

Non-Convex MIQCP in Gurobi 9.1: New Advances



GUROBI
OPTIMIZATION

The World's Fastest Solver

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Non-Convex MIQCP Performance

• Presolve		33%
• Detect parallel Q constraints	4%	
• Add Q equations to Q constraints to cancel quadratic terms	18%	
• Add Q equations to objective to cancel quadratic terms	0%	
• Improved bilinear probing code	3%	
• Accept small bound changes for variables that appear in quadratic terms	2%	
• Allow substitutions on variables that appear in linear part of Q constraints	1%	
• Feasibility-based bound tightening on variables in quadratic terms	2%	
• QCP to Bilinear Translation		36%
• Also disaggregate Q constraints with only positive squares plus linear terms	8%	
• Convert positive squares of objective into constraint	14%	
• Clean up translation code to save some work	1%	
• Reuse product variables across bilinear and convex Q constraints	9%	
• Node presolve		4%
• Tighten finite bound for variables with one infinite bound	1%	
• Re-propagate bilinear constraints if domain of mixed product term changed	1%	
• Faster propagation for disjoint product terms	1%	
• Exploit implied quadratic equations in propagation	1%	

Time limit: 10000 seconds
Speed-ups on 217 models that take at least 1 second

MINLP Performance – Summary

• Branching		15%
• Adjust balance of McCormick volume and violation scores	15%	
• Cuts		22%
• Tangent cuts for convex part of bilinear constraints	1%	
• Tilt tangent cuts to increase Euclidean violation	1%	
• Exploit implied quadratic equations in cuts	4%	
• PSD cuts	15%	
• Primal Heuristics		2%
• Randomize order for greedy Q term coverage in fix-and-dive	1%	
• Consider quadratic constraints in a sub-MIP heuristic	1%	
• Simplex/MIP Integration		2%
• Add bias to favor moving McCormick constraints into basis	2%	
• Other Improvements		50%
• Including effects of MIP/LP/QP/QCP improvements		
• Total		4.11x

Time limit: 10000 seconds
Speed-ups on 217 models that take at least 1 second

Parallel Quadratic Constraints

- **Identify quadratic constraints that are parallel to each other**

- Example from customer model:

$$\begin{aligned} 0.259286x_{155} + \dots + 0.259286x_{7563} - x_{18079} + [+2x_{18078}*x_{18079}] &\leq 1 \\ - 0.259286x_{155} - \dots - 0.259286x_{7563} + x_{18079} + [-2x_{18078}*x_{18079}] &\leq -1 \end{aligned}$$

- Can be merged into equation:

$$0.259286x_{155} + \dots + 0.259286x_{7563} - x_{18079} + [+2x_{18078}*x_{18079}] == 1$$

- Other case: discard identical or dominated constraint

- **Happens frequently in sub-MIPs solved by primal heuristics**

- **Detection is very similar to linear case**

- Hash function for linear and quadratic parts (normalize for sign/scaling)
- Pairwise comparison of constraints with identical hash value
- Very fast in practice

Parallel Quadratic Constraints

- **Affects about 20% of models in non-convex MIQCP test set**
 - 11% speed-up on those models
 - 4% speed-up overall
 - 12 consistent wins, 0 consistent losses

Substitute Identical Quadratic Part

- Different linear part, but identical quadratic part

- **Case 1**

- At least one constraint is an equation

$$\begin{aligned}a^1x + x^T Qx &= b^1 \\ a^2x + x^T Qx &\leq b^2\end{aligned}$$

- Subtract equation from other constraint turns other into linear constraint

$$\begin{aligned}a^1x + x^T Qx &= b^1 \\ (a^2 - a^1)x &\leq b^2 - b^1\end{aligned}$$

- **Case 2**

- Both constraints are inequalities

$$\begin{aligned}a^1x + x^T Qx &\leq b^1 \\ a^2x + x^T Qx &\leq b^2\end{aligned}$$

- Introduce auxiliary variable to represent quadratic part

$$\begin{aligned}a^1x + s &\leq b^1 \\ a^2x + s &\leq b^2 \\ x^T Qx - s &\leq 0\end{aligned}$$

Substitute Identical Quadratic Part

- **Affects about 27% of models in non-convex MIQCP test set**
 - 39% speed-up on those models
 - 18% speed-up overall
 - 22 consistent wins, 0 consistent losses

- **Reducing Q part of objective affects only 7 models (< 2%)**
 - 5% speed-up on those models
 - 0.1% speed-up overall
 - 0 consistent wins, 0 consistent losses

PSD Cuts

- **New cutting plane separator in Gurobi 9.1 for non-convex MIQCPs**
 - Controlled by PSDCuts parameter
- **Sherali and Fraticelli (2002):**
 - “Enhancing RLT relaxations via a new class of semidefinite cuts”
- **Qualizza, Belotti and Margot (2012):**
 - “Linear Programming Relaxations of Quadratically Constrained Quadratic Programs”

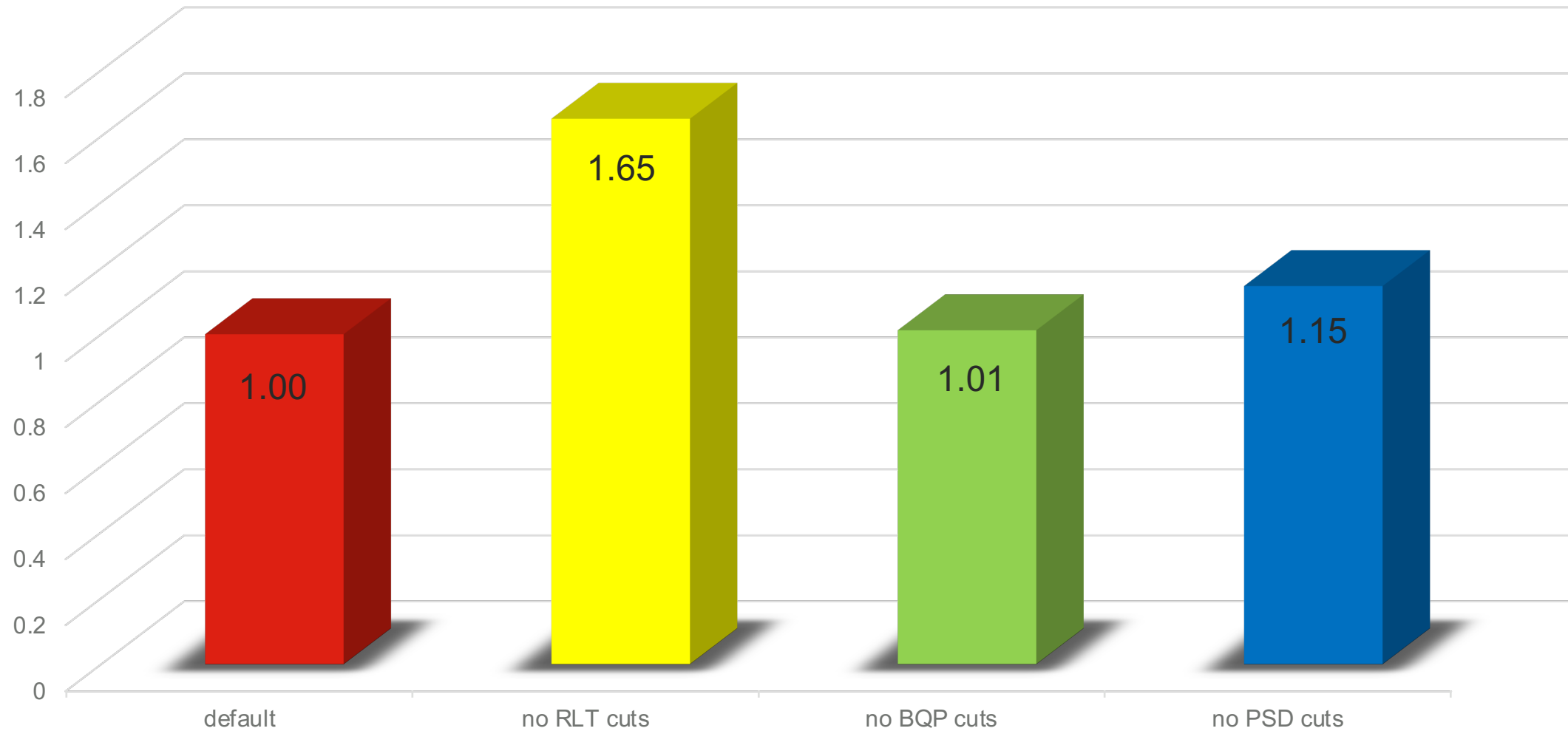
- **Bilinear constraints:** $y_{ij} = x_i x_j$
- **As matrix equation:** $Y = xx^T$
- **Relax to:** $Y \succeq xx^T \Leftrightarrow Y - xx^T \succeq 0$ (matrix is PSD)
- **Schur’s complement:** $Y - xx^T \succeq 0 \Leftrightarrow \begin{pmatrix} 1 & x^T \\ x & Y \end{pmatrix} \succeq 0$
- **Equivalent to:** $v^T \begin{pmatrix} 1 & x^T \\ x & Y \end{pmatrix} v \geq 0$ for all $v \in \mathbb{R}^{n+1}$
- **Separate cuts by finding v for which this is violated**
 - Eigenvectors corresponding to negative eigenvalues

- **We separate PSD cuts for up to 10 product variables**
 - Only use those sets where all mixed y_{ij} variables exist
 - Find cliques in graph with nodes x_j and edges y_{ij}
- **Example for single mixed product variable y_{ij}**
 - Find v with $v^T \begin{pmatrix} 1 & x_i^* & x_j^* \\ x_i^* & y_{ii}^* & y_{ij}^* \\ x_j^* & y_{ij}^* & y_{jj}^* \end{pmatrix} v < 0$ for current LP solution (x^*, y^*)
 - Find negative eigenvalue, let v be the corresponding eigenvector
 - Add cut $v^T \begin{pmatrix} 1 & x_i & x_j \\ x_i & y_{ii} & y_{ij} \\ x_j & y_{ij} & y_{jj} \end{pmatrix} v =$
$$v_1^2 + 2v_1v_2x_i + 2v_1v_3x_j + v_2^2y_{ii} + v_3^2y_{jj} + 2v_2v_3y_{ij} \geq 0$$

PSD Cuts

- **Affects about 34% of models in non-convex MIQCP test set**
 - 32% speed-up on those models
 - 15% speed-up overall
 - 28 consistent wins, 5 consistent losses

Non-Convex MINLP Cuts Summary



Bilinear Inequalities and Cuts

- Recall PSD cuts, formulated with y_{ij} variables

$$\bullet v^T \begin{pmatrix} 1 & x_i & x_j \\ x_i & y_{ii} & y_{ij} \\ x_j & y_{ij} & y_{jj} \end{pmatrix} v =$$

$$v_1^2 + 2v_1v_2x_i + 2v_1v_3x_j + v_2^2y_{ii} + v_3^2y_{jj} + 2v_2v_3y_{ij} \geq 0$$

- **But actually, one can view this as a two step process**
 - Formulate quadratic cut in x_j variables
 - $v_1^2 + 2v_1v_2x_i + 2v_1v_3x_j + v_2^2x_i^2 + v_3^2x_j^2 + 2v_2v_3x_ix_j \geq 0$
 - Substitute quadratic terms for y variables using $y_{ij} = x_ix_j$
 - But what if we only have $y_{ij} \leq x_ix_j$ or $y_{ij} \geq x_ix_j$?

Bilinear Inequalities and Cuts

- **Same question for RLT and PSD cuts**

- Given a quadratic cut

$$\sum a_{ij}x_ix_j \leq b$$

and relations

$$y_{ij} = x_ix_j, y_{ij} \leq x_ix_j, \text{ or } y_{ij} \geq x_ix_j$$

how can we derive a valid linear cut?

- **Need to look at signs of a_{ij} coefficients**

- $a_{ij} > 0$: can only use $y_{ij} = x_ix_j$ and $y_{ij} \leq x_ix_j$
- $a_{ij} < 0$: can only use $y_{ij} = x_ix_j$ and $y_{ij} \geq x_ix_j$
- If not compatible: need to relax term
 - E.g., by substituting bounds for x_i and x_j that minimize $a_{ij}x_ix_j$

- **Observation: bilinear equations help to find better cuts**

Implied Quadratic Equations

- Consider a quadratic inequality

$$a'x' + ax + y^T Qy \leq b$$

with

- the linear part partitioned into ax and $a'x'$, and
- the set of variables in the quadratic part being disjoint from the linear part
- **The inequality is an implied equation if**
 - for any x' and y we can always move ax upwards until we hit b , or
 - for any x' and x we can always move $y^T Qy$ upwards until we hit b .

Implied Quadratic Equations

- **Exploit implied quadratic equations in**
 - Cuts
 - Allows more substitutions of bilinear terms by product variables
 - RLT cuts
 - PSD cuts
 - BQP cuts
 - Propagation
 - Propagate constraint in opposite direction
 - Node presolve
 - Fix-and-divide heuristics
 - Branching
 - Update shadow costs of variable for both directions

Implied Quadratic Equations

- **Affects about 32% of models in non-convex MIQCP test set**
 - 9% speed-up on those models
 - 4% speed-up overall
 - 8 consistent wins, 1 consistent loss

Non-Convex MIQCP Performance

Gurobi 9.0 vs. 9.1

Total run-time over all 1524 models in log files: 7282968 sec = 2023.0 h = 84.3 d

Full set	Count	Loss/Win	NodeR	IterR	VMemR	TimeR	
all:	729	52/ 154	0.577	0.618	0.847	0.657	
>0s:	413	52/ 154	0.363	0.433	0.739	0.442	
>1s:	217	47/ 140	0.169	0.226	0.649	0.245	← 4.1x speed-up
>10s:	161	31/ 112	0.111	0.156	0.583	0.158	
>100s:	123	19/ 94	0.073	0.116	0.517	0.104	← 9.6x speed-up
>1000s:	85	14/ 66	0.062	0.103	0.462	0.081	

Unsolved: (31 / 4) + 316 for all solvers
- Time limit: (34 / 8) + 307 for all solvers
- Mem limit: (4 / 3) + 2 for all solvers
No feasible: (14 / 4) + 111 for all solvers

Non-Convex MIQCP Performance

Other Solvers vs. Gurobi 9.1

- **Comparison of other solvers vs. Gurobi 9.1** conducted by Prof. Hans Mittelmann on models from QPLIB
 - See <http://plato.asu.edu/bench.html>
 - Gurobi 9.0 results from 8 October 2020 (discrete non-convex) and 10 October 2020 (continuous non-convex)
 - Gurobi 9.1 results from 10 November 2020 (discrete non-convex) and 2 December 2020 (continuous non-convex)
 - Antigone, BARON, FSCIP, Couenne, Minotaur, SCIP, Octeract, Gurobi
- **Binary Non-Convex QPLIB Benchmark**
 - Not relevant here: translate into MILP
- **Convex Continuous QPLIB Benchmark**
 - Not relevant here: these are convex SOCPs
- **Convex Discrete QPLIB Benchmark**
 - Not relevant here: these are convex MIQCPs

Problem Class	#	Gurobi 9.0 solved	Gurobi 9.1 solved	Best Competitor	Competitor solved	Competitor vs. Gurobi 9.0	Competitor vs. Gurobi 9.1
Continuous non-convex	57	28	35	Antigone	29	1.59x	4.68x
Discrete non-convex	75	65	66	FSCIP	32	7.31x	10.5x

Solved by at least one solver

Thank You!

Implied Quadratic Equations

- Quadratic inequality $a'x' + ax + y^T Qy \leq b$
- Conditions for being able to move ax upwards:
 - $\inf\{a'x'\} + \sup\{ax\} + \inf\{y^T Qy\} \geq b$
 - For all integer feasible (x', y) there exists integer feasible x such that
$$ax + a'x' + y^T Qy = b$$
 - None of the x_j appear in equations
 - Each of the x_j appears in other inequalities only with opposite sign
 - $a_j > 0 \Rightarrow A_{ij} \leq 0$ for all other constraints i
 - $a_j < 0 \Rightarrow A_{ij} \geq 0$ for all other constraints i
 - Similar for objective function
 - $a_j > 0 \Rightarrow c_j \leq 0$
 - $a_j < 0 \Rightarrow c_j \geq 0$
 - Similar for other quadratic constraints and SOS constraints
- Similar conditions for being able to move $y^T Qy$ upwards