# Genetic Algorithm for Graph Layouts

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#### Idea

- Given a graph (directed or undirected), try to plot it on a 2D plane such that the number of edge crossings are minimized
- **Search Space**: vector of coordinates
  - Example:
    - $\blacksquare$  say we have a graph that looks like this  $o^A$ ----- $o^B$ ----- $o^C$
    - and we have a 100x100 pixel canvas to draw this graph
    - then there are  $(100^2)^3$  or a trillion different ways to plot this graph (points could end up on top of each other)
- Objective Function:

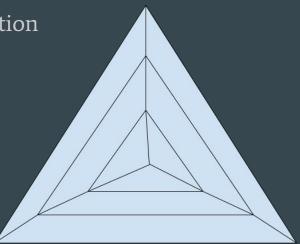
$$f(x) = {|E| \choose 2} - \sum_{p \in Pairs} {1 \text{ if p has intersection} \choose 0 \text{ if p has no intersection}}$$

# Variation Operators

- Representation:  $<(x_1, y_1), (x_2, y_2), .... (x_n, y_n) >$
- Used Identical methods presented in lecture 1 slides
  - **Recombination**: single point crossover
  - **Mutation**: randomly select (x,y) coordinate for random point if individual is to be mutated
  - **Selection**: fitness proportional selection

#### **Trial Run**

- Population: 300
- Generations: 1500
- Target Fitness: at most 1 edge pair with intersection
- Graph: 10 nodes, 18 edges
- Canvas Size: 800x800



### Results



# Room For Improvement

- evaluate fitness based on different criteria, for example:
  - o spacing between nodes or edges?
  - o for each node, how many of its own edges intersect with other edges (if none then I shouldn't move that node)?
- scale fitness better, ie, 99% is much better than 96%
- combine high fitness individuals
- If a node is in an "already near optimal position", maybe mutation should have a smaller effect
  - currently a good node may get pushed all the way across the canvas, and if it has a high degree,
    there is a higher chance that now its edges will intersect with other edges

# **Example Graph Layouts**