

Abstract

We report the PASCAL2 benchmark for DAOOPT and GUROBI on MPE task with 330 optimally solved instances from 8 benchmark domains. DAOOPT outperformed GUROBI in 3 domains, while GUROBI was faster than DAOOPT in the rest of the 5 domains. We show that DAOOPT performed well in domains where it could have high quality initial solutions for pruning the AND/OR search space, or skip search when the heuristic upper bounds were converged to the optimal due to MPLP/JGLP algorithms. GUROBI presented excellent performance if cutting planes were applied progressively and its heuristic algorithms could find the optimal solution at the root of branch-and-cut tree.

2011 PASCAL2 Inference Challenge

- Evaluate Inference Algorithms on Large-scale Difficult Problems
- Problems were provided as UAI Factor Table Format
- <http://www.cs.huji.ac.il/project/PASCAL/>

DAOOPT (Distributed AND/OR Optimization)

- DAOOPT was the winner of MAP/MPE Task
- Search Based on
 - Breadth Rotating AND/OR Branch-and-Bound
- Static Heuristic Evaluation From
 - Mini-Bucket Elimination with Moment Matching
 - Factor Graph/Join Graph Cost Shifting Schemes (MPLP/JGLP)
- Initial Lower Bound (Heuristic Solution) From
 - Guided Local Search++ (SLS4MPE Open Source)
- Variable Ordering From
 - Stochastic Greedy Ordering Scheme

GUROBI v5.5 (Currently, v5.6)

- Commercial Mathematical Programming Solver
- <http://www.gurobi.com>

ILP Formulation of MPE Task

- MPE Inference
 - Graphical Model $R = \langle \mathbf{X}, \mathbf{D}, \mathbf{F}, \otimes \rangle$
 - Random Variables $\mathbf{X} = (x_1, x_2, \dots, x_n)$
 - Domains of Variables $\mathbf{D} = (D_1, D_2, \dots, D_n), x_i \in D_i$
 - Factors $\mathbf{F} = \{P_1, P_2, \dots, P_r\}$
 - $S_j = \text{Scope}(P_j)$
- MPE Task $\max_{\mathbf{x}} P(\mathbf{x}) = \max_{\mathbf{x}} \prod_j P_j(\mathbf{x})$
- MPE to 0/1 ILP Conversion
 - Decision Variables $\mathbf{x}_j^k \in \{0,1\}$
 - k-th row of j-th Factor Table
 - Selecting a row for factor tables
 - Constraints
 - Mutual Exclusivity $\sum_{k=1}^{N_j} x_j^k = 1, \forall j \in \{1..r\}$
 - Cross Consistency $N_j = |D_{Var1} \times D_{Var2} \dots \times D_{Var|S_j|}|$
 - Objective Function $\sum_{k \sim S} x_j^k = \sum_{l \sim S} x_m^l, \forall j, m \in \{1..r\}$
- Objective Function $\max_{\mathbf{x}} \sum_{j \in \{1..r\}} \sum_{k=1}^{N_j} x_j^k (-\log(P_j^k)) = \min_{\mathbf{x}} \sum_{j \in \{1..r\}} \sum_{k=1}^{N_j} x_j^k f_j^k$

Benchmark Domains

- Benchmark Status
 - 12 Domains, 496 Instances Tested
 - We Report Results From 8 Domains with 328 Instances
 - Optimally Solved by DAOOPT or GUROBI)
 - Grid.Markov, Segmentation, WCSP Grouped by difficulty level

Terminated?	Yes	No	Total	Terminated?	Yes	No	Total
Grid.Bayes* (UAI08)	32	0	32	WCSP	18	43	61
Pedigree*(UAI08)	22	0	22	WCSP.Spot5	6	14	20
Grid.Markov.20	5	0	5	WCSP.Easy	8	0	8
Grid.Markov.40	8	0	8	WCSP.Hard	4	0	4
Grid.Markov.80	8	0	8	DBN	60	48	108
Segmentation.K2	50	0	50	Alchemy	1	0	1
Segmentation.K21	50	0	50	Protein-Protein	0	8	8
Promedas	68	0	86	Object Detection	0	37	37
Protein Folding	7	3	10	Image Alignment	0	10	10

Benchmark Problem Statistics

- Problem Statistics for Graphical Models and Integer Programming

Name	#	n _{min} /n _{max}	f _{min} /f _{max}	k _{min} /k _{max}	s _{min} /s _{max}	w _{min} /w _{max}	h _{min} /h _{max}	row _{max}	col _{max}
Grid(Bayes)	32	144/2500	145/2501	2	3	16/81	53/309	17102	19602
Pedigree	22	298/1015	335/1290	3/7	4/5	15/35	53/160	8429	9986
Grid(Markov).20	5	400	1161/1201	2	2	26/46	66/73	9201	4001
Grid(Markov).40	8	1600	4721/4801	2	2	53/95	144/149	36801	16001
Grid(Markov).80	8	6400	19041/19201	2	2	107/197	307/313	147201	64001
Segmentaion.K2	50	221/237	823/887	2	2	15/19	44/68	9699	3071
Segmentaion.K21	50	221/237	823/887	21	2	15/19	44/68	93451	291187
Promedas	68	196/1911	201/1928	2	3	4/116	33/162	9856	11565
WCSP.spot5	6	67/209	272/1395	4	2/3	6/26	15/87	88620	12390
WCSP.Easy	8	25/71	82/686	3/11	2	6/21	15/30	168583	9479
WCSP.Hard	4	16/179	208/7110	4/44	2	7/42	8/90	104103361	340305
ProteinFolding	7	337/1364	1360/5220	81	2	22/38	58/164	1030679	2699598
DBN	60	70	16167	2	2	29	29	12986347	34691

- #. Total Number of Instances
- n Number of Variables
- f Number of Factors
- k Maximum Domain Size
- s: Maximum Scope Size
- w: Width of the Best Ordering Found
- h: Height of the Pseudo Tree
- row/col: Size of Constraint Matrix

Benchmark Settings

- DAOOPT Input Parameters

PARAMETERS	SET1	SET2	SET3
MBE-MM i bound	Maximum i Selected by DAOOPT within 4GB Memory Limit		
MPLP Iteration	2 sec	30 sec OR 500 iter	60 sec OR 2,000 iter
JGLP Iteration	2 sec	30 sec OR 250 iter	60 sec OR 2,000 iter
CVO Iteration	3 sec OR 500 iter	60 sec OR 10,000 iter	180 sec OR 30,000 iter
SLS Time Limit	2 x 2 sec	10 x 6 sec	20 x 10 sec
BRAOBB Time	To Termination or Time-out Greater Than GUROBI Termination Time		

- GUROBI Input Parameters
 - Default, Except Explicitly Setting Root LP-Relaxation as Dual-Simplex
- Benchmark Platform
 - Turn-off Parallel Processing Capability, Disk Usage Algorithms

RESOURCE	TOTAL AVAILABLE	RESTRICTED FOR BENCHMARK
CPU	2 x 2.66 GHz Intel Core2 Duo	Single Thread
Memory	6 GB	MAX 4GB

Benchmark Results

- Runtime Summary Per Domains

Name	#	Param	i-bound*	A.M of Td	A.M of Tg	G.M of Td	G.M of Tg	A.M of (Tg/Td)
GUROBI WAS FASTER THAN DAOOPT								
Grid(Bayes)	32	Set 1	(16,22,31,25)	382.8	2.7	51.01	1.35	0.038
Pedigree	22	Set 2	(10,17,91,23)	232.4	27.3	181.41	9.66	0.096
Grid(Markov).20	5	Set 1	(20,20,20)	14.6	7.2	14.28	7.21	0.524
Grid(Markov).40	8	Set 2	(20,20,20)	212.1	84.1	212.1	78.22	0.396
Grid(Markov).80	8	Set 3	(18,18,18)	T-out	2125	T-out	1661	-
Segmentaion.K2	50	Set 1	(15,17,14,19)	7.14	0.26	7.12	0.26	0.037
Segmentaion.K21	50	Set 1	(4,4,4)	55.4	11.36	55.33	11.23	0.205
Promedas	68	Set 1	(4,20,6,27)	143.71	0.2	40.57	0.16	0.006
DAOOPT WAS FASTER OR SOLVED YET GUROBI WAS MEMORY OUT								
WCSP.spot5	6	Set 1	(6,12,19)	59.5	197.65 (5)**	18.91	47.31 (5)	23.23
WCSP.Easy	8	Set 1	(6,10,15)	43	620.33	19.88	94.99	14.83
WCSP.Hard	4	Set 1	(3,7,5,11)	1208 (3)	5388 (2)	729 (3)	5386 (2)	-
ProteinFolding	7	Set 3	(3,3,3)	3005	79 (1)	1004	79 (1)	-
DBN	60	Set 3	(28,28,28)	342.35	M-out	342.35	M-out	-

- DAOOPT was Faster or DAOOPT Solved yet GUROBI was Memory Out
 - WCSP (15/18), Protein Folding (6/7), DBN (60/60) Domains
- GUROBI Dominated GUROBI for Rest Domains
 - Grid.Bayes (32/32), Grid.Markov (21/21), Pedigree (22/22)
 - Segmentation (100/100), Promedas (68/68)
- Log Summary Per Domains
 - DAOOPT
 - SLS Found Optimal Solutions for Segmentation, Protein Folding, WCSP, DBN.
 - Exact Upper Bounds from MPLP+JGLP+MBE-MM for Segmentation.
 - GUROBI
 - Fast when GUROBI Solved Near at the Root Node
 - Grid Bayes, Grid Markov(average 0.5 Nodes), Grid.Markov(average 67 Nodes)

Name	SLS-OPT GAP	SLS-UB GAP	Mean OR	Mean Nodes	Mean Iter	Mean 0-1/2 Cuts
mean n,f,k,s,i,w,h	#s/s hit opt.	#s/s hit ub.	Mean AND	# solved at root	Mean Gomory	Mean Clique
Grid(Bayes) (649,650, 2, 3,22,35,121)	0.40%	0.95%	29779829.13	0.5	4804.16	71.69
	14/28	8/28	32069771.13	24/32	5.59	68.66
Pedigree (736,918,5,4,18,24,99)	0.01%	0.76%	14040647	210.32	41678.45	240.64
	15/22	4/22	17328814.82	1/22	5.95	14.32
Grid(Markov).20 (400,1185, 2, 2,20,38,70)	1.28%	1.39%	1326.2	0	4259.8	336.4
	0/5	0/5	1379.2	5/5	10.4	7.2
Grid(Markov).40 (1600,4761, 2, 2,20,74,147)	1.86%	1.90%	24042.38	1.13	16378.25	994.88
	0/8	0/8	24835.75	7/8	16.25	26
Grid(Markov).80 (6400,19121, 2, 2,18,152,310)	2.46%	2.59%	T-Out	176.88	101722.5	3828.5
	0/8	0/8	T-Out	4/8	42.75	513.13
Segmentaion.K2 (229,852, 2, 2,17,17,55)	0.00%	0.00%	133.34	0	1225	0
	50/50	50/50	134.34	50/50	0	0
Segmentaion.K21 (229,852, 21, 2,4,17,55)	0.00%	0.00%	96.08	0	4080.6	0
	50/50	50/50	97.08	50/50	0	0

References

- [1] L. Otten, A. Ihler, K. Kask, and R. Dechter. Winning the pascal 2011 map challenge with enhanced and/or branch-and-bound. In NIPS Workshop DISCML, 2012.
- [2] Z. Gu, E. Rothberg, and R. Bixby. Gurobi optimizer v5.5. <http://www.gurobi.com>.
- [3] The 2011 PASCAL2 Probabilistic Inference Challenge. <http://www.cs.huji.ac.il/project/PASCAL/>.
- [4] R. Marinescu. AND/OR Search Strategies for Combinatorial Optimization in Graphical Models. PhD thesis, University of California, Irvine, 2008.
- [5] L. Otten. Extending the Reach of AND/OR Search for Optimization in Graphical Models. PhD thesis, University of California, Irvine, 2013.
- [6] A. Globerson and T. S. Jaakkola. Fixing max-product: Convergent message passing algorithms for maplp-relaxations. In NIPS, pages 553-560, 2007.
- [7] R. Dechter A. Ihler, N. Flerova and L. Otten. Join-graph based cost-shifting schemes. 2012.
- [8] R. Dechter N. Flerova, A. Ihler and L. Otten. Mini-bucket elimination with moment matching. In NIPS Workshop DISCML, 2011.
- [9] Rina Dechter and Irina Rish. Mini-buckets: A general scheme for bounded inference. Journal of the ACM (JACM), 50(2):107-153, 2003.
- [10] L. Otten K. Kask, A. Gelfand and R. Dechter. Pushing the power of stochastic greedy ordering schemes for inference in graphical models. In AAAI, 2011.
- [11] Frank Hutter. SLS4MPE. <http://www.cs.ubc.ca/labs/beta/Projects/SLS4MPE/>.
- [12] D. Koller and N. Friedman. Probabilistic Graphical Models: Principles and Techniques, pages 577-579. The MIT Press, 2009.