

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1. Worst Case Complexity Analysis

We provide here a worst case complexity analysis of the proposed memetic algorithm. The algorithm consists of several main ingredients, including a pre-processing procedure, a probabilistic constructive procedure, a dynamic programming based neighborhood search (NS) procedure, a crossover operator associated with a pool updating strategy, whose computational complexities are respectively bounded as follows.

1. *Pre-processing procedure* (15): The pre-processing procedure is used to calculate and store all the possible values of $L(i, j, l)$, $i, j \in V, r_j > 0, 1 \leq l \leq H$ before the search (which would be fetched directly during the search, instead of calculating them repeatedly). Using a dynamic programming procedure (20), these computations can be achieved within a time of $O(|V| \times |PV| \times H) \leq O(|V|^2 \times H)$, where PV denotes the set containing all the profitable vertices (with $r_j > 0$) of graph G .

2. *Probabilistic constructive procedure* (15): Using the needed $L(i, j, l)$ values provided by the pre-processing step, the computations at each construction step can be finished within a time of $O(|V| \times |PV| + |V| \times H)$. Since at most $|PV|$ steps are needed to complete the

construction procedure, the total time for constructing a saturated BHS-tree is bounded within $O(|V| \times |PV|^2 + |PV| \times |V| \times H) \leq O(|V|^3 + |V|^2 \times H)$.

3. *Neighborhood search (NS) procedure:* Due to the possibility of special cases and the unknown number of NS iterations before termination, it is difficult to give an exact computational complexity of the NS procedure, thus we just provide a worst case analysis as follows. At first, at the beginning of each iteration of NS, we should re-execute the dynamic programming procedure of Eq. (2), whose complexity is bounded within $O(|V_{urp}| \times W_{max}) \leq O(|PV| \times B) \leq O(|V| \times B)$. Then, up to $O(|lv(T)|^2) \leq O(|PV|^2) \leq O(|V|^2)$ possible neighboring solutions should be examined by the estimation criterion (each one needs $O(H)$ time to get the available budget after deletion and $O(1)$ time to check if the examined neighboring solution could be directly discarded), among which only the hopeful ones identified by the estimation criterion should be actually generated. As explained in subsection 3.2.4, with the aid of the dynamic driven estimation criterion, in most cases only one or very few neighboring solutions should be actually generated. However, in the worst case (i.e., special case 1 occurs for every neighboring solution), it is possible that all the $O(|lv(T)|^2) \leq O(|V|^2)$ neighboring solutions should be actually generated. Now we discuss the time needed for actually generating a neighboring solution. At first, for preparation, we should store (and restore afterwards) the incumbent solution, both within $O(|V|)$ time. Then, the path deletion can be done within $O(H)$ time. After that, up to $|V_{urp}|$ new paths may be inserted, each one with $O(|V|)$ time to re-calculate a new hop-constrained shortest path between the chosen vertex for connection and the solution after path deletion, and up to $O(|V| \times H)$ time to insert the renewed hop-constrained shortest path, using the pre-calculated values of $L(i, j, l)$. Thus the time complexity of actually generating a neighboring solution is bounded within $O(|V|) + O(H) + O(|V_{urp}|) \times O(|V|) + O(|V_{urp}|) \times O(|V| \times H) \leq O(|V|^2 \times H)$. Based on the above information, the complexity of each iteration of NS is bounded within $O(|V| \times B) + O(|lv(T)|^2) \times O(H) + O(|lv(T)|^2) \times O(|V|^2 \times H) \leq O(|V| \times B + |V|^4 \times H)$. Given that it is difficult to exactly estimate the number of iterations before termination. Let I denote the maximum possible number of NS iterations (I is typically but unnecessarily in proportion to $|V|$), then the overall complexity of the NS procedure is bounded by $I \times O(|V| \times B + |V|^4 \times H)$.

4. *Crossover operator:* At first, the shared backbone of any two solutions could be identified (first step of the crossover operator) within a complexity of $O(|PV| \times H) \leq O(|V| \times H)$.

Subsequently, to complete the backbone and obtain a saturated BHS-tree, a complexity bounded by the probabilistic constructive procedure is needed, i.e., $O(|V|^3 + |V|^2 \times H)$, as analyzed above. Thus, the crossover operator can be achieved within $O(|V| \times H) + O(|V|^3 + |V|^2 \times H) = O(|V|^3 + |V|^2 \times H)$ time.

5. Pool updating strategy: Basically, the distance between any two solutions can be calculated within $O(|V|)$ time. Based on this, for the sake of a quick updating, we maintain in a $Q \times Q$ matrix the distances between any pair of solutions belonging to Pop (requiring a complexity of $O(|V| \times Q^2)$ for initialization). With this distance matrix, given a new offspring solution T^0 and the population Pop containing Q solutions, the goodness score of each solution of Pop can be calculated within $O(|V| + Q)$ time, and the goodness score of T^0 can be calculated within $O(|V| \times Q)$ time, meaning that the worst (with the lowest goodness score) solution of $Pop \cup \{T^0\}$ can be identified within a complexity of $O(Q) \times O(|V| + Q) + O(|V| \times Q) = O(|V| \times Q + Q^2)$. Finally, after replacing T^0 with some other existing solution, the distances matrix is updated within a complexity of $O(|V| \times Q)$. Overall, updating the solutions pool with a new offspring is done within a time of $O(|V| \times Q + Q^2) + O(|V| \times Q) = O(|V| \times Q + Q^2)$ (excluding the complexity needed for distances matrix initialization).

To summarize, the complexity of each round of the NS procedure followed by crossover and pool updating is bounded within $I \times O(|V| \times B + |V|^4 \times H) + O(|V|^3 + |V|^2 \times H) + O(|V| \times Q + Q^2) = O(I \times |V| \times B + I \times |V|^4 \times H + |V| \times Q + Q^2)$. Given that it is difficult to estimate the number of such rounds needed for a run of the memetic algorithm. If denote the maximum number of such rounds by G (G is typically but unnecessarily in proportion to M), the overall complexity (including the pre-processing procedure, the solutions pool initialization, as well as the distance matrix initialization) of each independent run of memetic is bounded within $O(|V|^2 \times H) + Q \times O(|V|^3 + |V|^2 \times H) + O(|V| \times Q^2) + G \times O(I \times |V| \times B + I \times |V|^4 \times H + |V| \times Q + Q^2)$. Again, we would like to point out this is only a worst case analysis which serves as a theoretical upper bound of computational complexity, being much higher than the actually needed computational efforts in most cases.

2. Supplementary results and additional comparisons

**Table 1 Results of the memetic algorithm with a reduced termination criterion using $M = 50$ (Memetic(50))
on the 56 most challenging instances of group G4, in comparison with previous fast heuristics D&R and TS(2000)**

Instance			D&R		TS(2000)		Memetic(50)		Instance			D&R		TS(2000)		Memetic(50)	
<i>Graph</i>	<i>b</i>	<i>H</i>	R^{best}	<i>t(s)</i>	R^{best}	<i>t(s)</i>	R^{best}	<i>t(s)</i>	<i>Graph</i>	<i>b</i>	<i>H</i>	R^{best}	<i>t(s)</i>	R^{best}	<i>t(s)</i>	R^{best}	<i>t(s)</i>
<i>C8_10</i>	20	5	208	1.32	≤ 228	14.2	228	3.02	<i>C8_10</i>	50	5	104	0.12	≤ 116	13.6	116	2.14
<i>C8_10</i>	20	15	303	9.51	≤ 308	42.0	324	6.86	<i>C8_10</i>	50	15	152	1.91	≤ 161	41.6	171	5.16
<i>C8_10</i>	20	25	311	16.23	≤ 310	65.6	329	8.87	<i>C8_10</i>	50	25	151	3.19	≤ 160	63.7	172	7.96
<i>C8_100</i>	20	5	2261	1.28	≤ 2365	13.5	2365	4.03	<i>C8_100</i>	50	5	1151	0.23	≤ 1204	13.5	1216	2.61
<i>C8_100</i>	20	15	3269	9.24	≤ 3237	43.1	3370	7.13	<i>C8_100</i>	50	15	1696	1.88	≤ 1675	41.8	1750	5.71
<i>C8_100</i>	20	25	3310	16.87	≤ 3337	66.8	3416	9.74	<i>C8_100</i>	50	25	1735	1.79	≤ 1742	64.3	1792	8.02
<i>C9_10</i>	20	5	274	2.26	≤ 279	16.0	302	5.60	<i>C9_10</i>	50	5	143	0.26	≤ 145	16.1	149	3.67
<i>C9_10</i>	20	15	354	11.35	≤ 349	42.4	372	9.01	<i>C9_10</i>	50	15	172	2.36	≤ 165	39.3	184	8.11
<i>C9_10</i>	20	25	353	9.45	353	63.0	375	12.81	<i>C9_10</i>	50	25	168	3.70	≤ 167	59.1	185	11.66
<i>C9_100</i>	20	5	2864	2.24	≤ 2933	16.4	3112	6.31	<i>C9_100</i>	50	5	1514	0.50	≤ 1509	15.8	1563	3.88
<i>C9_100</i>	20	15	3642	11.15	≤ 3588	41.5	3853	10.59	<i>C9_100</i>	50	15	1675	2.13	≤ 1742	38.6	1878	8.13
<i>C9_100</i>	20	25	3602	17.26	≤ 3644	62.1	3878	14.08	<i>C9_100</i>	50	25	1674	3.35	≤ 1745	59.4	1905	12.52
<i>C10_10</i>	20	5	341	2.07	≤ 371	14.4	385	7.73	<i>C10_10</i>	50	5	156	0.39	≤ 173	13.9	184	5.71
<i>C10_10</i>	20	15	505	13.07	≤ 505	41.4	540	19.66	<i>C10_10</i>	50	15	211	2.54	≤ 221	40.4	247	15.95
<i>C10_10</i>	20	25	482	19.86	≤ 501	67.2	550	27.51	<i>C10_10</i>	50	25	219	3.83	≤ 218	62.9	250	23.49
<i>C10_100</i>	20	5	3530	1.88	≤ 3811	15.5	4055	8.89	<i>C10_100</i>	50	5	1513	0.32	≤ 1836	14.0	1937	6.27
<i>C10_100</i>	20	15	5163	13.30	≤ 5227	45.5	5621	21.35	<i>C10_100</i>	50	15	2415	2.88	≤ 2365	39.7	2550	15.69
<i>C10_100</i>	20	25	5287	22.97	≤ 5286	64.9	5703	29.11	<i>C10_100</i>	50	25	2472	4.66	≤ 2432	63.8	2586	24.18
<i>C13_10</i>	100	5	242	2.67	≤ 243	26.6	251	3.40	<i>C13_10</i>	100	15	303	12.57	≤ 296	72.8	311	6.22
<i>C13_10</i>	100	25	302	22.19	≤ 298	107.3	311	8.62	<i>C13_100</i>	100	5	2507	2.48	≤ 2526	25.9	2599	3.90
<i>C13_100</i>	100	15	3064	11.64	≤ 3029	70.4	3244	6.95	<i>C13_100</i>	100	25	3064	22.72	≤ 3057	109.8	3251	9.41
<i>C14_10</i>	100	5	344	5.33	339	26.8	368	5.20	<i>C14_10</i>	100	15	377	17.72	≤ 370	71.4	392	9.32
<i>C14_10</i>	100	25	377	28.02	369	101.8	391	12.73	<i>C14_100</i>	100	5	3485	5.02	≤ 3416	26.3	3843	6.61
<i>C14_100</i>	100	15	3846	17.32	≤ 3872	71.5	4110	11.50	<i>C14_100</i>	100	25	3846	27.43	≤ 3856	102.2	4107	14.13
<i>C15_10</i>	20	5	1174	79.48	≤ 1192	28.9	1218	30.08	<i>C15_10</i>	100	5	401	4.94	≤ 427	26.2	450	10.55
<i>C15_10</i>	100	15	515	23.48	≤ 501	71.0	558	18.11	<i>C15_100</i>	100	25	520	38.60	510	107.7	559	25.74
<i>C15_100</i>	20	5	12078	83.31	≤ 12198	28.5	12353	40.13	<i>C15_100</i>	100	5	4180	5.17	≤ 4358	27.1	4741	11.11
<i>C15_100</i>	100	15	5355	24.04	≤ 5177	70.8	5776	21.98	<i>C15_100</i>	100	25	5393	41.07	5243	106.7	5792	27.41

Table 2 Comparison between Memetic/DP and the memetic algorithm for solving the 56 most challenging cases of group G4

Instance			Memetic/DP			Memetic			Instance			Memetic/DP			Memetic		
<i>Graph</i>	<i>b</i>	<i>H</i>	<i>R^{best}</i>	<i>R^{avg}</i>	<i>t(s)</i>	<i>R^{best}</i>	<i>R^{avg}</i>	<i>t(s)</i>	<i>Graph</i>	<i>b</i>	<i>H</i>	<i>R^{best}</i>	<i>R^{avg}</i>	<i>t(s)</i>	<i>R^{best}</i>	<i>R^{avg}</i>	<i>t(s)</i>
<i>C8_10</i>	20	5	230	229.4	446.53	230	230.0	19.06	<i>C8_10</i>	50	5	116	116.0	69.99	116	116.0	10.80
<i>C8_10</i>	20	15	328	326.4	539.48	330	327.8	42.37	<i>C8_10</i>	50	15	171	168.1	70.07	171	168.5	23.03
<i>C8_10</i>	20	25	331	329.3	548.89	330	329.0	35.60	<i>C8_10</i>	50	25	172	171.6	83.17	172	171.6	31.50
<i>C8_100</i>	20	5	2380	2367.7	525.52	2380	2374.0	24.37	<i>C8_100</i>	50	5	1216	1212.0	92.18	1216	1216.0	12.30
<i>C8_100</i>	20	15	3412	3384.6	636.87	3418	3407.6	47.39	<i>C8_100</i>	50	15	1774	1757.7	109.59	1774	1759.6	26.46
<i>C8_100</i>	20	25	3429	3413.9	620.59	3443	3426.5	48.16	<i>C8_100</i>	50	25	1792	1784.8	103.36	1792	1784.2	33.34
<i>C9_10</i>	20	5	301	298.4	599.20	302	299.3	29.31	<i>C9_10</i>	50	5	149	148.9	73.95	149	149.0	16.04
<i>C9_10</i>	20	15	374	372.5	406.81	376	372.4	42.45	<i>C9_10</i>	50	15	184	182.1	107.67	183	180.8	34.77
<i>C9_10</i>	20	25	380	374.9	444.21	377	374.7	54.65	<i>C9_10</i>	50	25	186	185.8	108.47	186	184.8	43.77
<i>C9_100</i>	20	5	3112	3088.2	845.56	3112	3092.8	38.16	<i>C9_100</i>	50	5	1563	1563.0	86.07	1563	1563.0	17.57
<i>C9_100</i>	20	15	3875	3860.6	820.02	3873	3854.5	56.49	<i>C9_100</i>	50	15	1899	1871.5	121.28	1879	1863.5	36.32
<i>C9_100</i>	20	25	3912	3887.6	621.32	3906	3881.4	71.83	<i>C9_100</i>	50	25	1918	1906.9	158.94	1909	1898.6	50.11
<i>C10_10</i>	20	5	388	387.3	1009.09	388	387.4	57.24	<i>C10_10</i>	50	5	185	184.3	142.94	185	184.1	26.66
<i>C10_10</i>	20	15	551	544.9	1557.19	553	545.2	113.51	<i>C10_10</i>	50	15	247	247.0	175.81	247	247.0	61.70
<i>C10_10</i>	20	25	558	553.7	1587.71	561	555.4	133.69	<i>C10_10</i>	50	25	256	250.6	210.59	254	249.6	78.51
<i>C10_100</i>	20	5	4088	4067.3	1402.74	4069	4047.9	67.42	<i>C10_100</i>	50	5	1940	1939.4	149.69	1940	1939.3	34.83
<i>C10_100</i>	20	15	5682	5620.9	2133.47	5686	5620.0	130.25	<i>C10_100</i>	50	15	2566	2551.6	233.34	2601	2552.1	71.46
<i>C10_100</i>	20	25	5838	5738.1	1914.11	5773	5694.6	139.72	<i>C10_100</i>	50	25	2586	2579.5	260.60	2632	2576.1	92.41
<i>C13_10</i>	100	5	254	251.8	380.54	253	251.1	22.97	<i>C13_10</i>	100	15	312	310.5	266.70	316	312.0	33.71
<i>C13_10</i>	100	25	314	311.9	342.20	314	311.6	37.38	<i>C13_100</i>	100	5	2609	2590.4	450.91	2622	2593.0	29.87
<i>C13_100</i>	100	15	3253	3242.3	395.94	3255	3242.7	41.05	<i>C13_100</i>	100	25	3257	3245.6	460.48	3260	3241.2	42.62
<i>C14_10</i>	100	5	368	365.5	810.57	370	366.0	29.95	<i>C14_10</i>	100	15	396	393.4	889.66	398	394.7	55.08
<i>C14_10</i>	100	25	396	393.2	928.70	397	393.8	53.99	<i>C14_100</i>	100	5	3843	3827.6	1032.97	3847	3823.7	36.42
<i>C14_100</i>	100	15	4156	4127.2	1155.39	4172	4150.2	72.46	<i>C14_100</i>	100	25	4161	4126.2	1199.43	4150	4121.6	76.56
<i>C15_10</i>	20	5	1218	1214.7	7214.87	1221	1218.2	188.94	<i>C15_10</i>	100	5	458	453.6	1834.20	462	456.0	71.42
<i>C15_10</i>	100	15	559	557.4	941.18	558	557.5	92.59	<i>C15_10</i>	100	25	558	558.0	1131.41	558	556.4	99.20
<i>C15_100</i>	20	5	12370	12341.1	7218.16	12392	12366.5	359.13	<i>C15_100</i>	100	5	4779	4668.5	2051.51	4807	4758.0	97.01
<i>C15_100</i>	100	15	5807	5786.7	1384.34	5807	5793.8	109.24	<i>C15_100</i>	100	25	5807	5788.1	1481.98	5822	5795.7	127.45

Table 3 Comparison between ILS and the memetic algorithm for solving the 56 most challenging cases of group G4

Instance			ILS			Memetic			Instance			ILS			Memetic		
<i>Graph</i>	<i>b</i>	<i>H</i>	<i>R^{best}</i>	<i>R^{avg}</i>	<i>t(s)</i>	<i>R^{best}</i>	<i>R^{avg}</i>	<i>t(s)</i>	<i>Graph</i>	<i>b</i>	<i>H</i>	<i>R^{best}</i>	<i>R^{avg}</i>	<i>t(s)</i>	<i>R^{best}</i>	<i>R^{avg}</i>	<i>t(s)</i>
<i>C8_10</i>	20	5	230	228.4	15.64	230	230.0	19.06	<i>C8_10</i>	50	5	116	115.8	8.89	116	116.0	10.80
<i>C8_10</i>	20	15	327	325.1	25.67	330	327.8	42.37	<i>C8_10</i>	50	15	168	167.5	12.77	171	168.5	23.03
<i>C8_10</i>	20	25	330	327.8	28.60	330	329.0	35.60	<i>C8_10</i>	50	25	172	171.2	15.36	172	171.6	31.50
<i>C8_100</i>	20	5	2373	2349.1	20.29	2380	2374.0	24.37	<i>C8_100</i>	50	5	1216	1206.0	10.67	1216	1216.0	12.30
<i>C8_100</i>	20	15	3414	3375.4	34.18	3418	3407.6	47.39	<i>C8_100</i>	50	15	1764	1750.9	16.28	1774	1759.6	26.46
<i>C8_100</i>	20	25	3414	3403.4	33.54	3443	3426.5	48.16	<i>C8_100</i>	50	25	1792	1783.4	18.10	1792	1784.2	33.34
<i>C9_10</i>	20	5	301	297.1	27.39	302	299.3	29.31	<i>C9_10</i>	50	5	149	149.0	11.88	149	149.0	16.04
<i>C9_10</i>	20	15	377	373.4	30.88	376	372.4	42.45	<i>C9_10</i>	50	15	184	182.7	21.34	183	180.8	34.77
<i>C9_10</i>	20	25	379	376.5	34.32	377	374.7	54.65	<i>C9_100</i>	50	25	186	184.1	24.66	186	184.8	43.77
<i>C9_100</i>	20	5	3112	3089.3	36.65	3112	3092.8	38.16	<i>C9_100</i>	50	5	1563	1563.0	11.52	1563	1563.0	17.57
<i>C9_100</i>	20	15	3918	3882.3	34.91	3873	3854.5	56.49	<i>C9_100</i>	50	25	1888	1875.0	21.44	1879	1863.5	36.32
<i>C9_100</i>	20	25	3932	3889.2	41.76	3906	3881.4	71.83	<i>C9_100</i>	50	25	1895	1885.8	27.73	1909	1898.6	50.11
<i>C10_10</i>	20	5	386	384.9	41.42	388	387.4	57.24	<i>C10_10</i>	50	5	184	184.0	17.65	185	184.1	26.66
<i>C10_10</i>	20	15	545	540.7	58.81	553	545.2	113.51	<i>C10_10</i>	50	15	247	247.0	30.27	247	247.0	61.70
<i>C10_10</i>	20	25	553	547.4	75.24	561	555.4	133.69	<i>C10_10</i>	50	25	251	249.3	43.51	254	249.6	78.51
<i>C10_100</i>	20	5	4079	4043.7	62.36	4069	4047.9	67.42	<i>C10_100</i>	50	5	1939	1938.8	22.59	1940	1939.3	34.83
<i>C10_100</i>	20	15	5662	5584.1	83.03	5686	5620.0	130.25	<i>C10_100</i>	50	15	2550	2542.1	37.42	2601	2552.1	71.46
<i>C10_100</i>	20	25	5743	5658.9	96.00	5773	5694.6	139.72	<i>C10_100</i>	50	25	2587	2569.9	50.27	2632	2576.1	92.41
<i>C13_10</i>	100	5	252	249.7	19.11	253	251.1	22.97	<i>C13_10</i>	100	15	313	311.1	18.62	316	312.0	33.71
<i>C13_10</i>	100	25	313	311.3	20.85	314	311.6	37.38	<i>C13_100</i>	100	5	2609	2584.6	26.62	2622	2593.0	29.87
<i>C13_100</i>	100	15	3243	3238.2	28.45	3255	3242.7	41.05	<i>C13_100</i>	100	25	3260	3241.7	31.29	3260	3241.2	42.62
<i>C14_10</i>	100	5	368	365.6	29.62	370	366.0	29.95	<i>C14_10</i>	100	15	396	391.2	36.06	398	394.7	55.08
<i>C14_10</i>	100	25	393	391.4	37.37	397	393.8	53.99	<i>C14_100</i>	100	5	3811	3803.3	44.91	3847	3823.7	36.42
<i>C14_100</i>	100	15	4151	4108.1	54.37	4172	4150.2	72.46	<i>C14_100</i>	100	25	4136	4103.8	49.72	4150	4121.6	76.56
<i>C15_10</i>	20	5															