

# COT 6936: Topics in Algorithms

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[http://www.cs.fiu.edu/~giri/teach/COT6936\\_S10.html](http://www.cs.fiu.edu/~giri/teach/COT6936_S10.html)

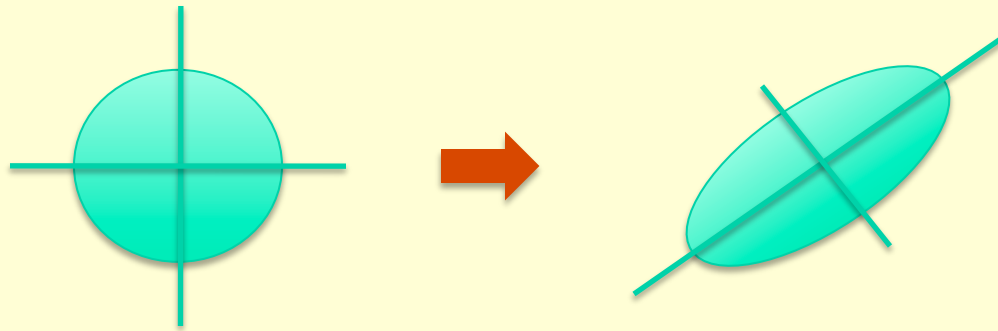
<https://online.cis.fiu.edu/portal/course/view.php?id=427>

# Spectral Methods

- Graph Connectivity problems
  - Google Page Rank
- Graph Partitioning problems
  - Clustering (even linearly non-separable case)
- Markov Chain Mixing problems
  - Random walks in graphs

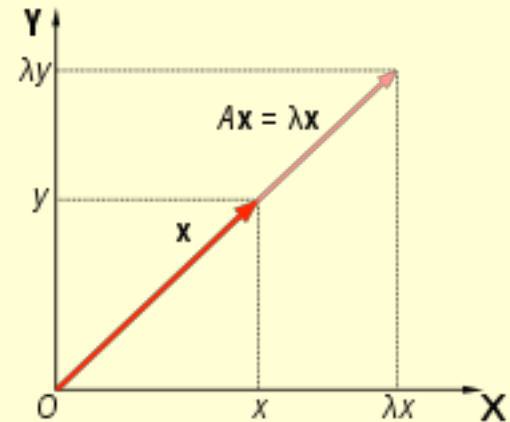
# Matrices and Eigenvalues

- Array of values
- Linear Transformation



- Eigenvalues and Eigenvectors

- $Ax = \lambda x$
- Under transformation  $A$ , eigenvectors only experience change in magnitude, not direction
- $A = Q \Lambda Q^{-1}$



# Graph Bisection

- Construct adjacency matrix  $A$
- Construct Laplacian  $L = D - A$ 
  - $D$  = diagonal matrix with degrees along diagonal
- $L$  is positive semi-definite (PSD); has non-neg eigenvalues; has smallest eigenvalue = 0
- Second eigenvector provides information about bisection.
  - Signs of 2<sup>nd</sup> eigenvector give a good bisection
  - Extreme case: Connected components have constant values in 2<sup>nd</sup> eigenvector

# Graph Bisection (Continued)

- Eigenvalues indicate strength of bisection
- How to get bisections with  $n/2$  vertices?
  - Use median value in second eigenvector
- How to get  $k$  partitions?
  - Perform bisections recursively
  - Use more eigenvectors

# Spectral Clustering: Strategy

- Given data points and a distance function, construct a weighted graph
- Let  $A$  be its adjacency matrix; let  $D$  be diagonal matrix with degrees along diagonal
- Construct Laplacian  $L$  (PSD, non-neg eigenv.)
  - Unnormalized:  $L = D - A$
  - Normalized symmetric:  $L = D^{-1/2}LD^{1/2}$
  - Random Walk:  $L = D^{-1}L$
- Matrix  $L_k$  has cols = first  $k$  eigenvectors of  $L$
- Cluster rows of  $L_k$

# Spectral Clustering

- Need distance measure (need not be a metric), i.e., triangle inequality not needed
- Not Model-based
- Global method
- Turns discrete problem into continuous

# Randomized Algorithm for MAX 3-SAT

- Assume each clause has 3 distinct literals
- Randomly assign 0/1 to all variables
  - Each clause is satisfied with prob  $7/8$
  - Expected number of clauses =  $7m/8$
  - There exists a truth assignment that satisfies  $7m/8$  clauses
- **Problem:**
  - How can we find a satisfying truth assignment with at least  $7m/8$  clauses satisfied?



# Derandomization

- Consider randomized algorithm from slide 8
- $E[S_\varphi] = \frac{1}{2} E[S_\varphi \mid x_1 = T] + \frac{1}{2} E[S_\varphi \mid x_1 = F]$
- $E[S_\varphi \mid x_1 = T], E[S_\varphi \mid x_1 = F]$  can be computed in polynomial time. **WHY?**
- If  $(E[S_\varphi \mid x_1 = T] \geq E[S_\varphi \mid x_1 = F])$ , then  
$$E[S_\varphi \mid x_1 = T] \geq E[S_\varphi] \geq 7m/8$$
- Set  $x_1 = T$ , and reduce  $\varphi$  to  $\varphi'$ .
- Find value for  $x_2$  and so on.

# How to compute the expected values

- $E[X] = \sum_i P(C_i = T)$
- For example, let
  - $C_i = (x_1 \vee \neg x_2 \vee x_3)$
- $P(C_i = T | x_1 = T) = 1$
- $P(C_i = T | x_1 = F) = 1 - P(C_i = F | x_1 = F) = \frac{3}{4}$