



Triple/Quadruple Patterning Layout Decomposition via Novel Linear Programming and Iterative Rounding

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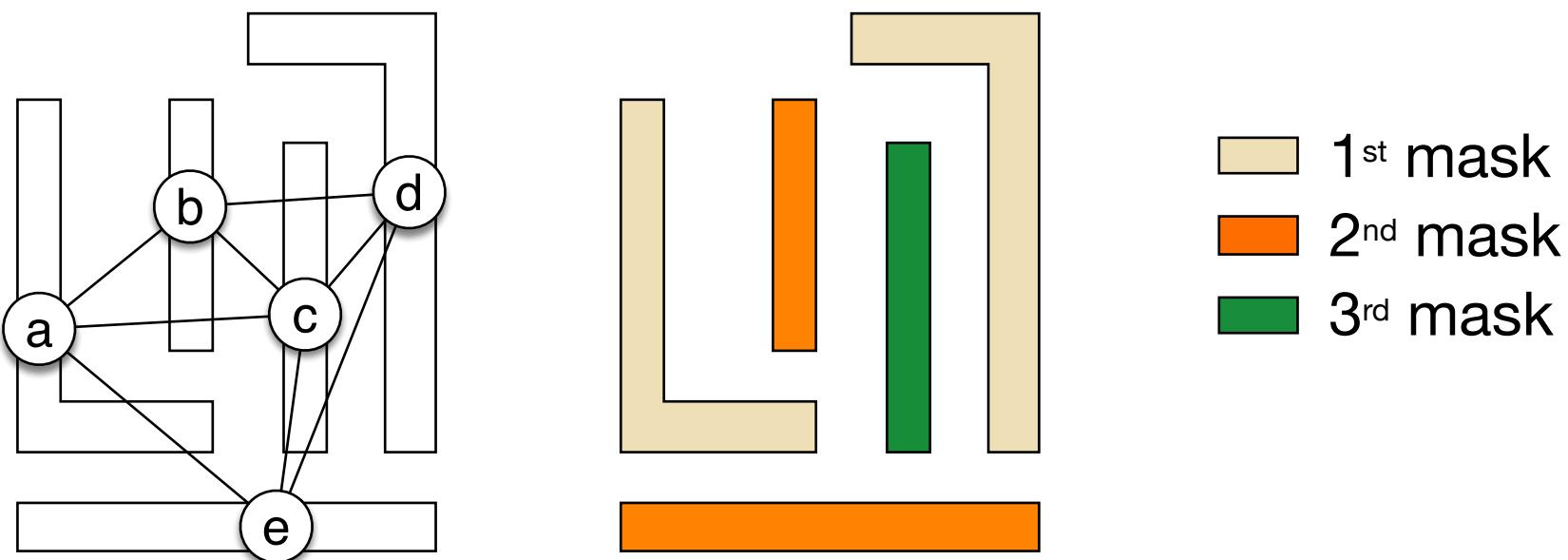
This work is supported in part by NSF and SRC

Outline

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- Introduction
 - A New Framework for Layout Decomposition
 - ILP \rightarrow LP relaxation with iterative rounding
 - Experimental Results
 - Conclusion

Triple Patterning Lithography (TPL)

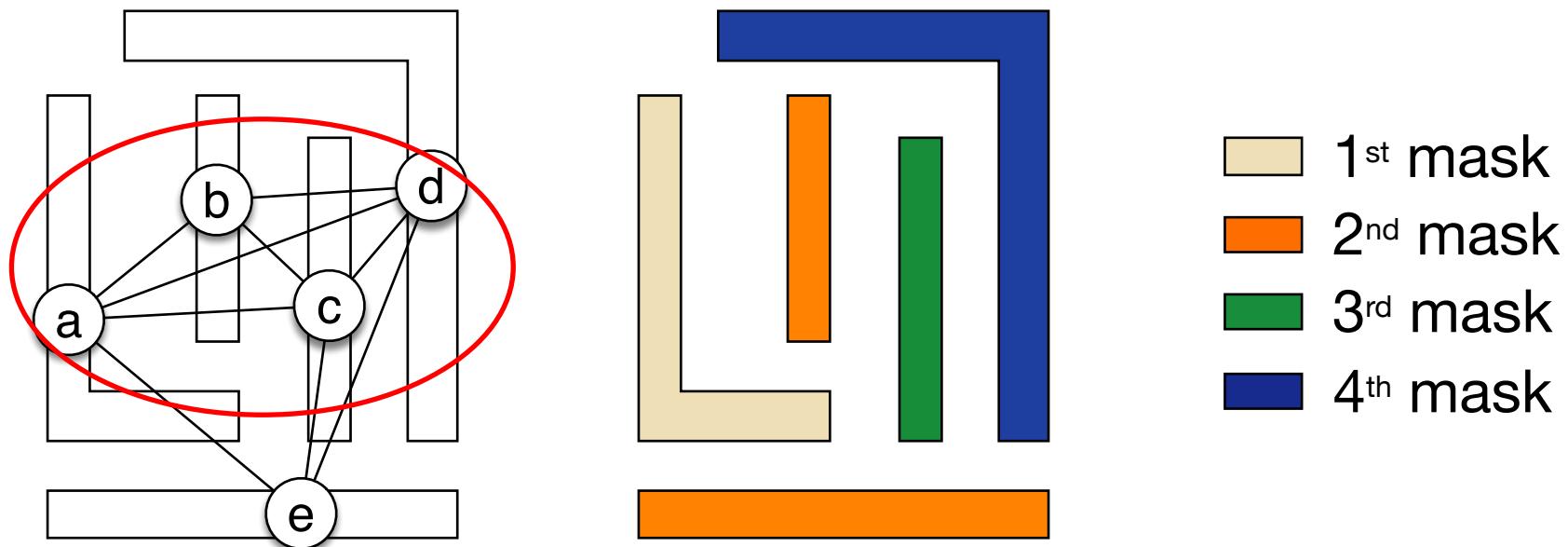
- An example of TPL conflict graph and decomposition
- Layout decomposition is a fundamental problem for multiple patterning



Quadruple Patterning Lithography (QPL)

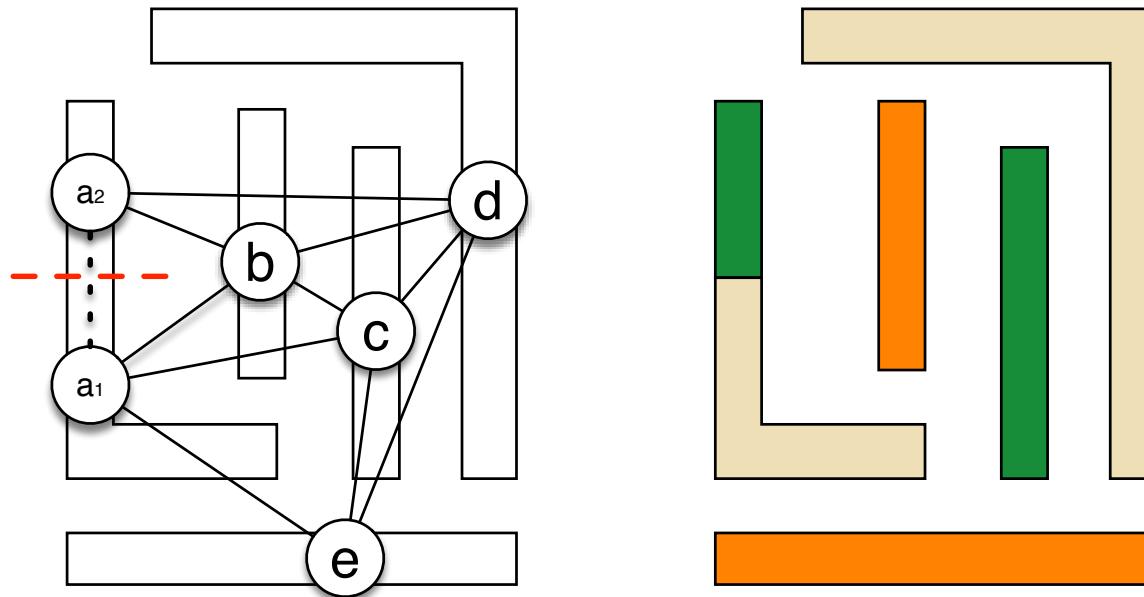


- An example of QPL layout decomposition (coloring) and conflict graph



Stitch Insertion

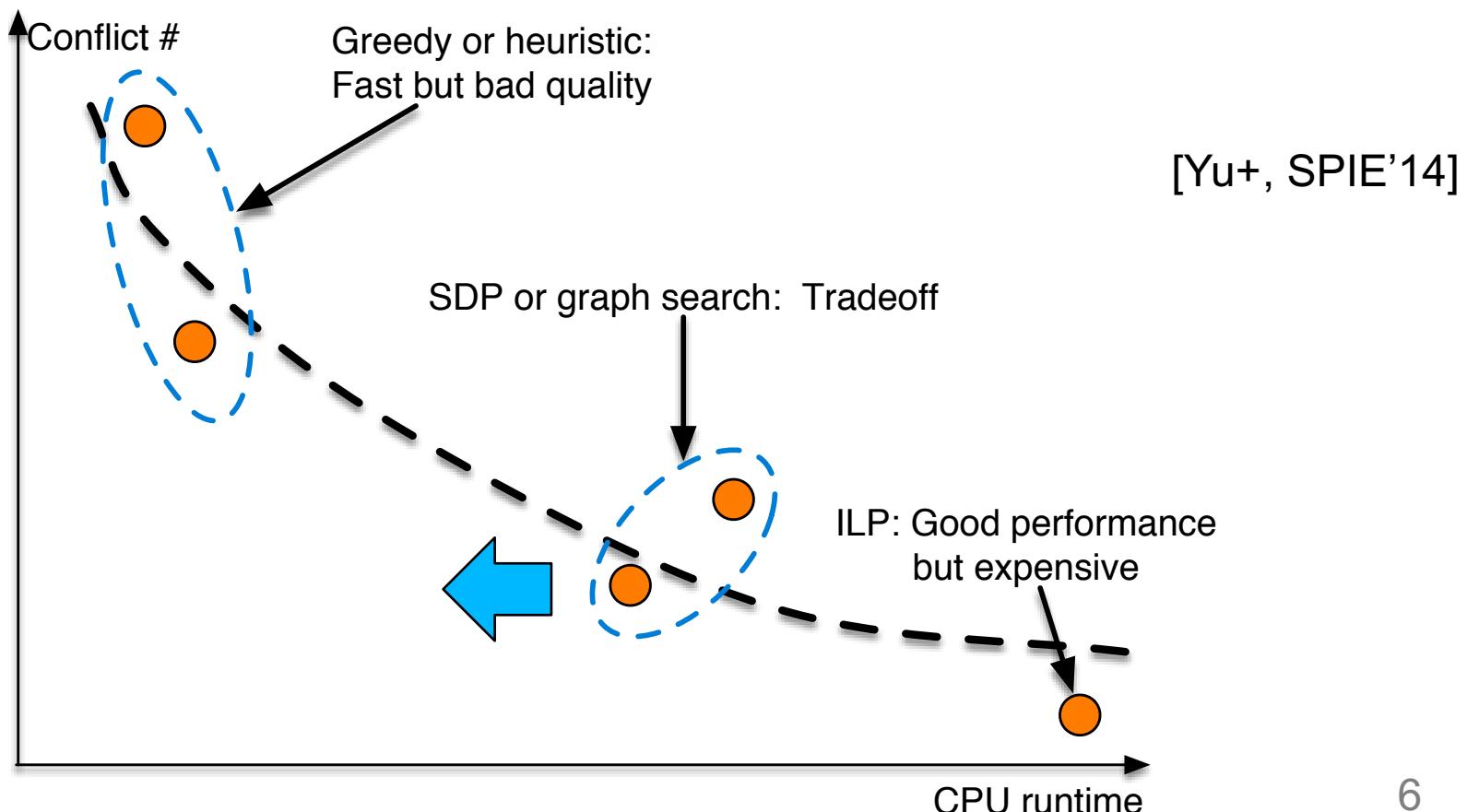
- Stitch may be inserted to resolve conflict



- However, strongly discouraged due to misalignment and yield loss
- In this work, we do not allow stitch insertion

Current State of MPL Decomposition

- **ILP or SAT:** [Cork+, SPIE'08], [Yu+, ICCAD'11], [Cork+, SPIE'13]
- **Greedy or heuristic:** [Ghaida+, SPIE'11], [Fang+, DAC'12], [Kuang+, DAC'13], [Fang+, SPIE'14]
- **SDP or graph search:** [Yu+, ICCAD'11], [Chen+, ISQED'13], [Yu+, ICCAD'13][Yu+, DAC'14]



Major Contributions of This Work

- A new layout decomposition framework for TPL/QPL
- ILP → novel linear programming (LP) based algorithm with iterative rounding scheme
- An odd-cycle based technique to enhance LP solution quality (which can be better mapped to ILP solution)
- Our experiments obtain comparable quality cf. previous state-of-the-art, but are 26x to 600x faster than ILP, and 1.8x to 2.6x faster than SDP

Problem Formulation



- Input
 - Uncolored layout patterns
 - Minimum coloring distance d_c
 - Number of colors available (TPL or QPL)
- Output
 - Decomposed layout with color assignment for each pattern
 - TPL/QPL friendliness
 - Stitch insertion is not allowed

Initial ILP Formulation

- Represent color with two binary variables

$(x_{i1}, x_{i2}) \rightarrow \text{color}$

$(0, 0) \rightarrow 0$

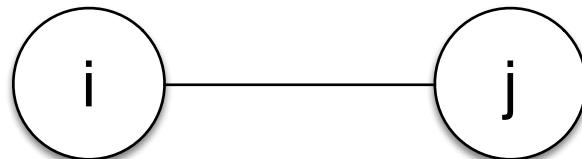
$(0, 1) \rightarrow 1$

$(1, 0) \rightarrow 2$

$(1, 1) \rightarrow 3$

(x_{i1}, x_{i2})

(x_{j1}, x_{j2})



$x_{i1}, x_{i2}, x_{j1}, x_{j2} \in \{0, 1\}$

$(0, 0)$

$(0, 0)$



$$x_{i1} + x_{i2} + x_{j1} + x_{j2} \geq 1$$

$(0, 1)$

$(0, 1)$



$$x_{i1} + (1 - x_{i2}) + x_{j1} + (1 - x_{j2}) \geq 1$$

Additional constraint for TPL

$$x_{i1} + x_{i2} \leq 1$$

...

ILP Formulation

- The goal is to meet all the constraints

min *Objective*

Only for TPL

(1a)

$$\text{s.t. } x_{i1} + x_{i2} \leq 1,$$

(1b)

$$x_{i1} + x_{i2} + x_{j1} + x_{j2} \geq 1,$$

$\forall e_{ij} \in E_c,$ (1c)

$$x_{i1} + \bar{x}_{i2} + x_{j1} + \bar{x}_{j2} \geq 1,$$

$\forall e_{ij} \in E_c,$ (1d)

$$\bar{x}_{i1} + x_{i2} + \bar{x}_{j1} + x_{j2} \geq 1,$$

$\forall e_{ij} \in E_c,$ (1e)

$$\bar{x}_{i1} + \bar{x}_{i2} + \bar{x}_{j1} + \bar{x}_{j2} \geq 1,$$

$\forall e_{ij} \in E_c,$ (1f)

$$\bar{x}_{i1} = 1 - x_{i1}, \quad \forall i \in V,$$

(1g)

$$\bar{x}_{i2} = 1 - x_{i2}, \quad \forall i \in V,$$

(1h)

$$x_{i1}, x_{i2} \in \{0, 1\}, \quad \forall i \in V.$$

(1i)

LP Relaxation

- Relax integer to continuous variables

$$\min \quad Objective \tag{1a}$$

$$\text{s.t. } x_{i1} + x_{i2} \leq 1, \tag{1b}$$

$$x_{i1} + x_{i2} + x_{j1} + x_{j2} \geq 1, \quad \forall e_{ij} \in E_c, \tag{1c}$$

$$x_{i1} + \bar{x}_{i2} + x_{j1} + \bar{x}_{j2} \geq 1, \quad \forall e_{ij} \in E_c, \tag{1d}$$

$$\bar{x}_{i1} + x_{i2} + \bar{x}_{j1} + x_{j2} \geq 1, \quad \forall e_{ij} \in E_c, \tag{1e}$$

$$\bar{x}_{i1} + \bar{x}_{i2} + \bar{x}_{j1} + \bar{x}_{j2} \geq 1, \quad \forall e_{ij} \in E_c, \tag{1f}$$

$$\bar{x}_{i1} = 1 - x_{i1}, \quad \forall i \in V, \tag{1g}$$

$$\bar{x}_{i2} = 1 - x_{i2}, \quad \forall i \in V, \tag{1h}$$

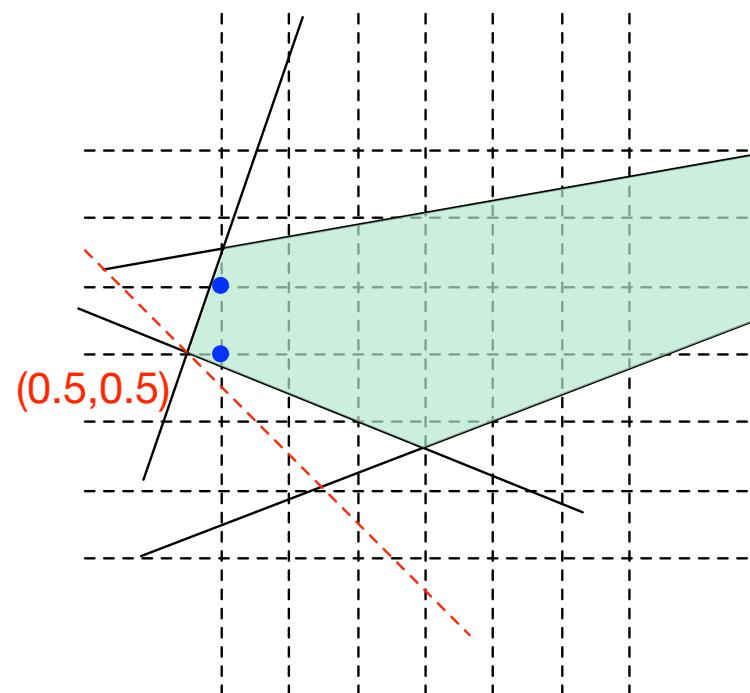
$$\underline{x}_{i1}, \bar{x}_{i2} \in \{0, 1\}, \quad \forall i \in V. \tag{1i}$$

$$0 \leq x_{i1}, x_{i2} \leq 1, \quad \forall i \in V$$



- Linear programming and iterative rounding (LPIR)
- Non-integer solutions
 - Fewer non-integers mean closer to optimal solutions of ILP
 - Prune non-integer solutions in the feasible set

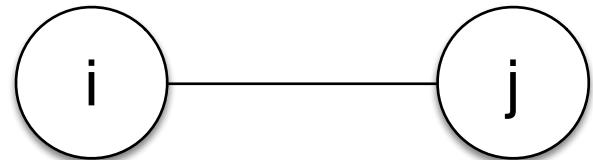
Reduce
non-integers



Simple Observation

- Suppose $x_{i1} = x_{j1} = 0$

$$(x_{i1}, x_{i2}) \quad (x_{j1}, x_{j2})$$



$$x_{i1} + x_{i2} + x_{j1} + x_{j2} \geq 1, \quad \forall e_{ij} \in E_c, \quad (1c)$$

$$x_{i1} + \bar{x}_{i2} + x_{j1} + \bar{x}_{j2} \geq 1, \quad \forall e_{ij} \in E_c, \quad (1d)$$

$$\bar{x}_{i1} + x_{i2} + \bar{x}_{j1} + x_{j2} \geq 1, \quad \forall e_{ij} \in E_c, \quad (1e)$$

$$\bar{x}_{i1} + \bar{x}_{i2} + \bar{x}_{j1} + \bar{x}_{j2} \geq 1, \quad \forall e_{ij} \in E_c, \quad (1f)$$

$$\bar{x}_{i1} = 1 - x_{i1}, \quad \forall i \in V, \quad (1g)$$

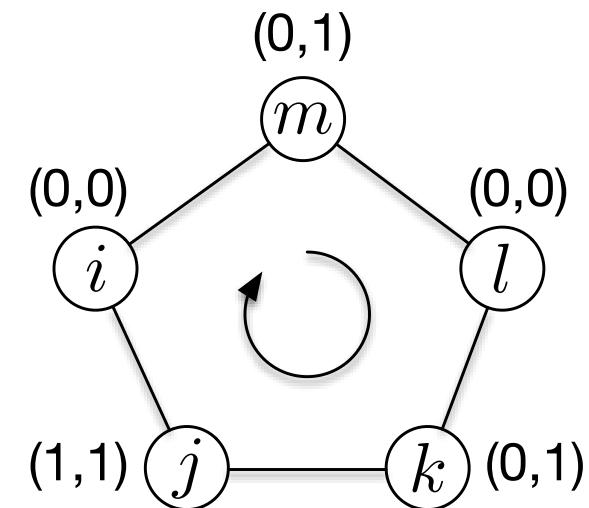
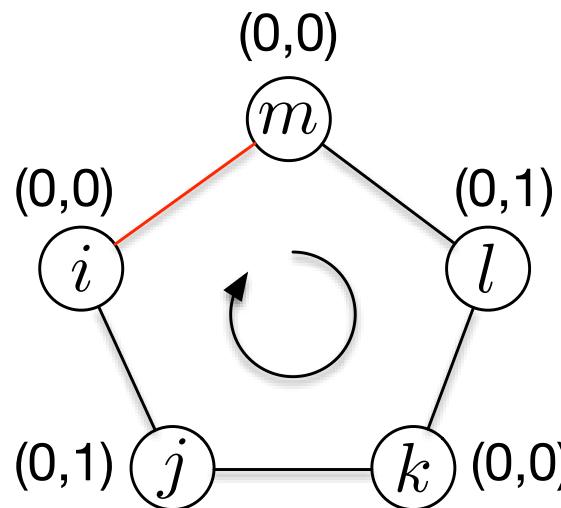
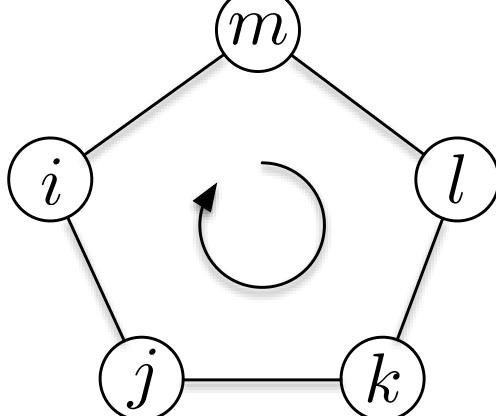
$$\bar{x}_{i2} = 1 - x_{i2}, \quad \forall i \in V \quad (1h)$$

→ $x_{i2} + x_{j2} = 1$

The second bits must be different

Non-integers along Odd Cycles

- Consider the constraints along an odd cycle
- Suppose $x_{i1} = x_{j1} = x_{k1} = x_{l1} = x_{m1} = 0$

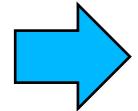


$$x_{i2} + x_{j2} = 1,$$

$$x_{j2} + x_{k2} = 1,$$

$$x_{k2} + x_{l2} = 1,$$

$$x_{m2} + x_{i2} = 1$$

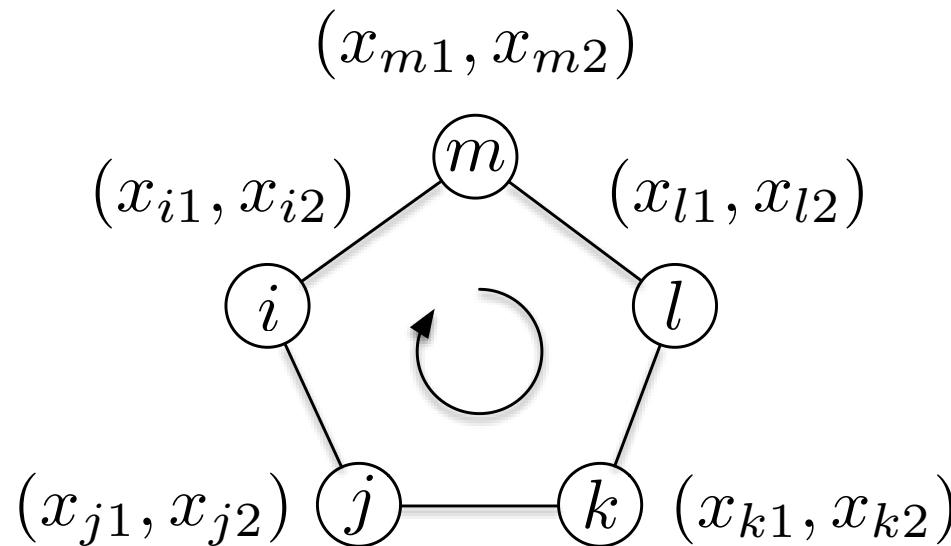


$$x_{i2} = x_{j2} = x_{k2} = x_{l2} = x_{m2} = 0.5$$

intuition?

LPIR – Add Odd Cycle Constraints

- Additional constraints
- Prune non-integer solutions from feasible set



s.t.

$$x_{i1} + x_{j1} + x_{k1} + x_{l1} + x_{m1} \geq 1,$$

$$(1 - x_{i1}) + (1 - x_{j1}) + (1 - x_{k1}) + (1 - x_{l1}) + (1 - x_{m1}) \geq 1$$

Help resolve potential non-integers in the second bits

LPIR – Objective Function Biasing

- Push non-integer solutions to integers by dynamically adapting the objective function
- If $x_i = 0.6$, it means x_i tends to be 1
- If $x_i = 0.4$, it means x_i tends to be 0

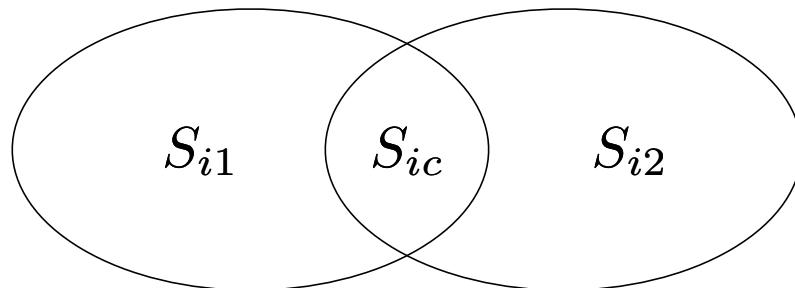
If $x_i > 0.5$, $\text{obj} \leftarrow \text{obj} + (1 - x_i)$.
If $x_i < 0.5$, $\text{obj} \leftarrow \text{obj} + x_i$.

Cannot handle (0.5, 0.5)

LPIR – Binding Constraints Analysis

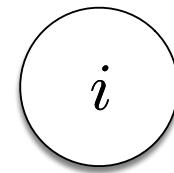
- Try to handle $(x_{i1}, x_{i2}) = (0.5, 0.5)$

Constraint set for x_{i1}



$$x_{i1} = 0.5$$

$$x_{i2} = 0.5$$



Constraint set for x_{i2}

S_{i1}

$$\begin{array}{l} \dots + x_{i1} + \dots \leq c_1 \\ \dots + x_{i1} + \dots \leq c_2 \\ \dots + x_{i1} + \dots \leq c_3 \\ \dots + x_{i1} + \dots \leq c_4 \end{array}$$

S_{i2}

$$\begin{array}{l} \dots + x_{i2} + \dots \geq c_5 \\ \dots + x_{i2} + \dots \geq c_6 \\ \dots + x_{i2} + \dots \geq c_7 \\ \dots + x_{i2} + \dots \geq c_8 \end{array}$$

S_{ic}

$$\begin{array}{l} \dots + x_{i1} + x_{i2} + \dots \leq c_9 \\ \dots + x_{i1} - x_{i2} + \dots \geq c_{10} \\ \dots + x_{i1} + x_{i2} + \dots \geq c_{11} \end{array}$$

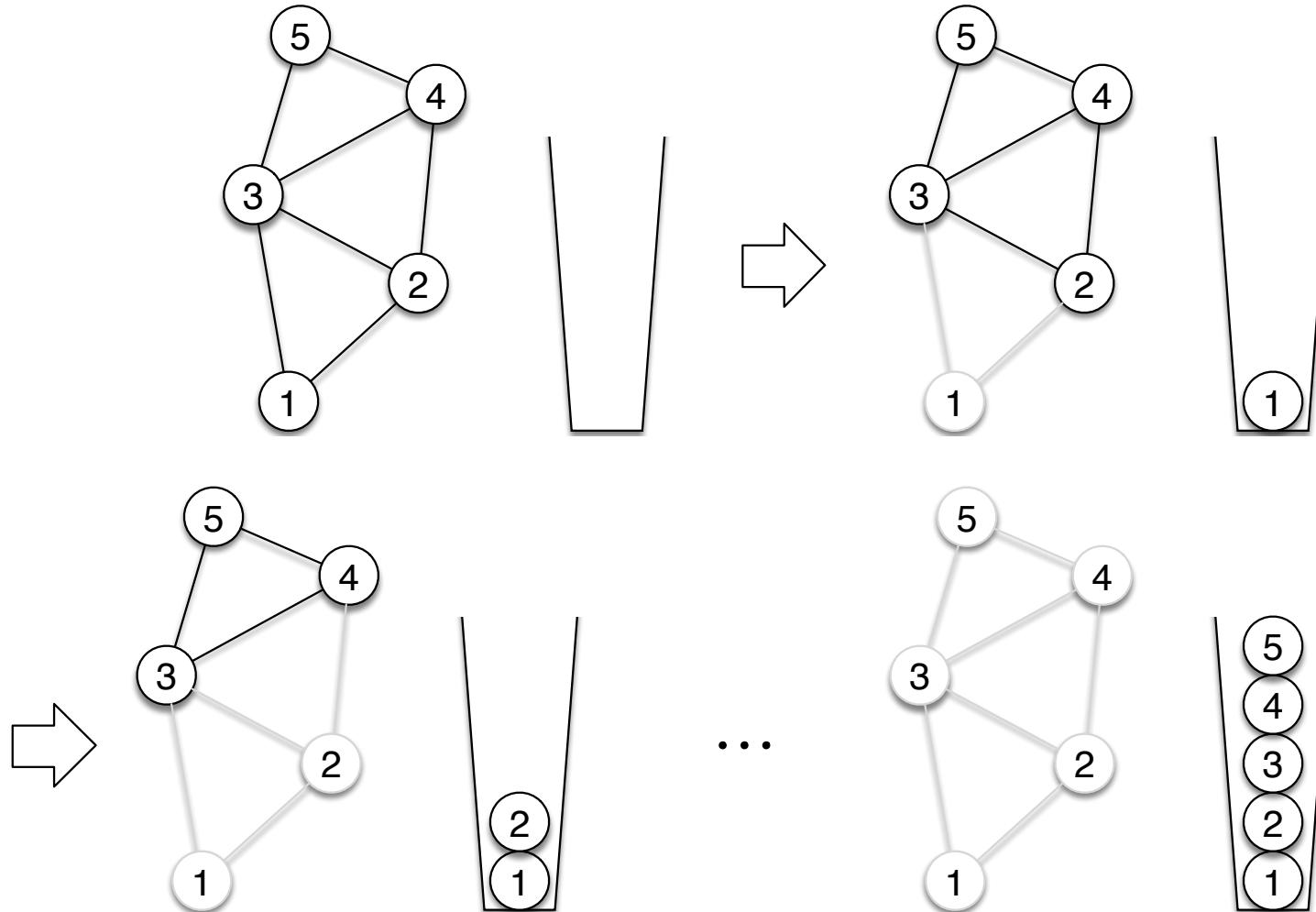
Try pushing x_{i1} to 0

Try pushing x_{i2} to 1

Check $(x_{i1}, x_{i2}) = (0, 1)$

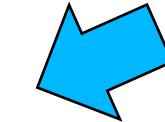
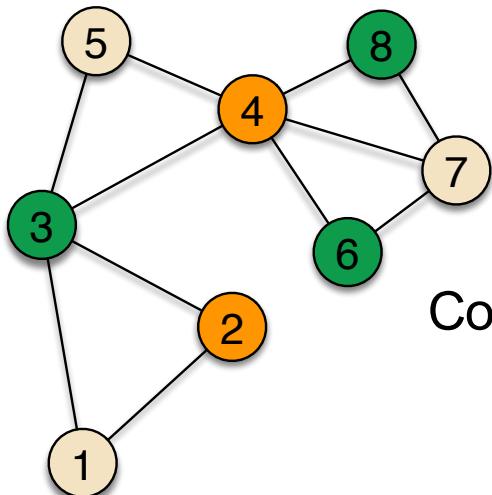
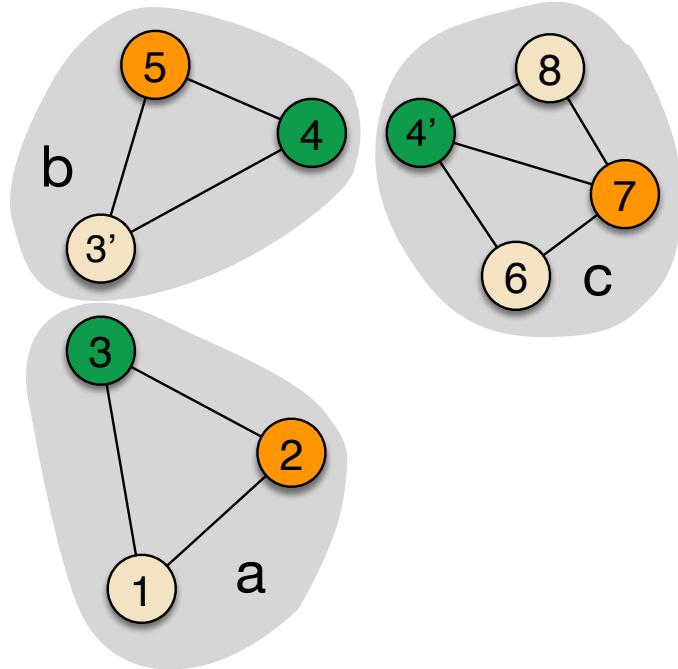
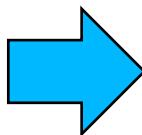
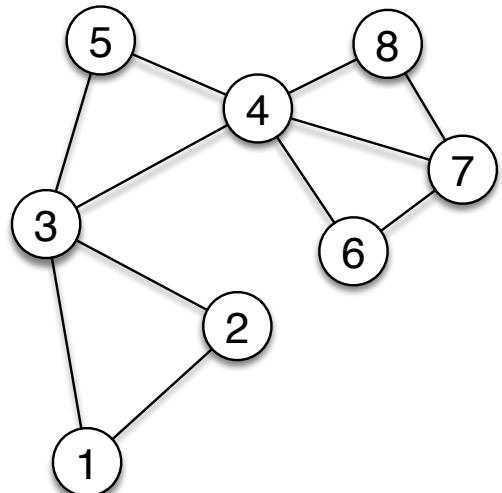
Graph Simplification – Iterative Vertex Removal

- Iterative vertex removal
- Density aware recovery



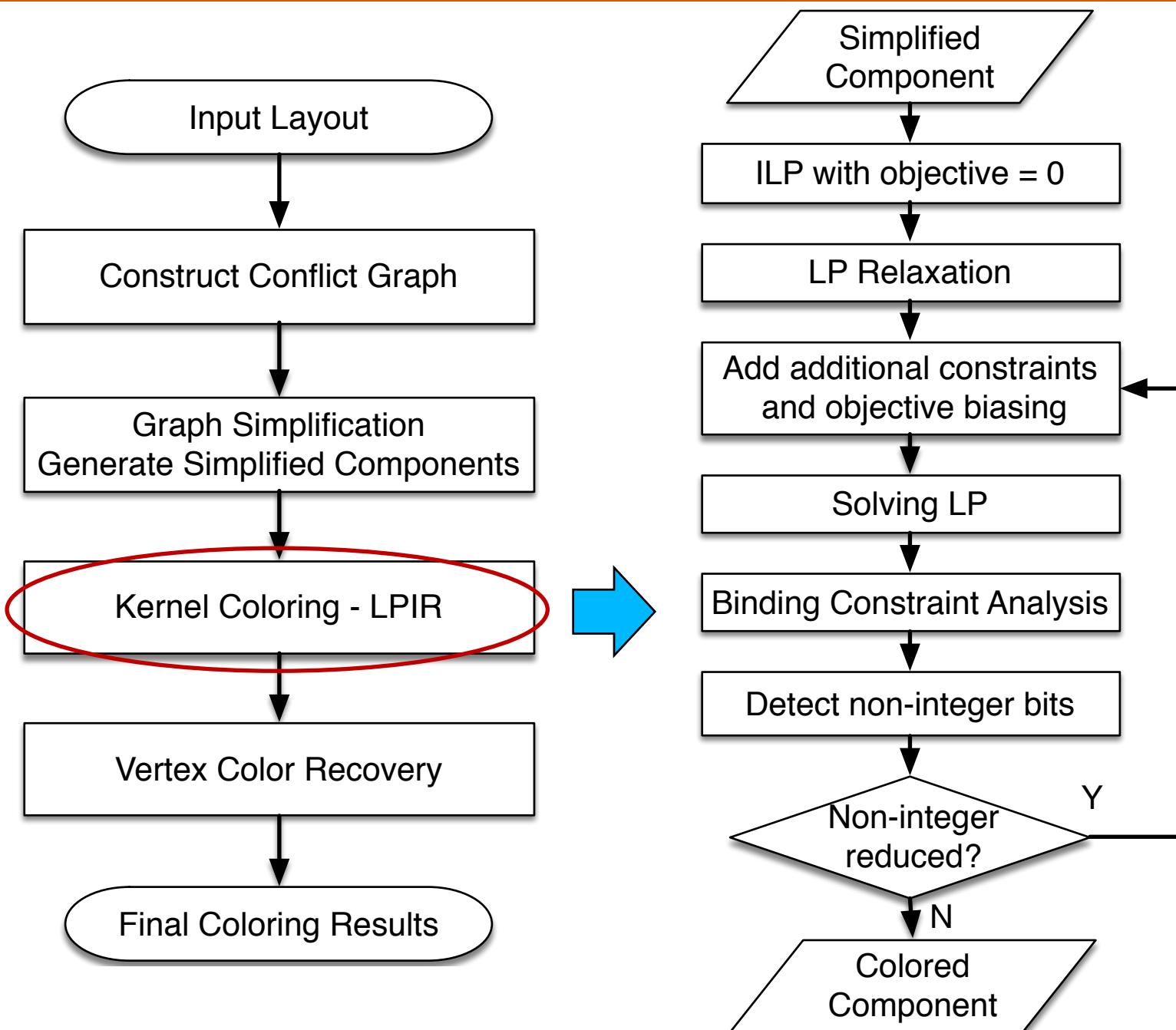
Graph Simplification: Bi-connected Component Extraction

- Color recovery
 - Color rotation on each component



Color rotation is needed

Overall Flow



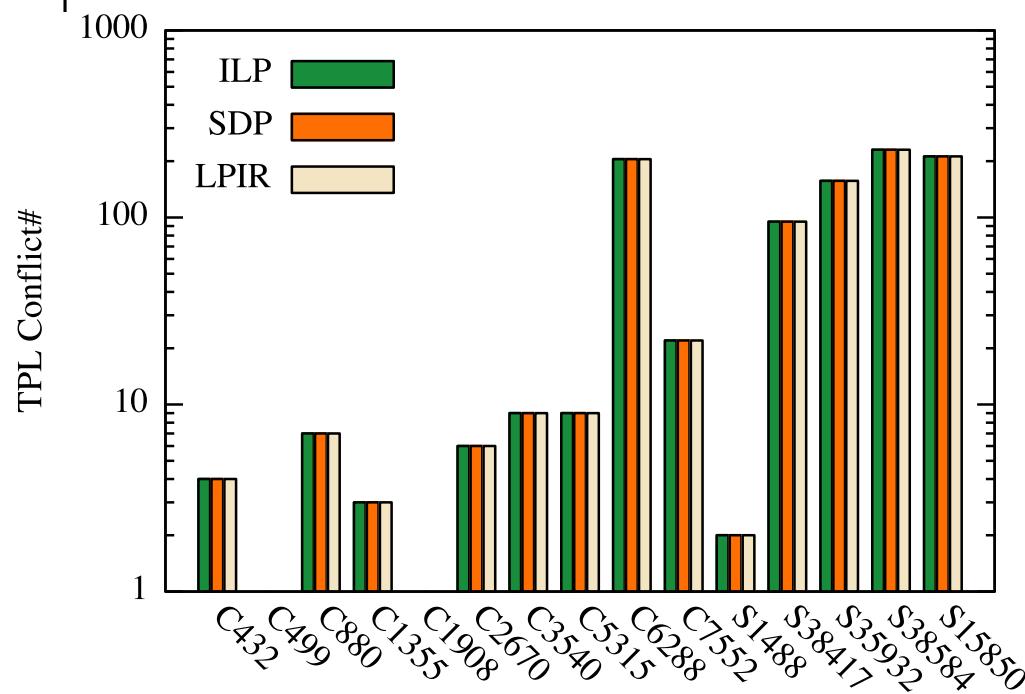
Experimental Environment Setup

- Implemented in C++
- 8-Core 3.4GHz Linux server
- 32GB RAM
- ISCAS benchmark from [Yu+, TCAD'15]
- LP solver Gurobi was used

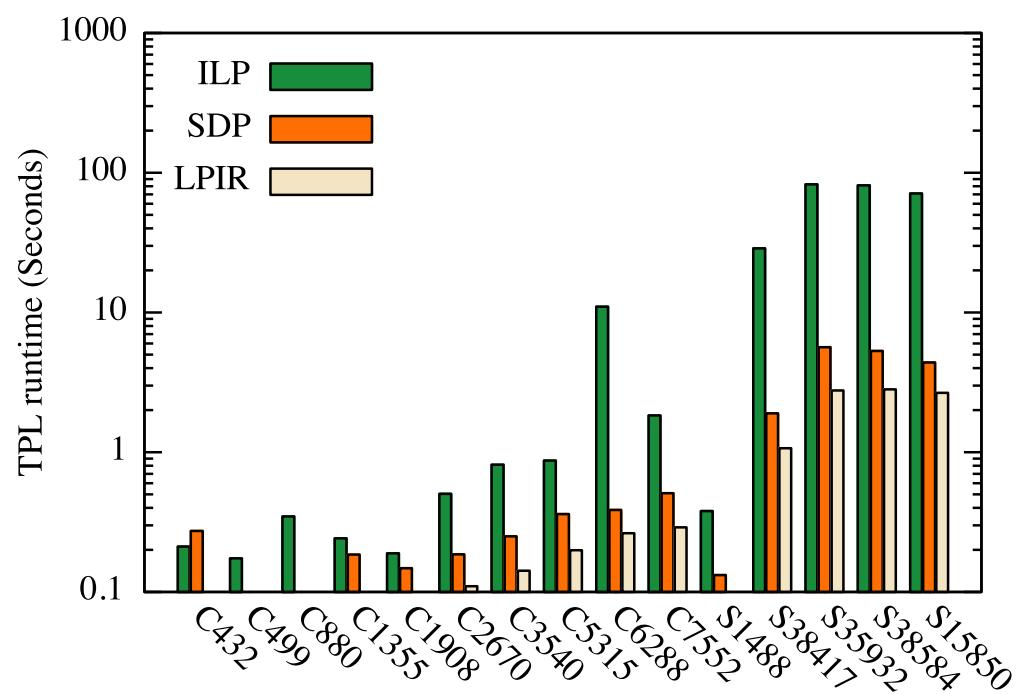
Experimental Results on TPL



TPL conflict#



TPL runtime



Baseline 1: ILP [Yu+, TCAD'15]

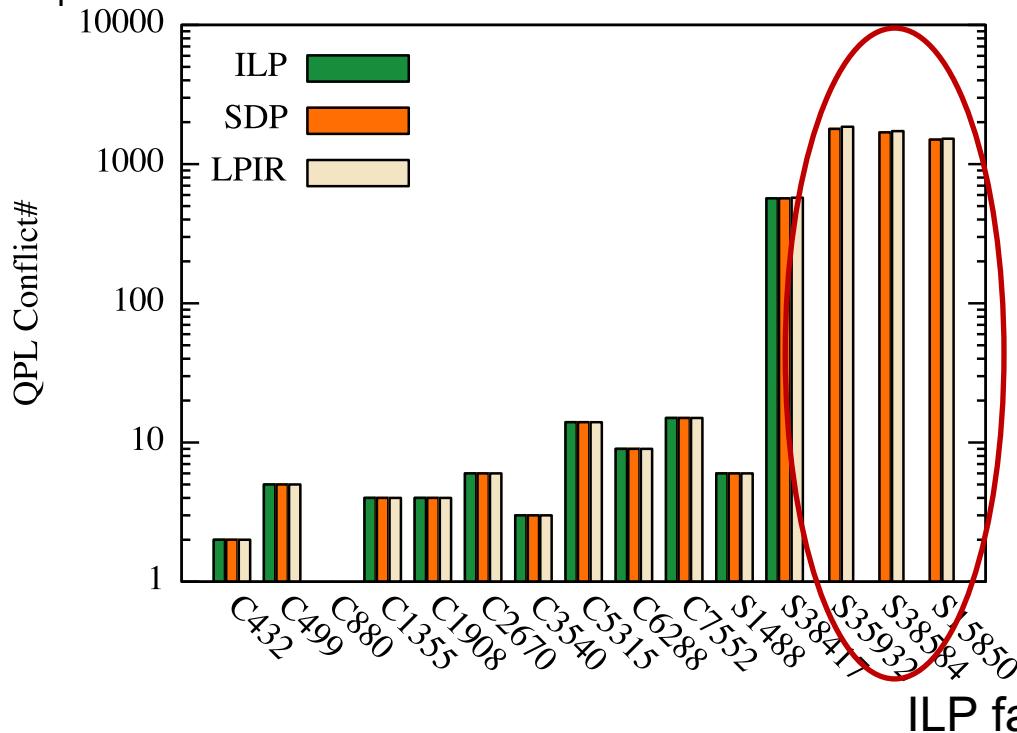
Baseline 2: SDP [Yu+, TCAD'15]

LPIR achieves almost the same conflict numbers as ILP and SDP,
but 26x faster than ILP and 1.8x faster than SDP

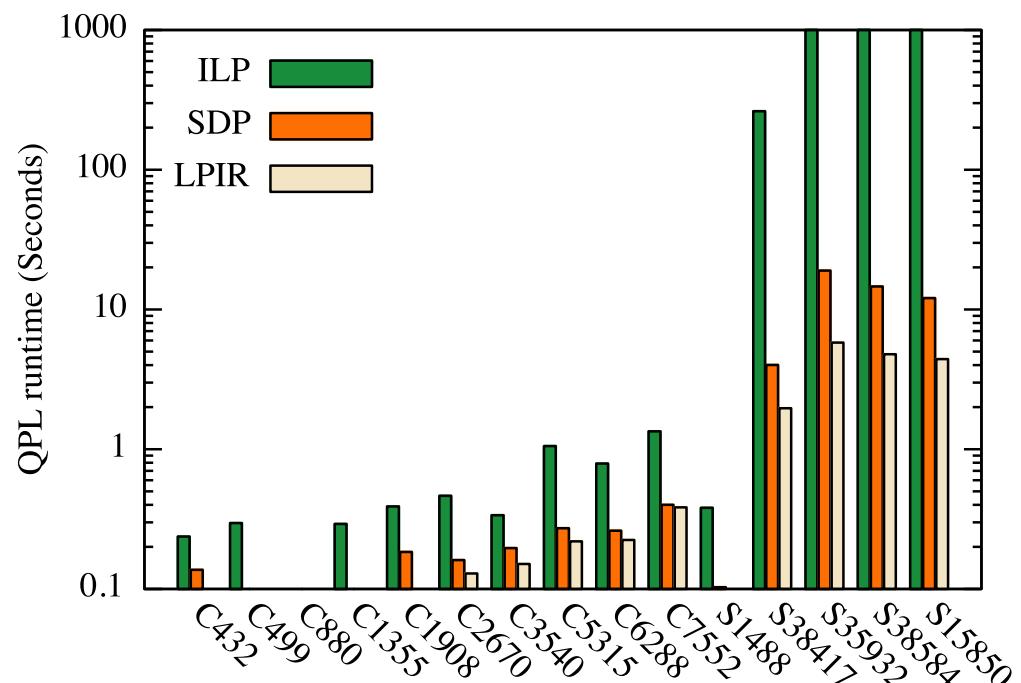
Experimental Results on QPL



QPL conflict#



QPL runtime



Baseline 1: ILP [Yu+, DAC'14]

Baseline 2: SDP [Yu+, DAC'14]

LPIR achieves less than 2% degradation in conflict numbers than SDP, but 600x faster than ILP and 2.6x faster than SDP

Conclusion

- This paper proposes a new layout decomposition framework for TPL/QPL
 - Novel linear programming (LP) based algorithm with iterative rounding
 - Odd-cycle based pruning technique to enhance LP quality
 - Very good results cf. previous state-of-the-art decomposer
- Future work
 - Lithography impacts (e.g., hotspots) from different decomposition solutions
 - Decomposition friendliness from early design stages like placement and routing



Thanks!