

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

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# Outline

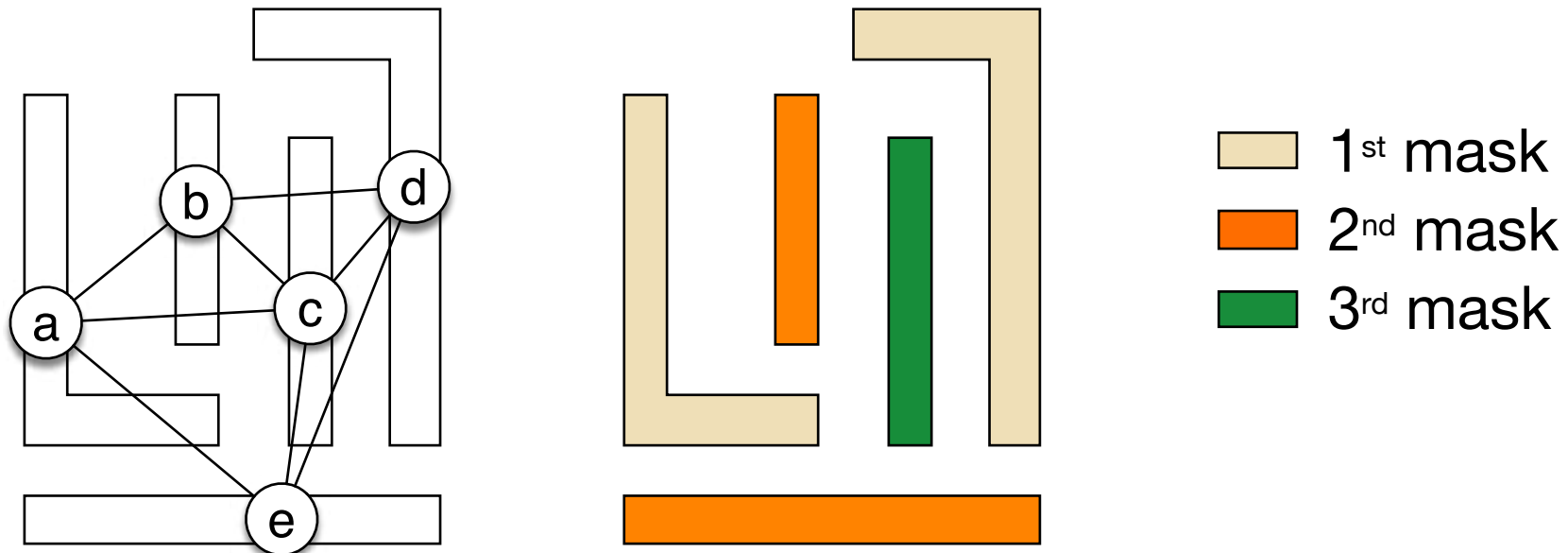


- Introduction
- A New Framework for Layout Decomposition
  - ILP  $\rightarrow$  LP relaxation with iterative rounding
- Experimental Results
- Conclusion

# Triple Patterning Lithography (TPL)



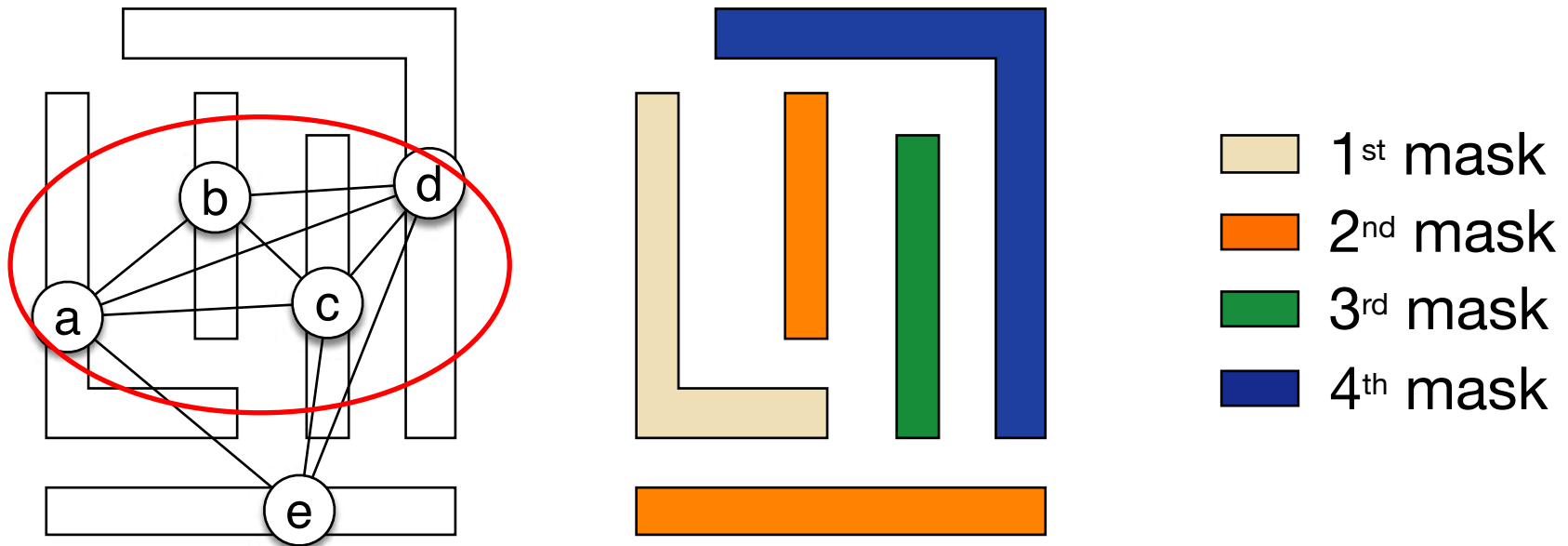
- An example of TPL conflict graph and decomposition
- Layout decomposition is a fundamental problem for multiple patterning



# Quadruple Patterning Lithography (QPL)



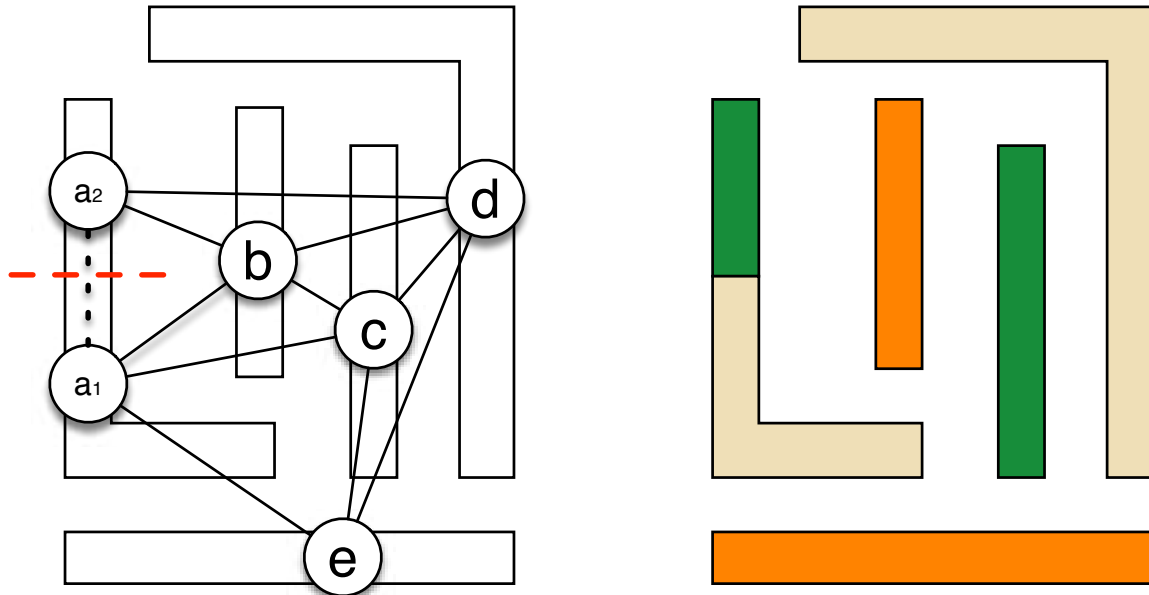
- An example of QPL layout decomposition (coloring) and conflict graph



# Stitch Insertion



- Stitch may be inserted to resolve conflict

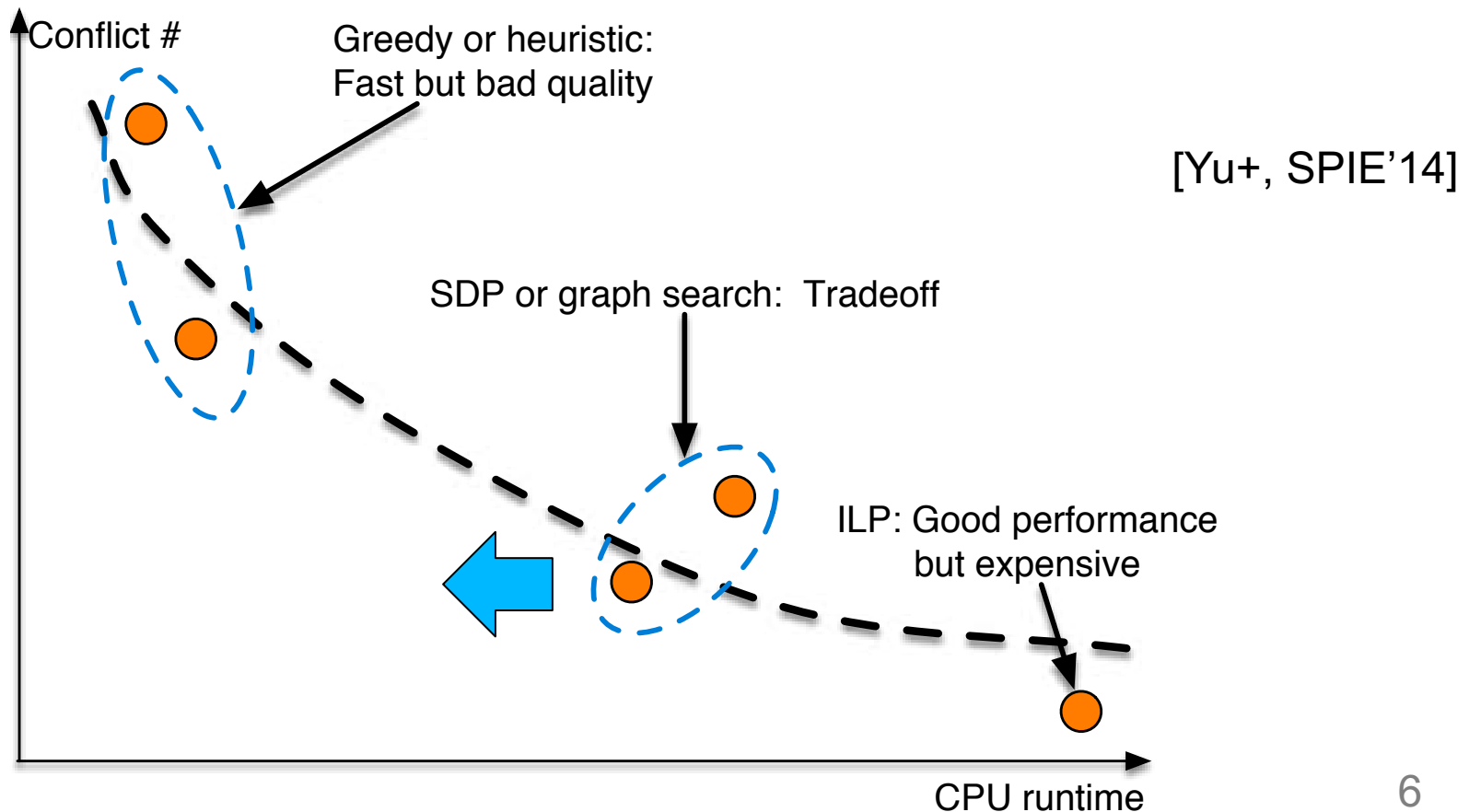


- However, strongly discouraged due to misalignment and yield loss
- In this work, we do not allow stitch insertion

# 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  $\rightarrow$  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

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

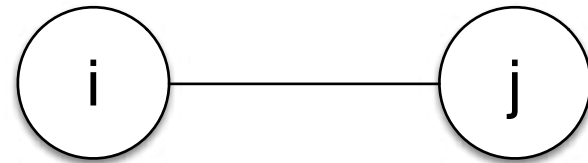
$(0, 0) \rightarrow 0$

$(0, 1) \rightarrow 1$

$(1, 0) \rightarrow 2$

$(1, 1) \rightarrow 3$

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



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

Additional constraint for TPL

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

$(0, 0)$      $(0, 0)$     **×**

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

$(0, 1)$      $(0, 1)$     **×**

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

...

# ILP Formulation



- The goal is to meet all the constraints

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

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

Only for TPL

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

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

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

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

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

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

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

# LP Relaxation



- Relax integer to continuous variables

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

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

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

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

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

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

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

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

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

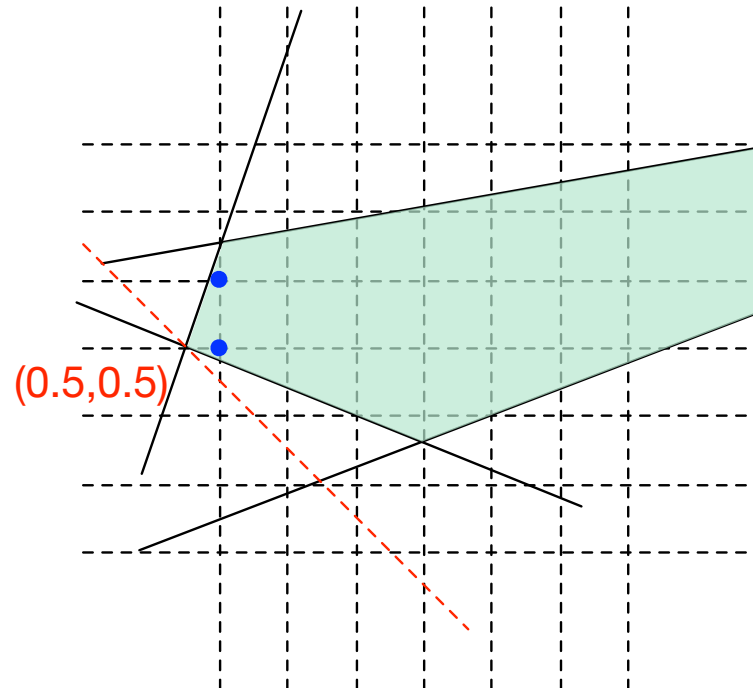
$$0 \leq x_{i1}, x_{i2} \leq 1, \quad \forall i \in V$$

# LPIR



- 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

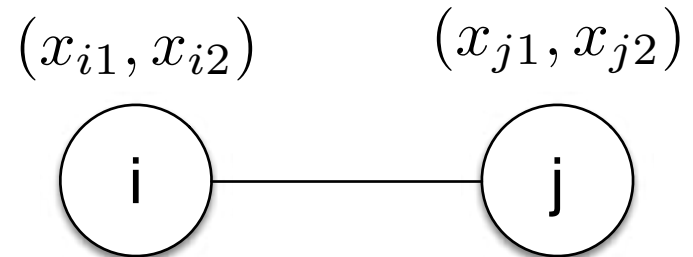
Reduce  
non-integers



# Simple Observation



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



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

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

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

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

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

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

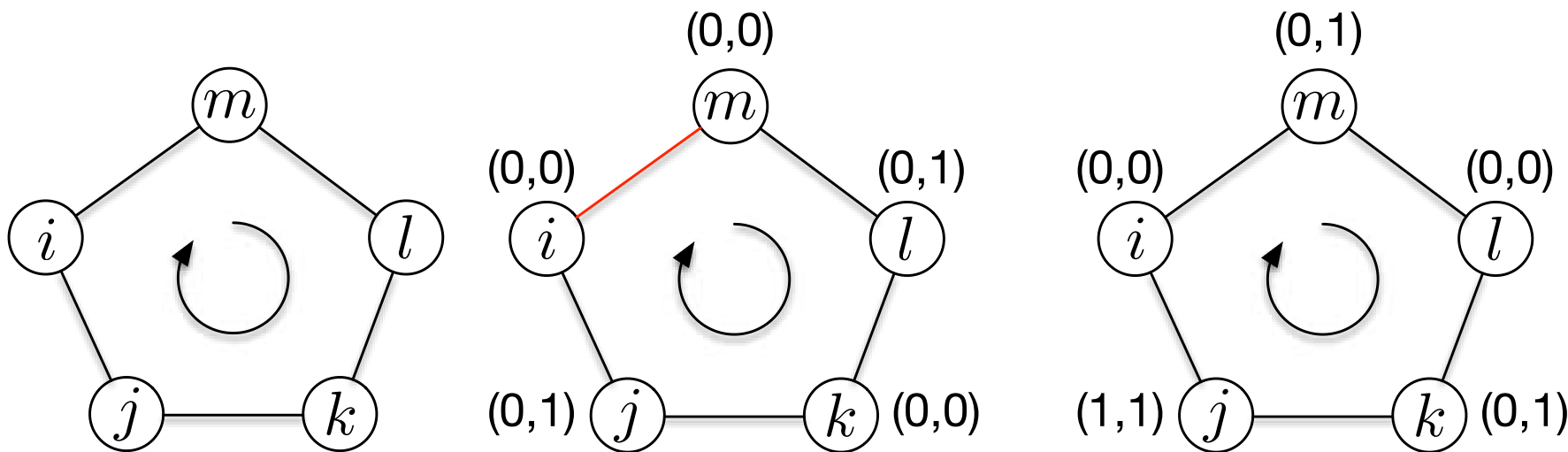
$x_{i2} + x_{j2} = 1$

The second bits must be different

# Non-integers along Odd Cycles



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

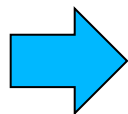


$$x_{i2} + x_{j2} = 1,$$

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

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

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



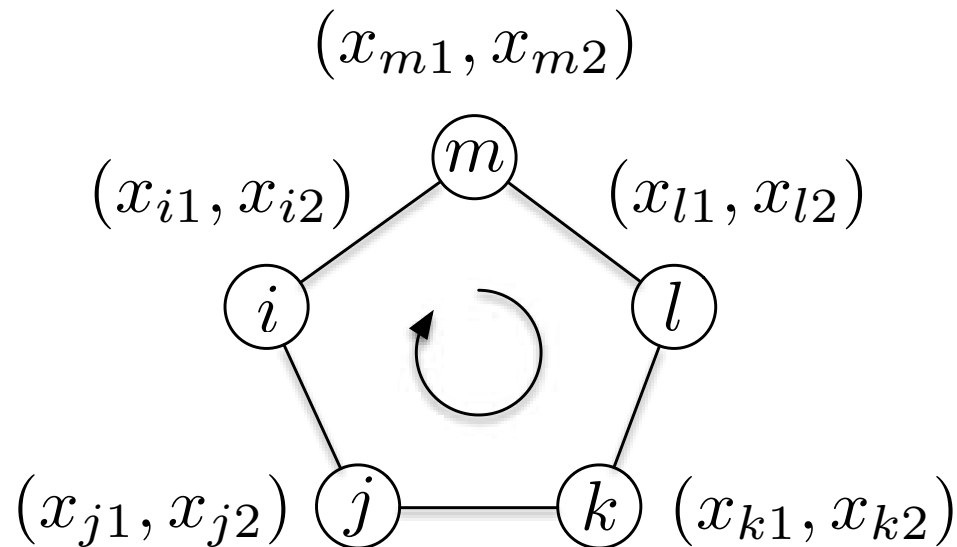
$$x_{i2} = x_{j2} = x_{k2} = x_{l2} = x_{m2} = 0.5$$

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)

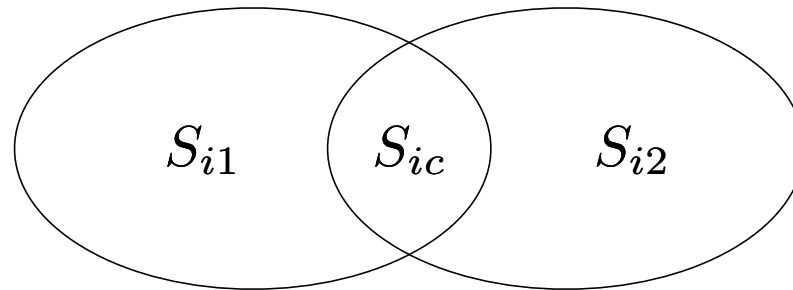


# LPIR – Binding Constraints Analysis



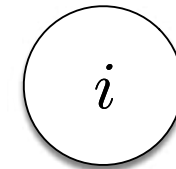
- Try to handle  $(x_{i1}, x_{i2}) = (0.5, 0.5)$

Constraint set for  $x_{i1}$



$$x_{i1} = 0.5$$

$$x_{i2} = 0.5$$



Constraint set for  $x_{i2}$

$S_{i1}$	$S_{i2}$	$S_{ic}$
$\dots + x_{i1} + \dots \leq c_1$ $\dots + x_{i1} + \dots \leq c_2$ $\dots + x_{i1} + \dots \leq c_3$ $\dots + x_{i1} + \dots \leq c_4$	$\dots + x_{i2} + \dots \geq c_5$ $\dots + x_{i2} + \dots \geq c_6$ $\dots + x_{i2} + \dots \geq c_7$ $\dots + x_{i2} + \dots \geq c_8$	$\dots + x_{i1} + x_{i2} + \dots \leq c_9$ $\dots + x_{i1} - x_{i2} + \dots \geq c_{10}$ $\dots + x_{i1} + x_{i2} + \dots \geq c_{11}$

Try pushing  $x_{i1}$  to 0

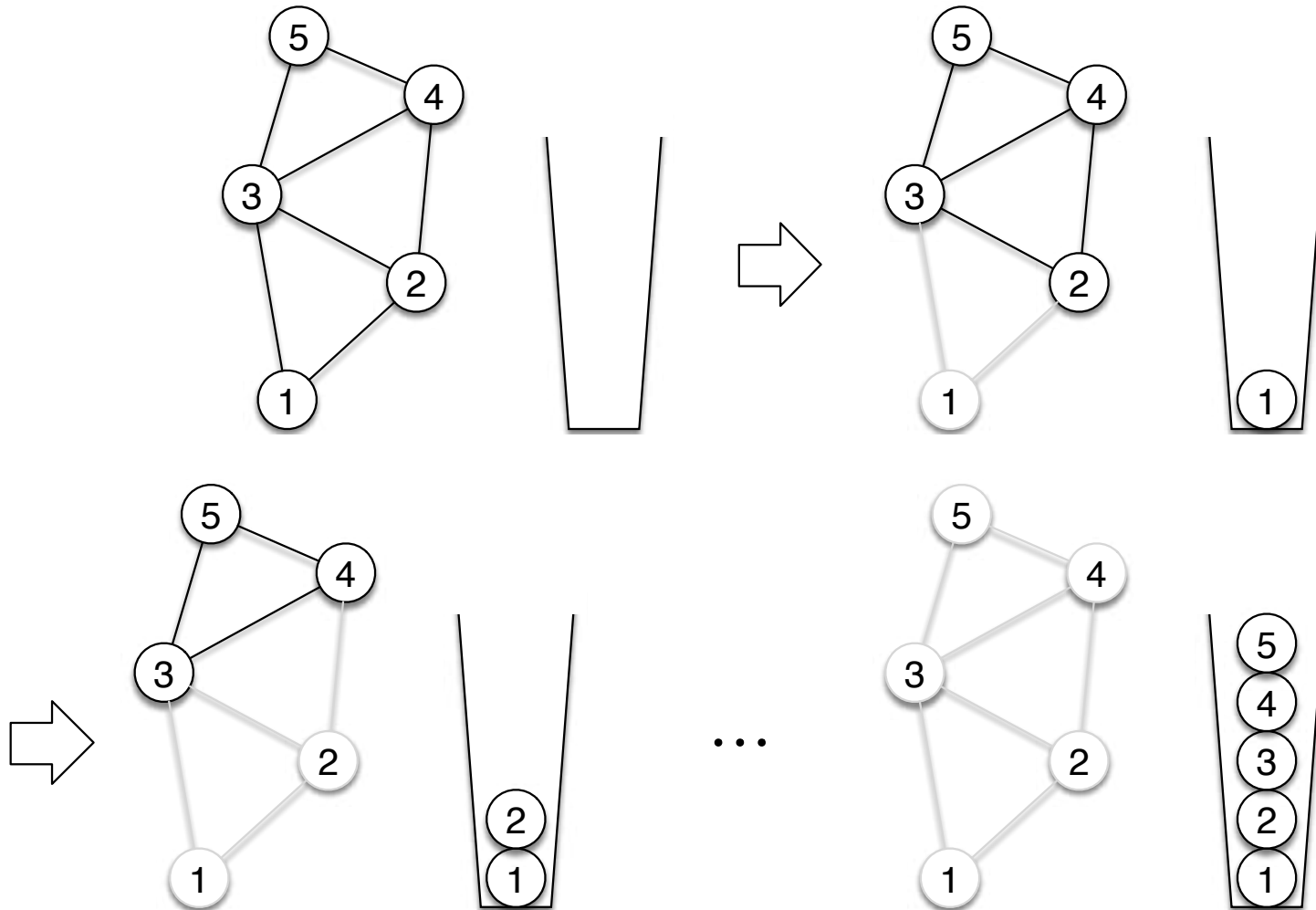
Try pushing  $x_{i2}$  to 1

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

# Graph Simplification – Iterative Vertex Removal



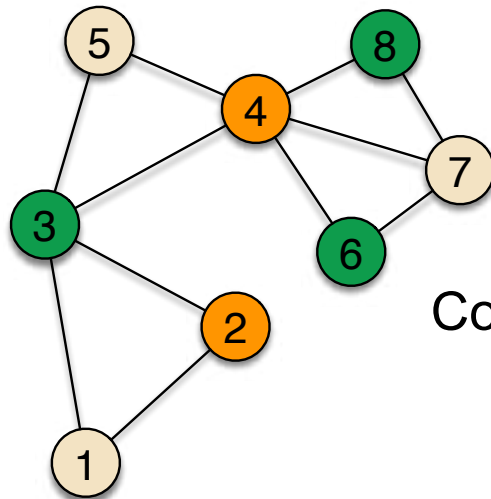
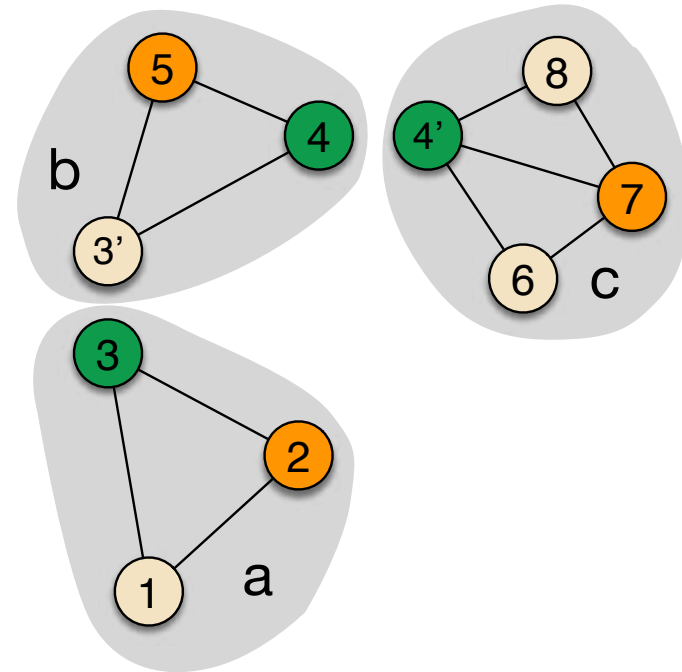
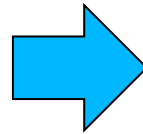
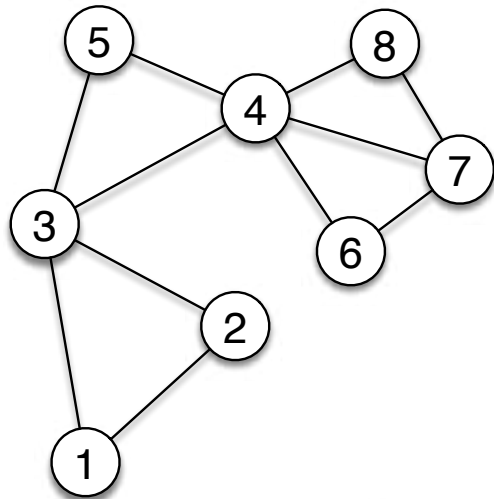
- Iterative vertex removal
- Density aware recovery



# Graph Simplification: Bi-connected Component Extraction

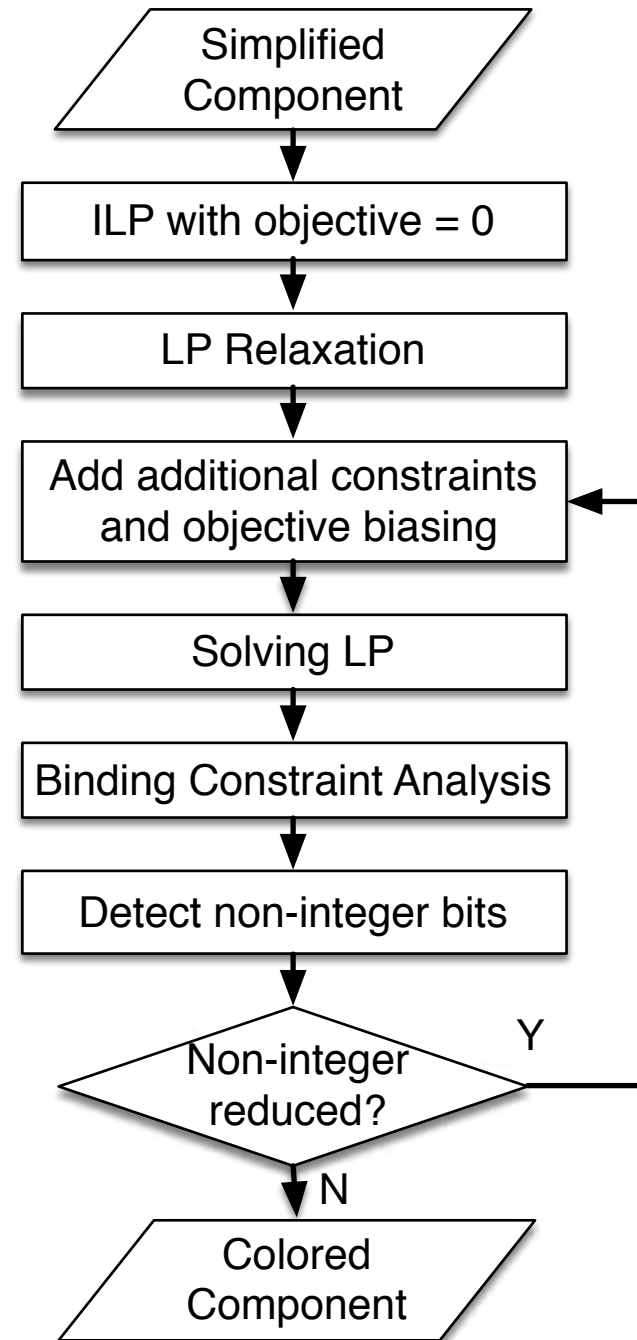
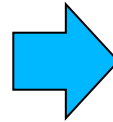
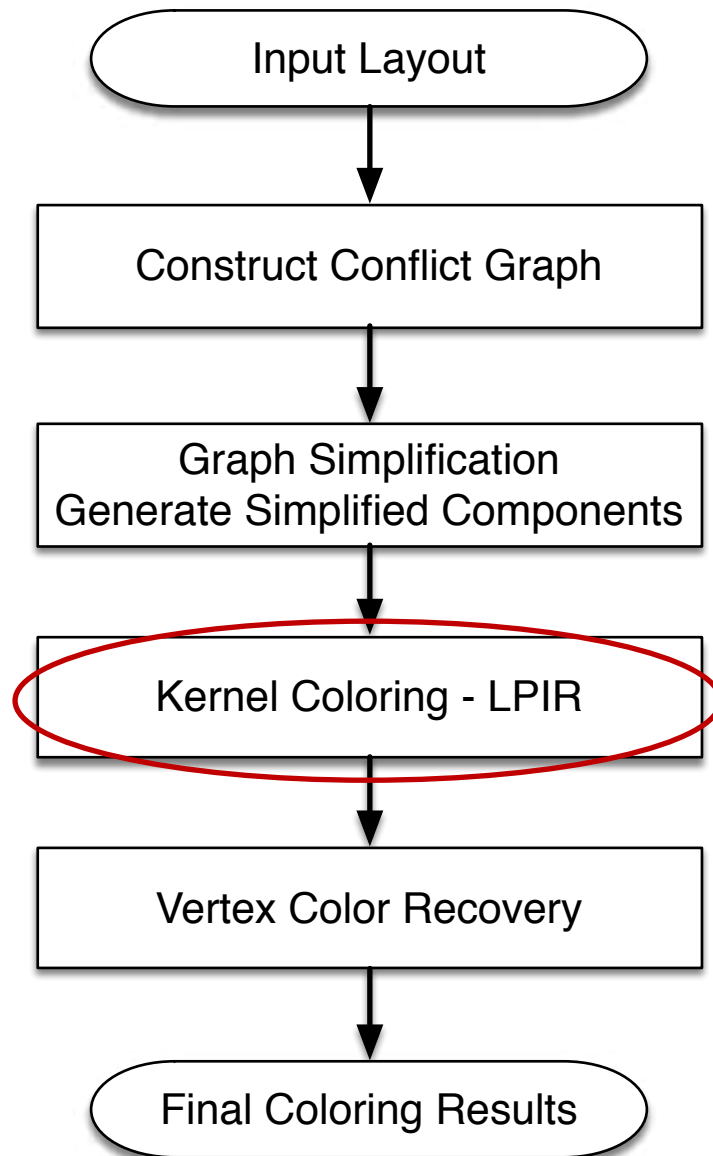


- Color recovery
  - Color rotation on each component



Color rotation is needed

# Overall Flow



# Experimental Environment Setup

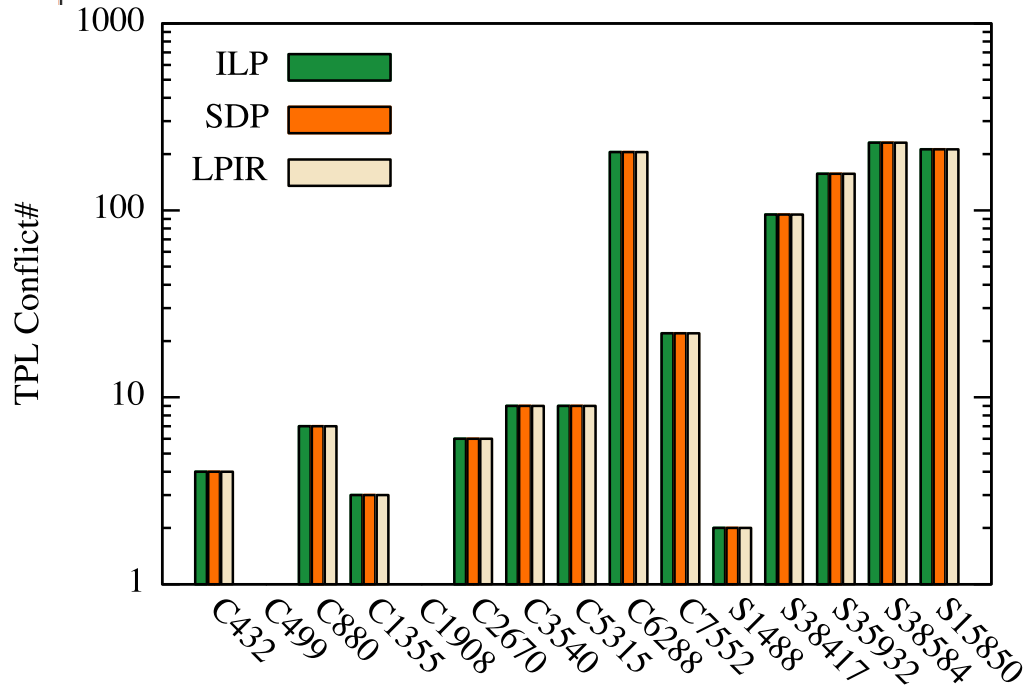


- Implemented in C++
- 8-Core 3.4GHz Linux server
- 32GB RAM
- ISCAS benchmark from [Yu+, TCAD'15]
- LP solver Gurobi was used

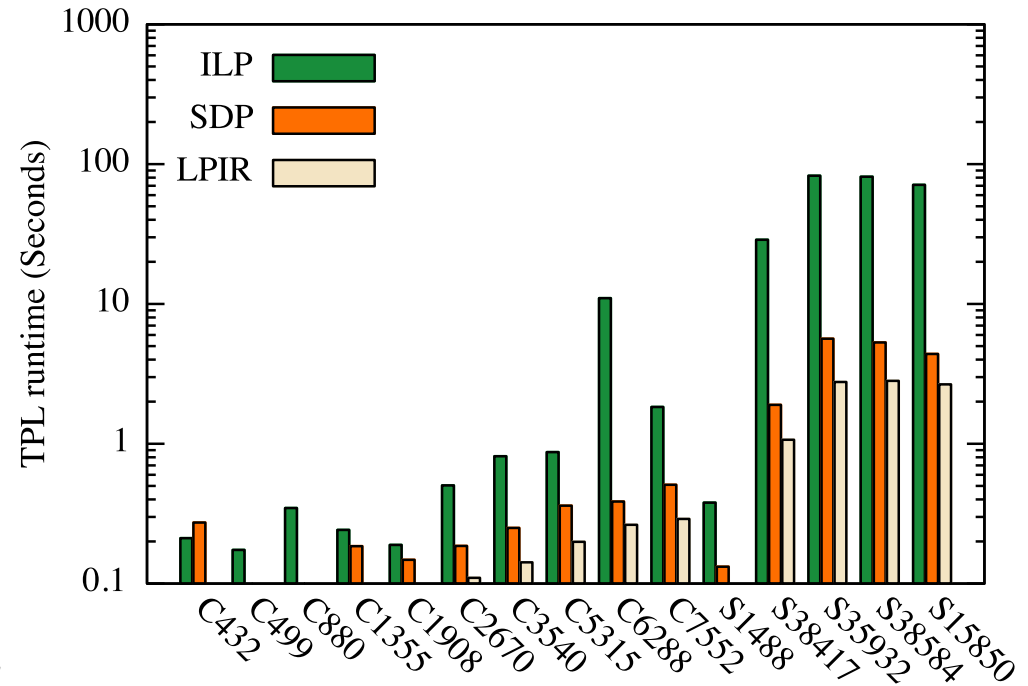
# Experimental Results on TPL



## TPL conflict#



## TPL runtime

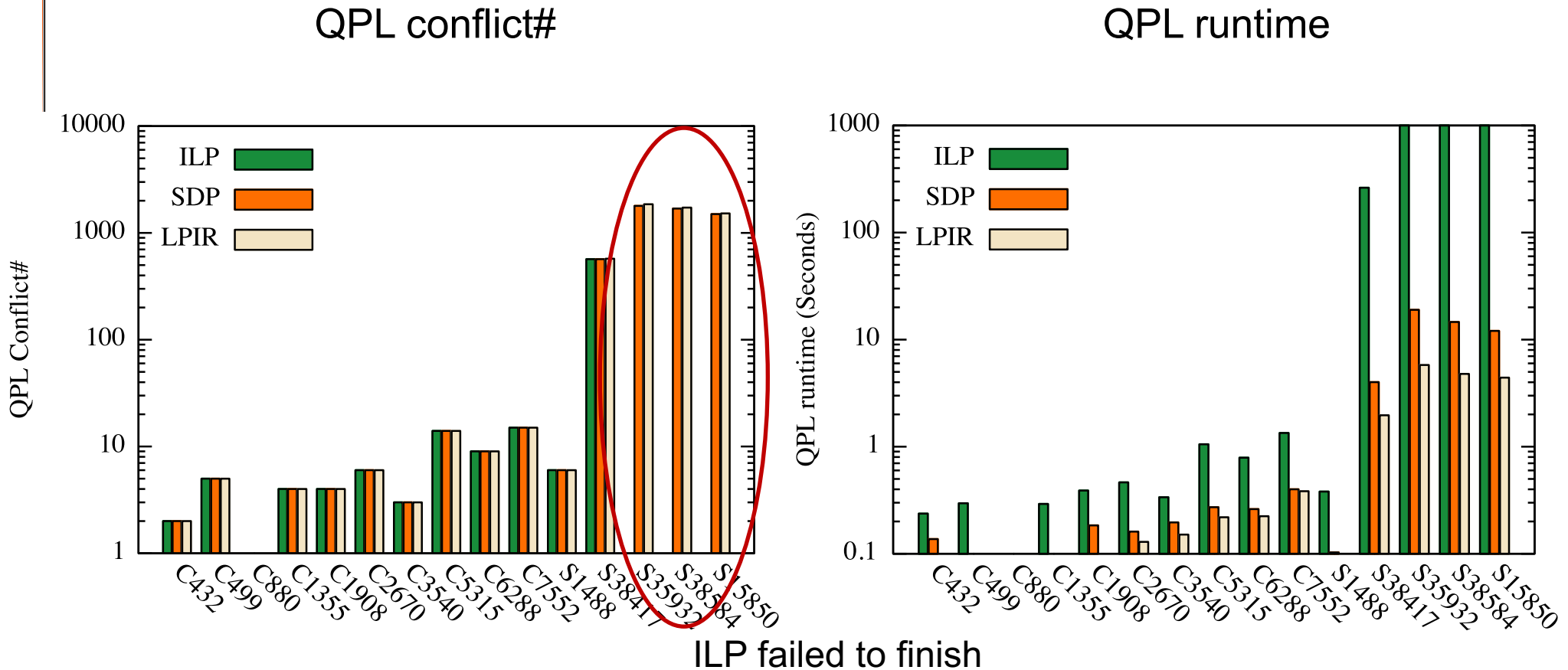


Baseline 1: ILP [Yu+, TCAD'15]

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



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!**