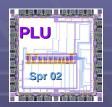


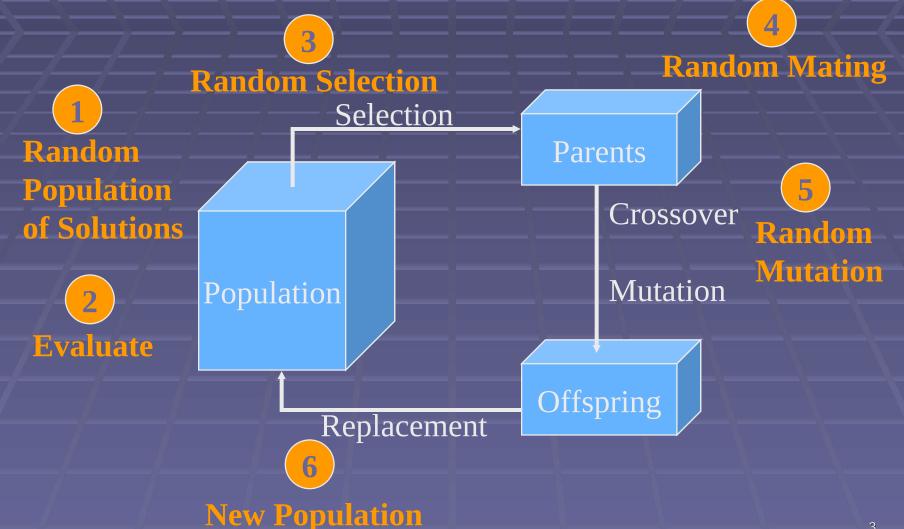
Definition

- Goldberg, 1989: "Genetic Algorithms are search algorithms based on the mechanics of natural selection and natural genetics."
- A genetic algorithms is a directed random search procedure

A genetic algorithm borrows ideas from biology to search a solution space for a target value.

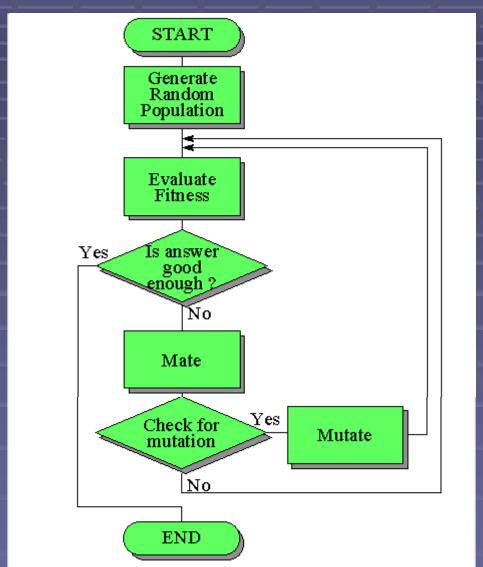


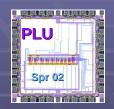
General Approach





GA Flow Chart



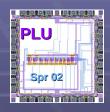


Parent Selection

- The process in which individual strings in the population are selected to contribute to the next generation is called parent selection
 - based on fitness
 - strings with a high fitness have a higher probability of contributing one or more offspring to the next generation
- Biased Roulette Wheel Selection

When you spin the wheel, items 1 and 5 have the greatest chance of coming up while item 2 has the smallest





Example

Given the following population of chromosomes, select two parents:

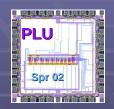
Chromosome	Fitness	% fitness	
(1 0 1 0 0 1)	23	0.28	
(1 1 1 0 0 1)	12	0.15	
(0 1 1 0 1 1)	25	0.30	
(0 1 0 1 1 0)	5		
(0 1 1 0 1 0)	17		
Total Fitness	82		

Find the total fitness of the population

Find the %fitness of each element

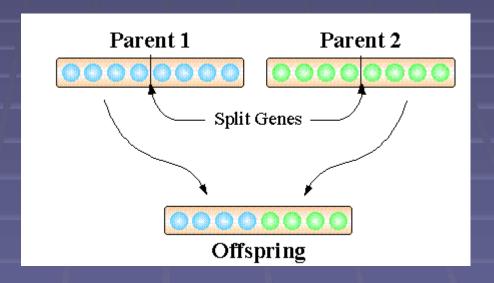
Find the cumulative fitness

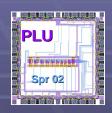
Now, throw a random number between 0 and 1, if it is in the range 0 to 0.28 select element 1, between 0.28 and 0.43 select element 2, . . .



Crossover

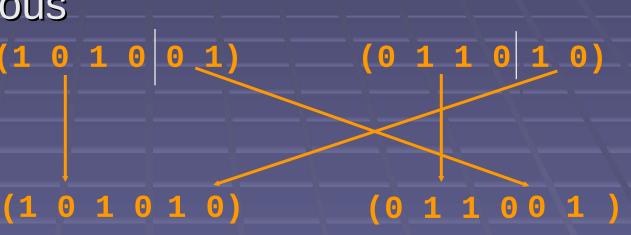
 Once two parents are selected, their chromosomes are mixed to create the children for the next generation





Example

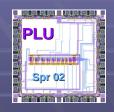
Assume the parents selected from the previous example are: (1 0 1 0 0 1)



These are the two children which are now part of the next generation

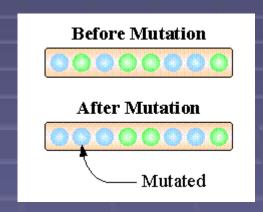
Find a random crossover point

Swap the bits after the crossover point



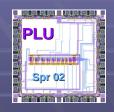
Mutation

A bit in a child is changed (from 1 to 0 or from 0 to 1) at random



This is a small probability event

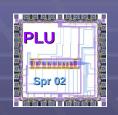
The effect is to prevent a premature convergence to a local minimum or maximum



GA Performance

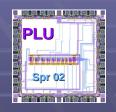
- Increasing diversity by genetic operators
 - mutation
 - Recombination

- Decreasing diversity by selection
 - of parents
 - of survivors



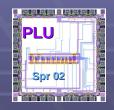
Effects of the Genetic Operators

- Using selection alone will tend to fill the population with copies of the best individual from the initial population
- Using selection and crossover will tend to cause the algorithm to converge on a good but suboptimal solution
- Using mutation alone induces a random walk through the search space
- Using selection and mutation creates a parallel, noise-tolerant, hill climbing algorithm



The Algorithm

- randomly initialize population(t)
- determine fitness of population(t)
- repeat
 - select parents from population(t)
 - perform crossover on parents creating population(t+1)
 - perform mutation on population(t+1)
 - determine fitness of population(t+1)
- until best individual is good enough



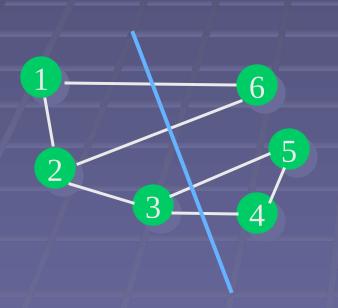
GA Applications

- GA's can be applied to several parts of the physical design problem
 - Partitioning
 - Placement
 - Other . . .
- Scope of the Partitioning problem
 - A standard layout benchmark suite has circuits ranging from 13,000 to 200,000 nodes.
 - The number of links range from 50,000 to 800,000



Representation

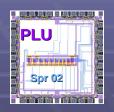
A graph partition is represented by a binary string



Each node is represented by a bit

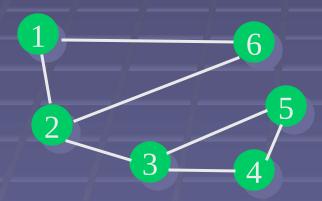
The 0 nodes are in one segment, the 1 nodes are in the other segment

 $(0\ 0\ 0\ 1\ 1\ 1)$



Population

A random population of binary strings is produced



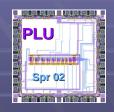
The fitness is the number of links

$$(0\ 0\ 0\ 1\ 1\ 1)$$
 4

$$(1\ 0\ 0\ 1\ 0\ 1)$$

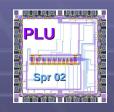
$$(1\ 0\ 1\ 1\ 0\ 0)$$
 5

$$(1\ 0\ 0\ 1\ 1\ 0)$$
 4



Crossover

- Parents are randomly selected (with a bias to the better fit elements)
- The parents are combined to create two children (single point crossover)



Mutation

- For a small number of the new population elements perform a mutation operation
 - Randomly select two nodes and swap their positions

