



AUT

COMP815 Nature Inspired Computing

*GA for Other Types of
Optimization Problems*

Example of Practical Optimization Problem

A butter production company wants to optimize the use of the machineries in its daily production of butter. Two types of butter are made – sweet and raw. One kilogram of sweet butter gives the company a profit of \$10 and one of raw a profit of \$15. Two machines are used in the production: a pasteurization machine and a whipping machine. The daily use time of the pasteurization machine is 3.5 hours and 6 hours for the whipping machine. The processing times (in minutes) for 1kg of butter are given below:

Machine	Sweet butter	Raw butter
Pasteurization	3	3
Whipping	3	6

Let's first refer to the General Problem Format!

Mathematical Description

Objective function

Minimize

$f(x)$

Parameters

$x = [x_1, x_2, \dots, x_N]$

Subject to

$x \in M$

Feasible Set

Example of Practical Optimization Problem

One kilogram of sweet butter gives the company a profit of \$10 and one of raw a profit of \$15.

The daily use time of the pasteurization machine is 3.5 hours (210 min) and 6 hours (480 min) for the whipping machine.

The processing times (in minutes) for 1kg of butter are given below:

Machine	Sweet butter	Raw butter
Pasteurization	3	3
Whipping	3	6

Objective function	<i>maximum the profit from two butters</i>	$\$10 * \text{sweet(kgs)} + \$15 * \text{raw(kgs)}$
Parameters		
Feasible set		

Example of Practical Optimization Problem

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Objective function	maximum the profit from two butters	$f(x) = 10 * x_1 + 15 * x_2$
Parameters	Sweet(kgs), raw(kgs)	x_1, x_2
Feasible set	Obviously	$x_1 \geq 0, x_2 \geq 0$
	daily use time of two machines: pasteurization and whipping	P: $3 * x_1 + 3 * x_2 \leq 210$ W: $3 * x_1 + 6 * x_2 \leq 480$

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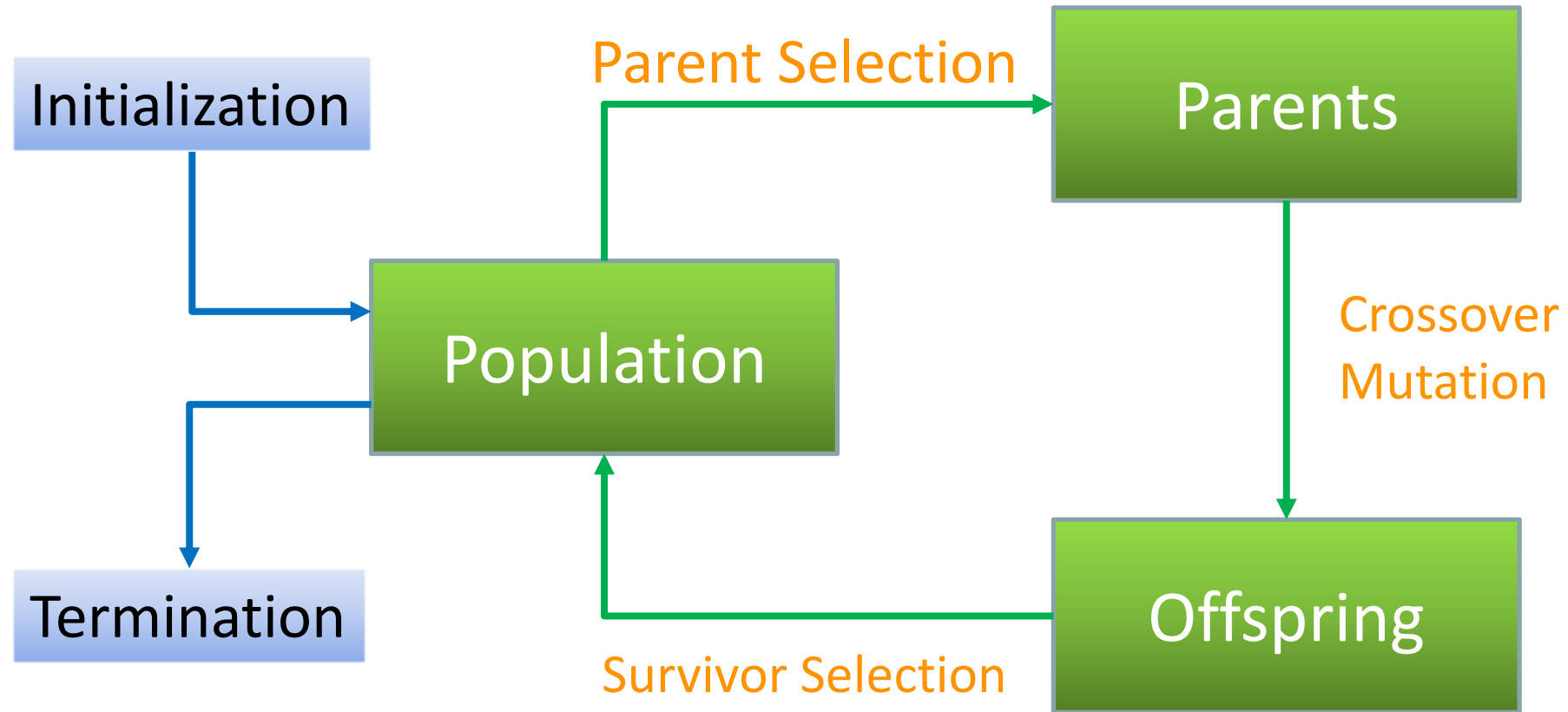
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Maximize $f(x) = 10 * x_1 + 15 * x_2$

Subject to $x_1 \geq 0, x_2 \geq 0,$

$3 * x_1 + 3 * x_2 \leq 210, \quad 3 * x_1 + 6 * x_2 \leq 480$

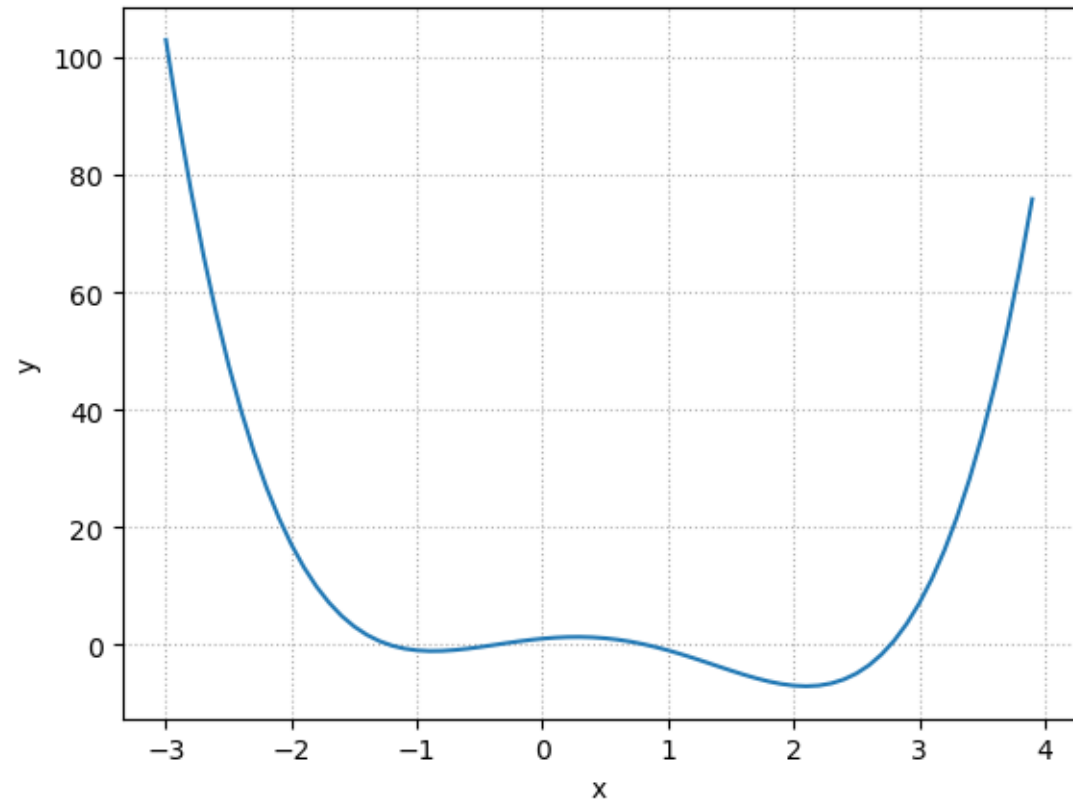
Last Lecturer Review



Last Lecturer Review

min $f(x) = x^4 - 2x^3 - 3x^2 + 2x + 1$

for $-3 \leq x \leq 4$



Chromosome Encoding

Each candidate solution encodes a value of $-3 \leq x \leq 4$

Each candidate solution is represented by a *chromosome*

n binary digits: $\{0,1\}^n$

Let $n = 10$

$n=10$ $a = -3$ $b = 4$

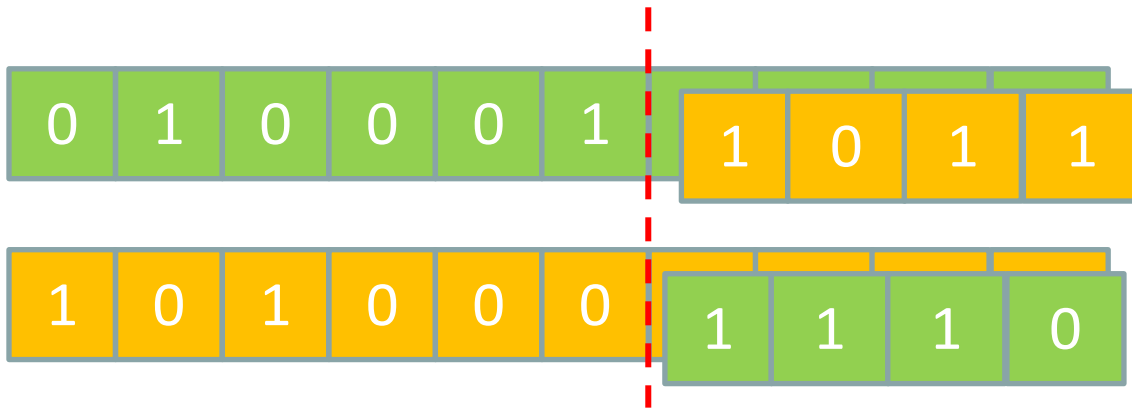
x	chromosome
-3	00 0000 0000
0	01 1011 0110
4	11 1111 1111



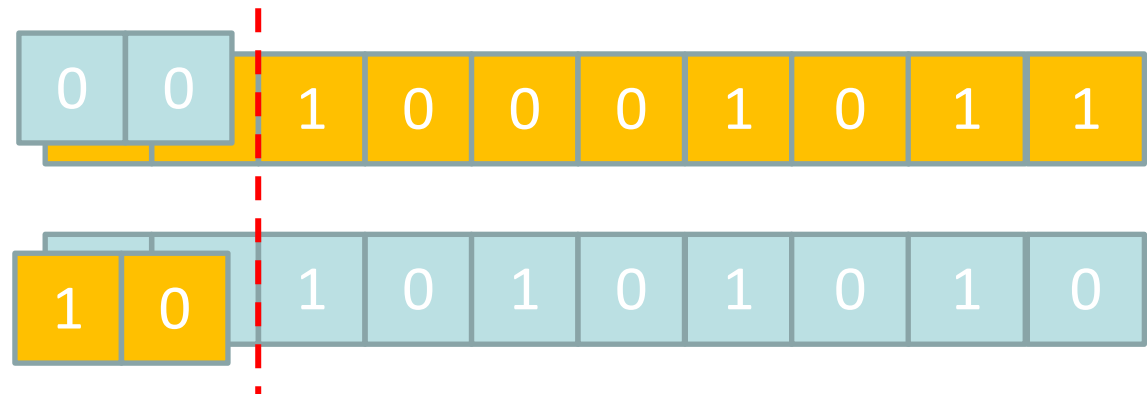
chromosome	x
00 0000 0000	-3
01 1011 0110	-0.003
11 1111 1111	4

Crossover

Parents chosen



Single-point crossover
Crossover point randomly selected



Single-bit Mutation

0 1 0 1 0 1 1 0 1 1

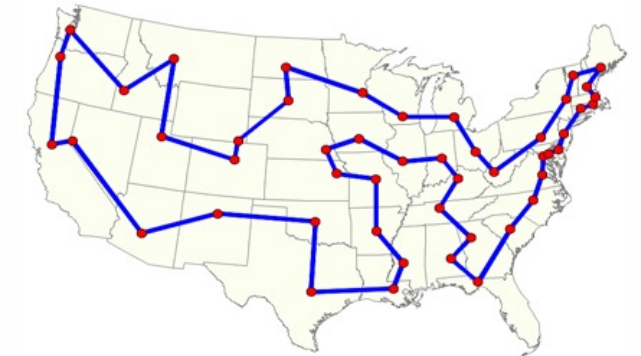
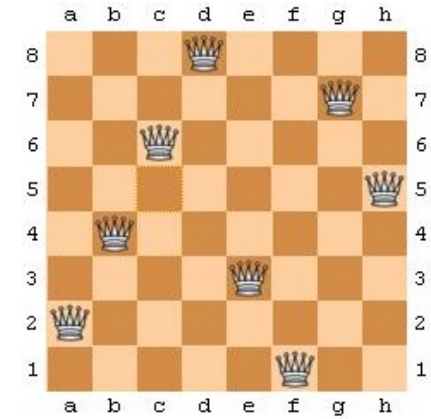
1 0 1 0 0 0 1 1 1 0

0 1 1 0 0 0 1 0 1 1

1 0 1 0 1 0 0 0 1 0

More Problems

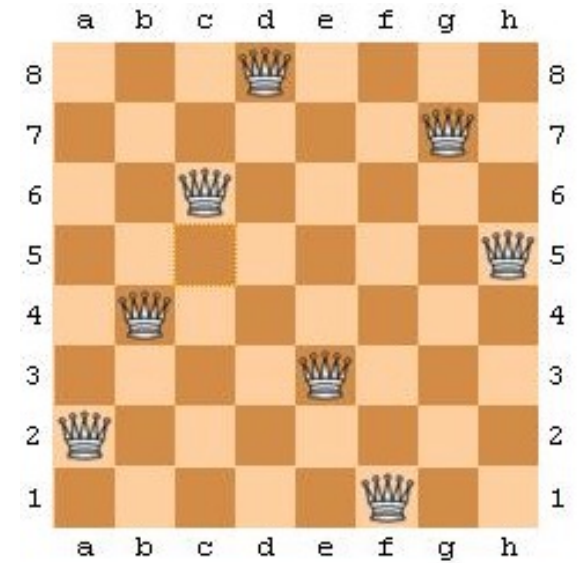
- N-Queen, 8-Queen
- Travelling Salesman Problem
- New Strategies for
 - Chromosome
 - Crossover
 - Mutation



N -queens Problem

- A constraint satisfaction problem
- Place N queens on an $N \times N$ chessboard so that no two of them can check each other

Design a GA for the 8-queens problem



A solution of the 8-queens problem

Chromosome Representation

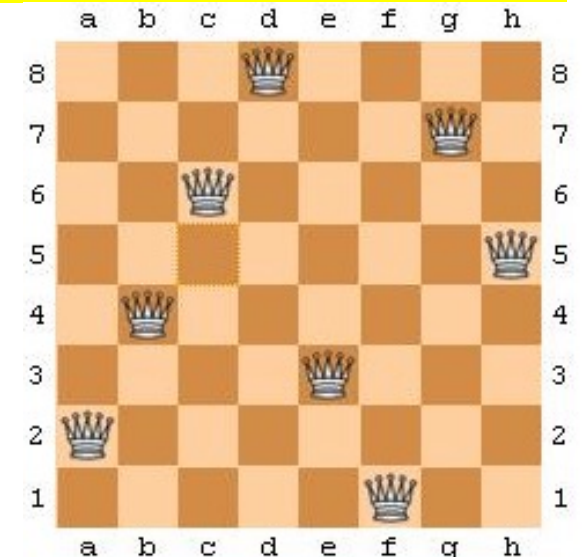
Could use a matrix (2D) of binary genes

Since we know that no two queens could be in the same row and the same column, we can use a permutation representation

Example:



A permutation of integers 1 to 8



Mutation Operator

Need to remain a permutation after mutation

Swap mutation:

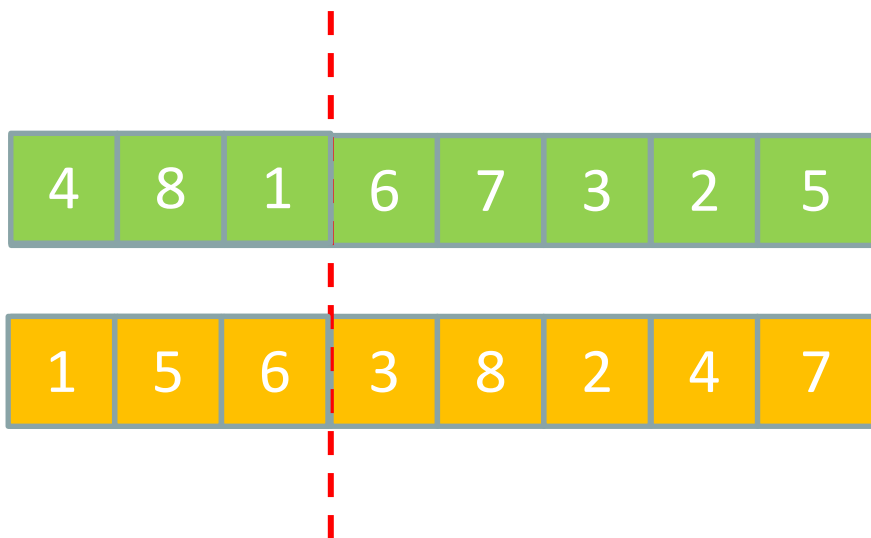
1. Randomly select two genes
2. Swap their positions



Crossover Operator

Cut-and-Crossfill

- Select a random crossover point
- Cut the parents into two segments at this position
- Copy the first segment of parent 1 into child 1 and the first segment of parent 2 into child 2
- Scan parent 2 from left to right and fill the second segment of child 1 with values from parent 2, skipping those that the first segment already contains
- Do the same for parent 1 and child 2



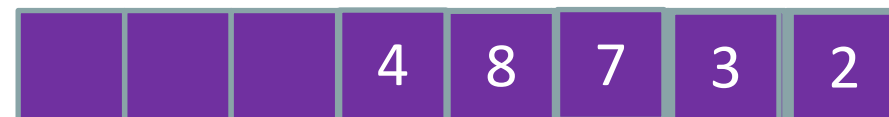
Parent 1

Parent 2

Child 1



Child 2



Parent Selection

- Let there be n individuals in the first generation
- In each cycle, select 2 parents to produce 2 children
 - Randomly select 5 individuals
 - Choose the best 2 as parents

Replacement Strategy

- Combine the population and the two offsprings
 - Total $n + 2$ individuals
- Rank their fitness
- Remove the worst two

Fitness

$q(x)$ = the number of queen pair violations by x

$$fitness = \frac{1}{q(x)}$$

Can be zero



$$fitness = \frac{1}{q(x) + \epsilon}$$

A small value

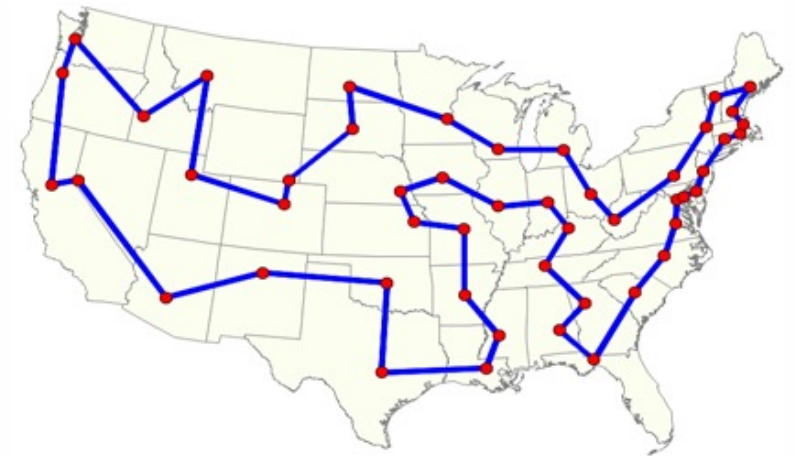


Travelling Salesman Problem

- Given N cities and the distances between each pair of them, find the shortest path that goes through each city once and returns to the starting city
- The number of possible paths that goes through all the cities is

$$\frac{(N - 1)!}{2}$$

Why?



Travelling Salesman Problem

- Given N cities and the distances between each pair of them, find the shortest path that goes through each city once and returns to the starting city
- The number of possible paths that goes through all the cities is

Four cities: A B C D

1. All Permutations: $N!$

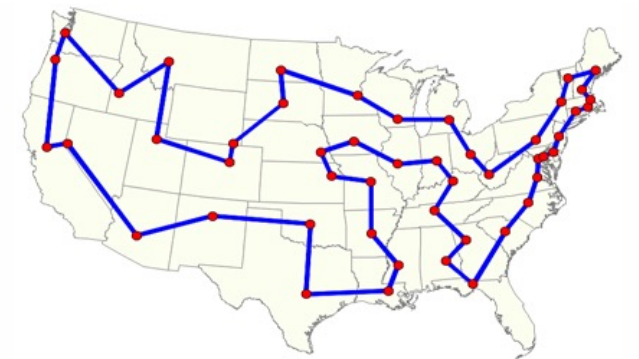
2. Circular

A-B-C-D-A \rightarrow B-C-D-A-B

Fix any starting: $(N-1)!$

3. Reversal

A-B-C-D-A \rightarrow A-D-C-B-A $(N-1)!/2$



Chromosome Design

Cities can be numbered $1, 2, \dots, N$

Important Criteria for Design:

- A valid tour must consist of all cities
- Each city should only appear once, except the starting city

Permutation representation could be used

Mutation

Swap mutation could be used

Other possible mutation operations:

Insertion

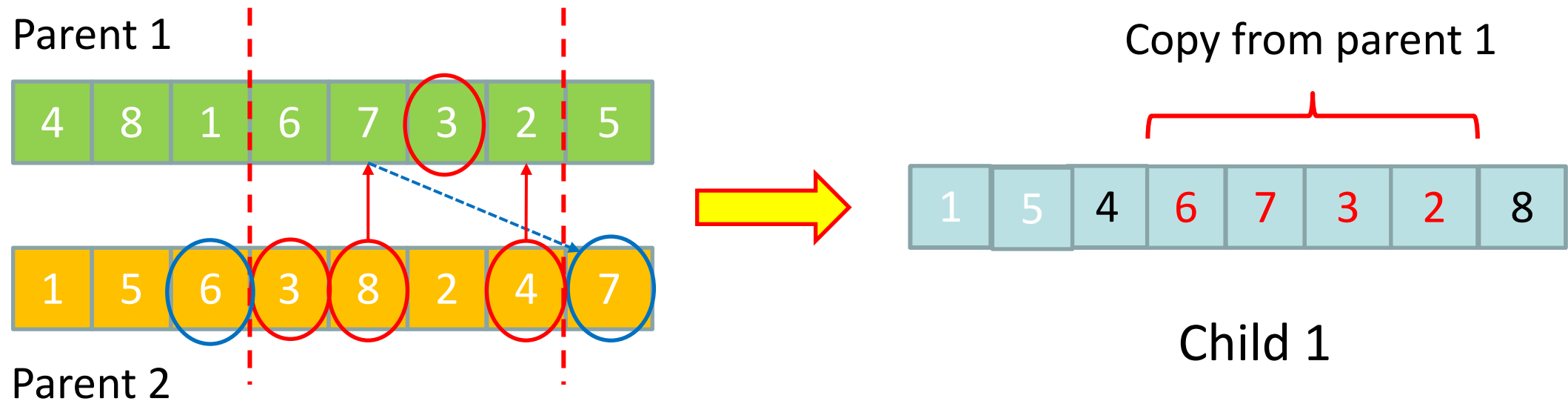


Inversion



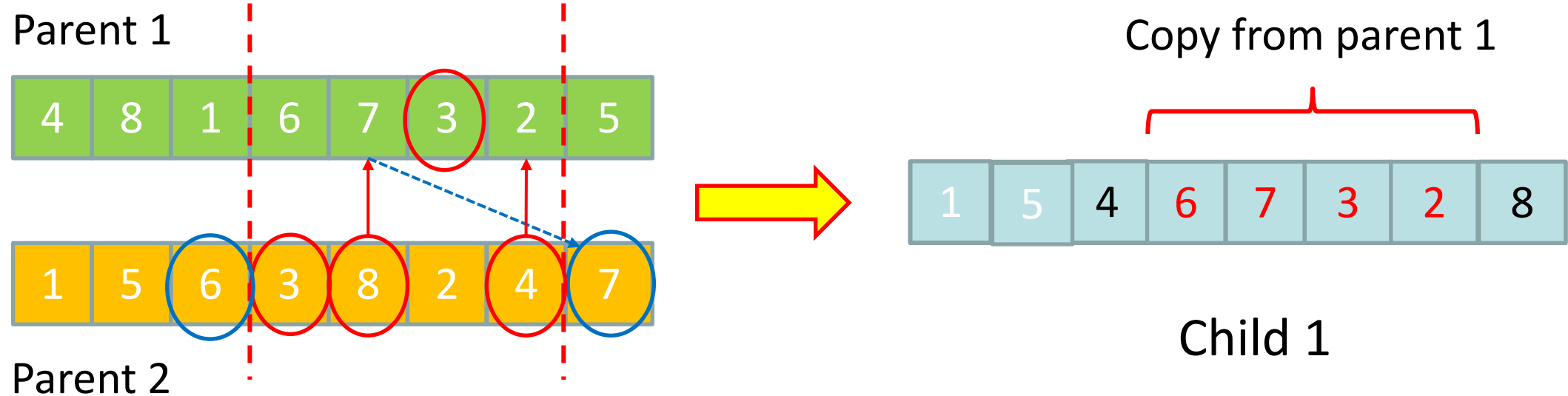
Crossover

Partially Mapped Crossover (PMX)



Crossover

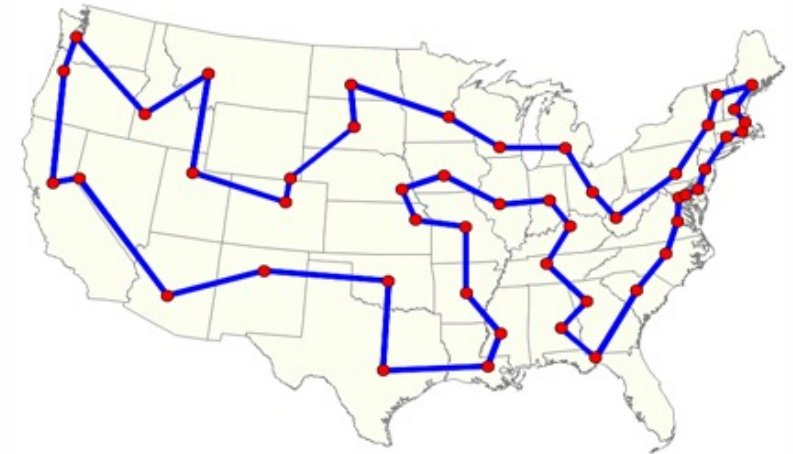
Partially Mapped Crossover (PMX)



Fitness of TSP

Measure if a path is good or not

Total distance:
 $\sum d_{ij}$



Parameter Tuning

- Difficulties:
 - Parameters interact:
Population Size, #Generations, Crossover Rate, Mutation rate, etc
 - Trying different combinations is time consuming
- Good tuning algorithms proposed around 2005:
 - SPO
 - F-race
 - REVAC
 - Meta-GA

Still not widely adopted

Approach

- Treat the design of an EA as a search problem
- The tuning method is a search algorithm (the set of parameters is the vector of values to be searched)
- Tuning algorithms can provide information about an EA:
 - Robustness
 - Distribution of solution quality
 - Sensitivity

Algorithm Quality

- Standard performance metrics
 - Mean best fitness (MBF)
 - Average number of evaluations to a solution (AES)
 - Success rate (SR)

Generate-and-Test

- Robustness
 - Of a set of EA parameters to variations of problem instances
 - Of solution quality to variations of parameters for a particular problem instance

Dealing with Constrained Optimization

- Make use of the encoding method, e.g. N -queens problem
- Apply GA as normal and assign zero fitness to any individual that violates the constraints
 - Inefficient search for smaller problems
 - With highly constrained problems:
 - Can produce many illegal individuals, wasting effort
 - Cause over-rapid convergence – too few legal individuals generated

Dealing with Constrained Optimization

- Three approaches
 - Use of **penalty functions**
 - Use of **repair operators**
 - Creation of tailored **diversity generation** operations

Penalty Functions

Altered fitness function

Original fitness function

penalty function

$$\max f^*(x) = f(x) - \text{pen}(x)$$

Weight of constraint i

$$\text{pen}(x) = \sum_{i=1}^m w_i b_i \quad b_i = \begin{cases} 1, & \text{if } x \text{ violates constraint } i \\ 0, & \text{otherwise} \end{cases}$$

Works for relatively few constraints

Repair Operators

Repair infeasible solutions
by moving them back into the feasible region

Tricky to design

Tailored Diversity-Generation Operators

Crossover and mutation operators are
agnostic to the problem domain

Crossover of two legal parents
may produce an illegal child

Mutation of a legal individual
may result in an illegal one

Design problem-specific crossover and mutation operators

Summary

- N-Queen 8-Queen Problem
- Travelling Salesman Problem
- Parameter Tuning for Genetic Algorithm