

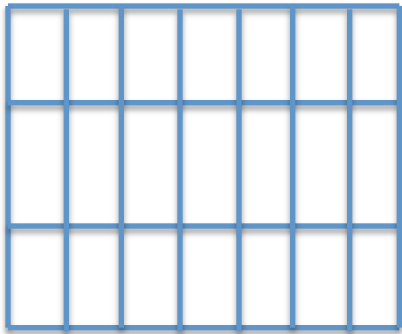
# Graph Partitioning

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# A load balancing problem

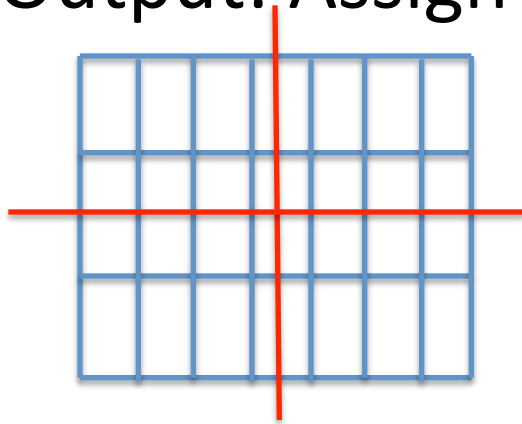
- Input:
  - undirected graph  $G$ . Vertices = processes. Edges = interprocess communication.
  - integer  $k$  = number of available processors.
- Output: Assign processes to processors



, 4

# A load balancing problem

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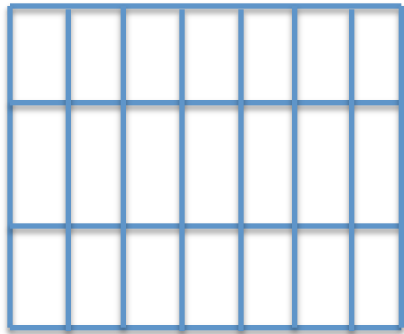


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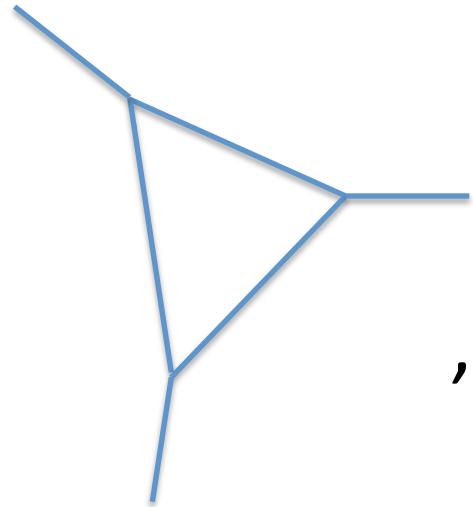
Give equal number of  
processes to each processor.  
Minimize actual communication

# The graph partitioning problem

- Input: undirected graph  $G$ , integer  $k$
- Output: Partition into subgraphs  $G_1, G_2, \dots, G_k$  with equal number of vertices, by removing minimum number of edges.



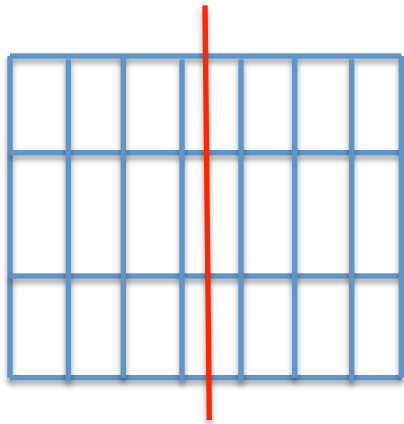
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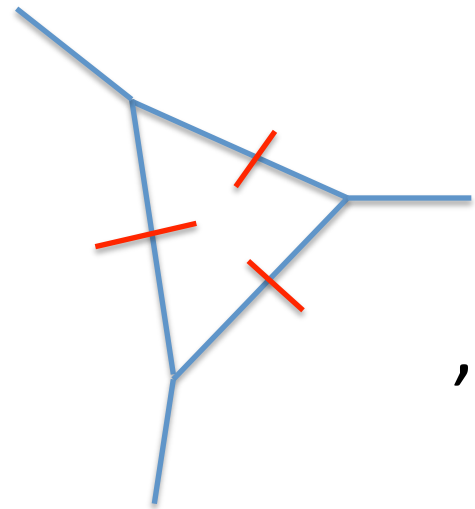
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# Other motivations

- Divide and conquer algorithms on graphs.
  - Partition into balanced subgraphs.
  - Recurse on subgraphs.
  - Cost of combine step : high if large number of edges between subgraphs.
- Identify bottlenecks in graph for purpose of fault tolerance etc.
- Nice connections to geometry.

Rest of this talk:  $k=2$ , i.e. **Graph bisection**

# Outline

- Basic results
- Partitioning special classes of graphs
  - Planar graphs
  - FEM (finite element method) graphs
- Partitioning arbitrary graphs
  - Spectral Methods
  - More advanced methods
  - Heuristics



# Some basic results

- NP-hard even for  $k=2$ .
  - Reduction from MAX-CUT.
- Polytime algorithm known for trees
  - Moderately hard exercise. Dynamic programming
- Other graph classes: next
- Approximation algorithm:  $O(\log^{1.5} n)$  factor known.
  - intro to some of the ideas: later

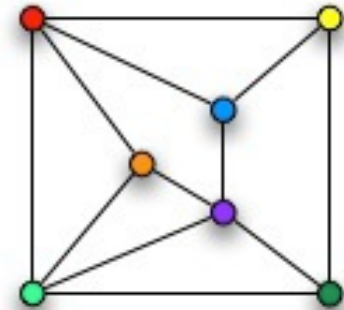
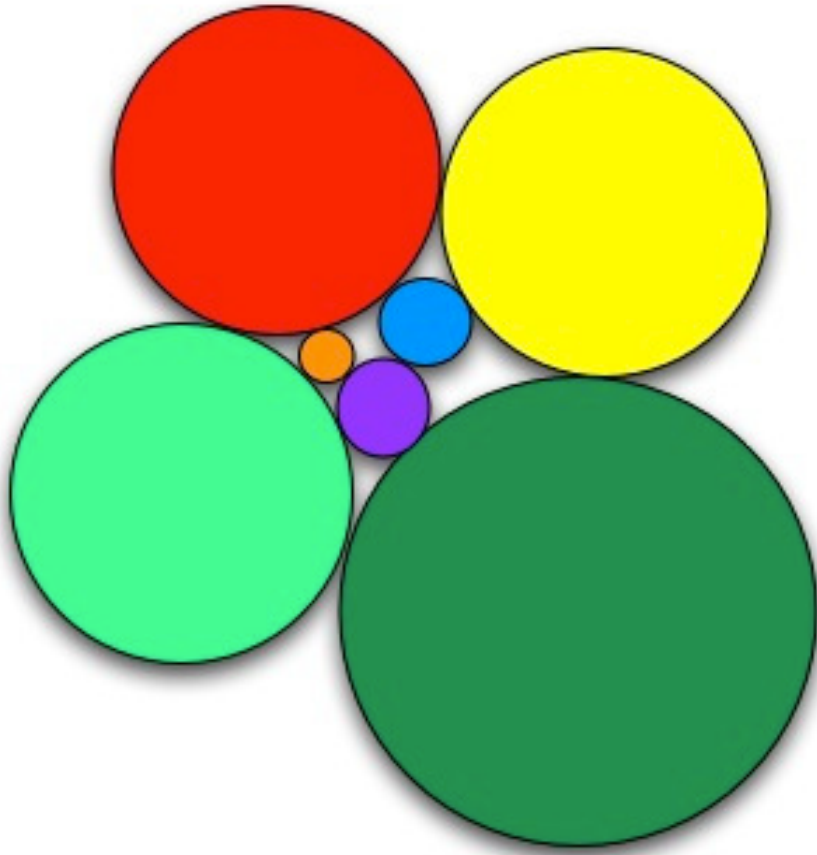
# Planar graphs

- NP-complete? Not known.
- Lipton-Tarjan Planar Separator Algorithm:
  - Any planar graph with  $n$  vertices can be bisected by removing  $O(\sqrt{n})$  vertices and incident edges.
  - $O(\sqrt{n})$  edges for bounded degree graphs.
  - Algorithm may give  $O(\sqrt{n})$  separator even if smaller exists.
  - Optimal for “Natural planar graphs” which cannot be bisected without removing  $\Omega(\sqrt{n})$  vertices.

# Planar separator theorem

- Every planar graph can be bisected by removing  $O(\sqrt{n})$  vertices.
- Lipton-Tarjan proof:
  - uses clever breadth first search.
- Today: Alternate proof
  - Elegant connection to geometry. Extends to other geometrically embedded graphs, e.g. FEM graphs.

# Representing planar graphs by “kissing” disks

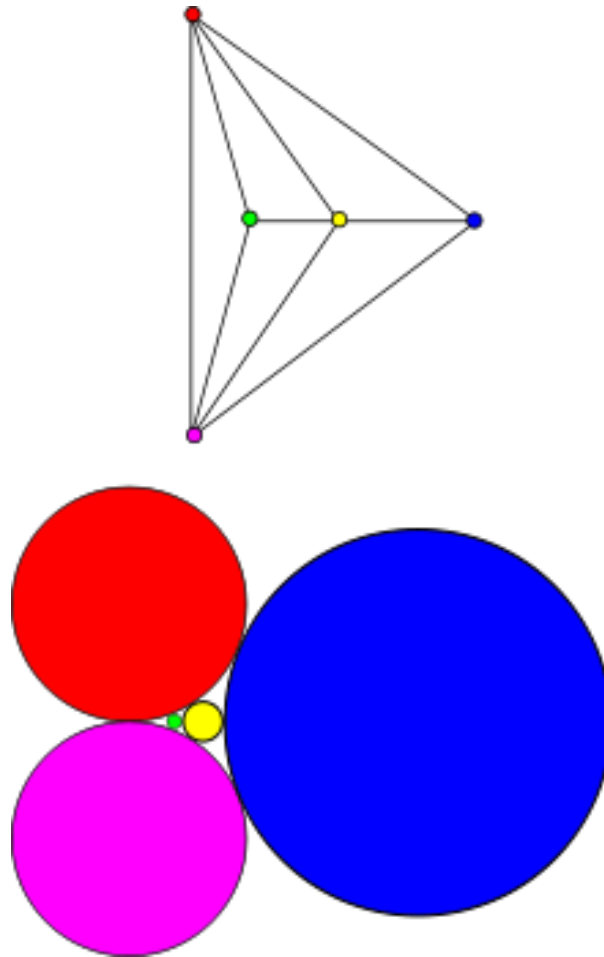


# Circle packing theorem

Theorem [Koebe 1936] For every planar graph with  $n$  vertices, there exist disjoint circles  $c_1, c_2, \dots, c_n$  such that  $c_i$  and  $c_j$  are tangential iff  $v_i, v_j$  are connected by an edge.

...Kissing disk embedding

# Circle packing example



(Wikipedia)

# Algorithm to find a Planar Separator

1. Get a kissing disk embedding.
2. Stereographically project it onto a sphere.
3. Find “center point” of disk centers.
4. Adjust so that center point moves to sphere center
5. Separator = random great circle through center.

Theorem:

Each hemisphere has  $\theta(n)$  vertices.

Expected number of edges cut =  $O(\sqrt{n})$ .

## Remark

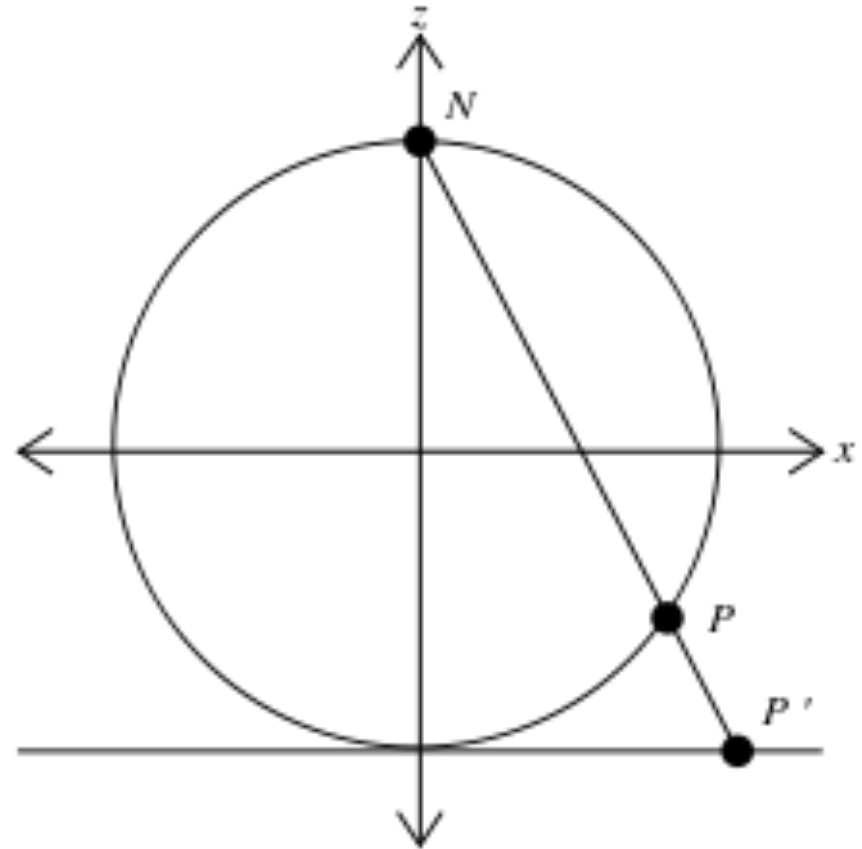
- Algorithm may not produce exact bisection, but say,  $\frac{1}{4}$ ,  $\frac{3}{4}$  split of vertices.
- Repeat on larger partition if balance not achieved.  $\frac{3}{4}$  splits into  $\frac{9}{16}$ ,  $\frac{3}{16}$
- Combine  $\frac{1}{4} + \frac{3}{16} = \frac{7}{16}$ .
- Now we have better balance:  $\frac{7}{16}$ ,  $\frac{9}{16}$ .
- **Repeating log times suffices.** Number of edges cut increase only by constant factor.



# Stereographic projection

Input: Disk embedding on horizontal plane.

- Place a sphere on the plane.
- Point  $P'$  on plane mapped to  $P$  on sphere such that  $PP'$  passes through North Pole.
- Circles map to circles!



(Wikipedia)

# Algorithm to find a Planar Separator

1. Get a kissing disk embedding. ✓
2. Stereographically project it onto a sphere. ✓
3. Find “center point” of disk centers.
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# Centerpoint C of n points

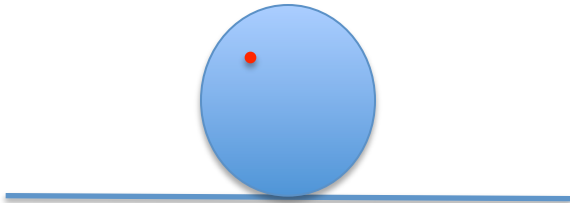
C = point such that any plane through it has at least  $n/4$  points on both sides.

- Centerpoints exist, and may not be unique.
- Approximate centerpoints are easily found.  
Will guarantee at least  $n/Q$  points on both sides, where  $Q$  maybe  $> 4$ .

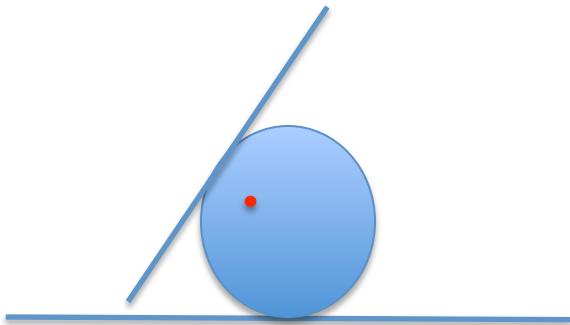
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# Adjusting the centerpoint



- Centerpoint shown in red.
- Project back onto inclined plane.
- Use a bigger sphere having same point of contact.
- How does red point move?

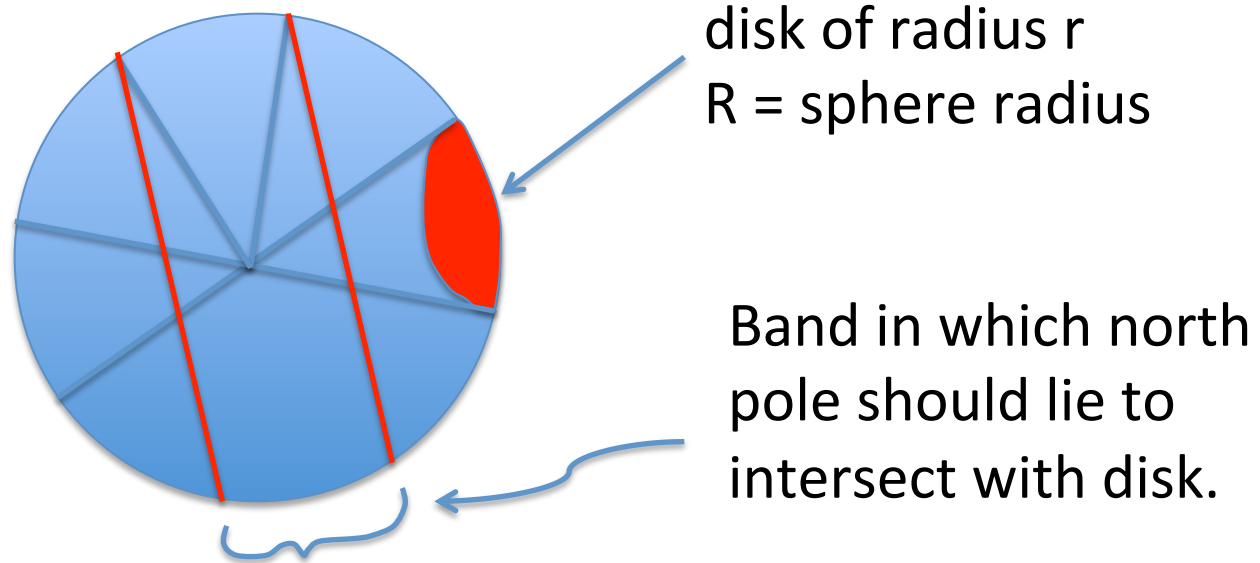


# Where are we?

- We have a sphere with a disk on it for each vertex.
- Disks touch if corresponding vertices have an edge.
- Center of sphere = centerpoint of vertices.
- What will happen if we partition using some great circle?
- Centerpoint:  $\theta(n)$  vertices will be on each side
- Will too many disks intersect with great circle?

# How many disks intersect with a random great circle?

- Pick a random great circle = pick a random point as its north pole.
- Probability = Band area/sphere area  
 $= 2\pi rR / 4\pi R^2 = r/2R$



# Expected number of intersections

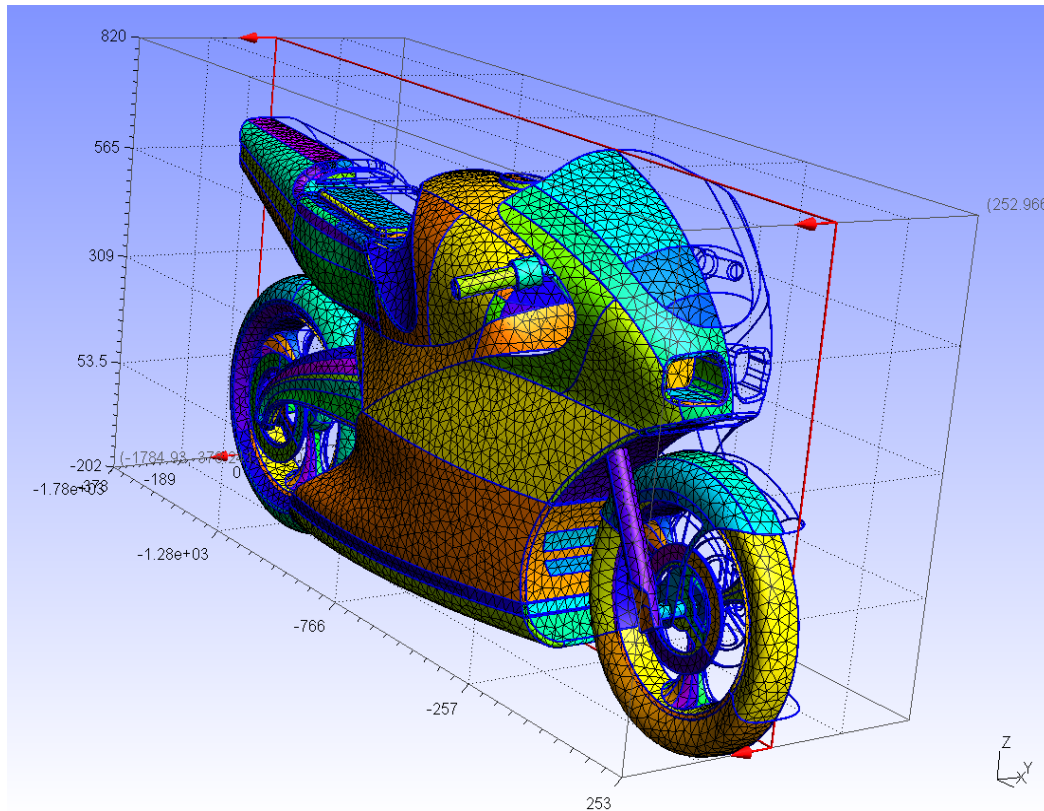
- $E = r_1/2R + r_2/2R + \dots r_n/2R$
- We know:  $\pi r_1^2 + \pi r_2^2 + \dots + \pi r_n^2 \leq 4\pi R^2$
- $E$  is maximized when all  $r_i$  are equal,  $r_i \leq 2R/\sqrt{n}$
- $E \leq n * (2R/\sqrt{n})/2R = \sqrt{n}$
- Plane has at least  $n/4$  vertices on each side
- By removing incident edges on  $\sqrt{n}$  vertices we get  $1/4 - 3/4$  partition of vertices.
- Clearly such a partition exists! **QED!**



# Remarks

- Nice connection between graph theory and geometry.
- Great circle when projected back, becomes a circle on the graph. “Circle separator”
- Do we need circles? Can we hope to get good partitions using a straight line?
  - There exist graphs for which this is not possible.
- In practice, Lipton-Tarjan’s original algorithm will be faster than above algorithm.

# Finite Element Method (FEM) Graphs



<http://www.geuz.org/gmsh/gallery/bike.png>

# FEM Graphs

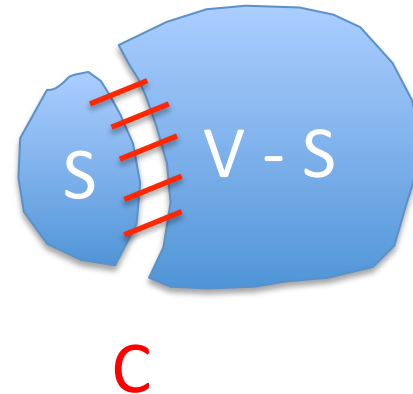
- 3 dimensional object represented by collection of volumes, “elements”
- Element size:
  - Small where physical properties vary a lot.
  - Large where physical properties vary less.
- “Aspect ratio” of elements is small, i.e. elements are not flat/narrow.
  - Useful for to reduce floating point overflow..

# FEM Graphs continued

- Effectively we have a “kissing disk” embedding of elements in 3 dimensional space.
- Theorem[Miller-Thurston 1990] A 3d FEM mesh with  $n$  elements of small aspect ratio can be bisected by removing  $O(n^{2/3})$  edges.
- Partitioning FEM graphs is useful for load balancing on parallel computers.

# Partitioning General Graphs

- Don't insist on bisection.
- Sparsity =  $|C|/|S|$
- Cut ratio =  $|C|/(|S| + |V-S|)$
- Sparsity/C.Ratio  $\approx |V|$
- We study algorithms that attempt to find cuts of minimum ratio.
- We can get bisection by repeating if necessary.



# Strategy

Our optimization problem is expressed in the language of graphs.

1. Pose it as a problem on integer variables.
  - NP-complete.
2. Find a similar problem on real variables that can be solved in polynomial time.
3. Exploiting the symmetry, get back an integer solution. **Hope: it is close to the optimal.**

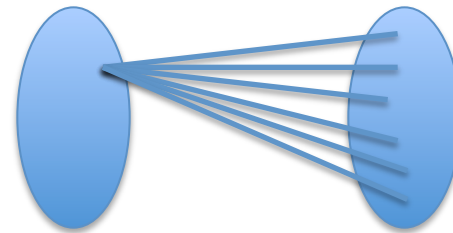
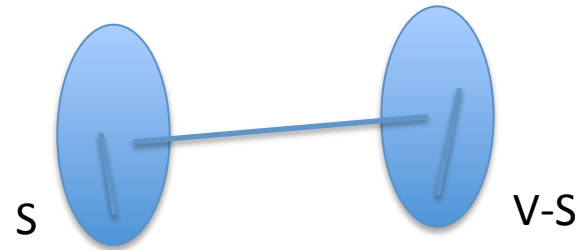
# Algebraic definition of cut ratio

- $x$  = bit vector.  $x_i = 1$  iff  $i$ th vertex is in  $S$ .

- No. of crossing edges:

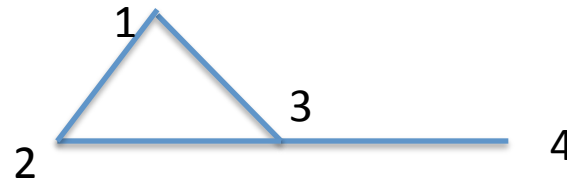
$$\sum_{(i,j) \in E} |x_i - x_j|$$

- $|S| |V-S| = \sum_{i < j} |x_i - x_j|$



- Want to minimize

$$\frac{\sum_{(i,j) \in E} |x_i - x_j|}{\sum_{i < j} |x_i - x_j|}$$



$$\min_{x_i \in \{0,1\}} \frac{|x_1 - x_2| + |x_1 - x_3| + |x_2 - x_3| + |x_3 - x_4|}{|x_1 - x_2| + |x_1 - x_3| + |x_1 - x_4| + |x_2 - x_3| + |x_2 - x_4| + |x_3 - x_4|}$$

# General Algorithmic Strategy:

## Relaxation

- Problem 1: minimize  $3x^2 - 5x$ ,  $x$  is an integer.
- Problem 2: minimize  $3x^2 - 5x$ ,  $x$  is real.
  - Easier to solve, make derivative = 0 etc.
  - $\min(\text{Problem 1}) \geq \min(\text{Problem 2})$
- Problem 2 = Relaxation of problem 1
- Strategy to solve problem 1: solve problem 2 and then take a nearby solution.
- Minimizing cut ratio: similar idea..



# Relaxation to “solve” for cut ratio

- Want to find  $r = \min_{x_i \in \{0,1\}} \frac{\sum_{(i,j) \in E} |x_i - x_j|}{\sum_{i < j} |x_i - x_j|} = \min_{x_i \in \{0,1\}} \frac{\sum_{(i,j) \in E} (x_i - x_j)^2}{\sum_{i < j} (x_i - x_j)^2}$

- Allow real values
  - Suppose we can solve for  $x$ .
  - We will “round” solution
  - high = 1. Low = 0.

$$r \geq \min_{x_i \in R} \frac{\sum_{(i,j) \in E} (x_i - x_j)^2}{\sum_{i < j} (x_i - x_j)^2}$$

- Insist that  $x_i$  add to 1

$$r \geq \min_{x \cdot 1 = 0} \frac{\sum_{(i,j) \in E} (x_i - x_j)^2}{\sum_{i < j} (x_i - x_j)^2}$$

# contd.

$$\text{Numerator} = \sum_{(i,j) \in E} (x_i - x_j)^2 = x^T L x$$

$L_{ij} = -1$  if  $(i,j)$  is an edge,  
= degree(i) if  $i=j$ ,  
=0 otherwise

$L$  = "Laplacian of the graph"

$$\text{Denominator} = \sum_{i < j} (x_i - x_j)^2 = \sum_{i < j} x_i^2 - 2x_i x_j + x_j^2 = (n-1) \sum_i x_i^2 - 2 \sum_{i > j} x_i x_j = n \sum_i x_i^2$$

$$r \geq \min_{x \cdot 1 = 0} \frac{\sum_{(i,j) \in E} (x_i - x_j)^2}{\sum_{i < j} (x_i - x_j)^2} = \min_{x \cdot 1 = 0} \frac{x^T L x}{n \sum_i x_i^2} = \frac{\sigma_2}{n}$$

$\sigma_2$  is second smallest "Eigenvalue of  $L$ "

Can be calculated quickly.

minimizing  $x$  can also be calculated.

# Algorithm for finding good ratio cut

1. Construct Laplacian matrix
2. Find second smallest eigenvector
3. Choose threshold  $t$ .
4. Partition =  $\{i \mid x_i \leq t\}, \{i \mid x_i > t\}$

How to choose  $t$ ? Try all possible choices, and take one having best ratio.

“Spectral Method” : eigenvectors involved.

# How good is the partition

- Cheeger's Theorem:

- Partition found by previous algorithm will have cut ratio at most  $\frac{\sqrt{8d\sigma_2}}{n}$

Difficult proof

- Optimal partition will have cut ratio at least  $\frac{\sigma_2}{n}$

Just proved!

- Seems to work well in practice

# Examples

- Cycle on  $n$  vertices: Spectral method finds optimal partition.
- Bounded degree planar graphs: Spectral method finds  $O(\sqrt{n})$  sized bisector.
- Binary hypercube: Spectral methods will find cut of ratio  $\sqrt{4 \log n} / n$ , whereas opt ratio =  $1/n$

# Remarks about intuition

- How did we steer the analysis towards the Laplacian?
  - Experience
  - Laplacian arises in solving differential equation, where second eigenvalue gives how graph may “vibrate”. Vibration mode points to cuts.
  - Laplacian is related to SVD, singular value decomposition. SVD of a point cloud identifies length, breadth, width of cloud. Small cuts are perpendicular to length.

# Advanced methods

- Starting point: 
$$r = \min_{x_i \in \{0,1\}} \frac{\sum_{(i,j) \in E} |x_i - x_j|}{\sum_{i < j} |x_i - x_j|}$$
- Different ways to relax.
  - If you relax too much, you may be able to solve the new problem more easily, but the solution will be too far.
- Arora-Rao-Vazirani 04 : stricter relaxation.  
Always give cut with ratio =  $O(\sqrt{\log n})$  times optimal.

# Heuristics

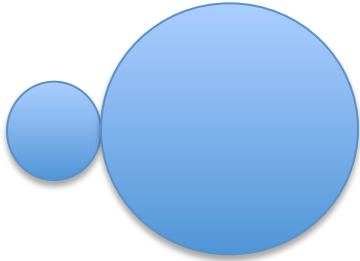
- Graph coarsening [Metis 04]:
  - Pick disjoint edges in graph.
  - Merge endpoints of each edge into a single vertex. (fast)
  - Repeat several times.
  - Find partition for final “coarse” graph using sophisticated method. (slow, but on small graph)
  - Final partition induces partition on original graph.



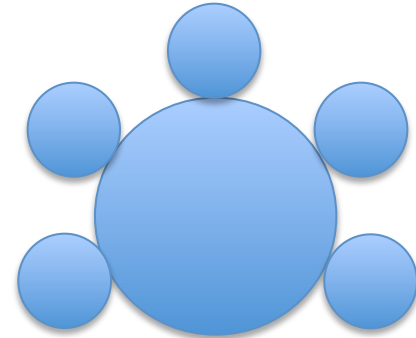
# Concluding Remarks

- Important problem
- Many applications
- Many interesting methods. Theory + Heuristic.
- Deep connections to geometry.

# Generalization of bisection



- We may need to cut many edges to bisect, but by cutting very few we may get  $\frac{1}{4}$  -  $\frac{3}{4}$  split.
- May be useful to know such “bottlenecks”



- Even for bisection, it might be easier to get one piece at a time

# c-balanced partitioning

- For a fixed  $c$ , remove as few edges as possible to get subgraphs of size  $cn$ ,  $(1-c)n$ .
- If  $c=O(1)$ , we can use repeatedly to achieve bisection. Example:  $c = 0.1$ 
  - After 1 application we have 0.1 – 0.9 split.
  - Apply once more on 0.9 : 0.1, 0.09, 0.81
  - Apply log times on larger piece and extract  $n/2$  vertices. Bisection!