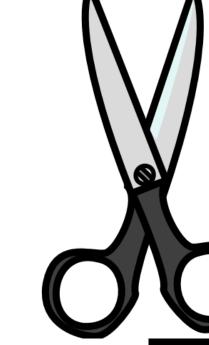
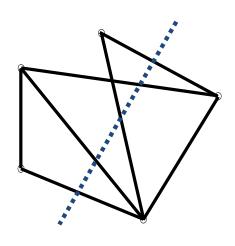
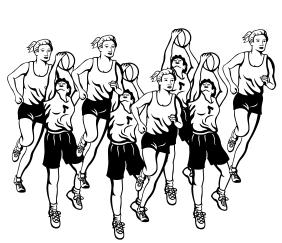
Partitioning
CPEN 513
2019/2020 Term 2











### Partitioning:

Often, circuits are bigger than a single chip -> Need to divide the circuit among chips

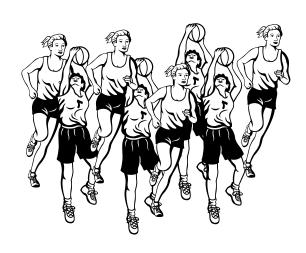
Also, may need to divide a circuit among two portions of a chip

#### Two approaches:

- 1. Multi-way Partitioning
- 2. Recursive Bi-partitioning

# Constructive Algorithms:

Back in Grade School Gym class, how did you divide into two teams?



# Constructive Algorithms:

```
Choose two seeds, put one in partition A and one in B
While there are still nodes left {
    Pick node "most tightly" connected to all nodes
        already in A and put it in partition A
    Pick node "most tightly" connected to all nodes
        already in B and put it in partition B
}
```

# Constructive Algorithms:

Problems with this approach:

- Very sensitive to choice of seeds
- For the first few nodes we are adding in each partition, we have very little information about what will end up in each partition at the end

We would like to make "bad moves" sometimes

# Simulated Annealing:

Cost Function could be |C| where C is the set of cut edges (in other words, we are trying to minimize the number of cut edges).

Problem: This will create very unbalanced partitions

It has been shown that this works well:

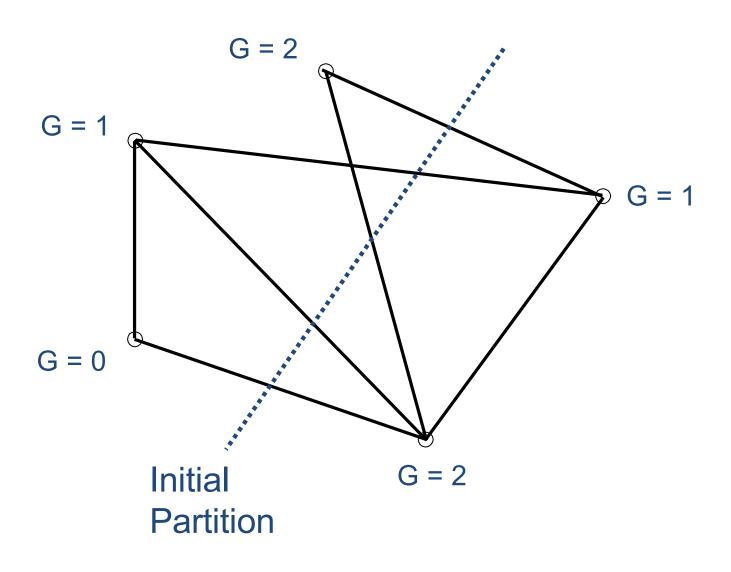
$$|C| + \lambda B$$

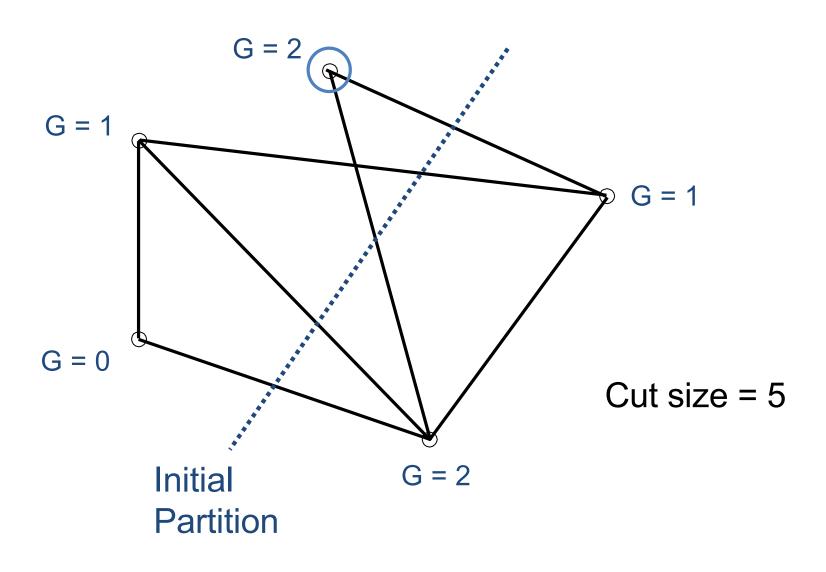
where B somehow quantifies inbalance

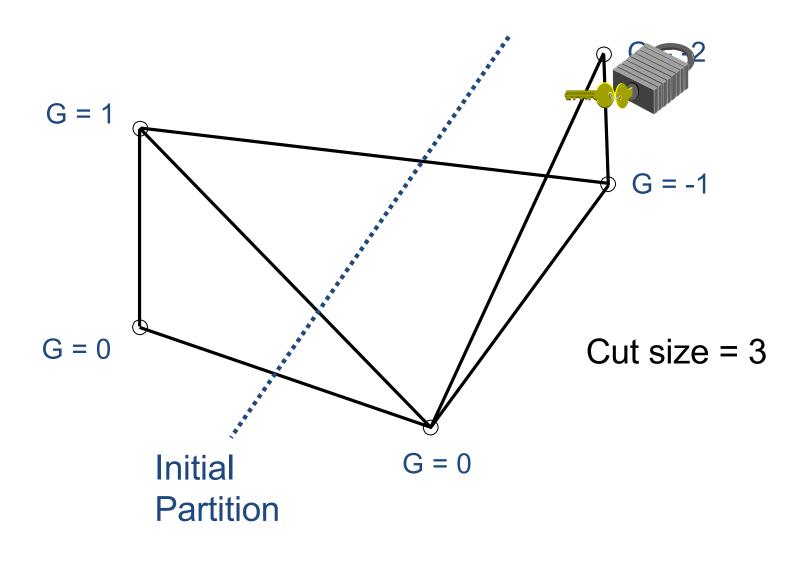
Main difference between this and SA: Moves are not chosen randomly.

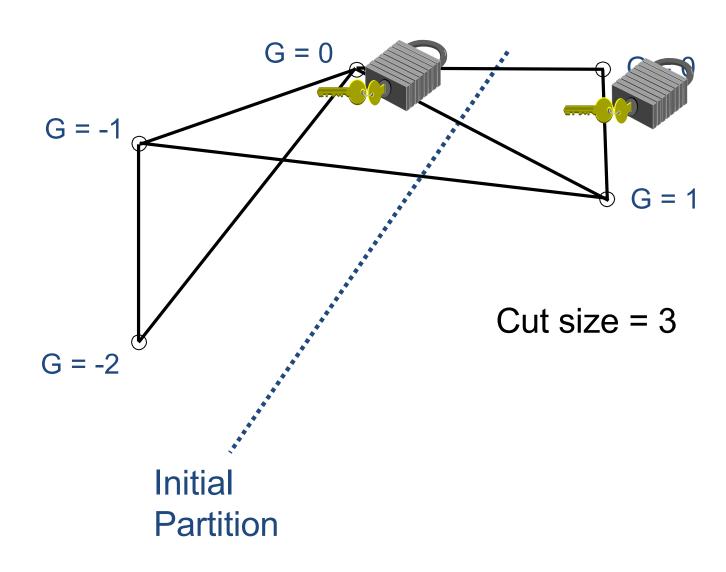
#### Definition: Gain(v)

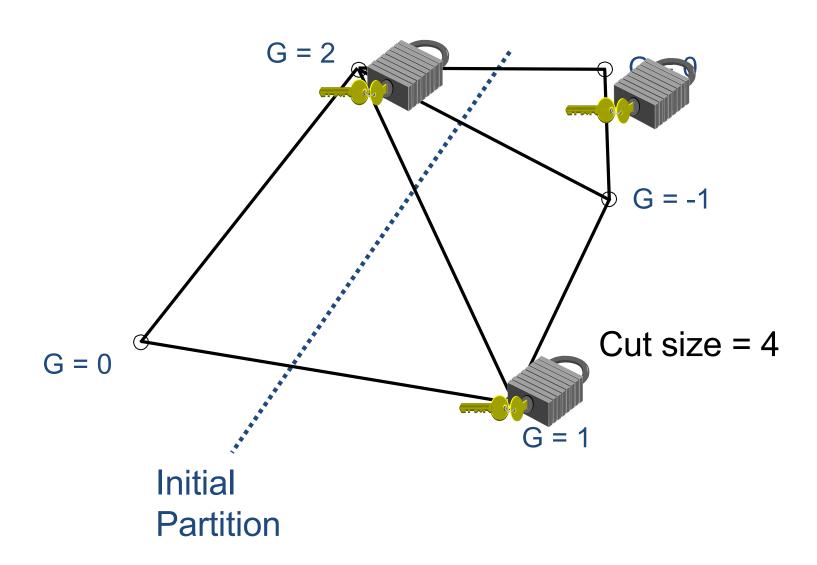
- Given a partition, and a node v, Gain(v) is defined as the improvement in cut-set size if v is swapped to the opposite partition
- Equivalently,

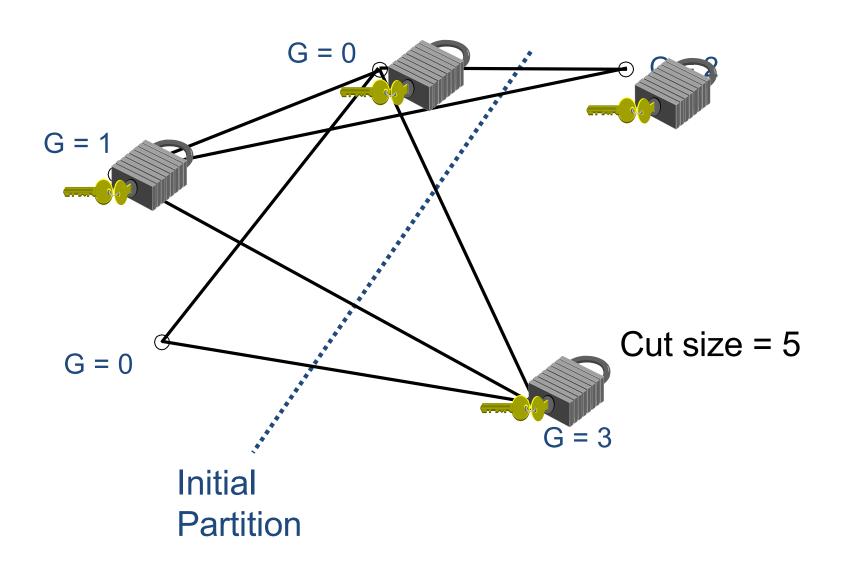


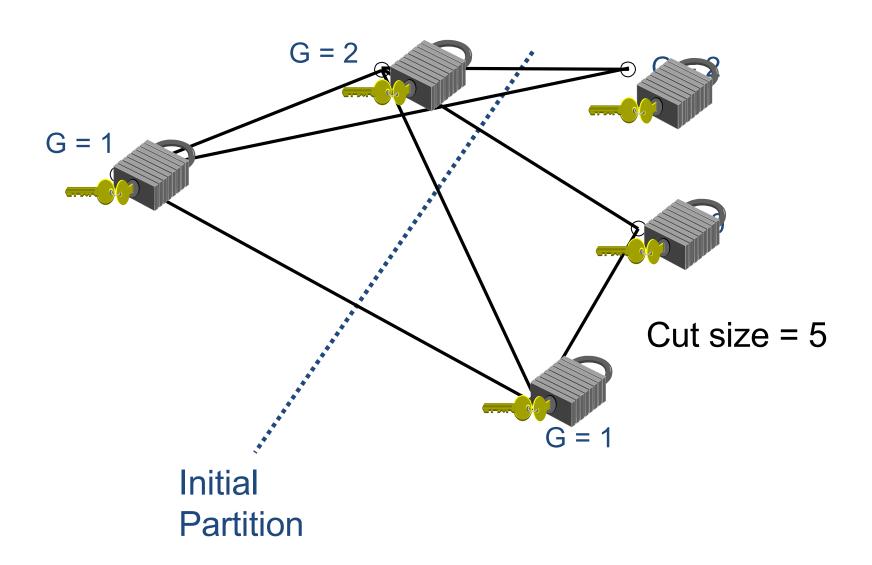






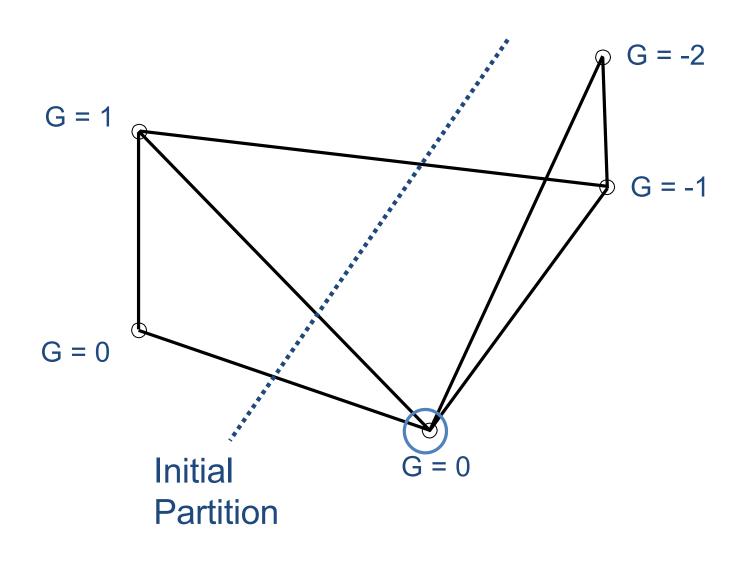






All the nodes are locked so we are done this iteration

Best cut size we saw was 3, so roll back to that and do another iteration



Repeat this for a small number of iterations (typically 4-5 is enough)

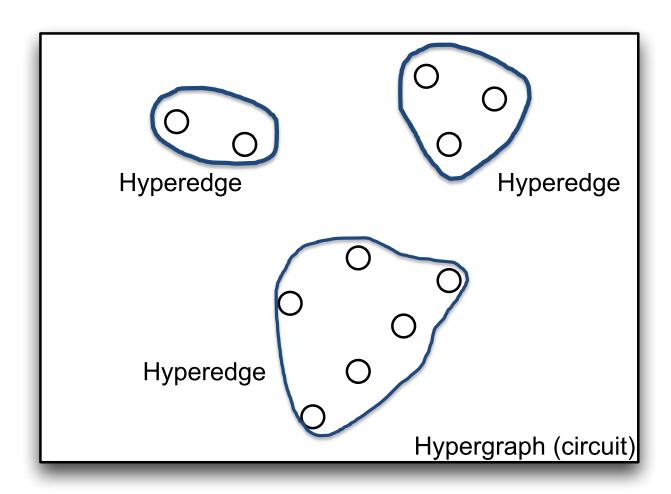
Choose the best cut-size found and that is our answer

By using clever data structures this algorithm is O(n)

#### Multi-sink nets

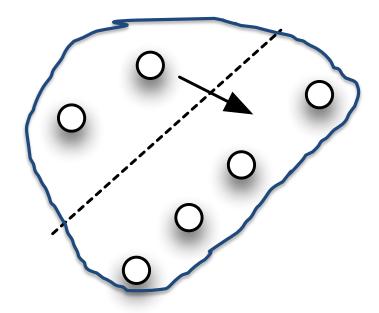
What about nets that have more than one sink?

- Each edge can have more than 2 endpoints



## Dealing with Multi-sink nets

Option 1: Use FM as normal:



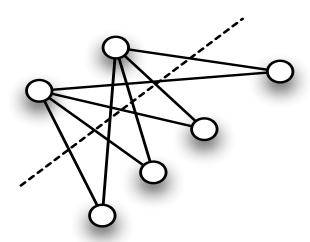
Problem: no motivation to make this move

### Dealing with Multi-sink nets

Option 2: Convert hypergraph to a graph and do FM on the graph usual technique: replace each hyperedge with a clique, and weight each edge on the clique as  $\frac{1}{|e|-1}$ 

where e is the number of

terminals



Problem: 1. Wrong Optimization Goal

2. Reduces "sparsity" of graph (increased runtime)

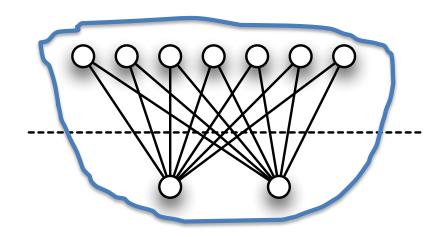
Other approaches: delete edges of the clique randomly

### Assignment 3: Due March 2

You will write a Kernighan-Lin / Fiduccia-Matheyses partitioner

Important: Cost function is # of nets that cross the partition

- Each net contributes either 0 or 1 to the sum
- Be careful for multi-terminal nets



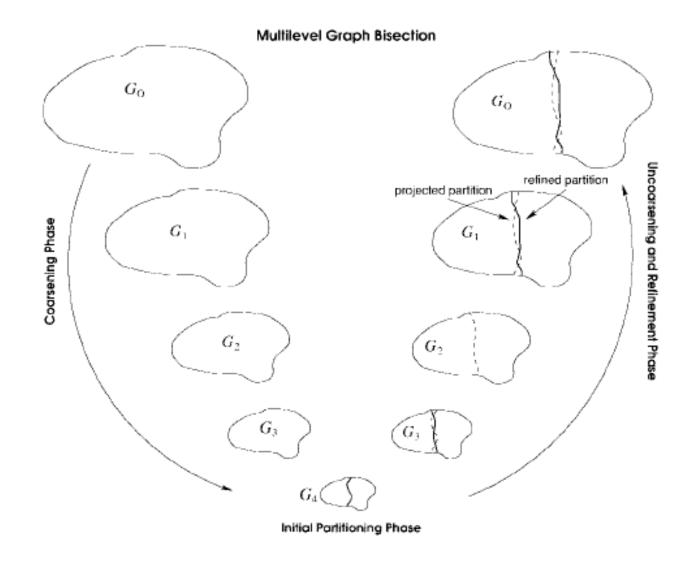
This net contributes 1 to the sum.

Board notes: Branch and Bound

**Board notes: Genetic Algorithms** 

See paper by Bui on the website – no paper review reqd

#### Combining with bottom-up Clustering: Hmetis



See paper by Karypis on the website – no paper review reqd

#### Summary of Partitioning

- 1. Constructive Algorithms
- 2. Simulated Annealing
- 3. Kernighan-Lin / Fiduccia-Matheyses
- 4. Branch and Bound (Assignment 3!)
- 5. Genetic Algorithms

Clearly, partitioning algorithms have applications outside of CAD.