

# 1008 Meeting

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# Global Setting

- Running on AMD EYYC 9654 / 9684X with A100, thread setting = 192
- 50 seeds on  $ER(500, 0.5)$  for testing with fixed number of initialization (10 by default), and 10 seeds on  $ER(n, 0.5)$  for testing with fixed time.
- Maximum iteration set to 15000 for fixed initialization testing and 150 for MGD fix time testing when  $n \leq 2000$  and 300 otherwise. For Projected Newton and L-BFGS-B, we use 15000 for all experiments.

# Methods

- CP-SAT: Widely used as baselines in combinatorial optimization studies. Take the integer programming form of the maximum independent set problem and solve.
- MGD (pCQO) / Projected Newton / Scipy's L-BFGS-B: Solve a relaxed optimization problem of maximum independent set that is continuous and differentiable with gradient based methods. Objective function are taken from pCQO paper with the clique informed penalty term.

# Methods

- Newton-Ours (**single threaded**): Projected Newton with a ‘free-set’ masking to determine points that can move legally. Only these points are computed in the algorithm. Maintain a sub-Hessian matrix for computation efficiency and recomputes only when the set of legal points or the damping matrix change. Armijo update with initial steps and backtracking scalar finetuned by grid search.
- L-BFGS-Ours (**single threaded**): Handcrafted L-BFGS also with the free set. Armijo update with initial steps and backtracking scalar finetuned by grid search.

Graph, Time	MGD	CP-SAT	Projected Newton	L-BFGS-B	Newton-Ours	L-BFGS-Ours-10	L-BFGS-Ours-20	Yongliang's -5 (with / without clique penalty)	Yongliang's -10 (with / without clique penalty)
ER(500,0.5), 5 sec	12.1	12.1	11.5	10.6	11.9	12.1	12.0	11.3 / 11.3	12.1 / 12.0
ER(1000,0.5), 5 sec	12.4	12.0	11.6	11.2	12.8	11.8	11.8	12.2 / 12.6	13.0 / 13.1
ER(1000,0.5), 10 sec	12.7	12.5	11.9	11.8	13.0	12.2	12.2	12.7 / 12.6	13.3 / 13.3
ER(2000,0.5), 10 sec	12.9	0	7.2	12.3	13.2	12.5	12.5	13.8 / 14.0	13.5 / 13.9
ER(2000,0.5), 20 sec	13.2	11.2	10.7	12.4	13.4	12.7	12.7	14.1 / 14.1	13.7 / 13.9
ER(3000,0.5), 30 sec	13.3	0	7.9	0	13.9	13.1	13.4	14.1 / 14.2	14.6 / 15.0
ER(5000,0.5), 50 sec	13.3	0	5.1	0	14.4	13.8	13.9	14.7 / 15.0	15.1 / 15.3

# Gorubi Doesn't Perform

- Compared to CP-SAT, Gorubi performs worse for the task of solving integer programming here unless the given time is too little for CP-SAT to produce even an initial sub-optimal solution.

Graph, Time	CP-SAT	Gurobi
ER(500,0.5), 5 sec	12	8.7
ER(1000,0.5), 5 sec	0	9.5
ER(1000,0.5), 10 sec	11.4	9.5

This comparison is run on my local environment.

# Free-Mask for Our Newton / L-BFGS

- Determine variables that are able to make meaningful moves.

Build a set  $F$ , for each coordinate  $i$ :

1. Interior: If  $0 < x_i < 1$ , *then*  $x_i \in F$
2. At boundaries:  $x_i$  at boundaries and the gradient is moving inward, *then*  $x_i \in F$
3. Otherwise,  $x_i \notin F$ .

# Free-Mask for Our Newton

- In normal Newton, we compute  $(H + \lambda I)d = -g$ .
- Here, we consider  $S_F \in \mathbb{R}^{n \times |F|}$  whose columns are the standard basis vectors  $e_i$  for  $i \in F$ .
- $d = S_F d_F$  where  $d_F \in \mathbb{R}^{|F|}$  is the step on free variables.
- Our method do  $(H_{FF} + \lambda I)d_F = S_F^T (H + \lambda I) S_F d_F = -S_F^T g = -g_F$



# Free-Mask for Our Newton

Now, caching the Cholesky of  $H_{FF} + \lambda I$  for each pair of  $(F, \lambda)$ .

Define  $M_F(\lambda) = H_{FF} + \lambda I$ , compute  $M_F(\lambda) = LL^T$  once, which unlock the ability to compute  $d_F = L^{-T} L^{-1}(-g_F)$  for steps with the same pair of  $(F, \lambda)$ .

Often,  $|F| \ll n$ , so even if we must refactor often, it's still cheaper than doing computation on the full space.

# Free-Mask for Our L-BFGS

- Similarly, with free mask we are doing the problem on the active coordinates.
- Avoid meaningless or harmful computation; projections break the secant equation at boundaries, and we don't need to compute steps that we are going to zero out anyway.

# Scipy's L-BFGS-B and Our L-BFGS

- Comparison by solution size: (L-BFGS-B indicates Scipy's solver)

	L-BFGS-B	Ours
Average solution size	11.28	10.80

- Comparison by iteration number/time:

	L-BFGS-B	Ours
Average iteration number	167.1	23.1
Average time	2.2564	0.0613

# Projected Newton and Our Newton

- Comparison by solution size: (L-BFGS-B indicates Scipy's solver)

	Projected Newton	Ours
Average solution size	11.08	11.1

- Comparison by iteration number/time:

	Projected Newton	Ours
Average iteration number	76.91	20.69
Average time	0.1665	0.0155

# Next Meeting

- Determining how to choose a good max iter
- Determining or controlling memory length for L-BFGS if using
- Threading for our methods
- ReduMIS / iSCO baseline comparison
- Try on Max Cut / TSP

# Appendix

# Outlier Case

- We have one run taking much more steps

```
init 0 → MIS size 9, iters 37, time 0.0278s
init 1 → MIS size 8, iters 19, time 0.0152s
init 2 → MIS size 9, iters 29, time 0.0216s
init 3 → MIS size 10, iters 18, time 0.0123s
init 4 → MIS size 10, iters 31, time 0.0220s
init 5 → MIS size 10, iters 14, time 0.0099s
init 6 → MIS size 10, iters 21, time 0.0140s
init 7 → MIS size 10, iters 15, time 0.0101s
init 8 → MIS size 10, iters 16, time 0.0109s
init 9 → no MIS flagged (iters 15000, time 23.4302s)
```

# Single Thread

- The same piece of Newton / L-BFGS code.
- Average time with same setting

Environment / Method	Newton	Scipy's L-BFGS-B
My Local Env	0.1350	0.7411
AMD24	0.1665	2.2564