

## **Advanced Flow-Based Multilevel Hypergraph Partitioning**

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- Hypergraph  $H = (V, E, c, \omega)$ 
  - vertex set V = {1, ..., n}
  - edge set  $E \subseteq \mathcal{P}$  ( V)  $\setminus \emptyset$
  - incident edges  $\Gamma(u) = \{ e \in E \mid u \in e \}$
  - vertex weights  $\varphi: V \to \mathbb{R}_{\geq 1}$
  - edge weights  $\omega: E \to \mathbb{R}_{\geq 1}$







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Partition into k disjoint blocks  $\Pi = \{V_1, \ldots, V_k\}$ 

blocks V<sub>i</sub> have roughly equal weight:

$$\varphi(V_i) \leq (1 + \varepsilon) \left\lceil \frac{\varphi(V)}{k} \right\rceil$$







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minimize connectivity objective:

$$con = \sum_{e \in E} (\lambda(e) - 1) \omega(e)$$







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С

 $\lambda(e) = |\{V_i \mid V_i \cap e \neq \emptyset\}|$ # blocks overlapping with *e* 



## **Applications**



### **Distributed Databases**



### **VLSI** Design





### **Route Planning**





## **Multilevel Algorithms**







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slide kindly provided by Sebastian Schlag







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X get **stuck** in local optima



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X large edges → **zero** gain moves









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#### Algorithm 1: FM Local Search



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select two adjacent blocks for refinement



build graph-based flow model



slide kindly provided by Sebastian Schlag







build graph-based flow model

- either: restrict flow model size so that balance is guaranteed
- or: make it a little larger, hope for balance. if not ~> scale down again







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### The new KaHyPar-HFC





select two adjacent blocks for refinement



flows directly on hypergraph



use FlowCutter [Hamann, Strasser JEA18] [Gottesbüren, Hamann, Wagner ESA19]

### naturally built-in

find more balanced minimum cut



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•  $\widetilde{f}(u, e) < 0 \Rightarrow u$  receives flow from e







$$\operatorname{rcap}(e, u, v) = c(e) - f(e) + \widetilde{f}(u, e)^{-} + \widetilde{f}(v, e)^{+} = (12 - 7) + 2 + 4 = 11$$
 $\operatorname{max}(0, -\widetilde{f}(u, e)) \quad \operatorname{max}(0, \widetilde{f}(v, e))$ 





























- $\widetilde{f}(u, e) > 0 \Rightarrow u$  sends flow into *e*
- $\widetilde{f}(u, e) < 0 \Rightarrow u$  receives flow from e

$$\operatorname{rcap}(e, u, v) = c(e) - f(e) + \widetilde{f}(u, e)^{-} + \widetilde{f}(v, e)^{+} = (12 - 7) + 2 + 4 = 11$$
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can implement *any* flow algorithm by treating nets like vertices
 we use Dinic



### Dinic



- residual BFS to compute distance labels
- residual DFS to find edge-disjoint shortest augmenting paths
- repeat until no flow augmented



## Dinic

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Distance labels for hypergraphs

- d[u] for vertices
- $d_i[e]$  for pushing flow to flow-sending pins  $\tilde{f}(v, e) > 0$
- $d_o[e]$  for pushing flow to all pins. set if  $c(e) f(e) + \tilde{f}(u, e)^- > 0$



## Dinic



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Optimizations

- capacity scaling
  - require  $\geq \alpha$  residual capacity
  - if no flow augmented try with  $\alpha \leftarrow \alpha/2$
- iterative DFS with stored iterators
- no allocations
- range of active values trick ~> fast resets













### 1. Augment flow







1. Augment flow

#### 2. Find min s- and t-cut







1. Augment flow 2. Find min *s*- and *t*-cut

#### 3. Pick smaller side






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3. Pick smaller side 4. Assimilate side







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### **Experimental Setup**



- Intel Xeon E5-2670 @ 2.6 Ghz, 64 GB RAM, 20 MB L3, 256KB L2
- # Hypergraphs: [publicly available]
  - SuiteSparse Matrix Collection 184
  - SAT Competition 2014 (3 representations) 92.3
  - ISPD98 & DAC2012 VLSI Circuits 28
- $k \in \{2, 4, 8, 16, 32, 64, 128\}$  with imbalance:  $\varepsilon = 3\%$

10 random seeds



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#### 10 random seeds

- Comparing KaHyPar-HFC\* and KaHyPar-HFC with:
  - KaHyPar-MF
  - hMetis-R(ecursive Bisection) & hMetis-K (direct -kway)
  - PaToH-D(efault) & PaToH-Q(uality)
  - Mondriaan
  - Zoltan-AlgD
  - HYPE













Performance ratio  $r(a, i) = \frac{con(a, i)}{\min\{con(a', i) \mid a' \in A\}}$  of algorithm *a* on instance *i* 







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- y-axis = fraction of instances with smaller performance ratio than value on x-axis
- **x** = 1  $\Rightarrow$  fraction of instances on which algorithm is the best
- higher is better







| Algorithm    | t[s]  |
|--------------|-------|
| KaHyPar-MF   | 67.07 |
| KaHyPar-HFC* | 62.49 |
| KaHyPar-HFC  | 44.84 |









### **Comparison with other Partitioners**







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## **Detailed Running Time**







#### Conclusion



KaHyPar-HFC – better and faster


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In the TR

- configuration study / assess impact of components
- different k / instance classes  $\Rightarrow$  improves a lot on dual SAT and large k
- earlier balance with subset sum for special vertices
- distance-based piercing
- flow routing

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https://kahypar.org

https://github.com/kahypar/kahypar

https://github.com/larsgottesbueren/WHFC



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