Report of Assignment – 3

Solving MAX_CUT Using GRASP

The maximum cut (MAX-CUT) problem

Given an undirected graph, G = (V, U), where V is the set of vertices and U is the set of edges, and weights w_{uv} associated with each edge $(u, v) \in U$, the maximum cut (MAX-CUT) problem consists in finding a nonempty proper subset of vertices $S \subset V$ ($S \neq \emptyset$), such that the weight of the cut (S, \overline{S}) , given by $w(S, \overline{S}) = \sum_{u \in S, v \in S} w_{uv}$, is maximized.

To find MAX_CUT of a graph, I used these 8 algorithms:

Constructive Algorithm:

- Random
- Greedy
- Semi-greedy

Local Search:

- LS:1 with random initialization
- LS:2 with greedy initialization,
- LS:3 with semi-greedy (α = 0.7)

GRASP:

- GRASP-1 with maximum iteration = 3
- GRASP-2 with maximum iteration = 50

P	roblem	1	Construc	tive Algor	ithm	L	ocal sear	ch				GRASI	1		My Best Value	Upper Bound
lame	V	E	Randomize Greedy			Local-1(random)		Local-2(greedy		Local-3(alpha=0.		GRASP-1(max_it=3)	GRASP-2(max_it=5			
							est Value		Best value		Best value		ue Iteration	Best value		
51		19176	11032	11268	11108		11316				11382	286 114			11466	12078
62		19176	11091	11292	11203		11344		11365		11426	282 114			11463	12084
3		19176	11049	11295	11229		11316	108	11447	121	11388	281 114			11469	12077
64		19176	10974	11231	11199		11391	69	11400		11421	261 114			11489	
55		19176	11070	11264	11277		11400	64	11407		11426	258 113			11468	
66		19176	1501	1744	1722		1851	121	1910		1924				2009	
67	800	19176	1413	1601	1502	154	1708	83	1797	131	1721	324 17	80 5216	1854	1854	
86	800	19176	1410	1609	1585	195	1736	92	1801	158	1787	274 18	08 4969	1868	1868	
9	800	19176	1471	1631	1515	180	1783	92	1807	120	1800			1904	1925	
10		19176	1247	1599	1561		1746		1759		1809		85 5033		1841	
11		1600	390	480	428		426		480		470		78 392		494	627
12		1600	406	474	434		418		494		452		70 407		494	621
13		1600	434	510	470		460	3	508		472		00 422		510	645
14		4694	2865	2949	2944		2921	21			2973		73 1210		2988	3187
15		4661	2842	2915	2918		2921	18	2932		2944		56 1248		2976	3169
			2855	2915	2923			29			2944				2980	3172
16		4672					2917					70 29				31/2
17		4667	2875	2925	2917	46	2933	22	2948		2947		48 1263		2972	
18		4694	700	816	744		782		866		840		59 2169		911	
19		4661	587	724	668		778				784		89 2073		824	
20		4672	628	796	751		755				748		28 2082		858	
21		4667	622	748	716		813		824		766		97 2168		849	
22		19990	12460	12810	12784		12839		12968		12911	362 129			13016	14123
23		19990	12396	12790	12729		12852	126	12965		12958	311 129			13005	14129
24		19990	12290	12820	12719		12877	91	12906		12882	348 129			13023	14131
25		19990	12311	12773	12768	256	12865	100	12943		12980	306 129	63 5500		13013	
26	2000	19990	12332	12772	12698	250	12801	115	12953	125	12947	339 129	59 5438	12990	12990	
27		19990	2305	2712	2604		2782	133	3005	196	2839	447 29			3005	
28		19990	2281	2670	2474	305	2766	106	2861	165	2818		84 7535		2938	
29		19990	2314	2734	2614		2928	103	2962		2881		94 7830		3045	
30		19990	2390	2808	2599		2815		2950		2974		07 7549		3037	
31		19990	2340	2647	2599		2844	103	2759		2812		98 7688		2982	
32		4000	1032	1196	1112		1086	16	1218		1132		98 916		1230	1560
33		4000	998	1214	1024		1074		1194		1128		82 922		1214	1537
34	2000	4000	998	1176	986		1074		1194		1144		80 945		1192	1541
35			7202	7373	7366		7362	51	7450		7466		63 2823		7476	8000
		11778														
36		11766	7170	7385	7371		7351	54	7427		7454		32 2805		7468	7996
37		11785	7209	7380	7351	138	7382	63	7454		7458	179 74			7479	8009
38		11779	7196	7369	7364		7349		7460		7428		43 2954		7480	
39		11778	1705	2002	1832		2038		2147		2060		43 5724		2171	
40		11766	1715	2005	1884		1995	98	2130		2052		37 5185		2149	
41		11785	1679	1934	1816		1987	73	2136		2031		80 4746		2143	
42		11779	1773	2129	1801		2067	48	2184		2113		84 5194		2238	
43		9990	6170	6392	6347	102	6361	74	6461		6492		05 2922		6511	7027
44		9990	6116	6421	6343	140	6400	52	6455		6406	191 64	71 3109	6514	6514	7022
45		9990	6196	6380	6359	183	6410	42	6446		6464		83 2912		6512	7020
16		9990	6205	6389	6385	129	6411	35	6425	44	6452		83 2997		6494	
17		9990	6134	6391	6375		6389	49	6493		6460		72 3019		6508	
48		6000	5004	6000	5816		5096	1	6000	2	5742		90 143		6000	6000
49		6000	4952	6000	5810		5148	1	6000	4	5738		94 121		6000	6000
50		6000	4828	5880	5798		5156		5880		5716		88 120		5880	5988
51		5909	3609	3679	3679		3671	29	3725	26	3712	94 37			3751	2300
52		5916	3627	3711	3690		3708		3743		3734		38 1509		3761	
553		5914	3623	3685	3677		3688	25	3740		3732		42 1391		3757	
554	1000	5916	3615	3706	3690	58	3675	20	3733	27	3724	88 37	40 1461	3754	3754	

Figure: The resultant table

From the table it is clear, that the best solutions among all these algorithms are almost near Upper Bounds.

In fact, 2 cases (G48 and G49) were able to achieve the upper bound.

Name	V	E	My Best Value	Upper Bound
G48	3000	6000	6000	6000
G49	3000	6000	6000	6000

We also can see, in 46 cases out of 54 cases the best Maximum Cut value among all the algorithms came from **GRASP-2 with maximum iteration=50** which is colorless section in **MY BEST VALUE** column.

The 8 exceptions are highlighted using red color in the MY BEST VALUE column.

The most noticeable trend in most cases is the improvement of Maximum Cut value as we go from left side of the table to Right table of the table. So, our most efficient algo sequence:

- 1. GRASP-2 with maximum iteration = 50
- 2. GRASP-1 with maximum iteration = 3
- 3. LS:3 with semi-greedy ($\alpha = 0.7$)
- 4. LS:2 with greedy initialization
- 5. LS:1 with random initialization
- 6. Semi-greedy
- 7. Greedy
- 8. Random

In general,

GRASP > Local Search > Constructive Algorithms

Conclusion: This result is predictable as constructive algorithm always gives the best solution in one iteration. Local search is an improvement over constructive algorithm as it always takes us to local maximum. GRASP is further improvement of local search as it random initialize and start local search a few times.