Lecture 2: Genetic Algorithms Introduction

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Biological Evolution Process

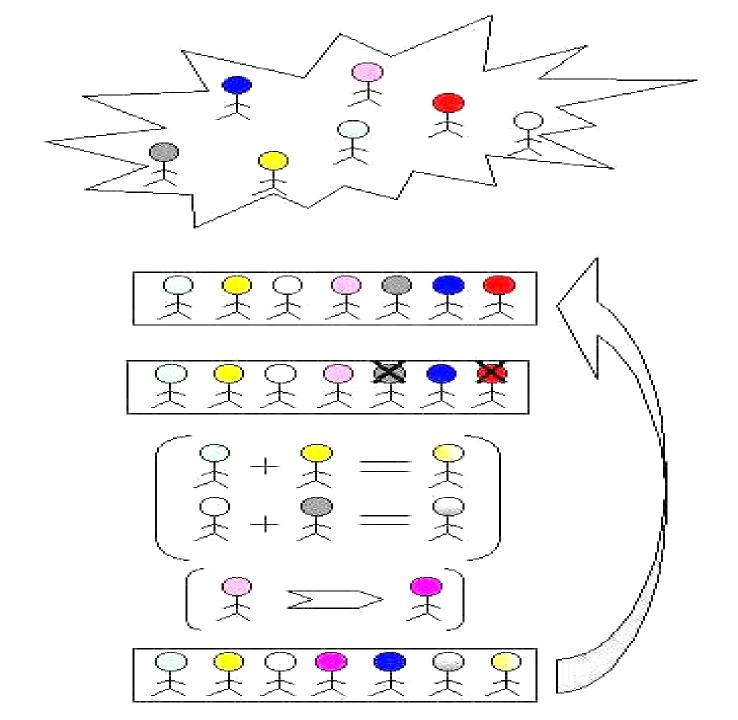
"Natural Selection" + "Genetic Inheritance"

Basic Idea of Principle of Natural Selection

"Select The Best, Discard The Rest"

Remember

- An individual is characterized by a set of parameters:
 Genes
- The genes are joined into a string: Chromosome
- The chromosome forms the genotype
- The genotype contains all information to construct an organism: the phenotype
- Reproduction is a "dumb" process on the chromosome of the genotype
- Fitness is measured in the real world of the phenotype

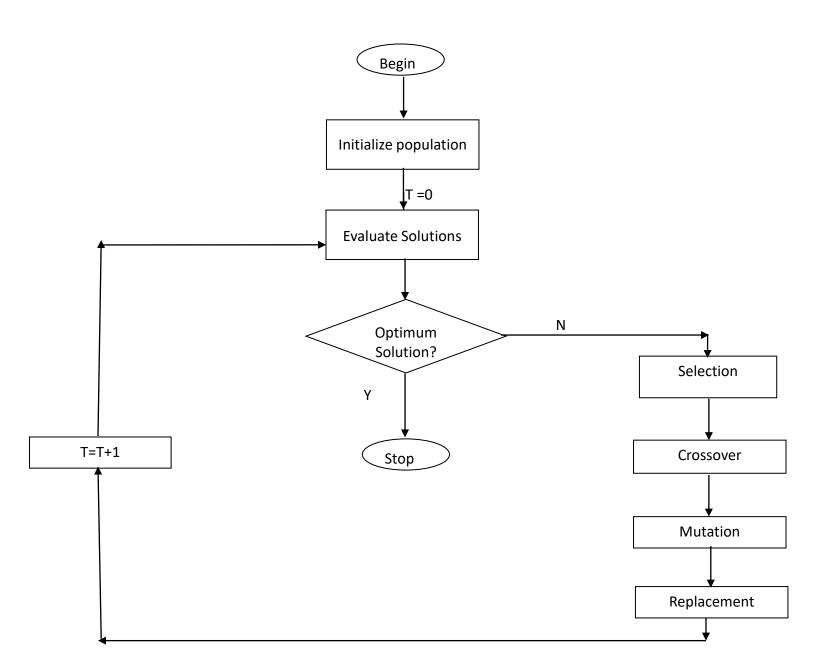


What Are Genetic Algorithms (GAs)?

 Genetic Algorithms are search and optimization techniques based on principle of Biological Evolution Process.

 Genetic Algorithms implement optimization strategies by simulating evolution of species through Natural Selection.

Working Mechanism Of GAs



Simple Genetic Algorithm

```
Simple Genetic Algorithm()
{
   Initialize the Population;
   Calculate Fitness Function; // Evaluate Individuals ...
   While (Fitness Value != Optimal Value) // Continue Evolution ...
   {
        Selection; //Natural Selection, Survival Of Fittest ...
        Crossover; //Reproduction, Propagate favorable characteristics ...
        Mutation; //Mutation ...
        Replacement
        Calculate Fitness Function;
```

Nature to Computer Mapping

Nature	Computer
Environment	Problem to be optimized.
Population	Set of candidate solutions.
Individual	One candidate solution to a problem.
Fitness/Survival Degree	Quality of a solution.
Chromosome	Encoding of a Solution to another format
	(For example, bit string, floating point array).
Gene	Part of the encoded solution.
Reproduction	Crossover
Copying errors	Mutation

Components of a GA

A problem to solve, and ...

- Encoding technique (gene, chromosome)
- Initialization procedure (creation)
- Evaluation function (environment)
- Selection (reproduction, environment)
- Genetic operators (crossover, mutation)
- Parameter settings (practice and art)

Many parameters to set

- Any GA implementation needs to decide on a number of parameters: Population size (N), mutation rate (m), crossover rate (c)
- Often these have to be "tuned" based on results obtained - no general theory to deduce good values
- Typical values might be: N = 50, m = 0.05, c = 0.9

Example 1: A Very Simple Example

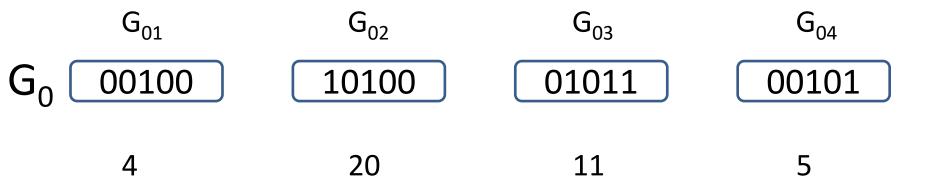
- Generate a bit string for the decimal number 31.
- Traditional Technique:
 - Start by 00000 till reaching 11111
- GA Steps:
 - 1. Build an initial population of random individuals (chromosomes or initial solutions)
 - 2. Evaluate fitness of individuals
 - 3. Select some individuals to
 - Apply Crossover and Mutation
 - 4. Build next generation
 - 5. Go to step 2.

Build an initial population of <u>random</u> individuals



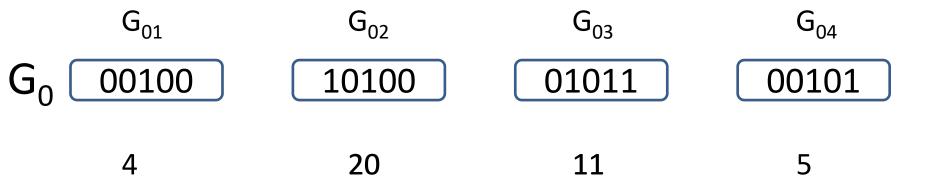
Generation 0

Evaluate fitness of individuals



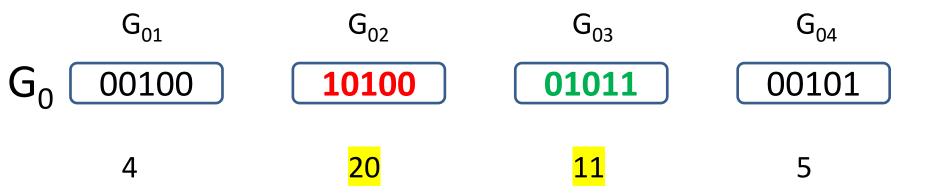
Fitness can be calculated as the decimal value of each bit string

Select some individuals to apply crossover and mutation



• Individual with good fitness should be selected for crossover and mutation.

Select some individuals to apply crossover and mutation



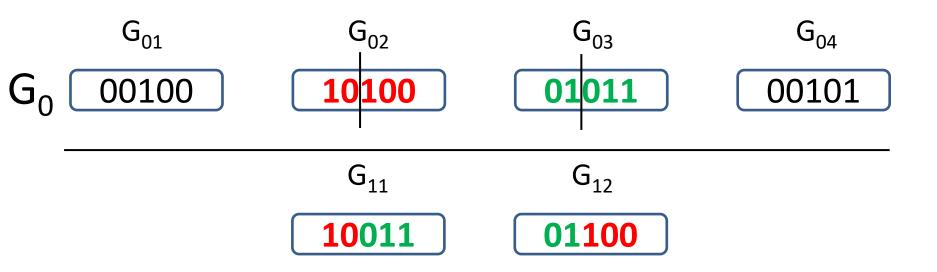
 Individual with good fitness should be selected for crossover and mutation.

Select some individuals to apply crossover and mutation



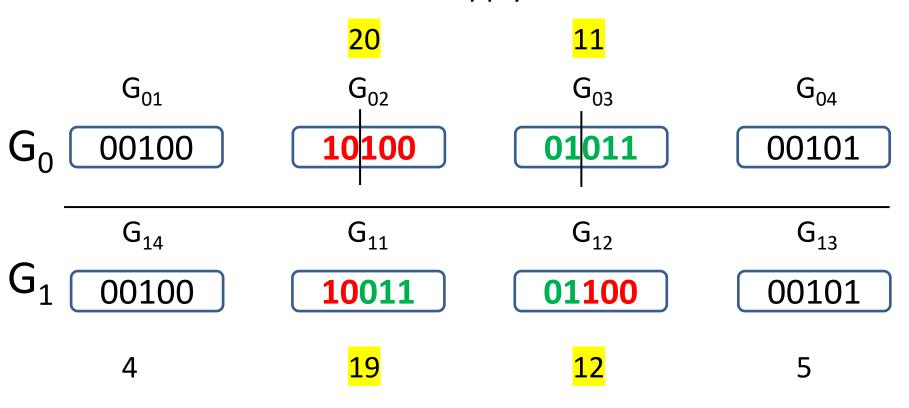
- How to apply crossover between these 2 parents?
 - Single point? Multiple point? Location?
 - It depends on your implementation parameters
 - Crossover may happen or not according to the probability of crossover

Select some individuals to apply crossover and mutation



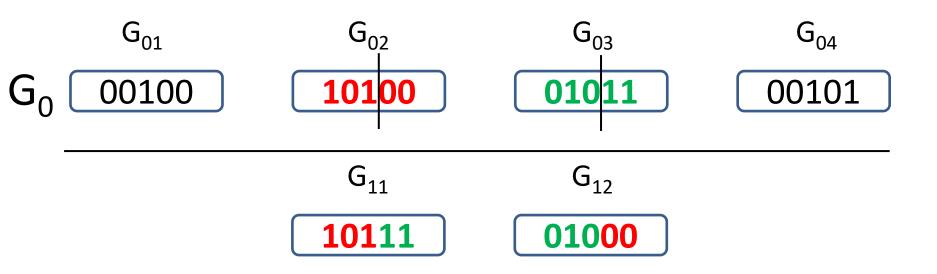
Single point crossover after 2 bits (Genes)

Select some individuals to apply crossover and mutation



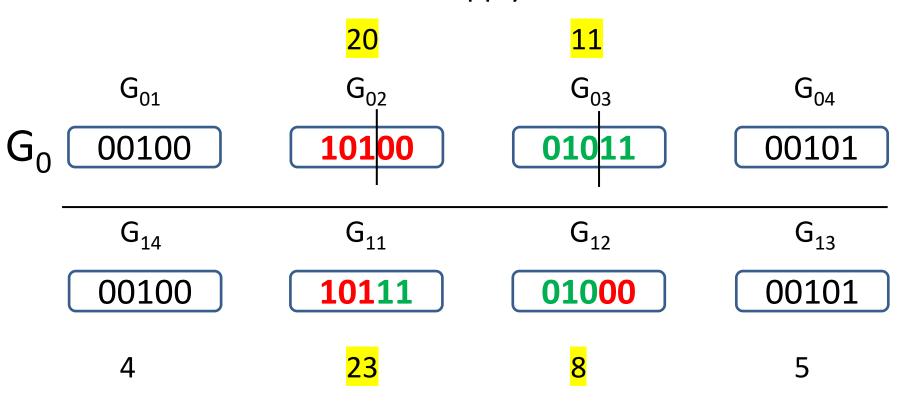
Fitness?

Select some individuals to apply crossover and mutation



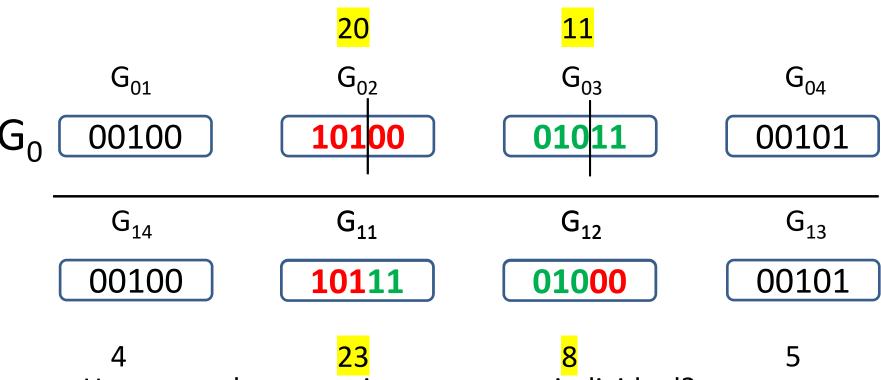
Single point crossover after 3 bits

Select some individuals to apply crossover and mutation



Fitness?

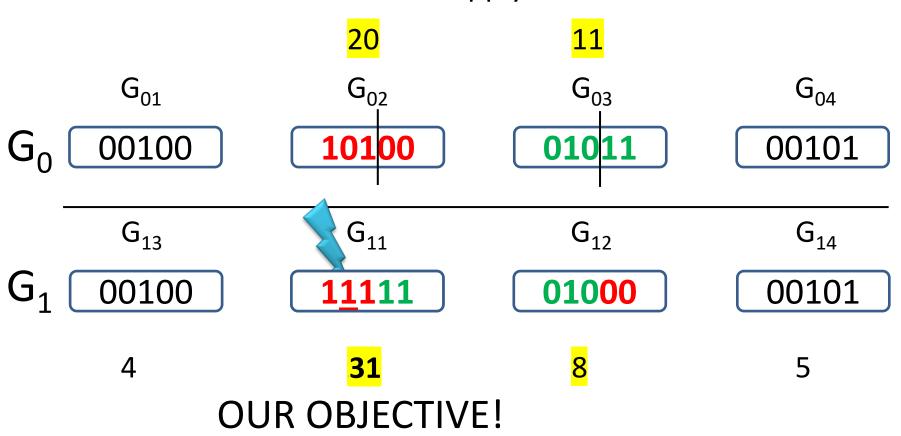
Select some individuals to apply crossover and mutation



- How to apply a mutation over some individual?
- For bit strings, and according to the mutation probability of your
 implementation, one bit might be flipped!

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Select some individuals to apply crossover and mutation

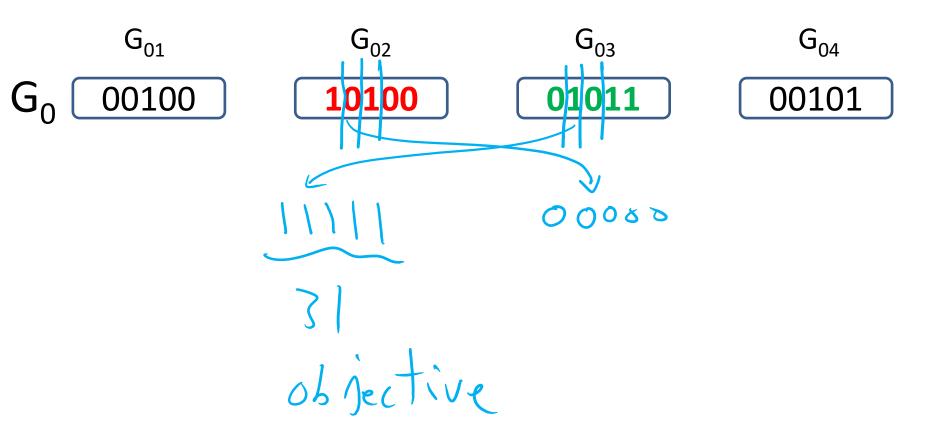


Select some individuals to apply crossover and mutation

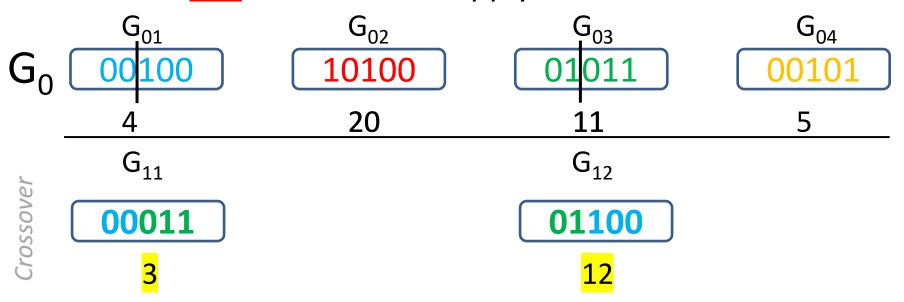


Applying multiple point crossover!

 G_{01} G_{02} G_{03} G_{04} G_{04} G_{05} G_{05} G

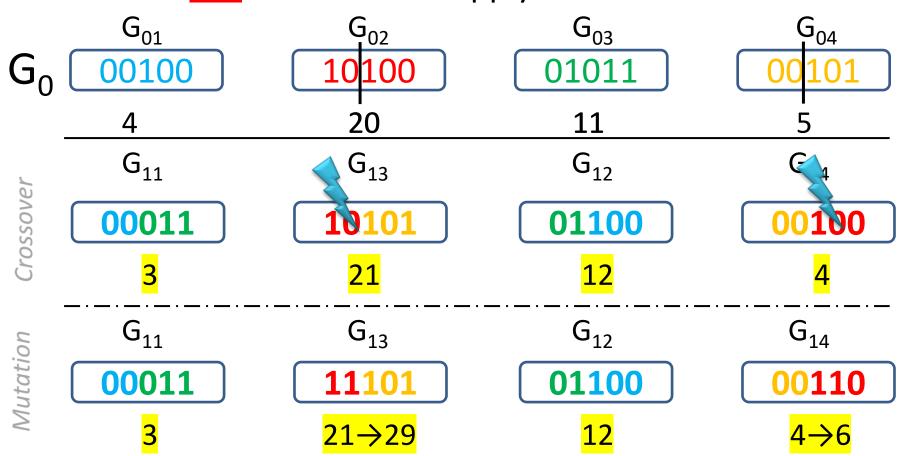


Select <u>ALL</u> individuals to apply crossover and mutation

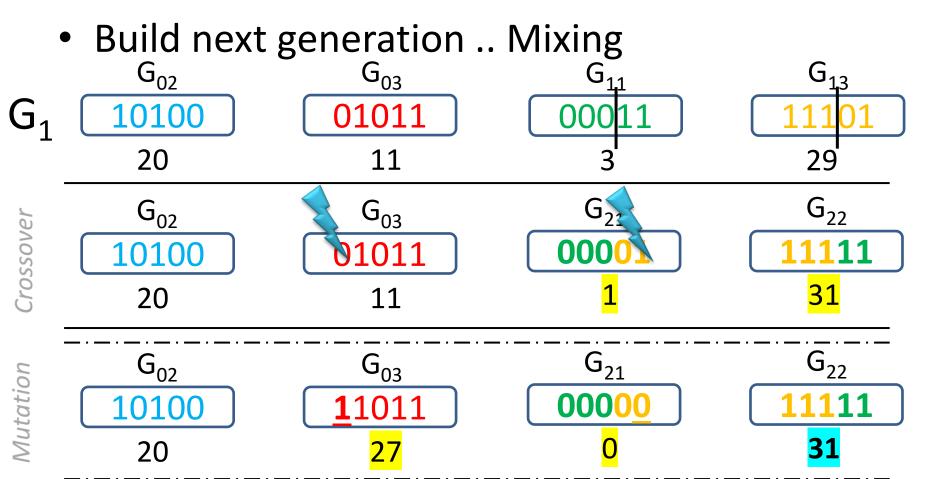


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Select <u>ALL</u> individuals to apply crossover and mutation



Build next generation .. Full Replacement or Mix? G₀₃ G_{02} G_{01} G_{04} 20 11 G_{14} G_{11} G_{13} G_{12} rossover 00011 **10101** 01100 00100 G_{13} G_{14} G_{11} G_{12} Mutation 01100 00011 29



OUR OBJECTIVE!

Does it really cost one iteration?

- Real world problems are not that easy.
- You could have:
 - Hundreds of individuals (chromosomes)
 - Huge search space
- You might perform thousands of iterations (have thousands of generations) before reaching your objective!

How do you encode a solution?

- Obviously this depends on the problem!
- GA's *often* encode solutions as fixed length "bitstrings" (e.g. 101110, 111111, 000101)
- Sometimes, your solutions (chromosomes)
 will be used in their original format (floating
 point numbers or alphabetical strings)

Why does crossover work?

- The idea is that crossover preserves "good bits" from different parents, combining them to produce better solutions
- The fitness may be improved or get worst after mutation
- A good scheme would therefore try to preserve "good bits" during crossover and mutation

Which problems can GAs solve?

- The solution can be encoded.
- Each gene represents some aspect of the proposed solution to the problem (the solution can be decomposed into smaller parts)
- The ability to "test" any solution and get a "score" indicating how "good" that solution is (Fitness).