



《芯片设计自动化与智能优化》 Legalization

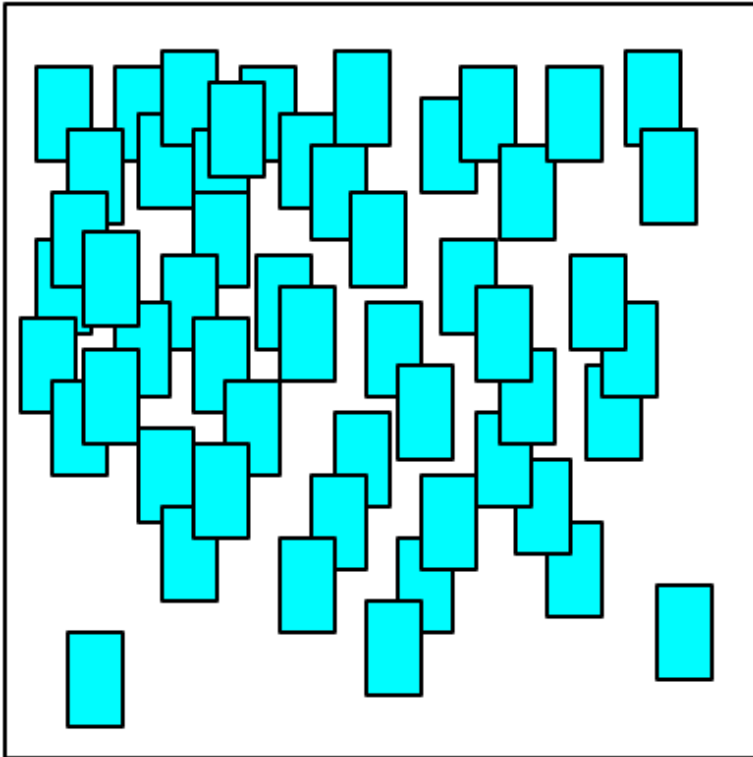
The slides are partly based on Prof. David Z. Pan's lecture notes at UT Austin.

Yibo Lin

Peking University

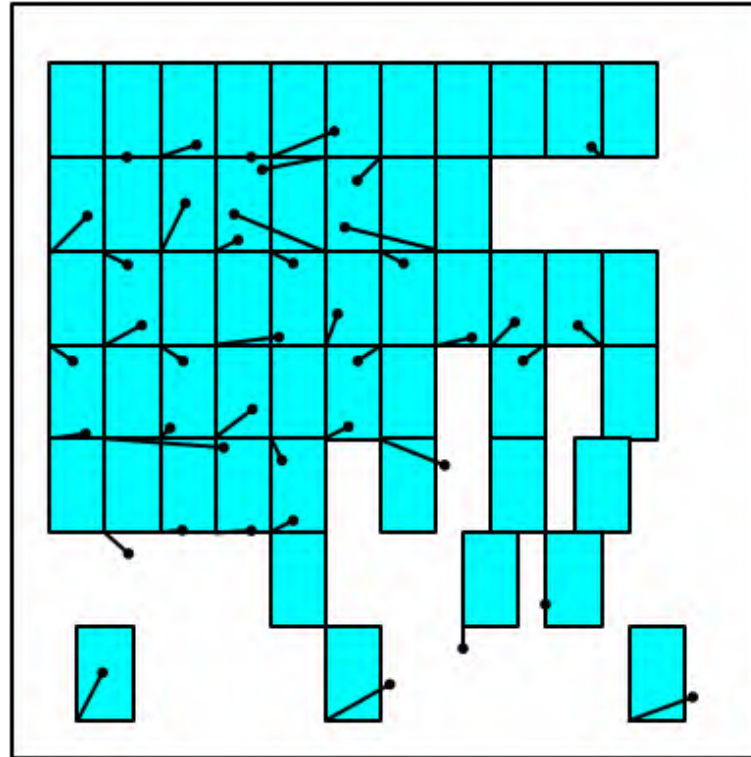
Typical Placement Flow

WL: 1.00e+6



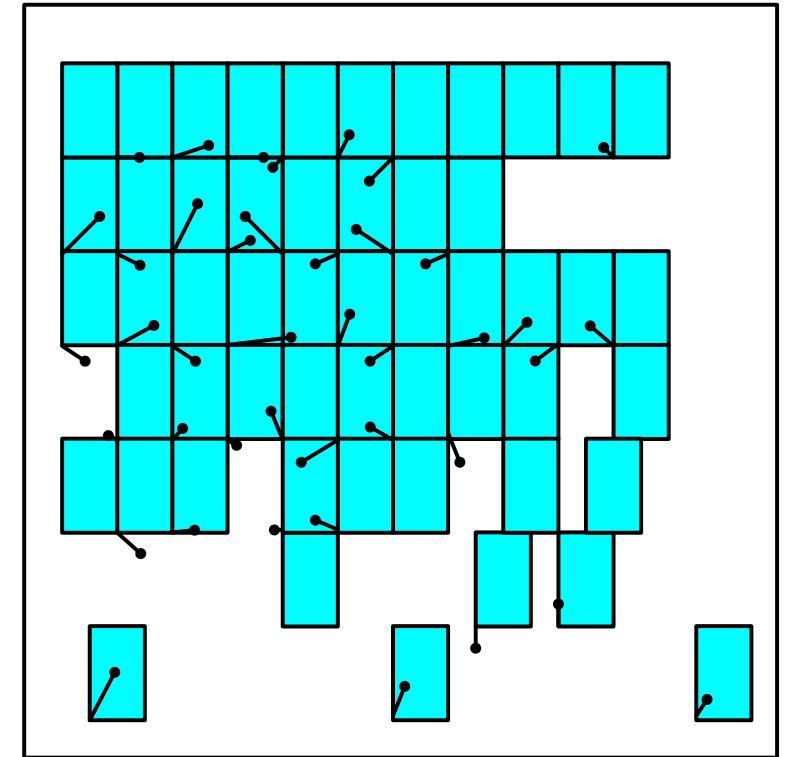
Global placement

WL: 1.05e+6



Legalization

WL: 1.02e+6



Detailed Placement

Outline

- What is placement
- History of placement algorithms
- Global placement
 - Quadratic placement: FastPlace & SimPL
 - Nonlinear placement: NTUplace & ePlace
- **Legalization**
 - Tetris
 - Row-based algorithms: Abacus, DP, LP, MCF
 - Integer linear programming
- Detailed placement
 - Global move & swap
 - Independent set matching
 - Local reordering
 - Row-based algorithms: DP, LP, MCF
- Other topics
 - Routability-driven placement
 - Timing-driven placement
 - Macro placement

Legalization – Problem Formulation

➤ Input

- Global placement solution

➤ Output

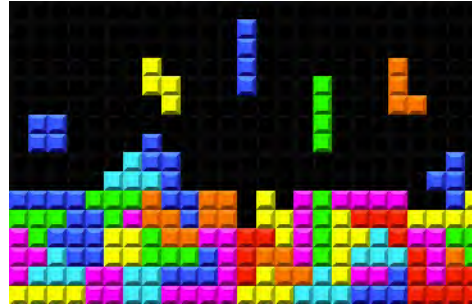
- Determine block locations
- Remove overlaps between blocks
- Satisfy all design rules

➤ Objective

- Minimize total displacement

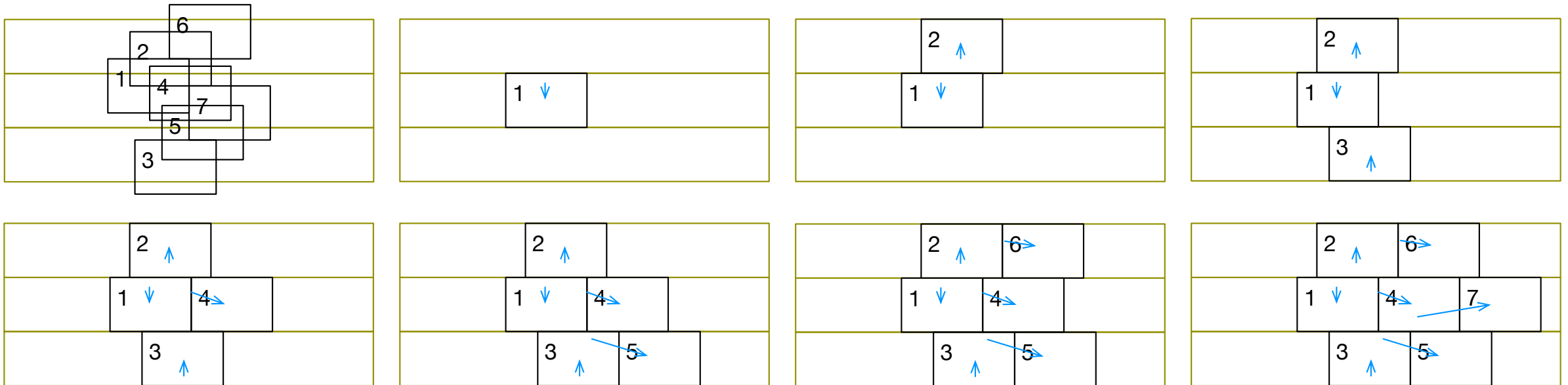
Tetris – Greedy Approach

- Sort cells from left to right
- Place cells one-by-one

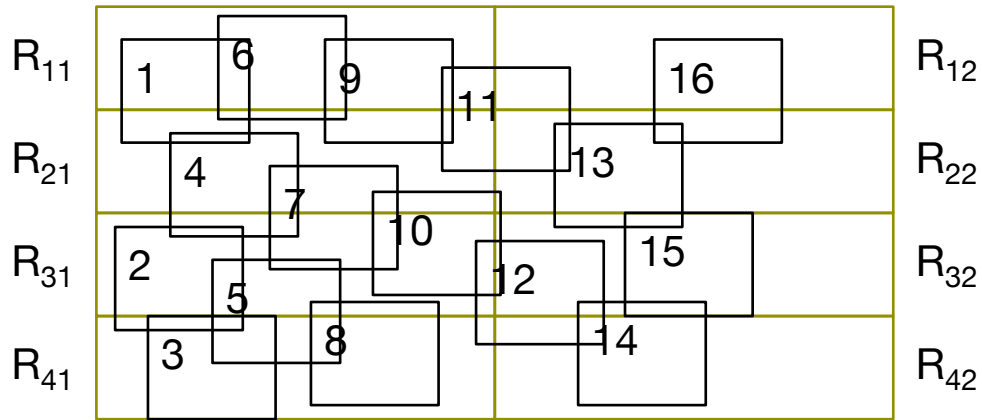


Simple, but effective
[NTUplace3, TCAD2018]

Global placement

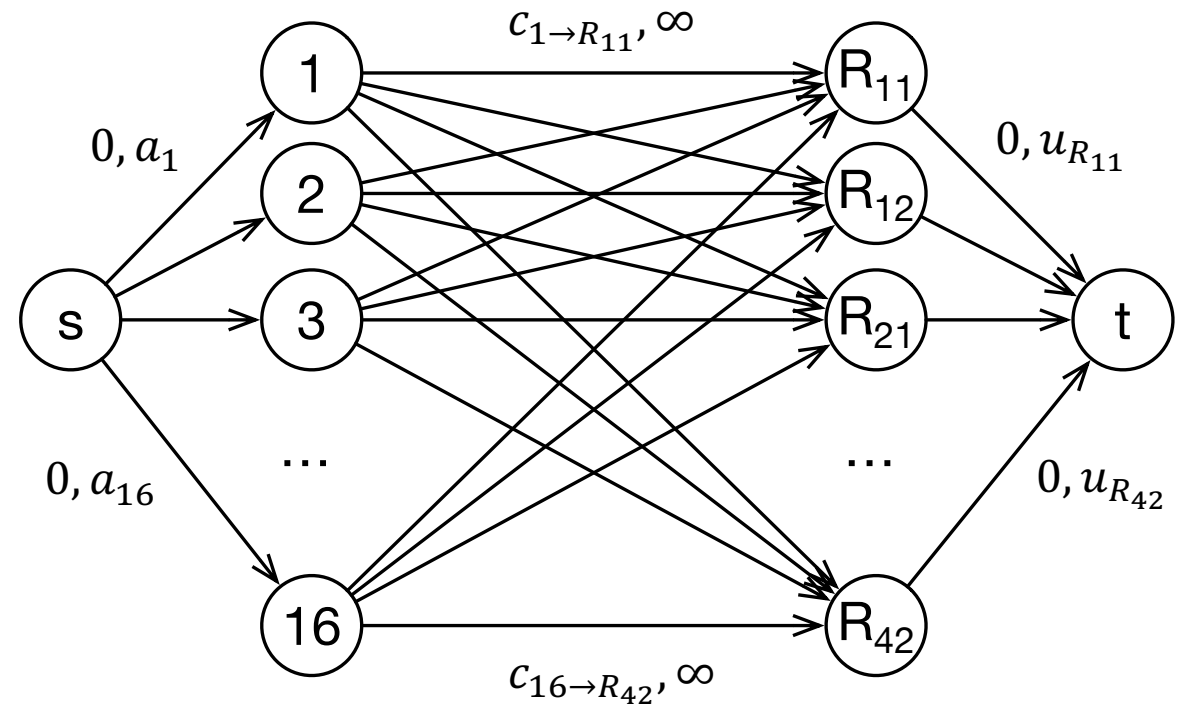


Region Assignment with Min-Cost Flow



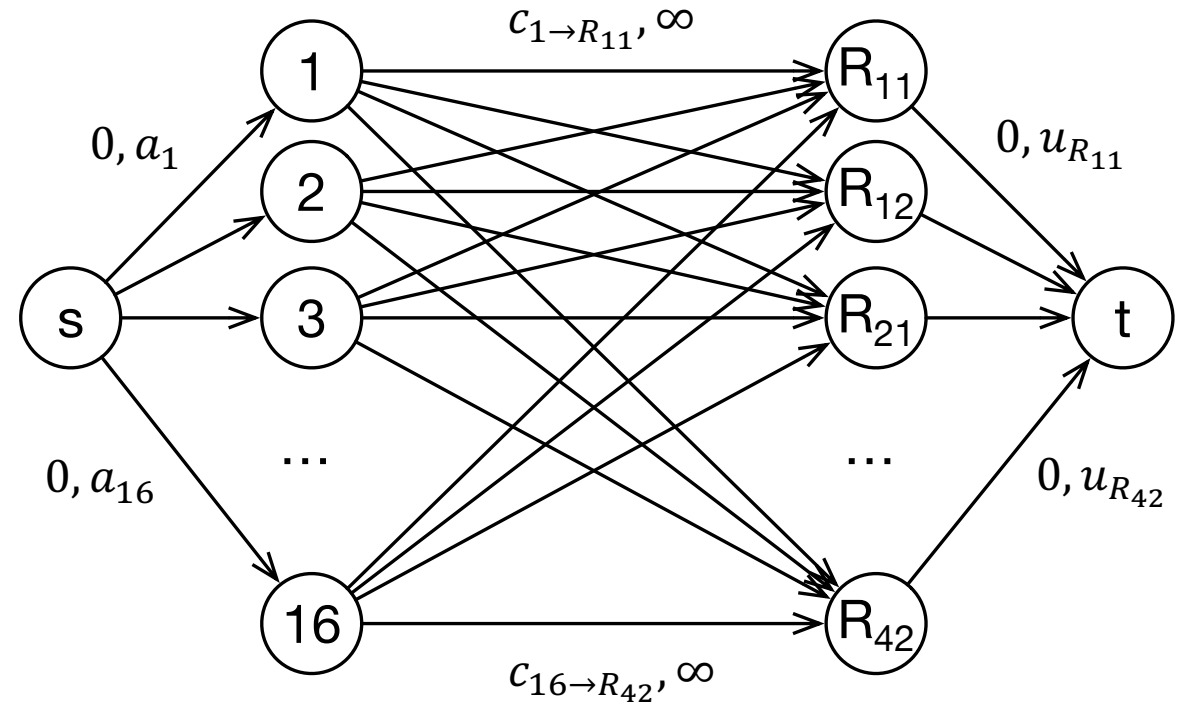
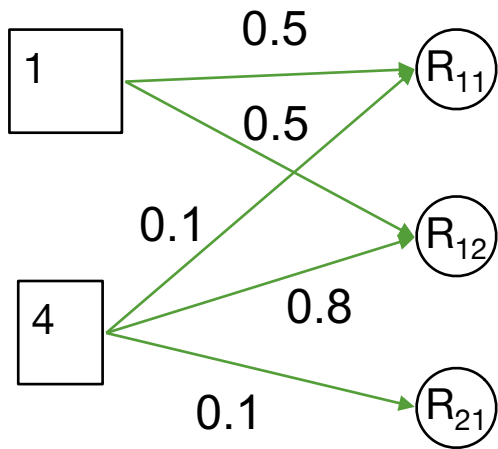
Edge cost
 $c = \widehat{disp}$

Center-to-center distance?
 Sensitive to region sizes?

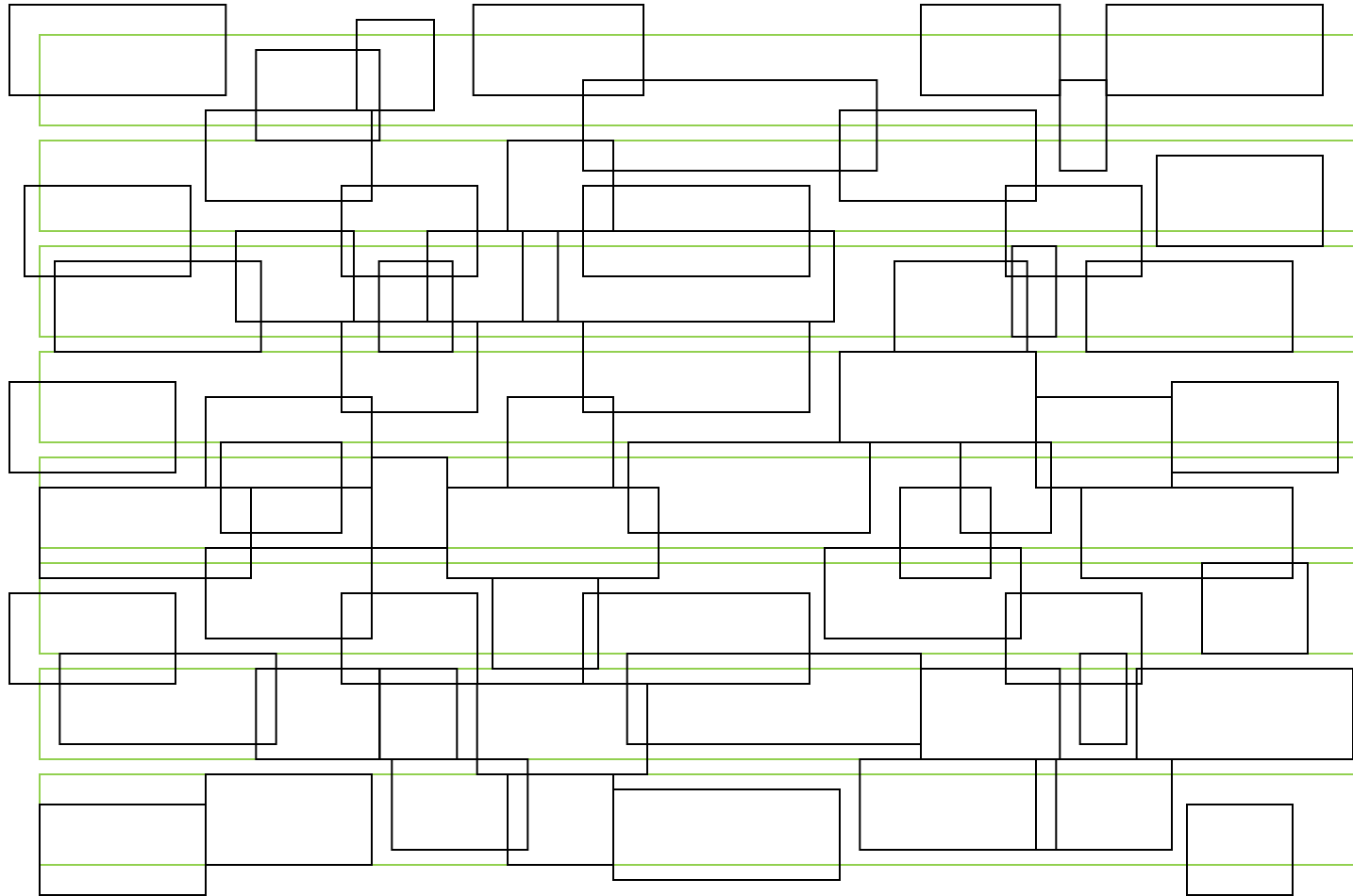


Region Assignment with Min-Cost Flow

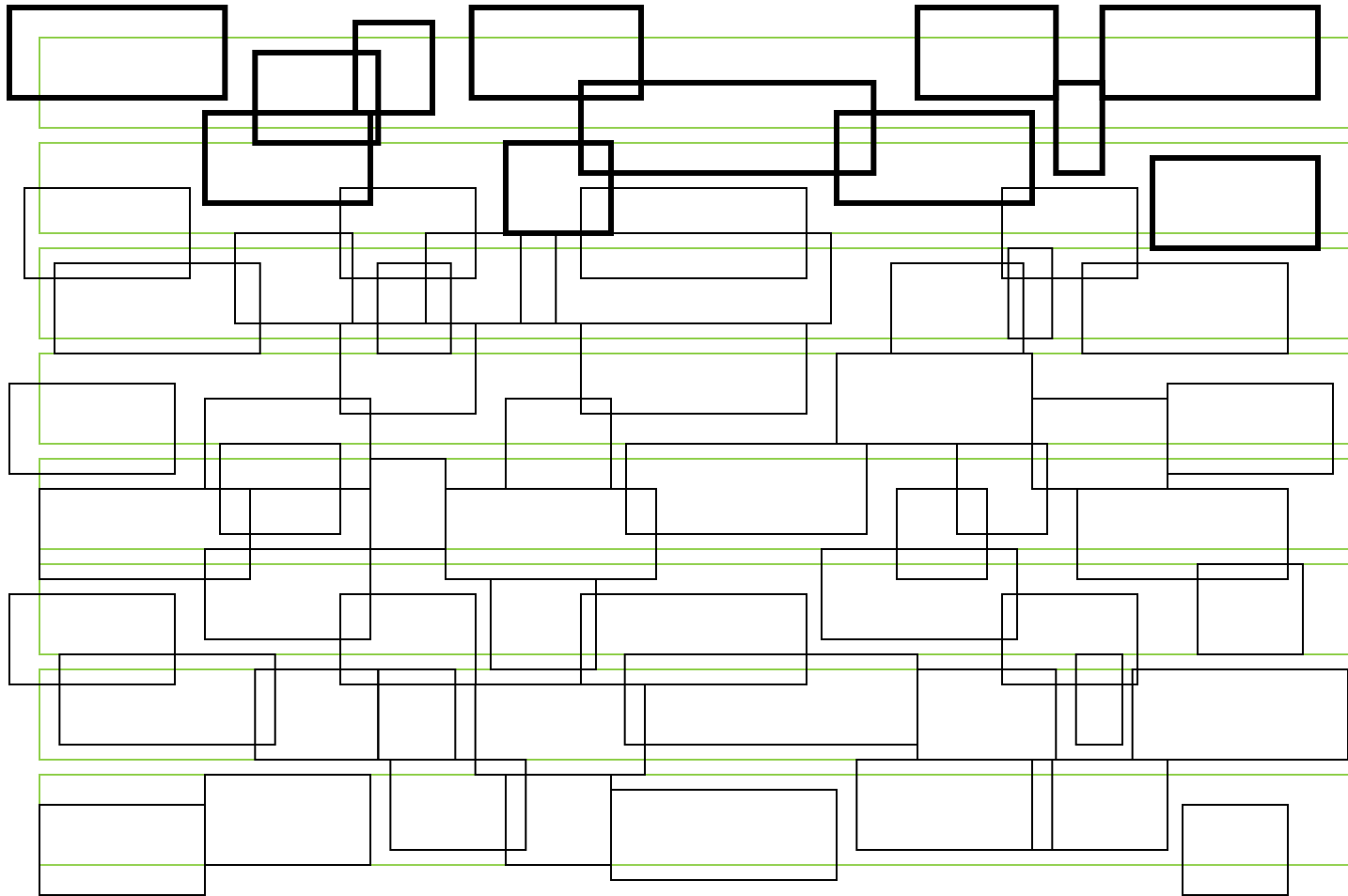
A cell may be assigned to multiple regions



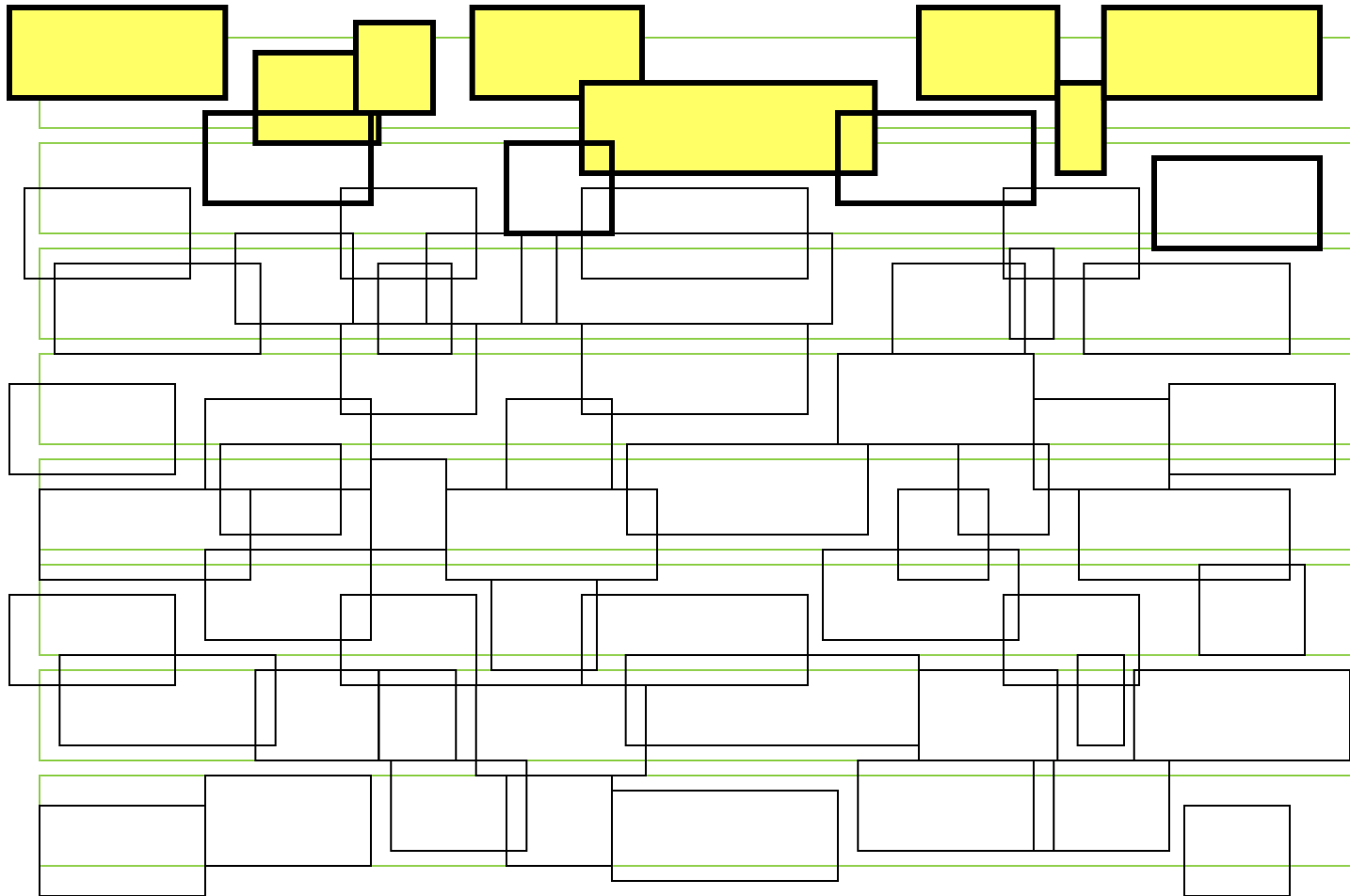
Row-by-Row Legalization



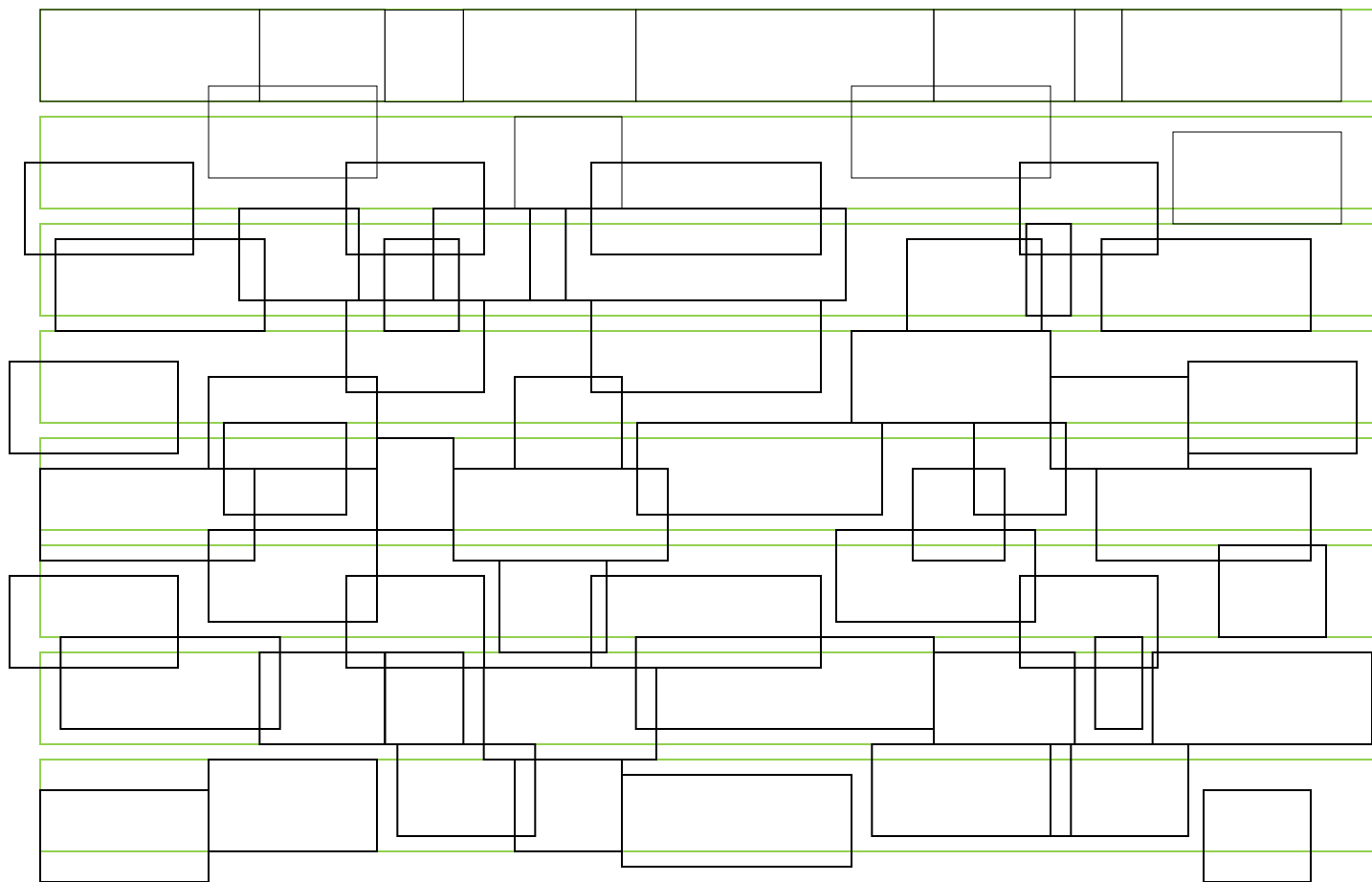
Row-by-Row Legalization



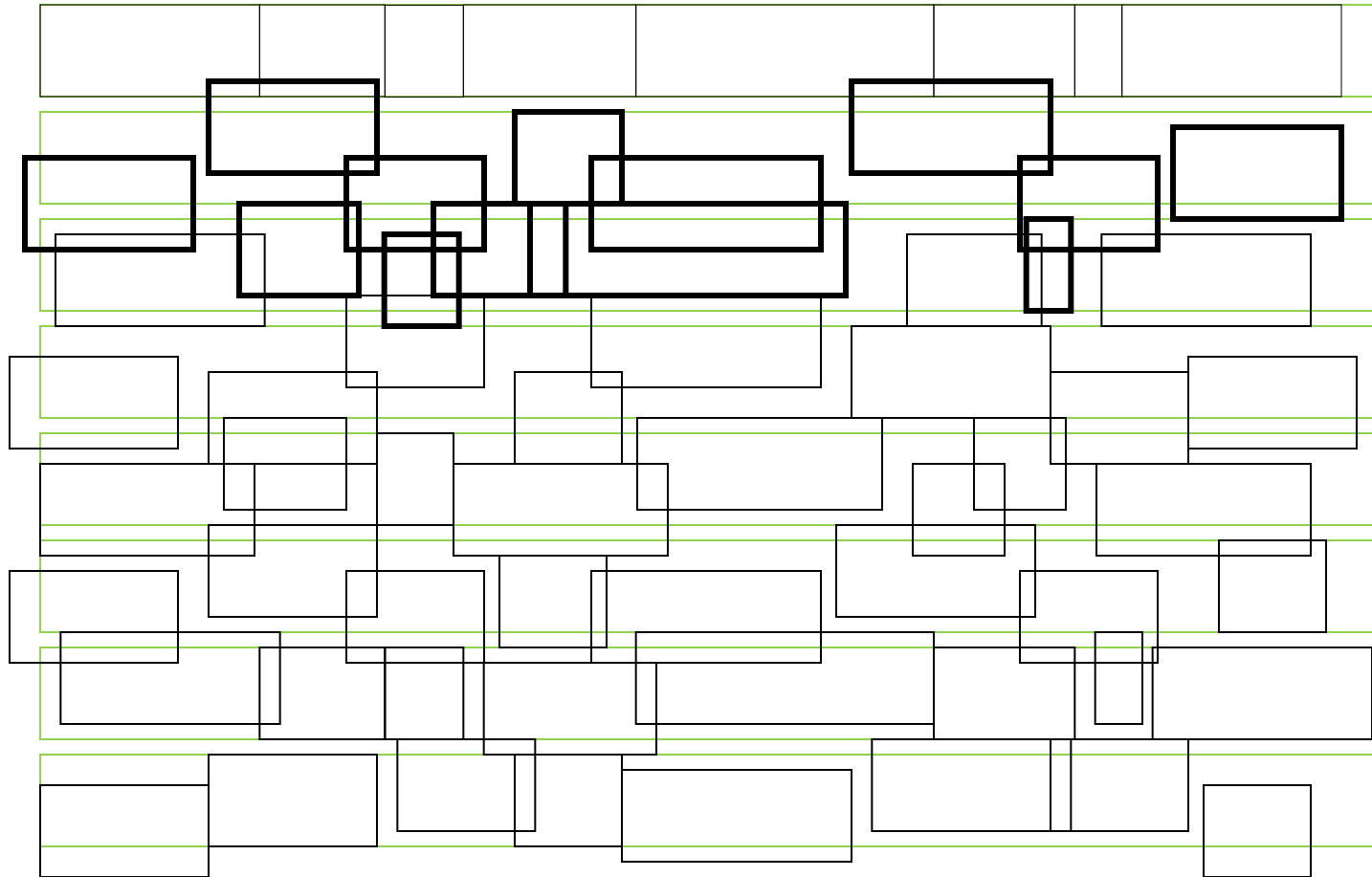
Row-by-Row Legalization



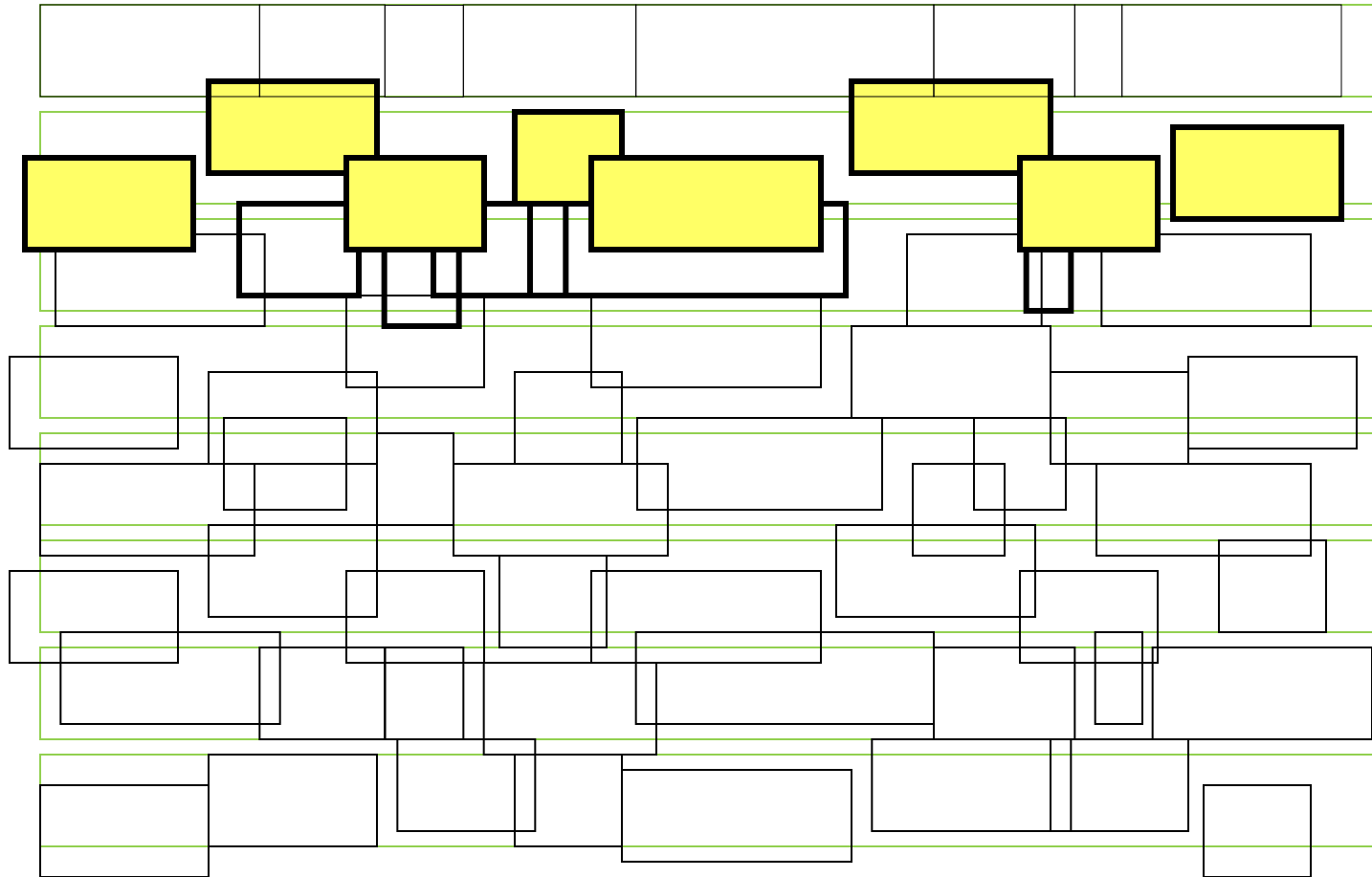
Row-by-Row Legalization



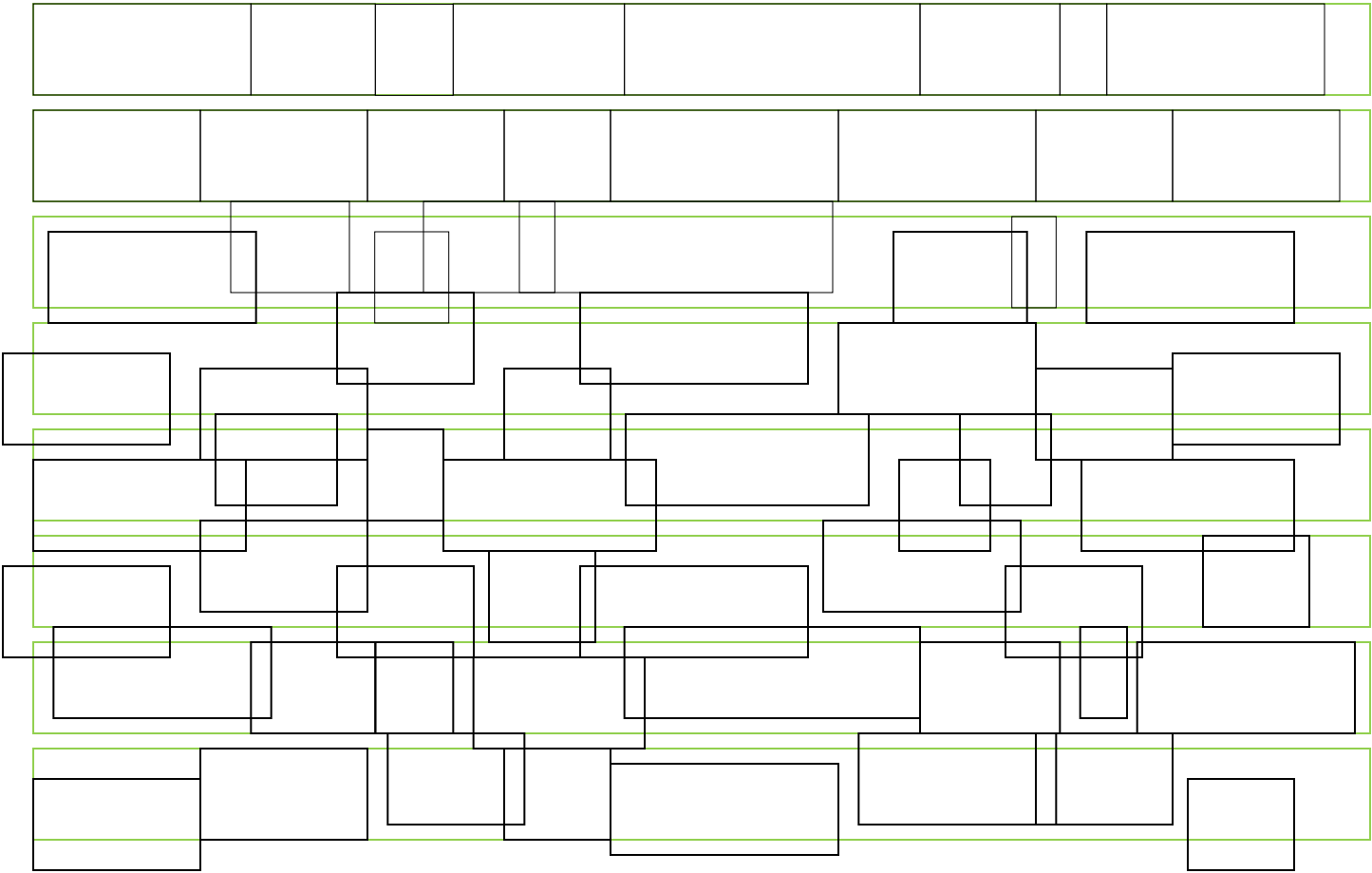
Row-by-Row Legalization



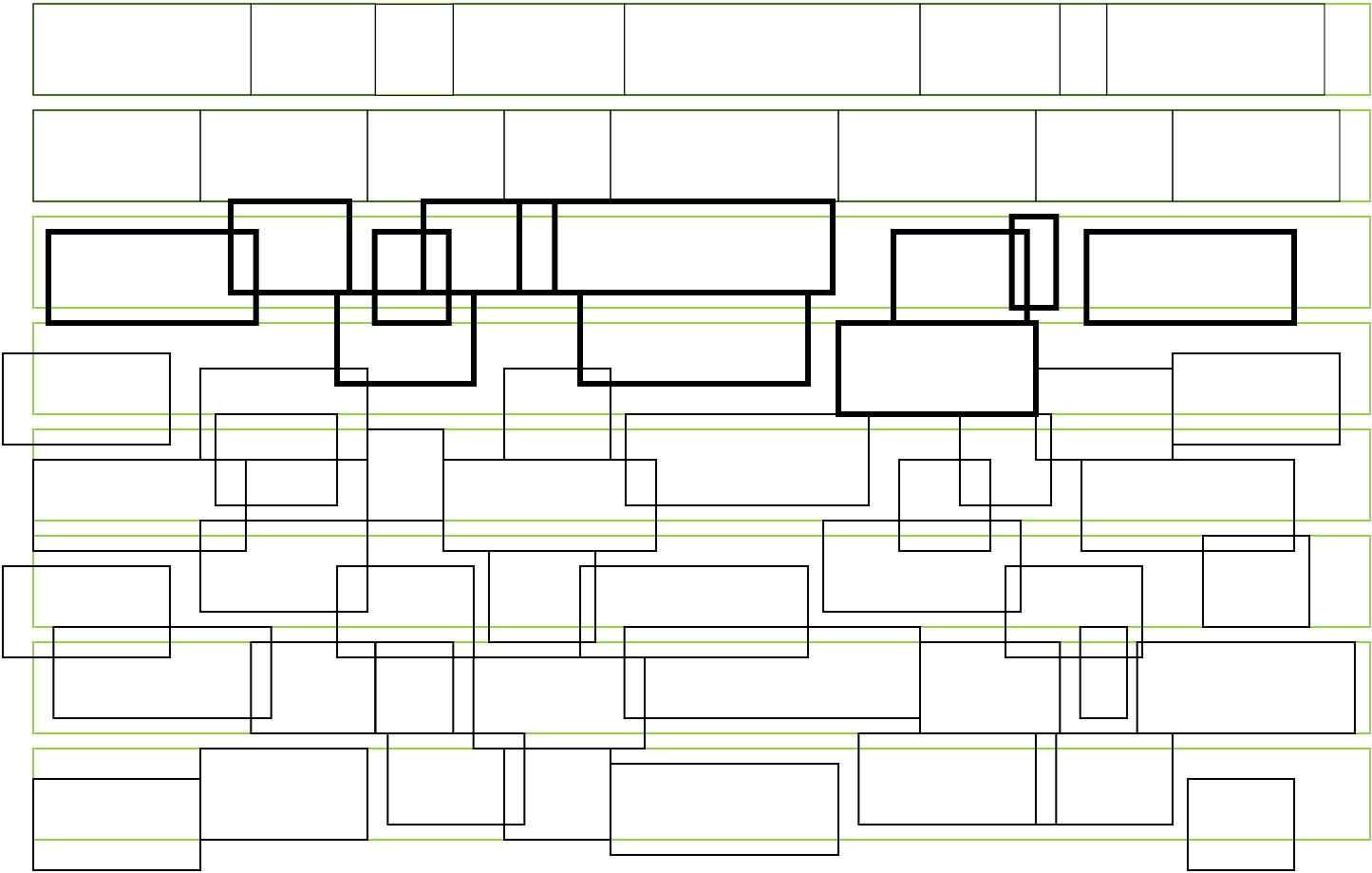
Row-by-Row Legalization



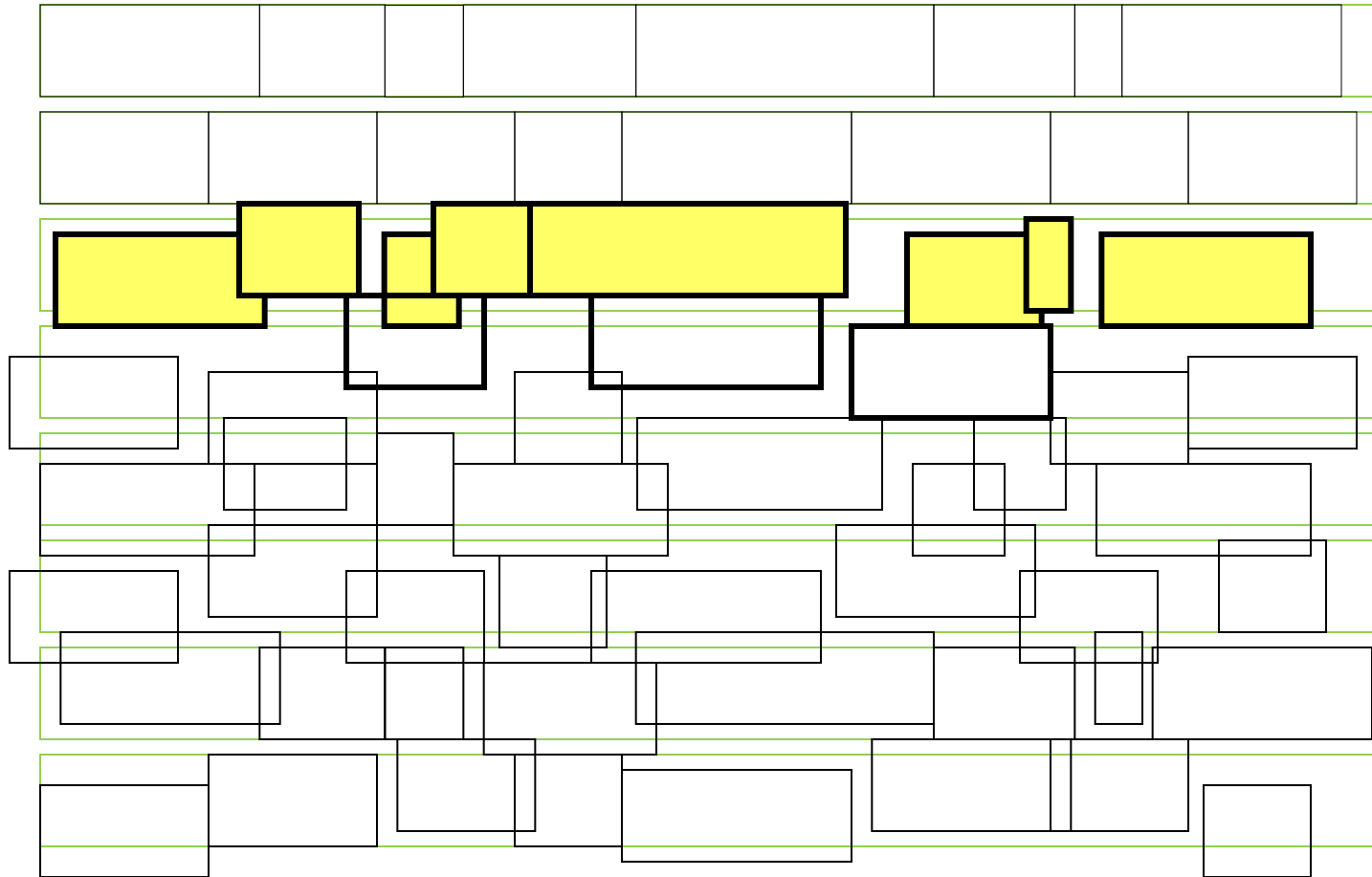
Row-by-Row Legalization



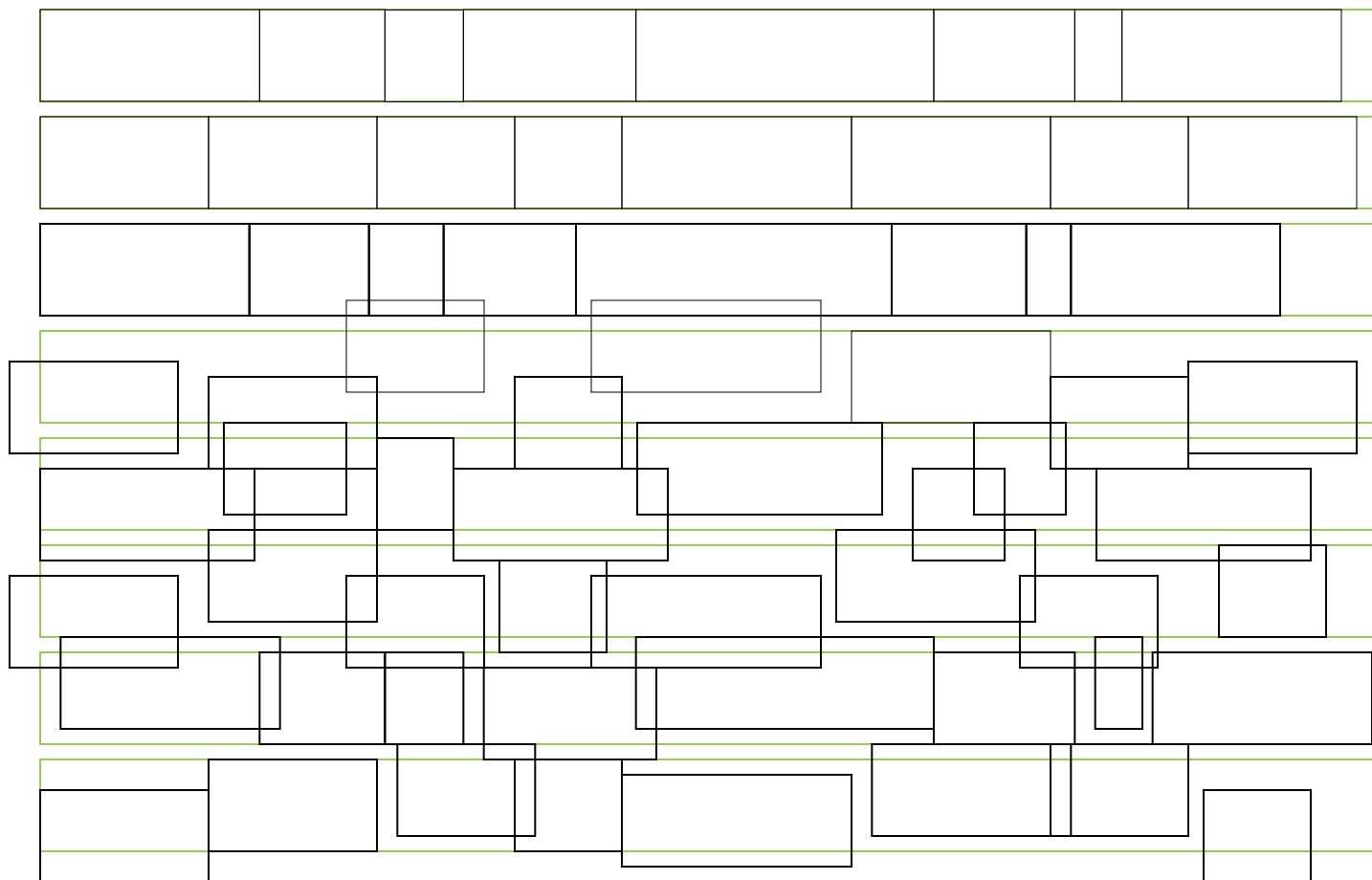
Row-by-Row Legalization



Row-by-Row Legalization



Row-by-Row Legalization



Row-by-Row Legalization

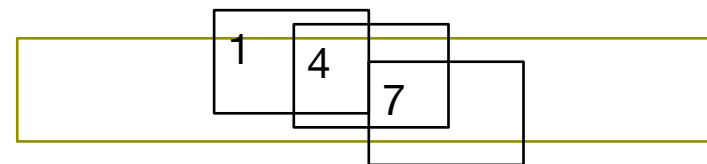
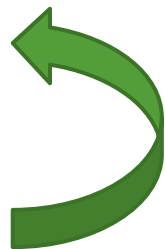
[illegible]

Row-based Algorithms – Abacus

- Assume the y coordinates of cells have been determined
 - Keep the **relative order** of cells
 - Only compute the x coordinates of cells

- Consider a row of cells (sorted from left to right)

- $\min_x \sum_{i=1}^{N_r} e_i (x_i - x_i^{gp})^2,$
- s.t. $x_i - x_{i-1} \geq w_{i-1}, \quad i = 2, \dots, N_r$



- Further simplify \geq to =

- $x_i = x_1 + \sum_{k=1}^{i-1} w_k$
- $\underbrace{\sum_{i=1}^{N_r} e_i x_1}_{e_c} - \underbrace{\left[e_1 x_1^{gp} + \sum_{i=2}^{N_r} e_i (x_i^{gp} - \sum_{k=1}^{i-1} w_k) \right]}_{q_c} = 0$
- $x_1 e_c - q_c = 0 \Leftrightarrow x_1 = \frac{q_c}{e_c}$

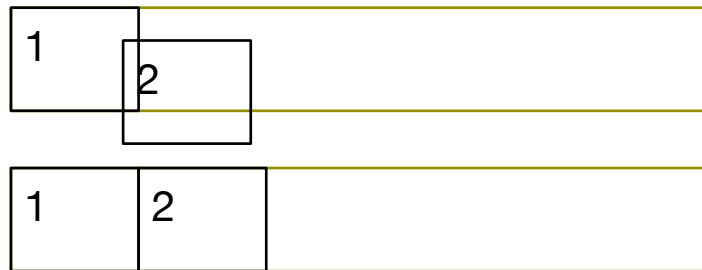
Init	Iteration ($i = 1, 2, \dots, N_r$)
$e_c = 0$	$e_c \leftarrow e_c + e_i$
$q_c = 0$	$q_c \leftarrow q_c + e_i (x_i^{gp} - w_c)$
$w_c = 0$	$w_c \leftarrow w_c + w_i$

Row-based Algorithms – Abacus

- Handling inequality constraints with clustering

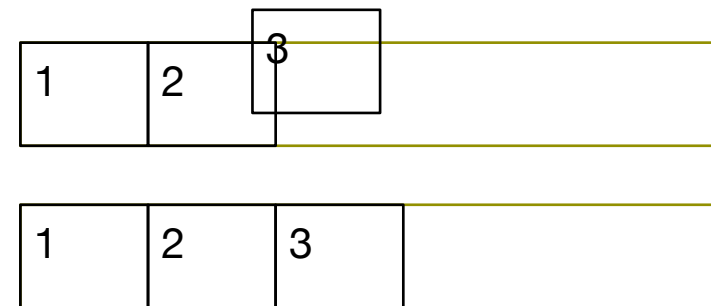


Place cell 1 at $x_1^* = \frac{q_c}{e_c}$



x_2^* overlaps with x_1^*
Merge 1 and 2 as one cluster

$$x_{\{1,2\}}^* = \frac{q_c}{e_c}$$

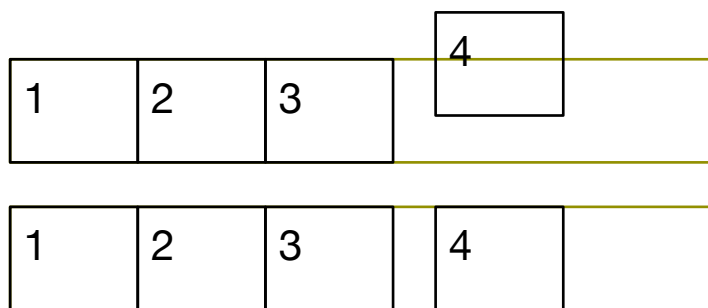


x_3^* overlaps with $x_{\{1,2\}}^*$
Merge 3 and $\{1,2\}$ as one cluster

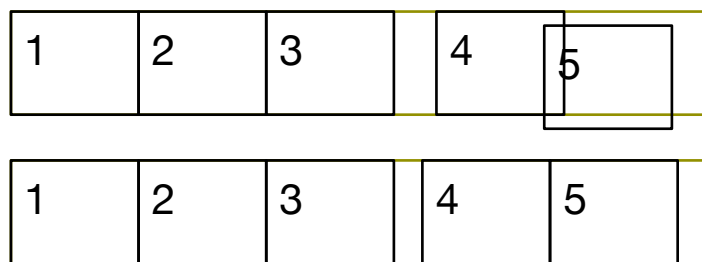
$$x_{\{1,2,3\}}^* = \frac{q_c}{e_c}$$

Row-based Algorithms – Abacus

► Handling inequality constraints with clustering

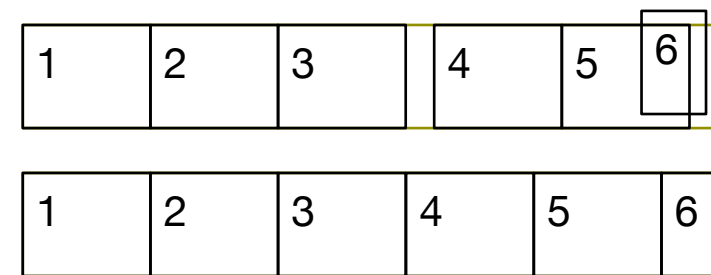


x_4^* does not overlap with $x_{\{1,2,3\}}^*$
Put cell 4 at x_4^*



x_5^* overlaps with x_4^*
Merge 5 and 4 as one cluster

$$x_{\{4,5\}}^* = \frac{q_c}{e_c}$$



x_6^* overlaps with $x_{\{4,5\}}^*$
Merge 6 and $\{4,5\}$ as one cluster

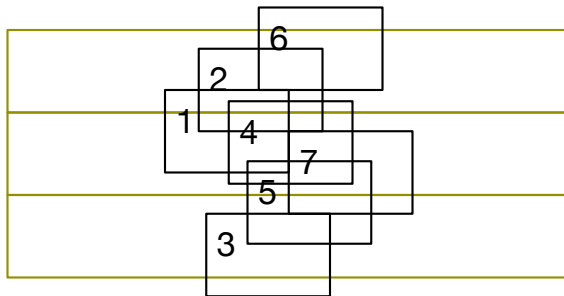
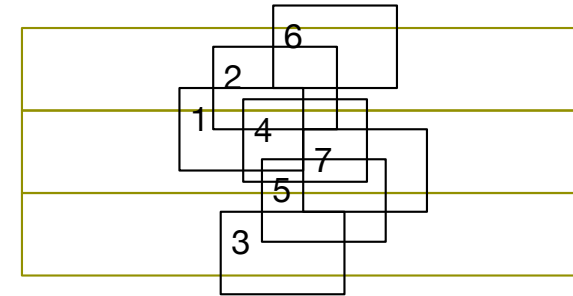
$$x_{\{4,5,6\}}^* = \frac{q_c}{e_c}$$

$x_{\{4,5,6\}}^*$ overlaps with $x_{\{1,2,3\}}^*$
Merge $\{4,5,6\}$ and $\{1,2,3\}$ as one cluster

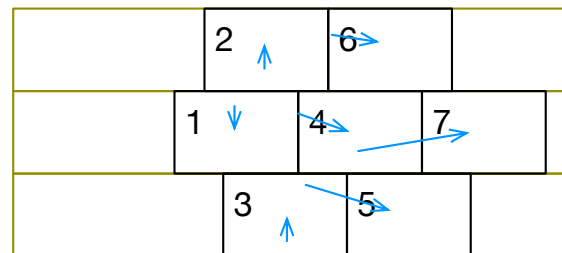
$$x_{\{1,2,3,4,5,6\}}^* = \frac{q_c}{e_c}$$

Row-based Algorithms – Abacus

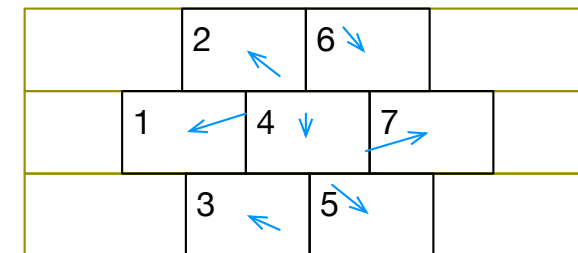
- Extend single-row legalization to multiple rows
- Sort all cells from left to right
- For each cell
 - Try assigning it to each row and legalize that row
 - Choose the best row to assign



Global placement



Tetris
Displacement = 3.752



Abacus
Displacement = 2.915

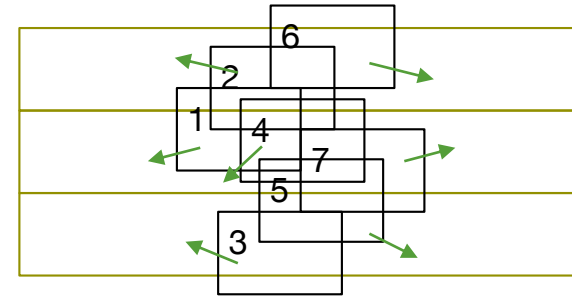
Row-based Algorithms – Linear Programming (LP)

► Pre-assign cells to rows

- Assign {2, 6} to the 1st row
- Assign {1, 4, 7} to the 2nd row
- Assign {3, 5} to the 3rd row

► Mathematical programming

- Keep the **relative order** of cells
- $\min_x \sum_i |x_i - x_i^{gp}|$
- s. t. $x_6 - x_2 \geq w_2,$
- $x_4 - x_1 \geq w_1, x_7 - x_4 \geq w_4,$
- $x_5 - x_3 \geq w_3,$
- $x_1 \geq L, x_2 \geq L, x_3 \geq L, x_5 \leq R, x_7 \leq R, x_6 \leq R$



Convert $|\cdot|$ to linear constraints
Introduce additional variables d_i

$$\begin{aligned} \min_{x,d} \quad & \sum_i d_i \\ \text{s. t.} \quad & x_i - x_i^{gp} \leq d_i, \\ & x_i^{gp} - x_i \leq d_i \end{aligned}$$

Row-based Algorithms – Dual Min-Cost Flow

- ▶ LP with differentiable constraints only
 - The duality of this LP is a min-cost flow problem (strong duality)
 - Min-cost flow can be solved with network simplex, much faster than regular simplex solvers for LP

$$\mathcal{P} : \min \sum_{i \in N} b_i \pi_i + \sum_{(i,j) \in E} u_{ij} \alpha_{ij},$$

$$\text{s.t. } \pi_i - \pi_j - \alpha_{ij} \leq c_{ij}, \quad \forall (i,j) \in E,$$

$$\alpha_{ij} \geq 0,$$

Slack variable



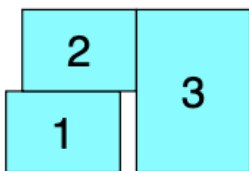
$$\mathcal{D} : \min \sum_{(i,j) \in E} c_{ij} f_{ij},$$

$$\text{s.t. } \sum_{j:(i,j) \in E} f_{ij} - \sum_{j:(j,i) \in E} f_{ji} = -b_i, \forall i \in N,$$

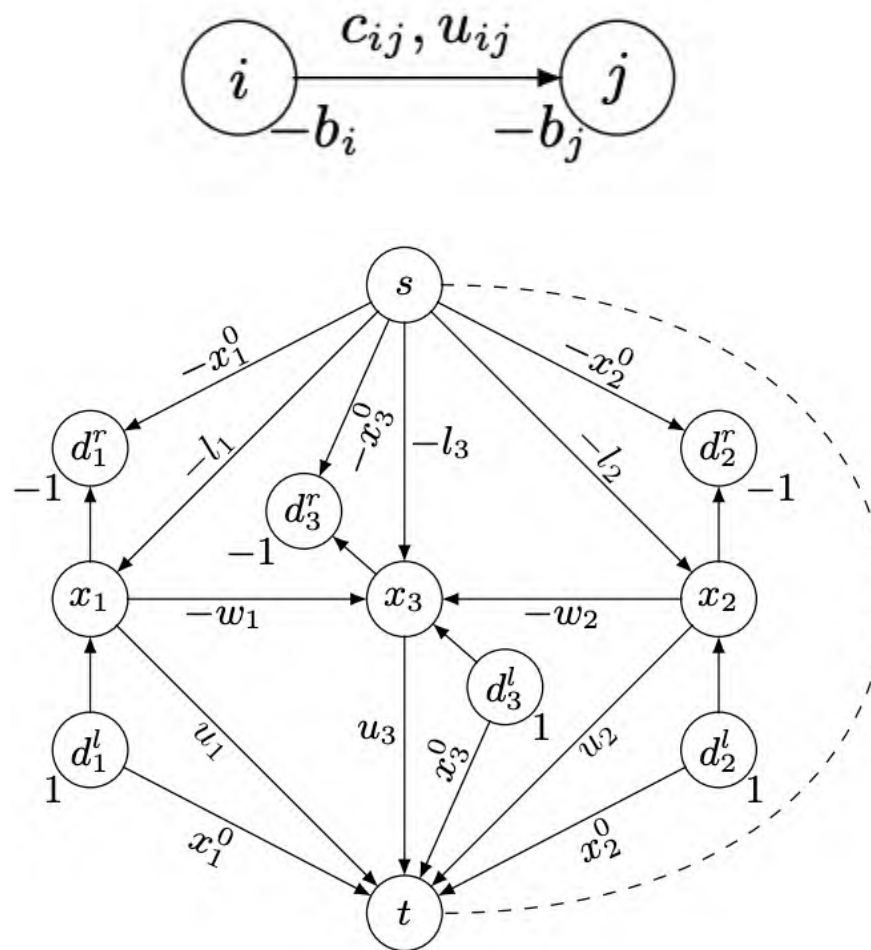
$$0 \leq f_{ij} \leq u_{ij}, \forall (i,j) \in E,$$

Row-based Algorithms – Dual Min-Cost Flow

$$\begin{aligned}
 \mathcal{P} : \min \quad & \sum_{i \in N} b_i \pi_i + \sum_{(i,j) \in E} u_{ij} \alpha_{ij}, \\
 \text{s.t.} \quad & \pi_i - \pi_j - \alpha_{ij} \leq c_{ij}, \quad \forall (i,j) \in E, \\
 & \alpha_{ij} \geq 0, \quad \forall (i,j) \in E,
 \end{aligned}$$



More extension with quadratic objective
 Linear complementary problem (LCP)
 [Chen+, DAC2017] Best Paper



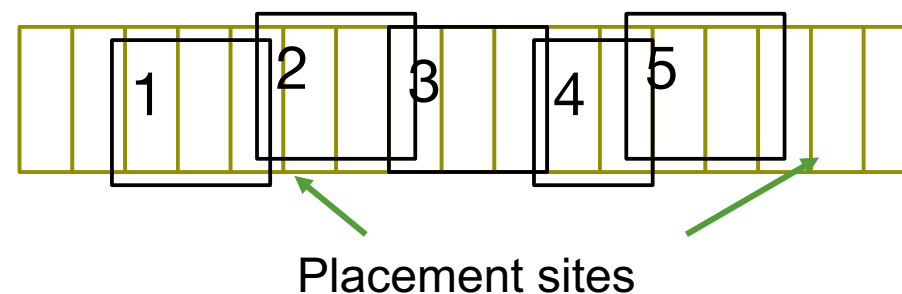
Site-based Algorithms

► Dynamic programming

- Taghavi, Taraneh, et al. "[New placement prediction and mitigation techniques for local routing congestion](#)." ICCAD 2010.
- Lin, Yibo, et al. "[Stitch aware detailed placement for multiple e-beam lithography](#)." Integration 2017. (Best Paper Award)

► Integer linear programming

- Li, Shuai, and Cheng-Kok Koh. "[Mixed integer programming models for detailed placement](#)." ISPD 2012.



Legalization – Summary

- Tackle legalization problem from different angles
- Tetris
- Region assignment
- Row-based legalization
 - Abacus
 - Linear programming
 - Min-cost flow
- Site-based legalization
 - Dynamic programming
 - Integer linear programming



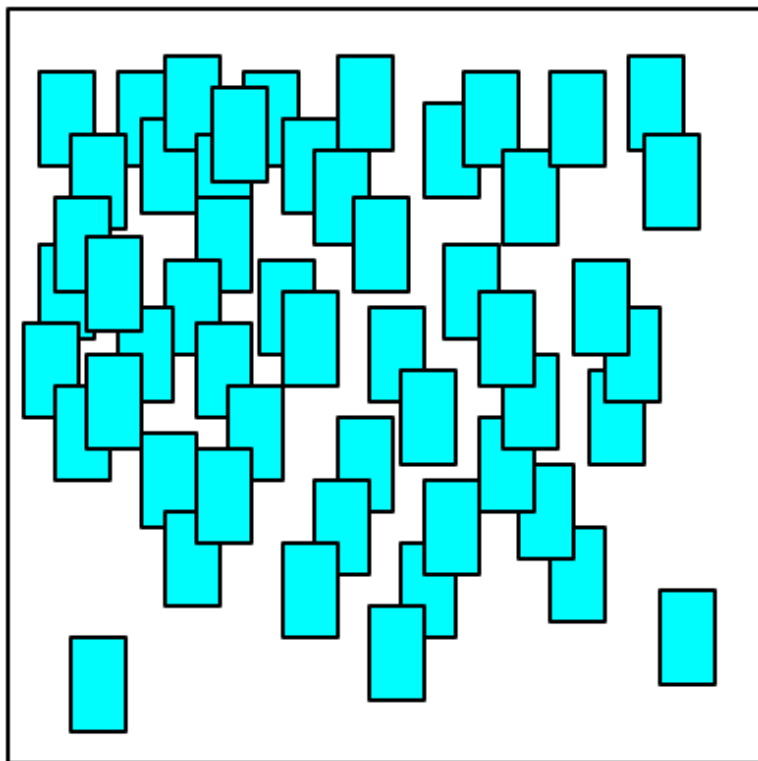
Continuous Problem



Discrete Problem

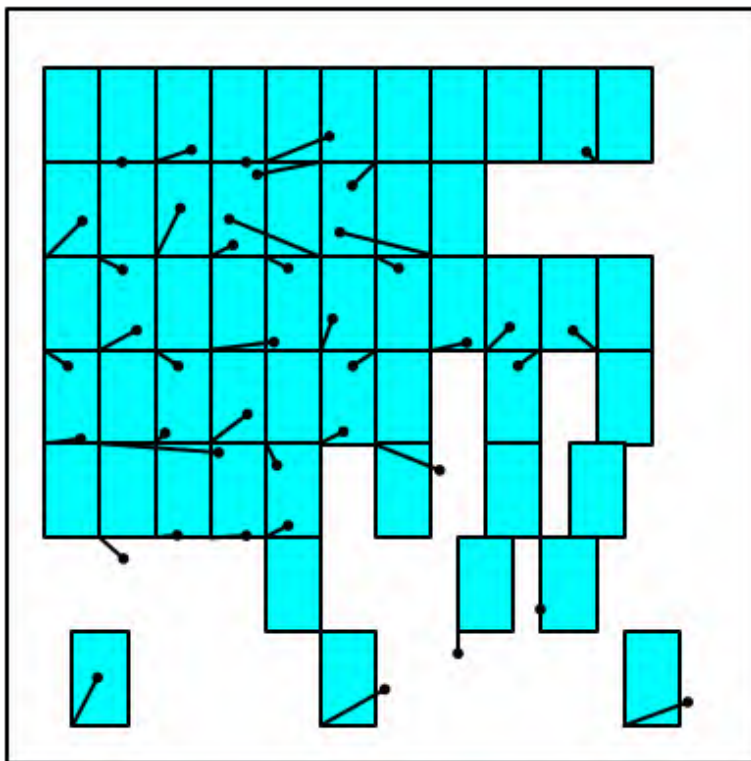
Typical Placement Flow

WL: 1.00e+6



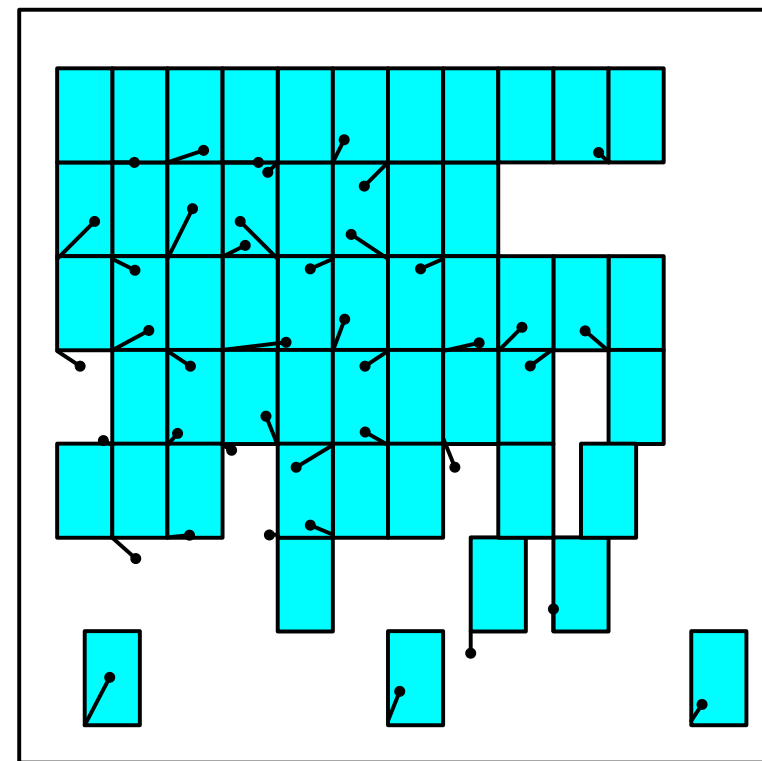
Global placement

WL: 1.05e+6



Legalization

WL: 1.02e+6



Detailed Placement