

# 8 Representation, Mutation and Recombination Part 3: Permutations and Trees

# Permutation Representations

- ordering and sequencing problems form a special type
- the task needs to be solved by arranging some objects in a certain order
- example: production scheduling:
  - important thing is which tasks are scheduled before others (**order**)
- example: Travelling Salesman Problem (TSP):
  - important thing is which elements occur next to each other (**adjacency**)
- since we only want each task to happen once, or each city to be visited once, we express these problems as a **permutation**
- if there are **n variables then the representation is a list of n integers, each of which occurs exactly once**

# Permutation Representation: Example: Back to the TSP

## *problem:*

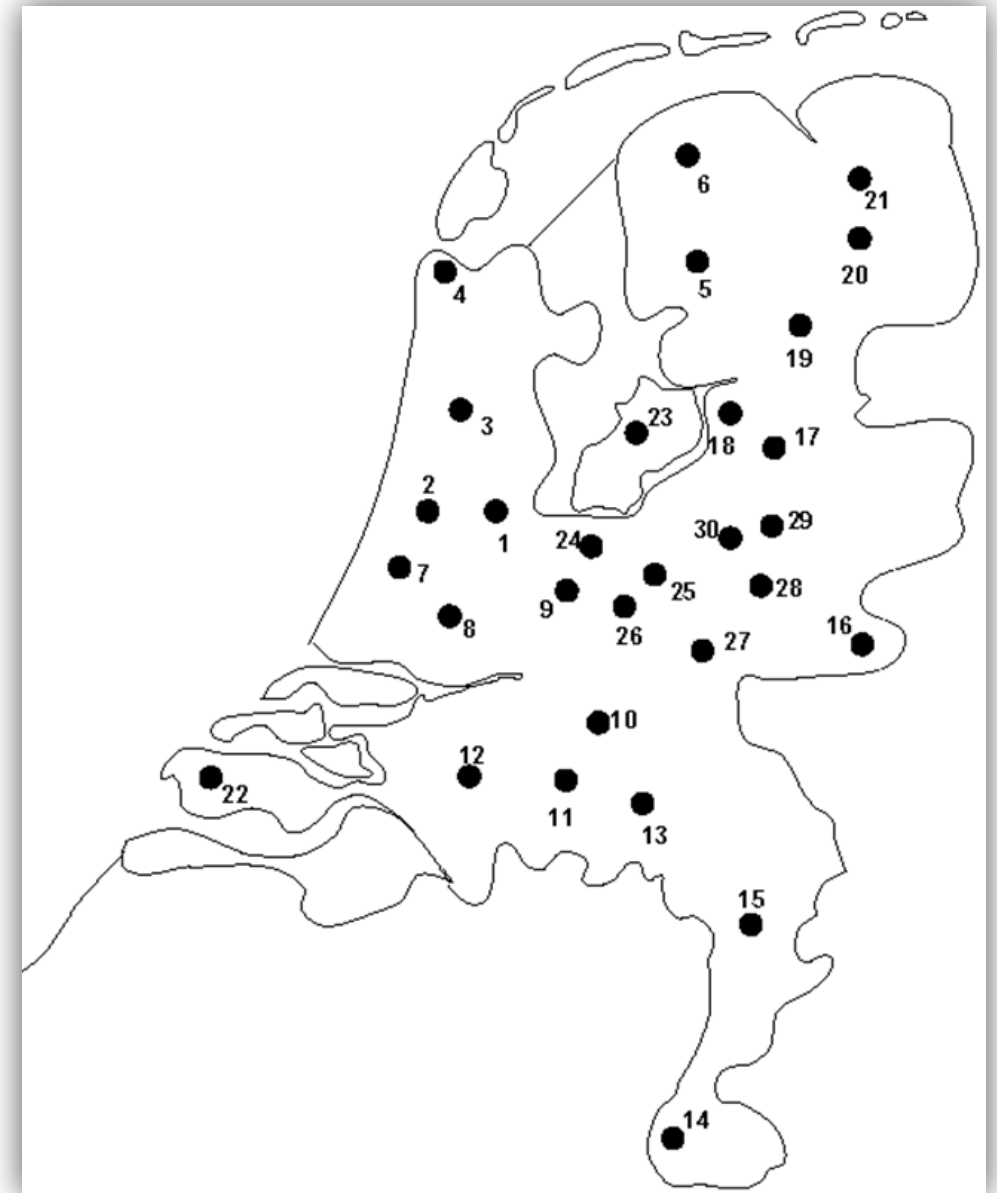
- given  $n$  cities, find a complete tour with minimal length

## *encoding:*

- label the cities  $1, 2, \dots, n$
- one complete tour is one **permutation**
- examples:  $[1, 2, 3, 4], [3, 4, 2, 1]$

## *HUGE search space:*

- for 30 cities there are  $30! \approx 10^{32}$  possible tours

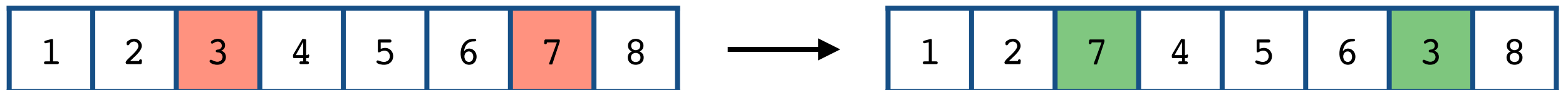


# Permutation Representations: Mutation

- the mutation operators often used for other number representations lead to inadmissible solutions
- why?
- therefore we must change a set of gene values at the same time
- so the mutation parameter now reflects the probability that some operator is applied once to the **whole genotype**, rather than individually in each position

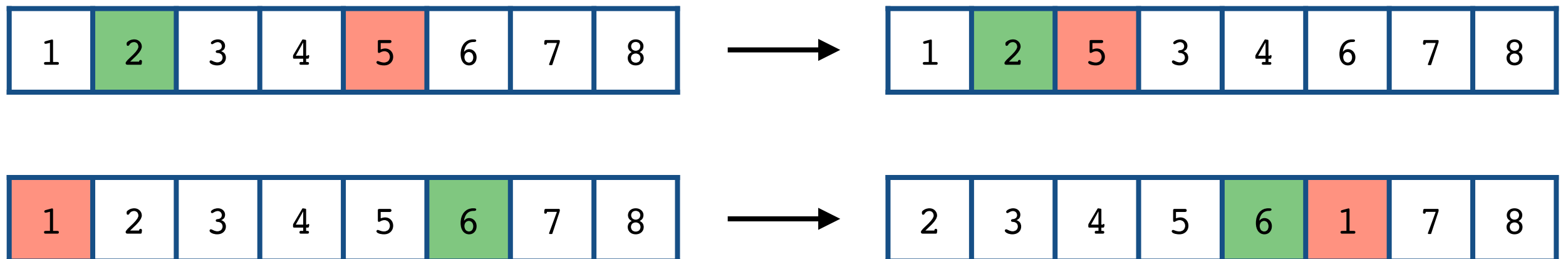
# Permutation Representations: Swap Mutation

- pick two alleles at random and swap their positions



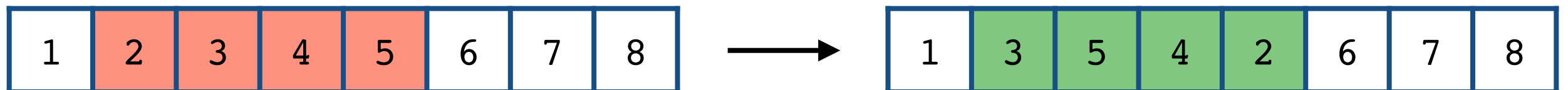
# Permutation Representations: Insert Mutation

- pick two different genes at random
- move the **second chosen** to follow the **first**
- shift the rest along to accommodate
- this preserves most of the order and adjacency information



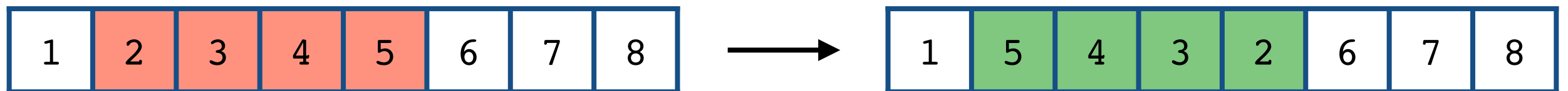
# Permutation Representations: Scramble Mutation

- pick a random subset of genes
- randomly rearrange the alleles in those positions



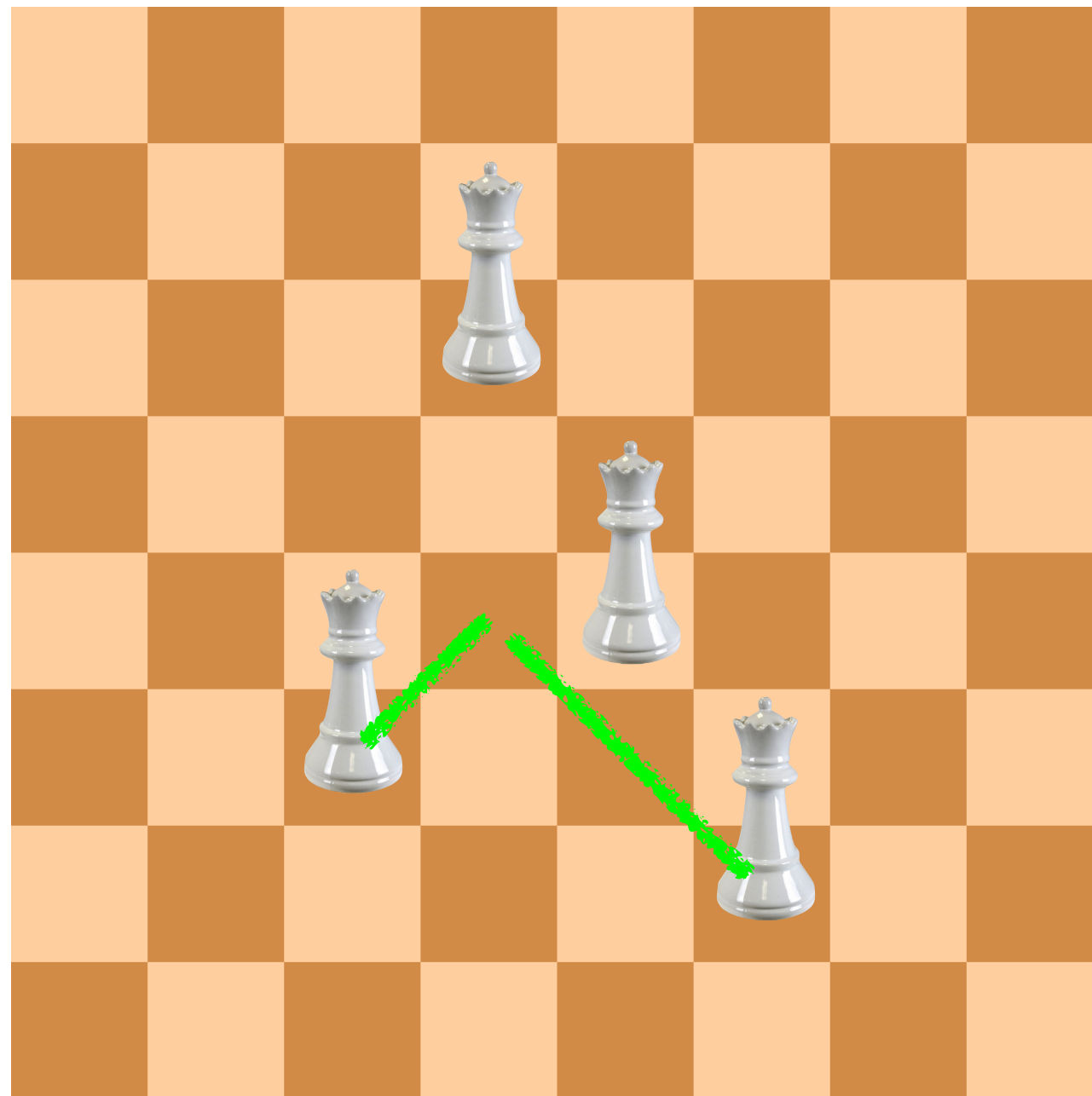
# Permutation Representations: Inversion Mutation

- pick two alleles at random and invert the substring between them
- preserves most adjacency information
  - so good for TSP
- but disrupts order information
  - so bad for production scheduling





# Swap versus Inversion Example: N-Queens



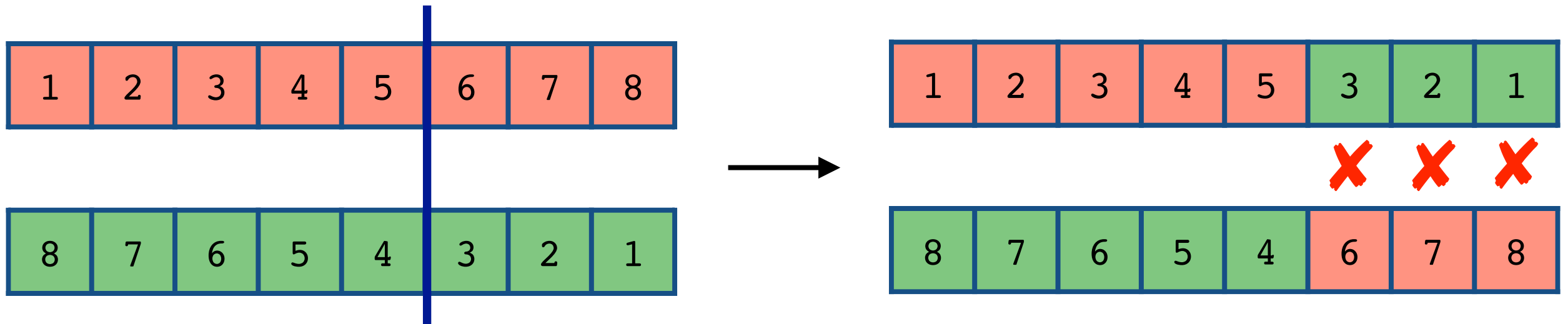
with an adjacency  
problem...

swap can damage a  
good sub-solution...

...while inversion can  
preserve it

# Permutation Representations: Crossover Operators

- normal crossover operators will often lead to inadmissible solutions:



- so many specialised operators have been devised which focus on combining order or adjacency information from the two parents

# Permutation Representations: Order 1 Crossover

- copy **arbitrary part** from first parent

1	2	3	4	5	6	7	8
---	---	---	---	---	---	---	---

5	7	1	3	8	6	2	4
---	---	---	---	---	---	---	---



			4	5	6		
--	--	--	---	---	---	--	--

- copy **rest of values** from second parent in order,  
beginning at the '2'

1	2	3	4	5	6	7	8
---	---	---	---	---	---	---	---

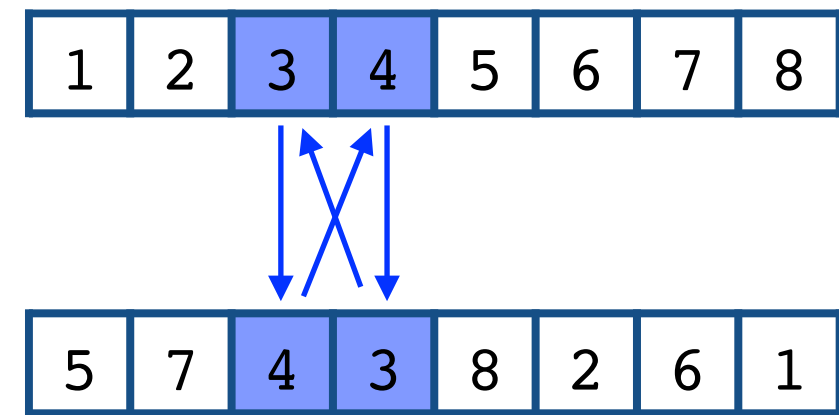
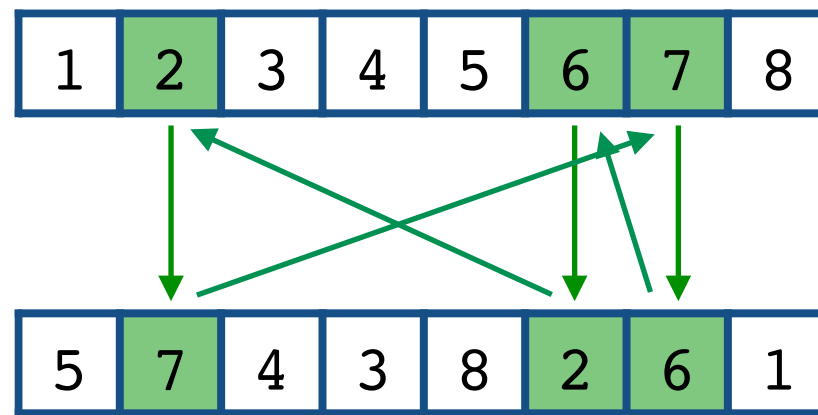
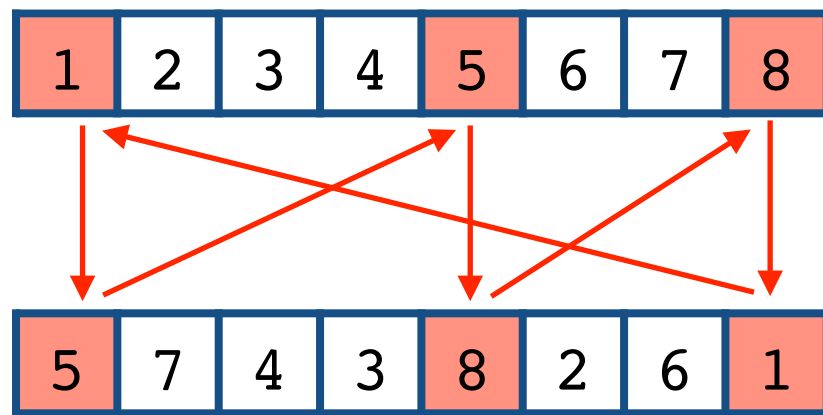
5	7	1	3	8	6	2	4
---	---	---	---	---	---	---	---



1	3	8	4	5	6	2	7
---	---	---	---	---	---	---	---

# Permutation Representations: Cycle Crossover

- identify all cycles in parents:



- copy alternate cycles into offspring:



# Permutation Representations: Edge Crossover

- intention: offspring should be created as far as possible using only the 'edges' present in one of the parents
- edges are when two values neighbour each other (wraparound)
- so intended for adjacency-based permutation problems
- creates only one child
- works by first constructing a table that lists which edges are present in the two parents
- if an edge is common to both parents it is marked with a '+'
- and common edges should be preserved in the child

# Permutation Representations: Edge Crossover: Construct Edge Table

parent 1

1	2	3	4	5	6	7	8
---	---	---	---	---	---	---	---

parent 2

5	7	4	3	8	2	6	1
---	---	---	---	---	---	---	---

edge table

element	edges	element	edges
1	2, 8, 6, 5	5	4, 6, 7, 1
2	1, 3, 8, 6	6	5, 7, 1, 2
3	2, 4+, 8	7	6, 8, 5, 4
4	3+, 5, 7	8	7, 1, 3, 2

+ because 3 and 4 neighbour each other in both parents

# Permutation Representations: Edge Crossover

- informal procedure:
  - construct the edge table (done ✓)
  - select an initial element, `entry`, at random and put it in the offspring
  - set the variable `current` = `entry`
  - remove all references to `current` from the lists of edges in the table
  - examine the list of edges for `current`:
    - if there is a common edge (marked with +), pick that to be the next current element
    - otherwise pick the entry in the list which itself has the shortest list of edges
    - ties are split at random
  - in the case of reaching an empty list:
    - choose a new current element at random

# Permutation Representations:

## Edge Crossover: Create Child

element	edges	element	edges
1	2, 8, 6, 5	5	4, 6, 7, 1
2	1, 3, 8, 6	6	5, 7, 1, 2
3	2, 4+, 8	7	6, 8, 5, 4
4	3+, 5, 7	8	7, 1, 3, 2

choices	element selected	reason	partial result (child)
all	5	initial random choice	5
4, 6, 7, 1	4	shortest list	5, 4
3, 5, 7	3	common edge	5, 4, 3
2, 8	8	random choice (tie)	5, 4, 3, 8
7, 1, 2	7	shortest list	5, 4, 3, 8, 7
6	6	only choice	5, 4, 3, 8, 7, 6
1, 2	1	random choice (tie)	5, 4, 3, 8, 7, 6, 1
2	2	only choice	5, 4, 3, 8, 7, 6, 1, 2



# Permutation Representations: Edge Crossover: Parents versus Child

parent 1

1	2	3	4	5	6	7	8
---	---	---	---	---	---	---	---

parent 2

5	7	4	3	8	2	6	1
---	---	---	---	---	---	---	---

child

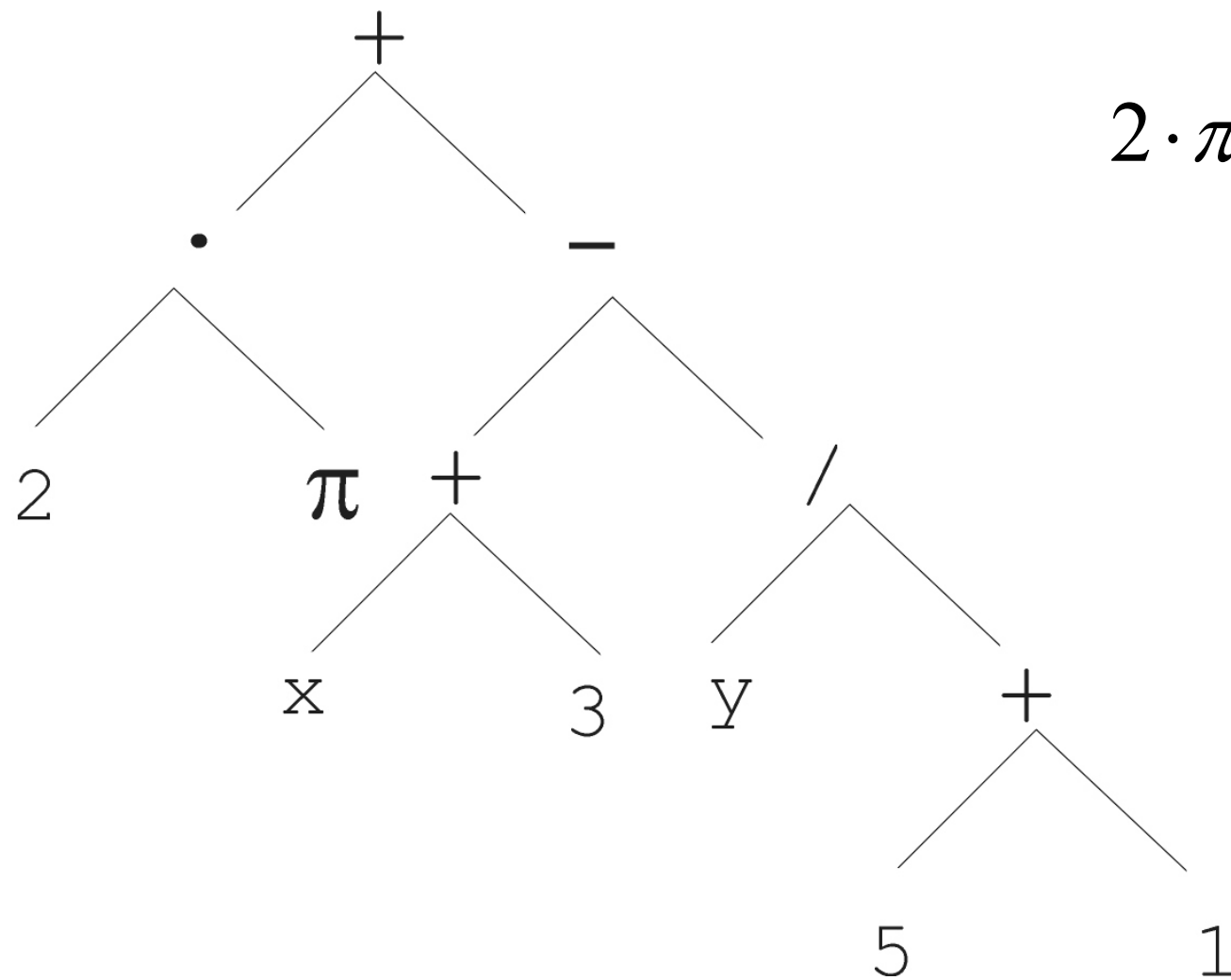
5	4	3	8	7	6	1	2
---	---	---	---	---	---	---	---

notice that many adjacencies between values have been preserved

# Tree Representation

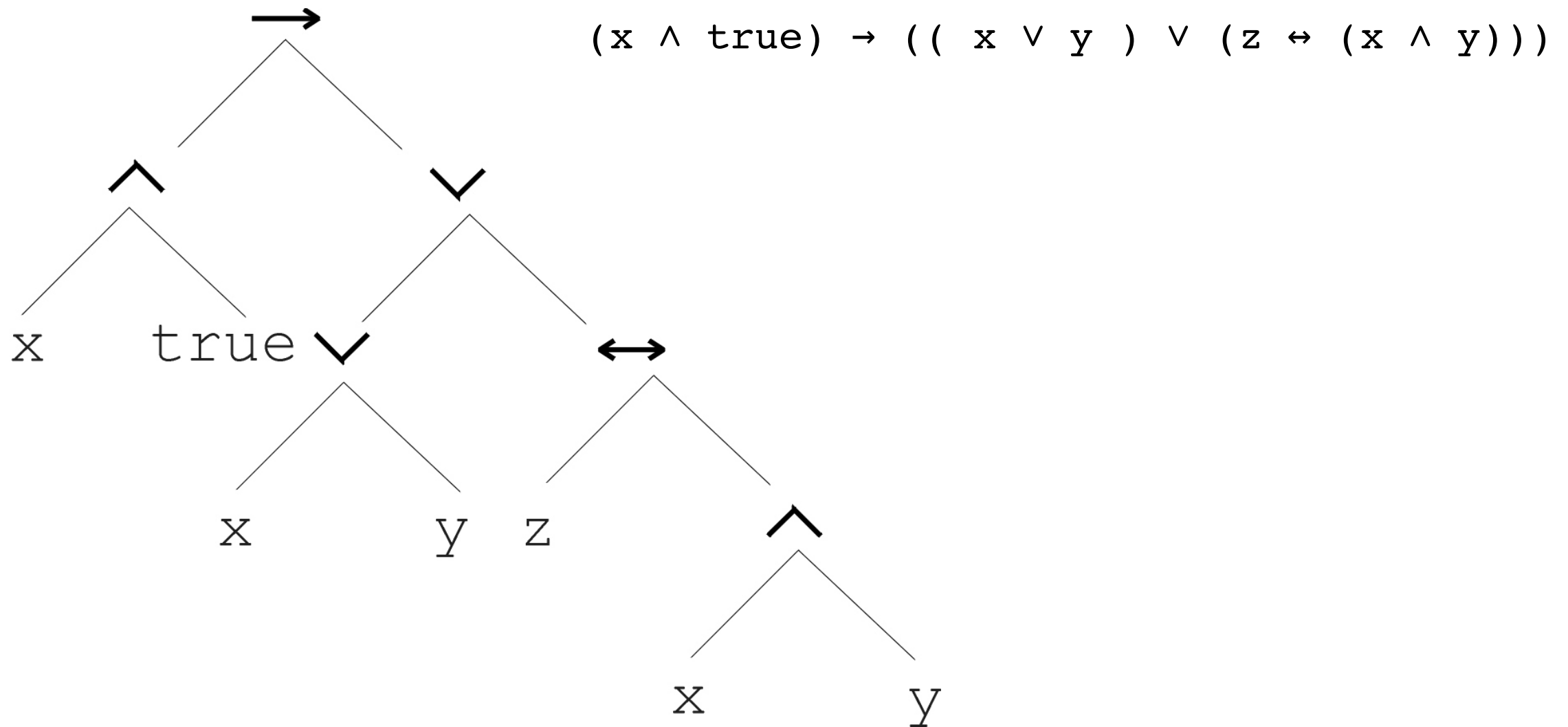
- trees are one of the most general structures for representing objects in computing
- they can be used to represent:
  - arithmetic formulae
  - logical formulae
  - parse trees
  - programs
  - and many other concepts
- they allow us to represent non-linear genotypes
- form the basis of genetic programming (GP)
  - where they can vary in depth and width

# Tree Representation

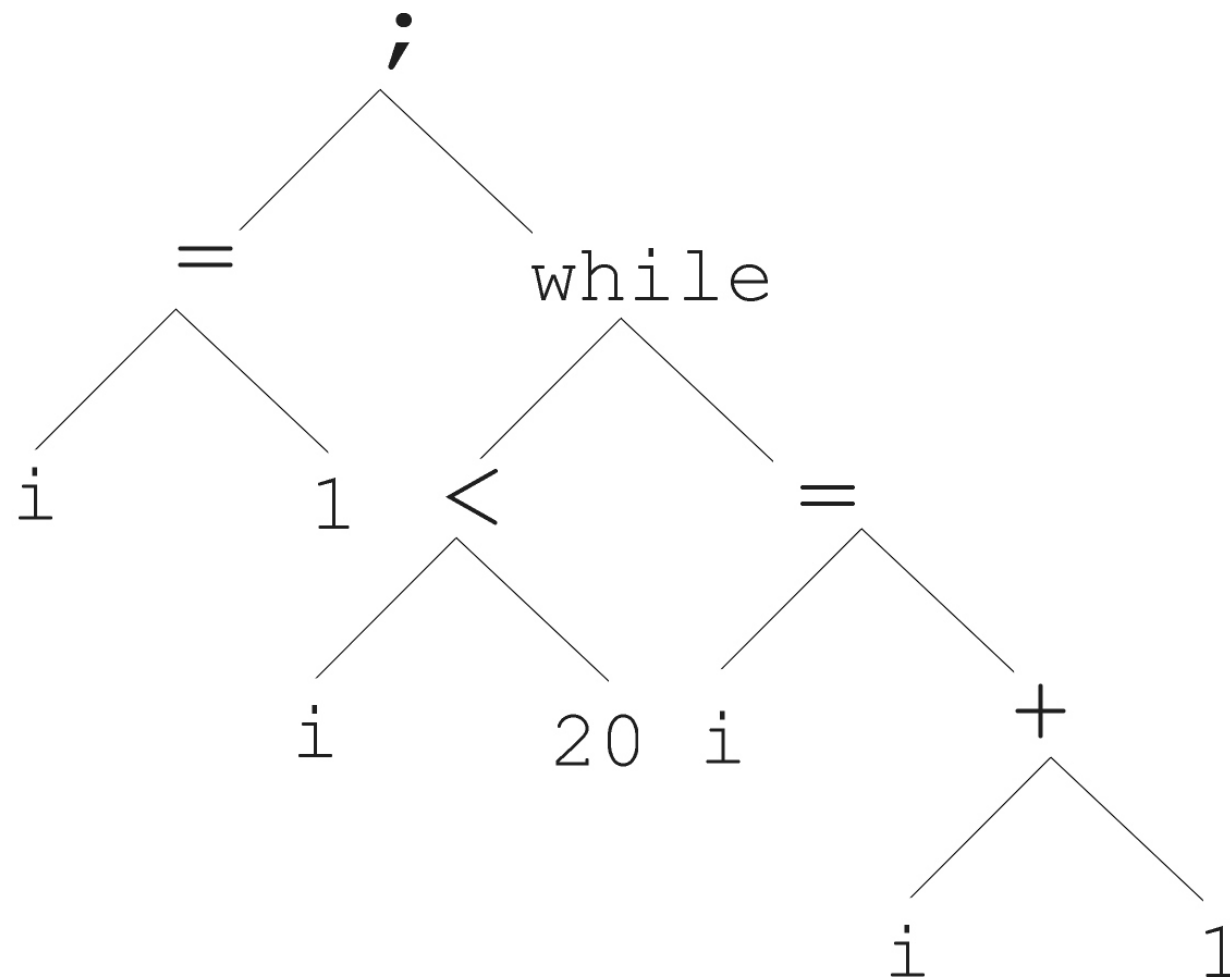


$$2 \cdot \pi + \left( (x + 3) - \frac{y}{5 + 1} \right)$$

# Tree Representation



# Tree Representation



```
i = 1;
while (i < 20)
{
    i = i + 1
}
```

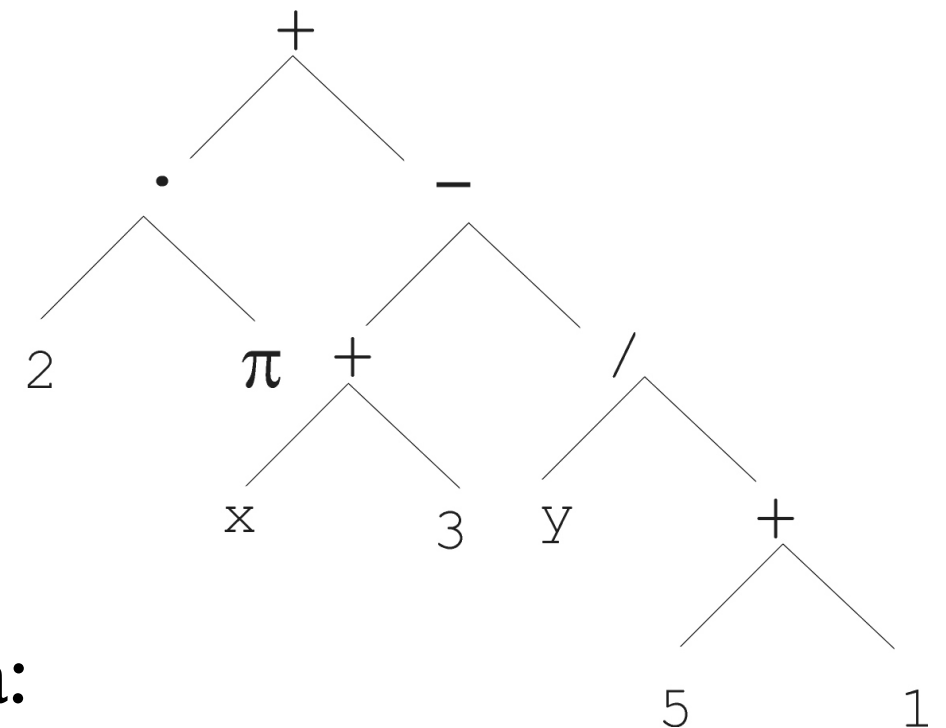
# Tree Representation

- symbolic expressions can be defined by

- terminal set  $T$ 
  - form the leaves
- function set  $F$ 
  - form the internal nodes

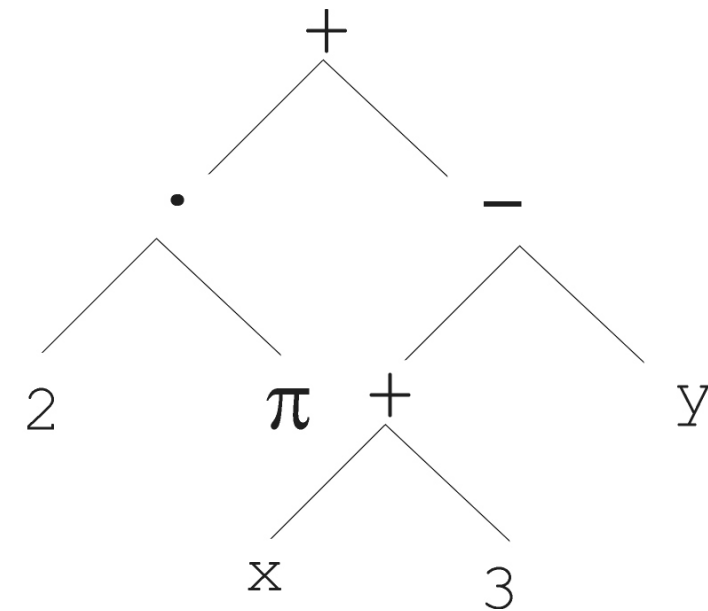
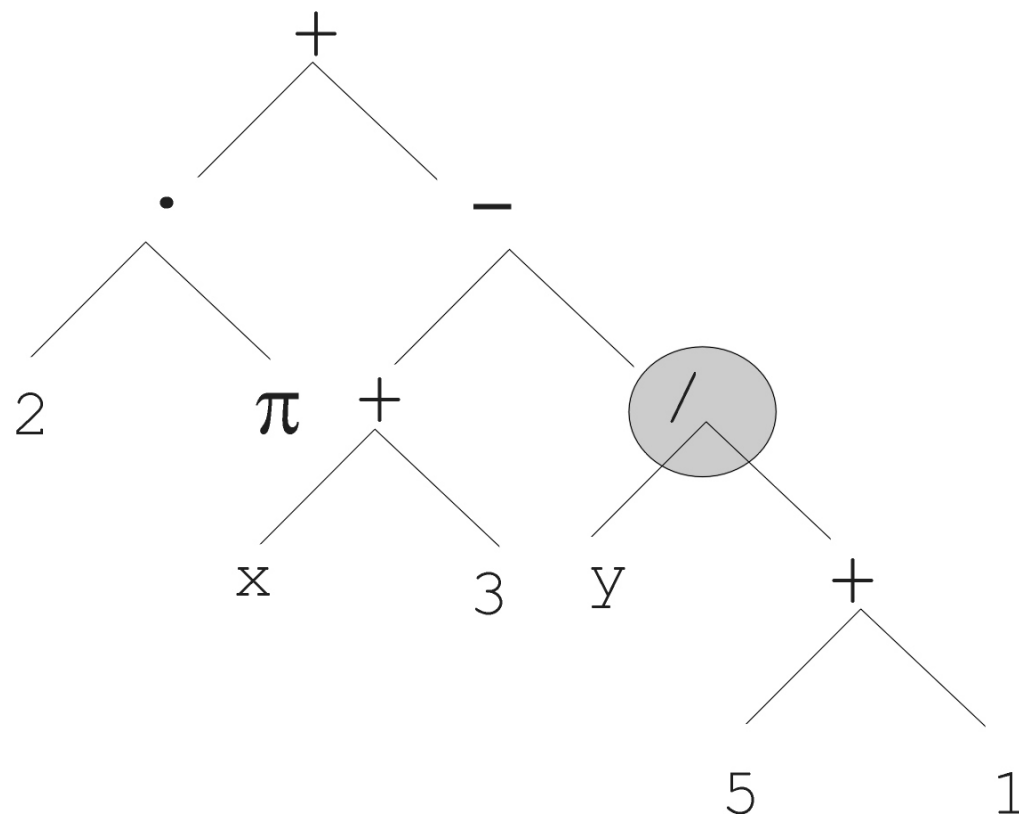
- for example, for an arithmetic formula:

function set	$\{+, -, \cdot, /\}$
terminal set	$\mathbb{R} \cup \{x, y\}$

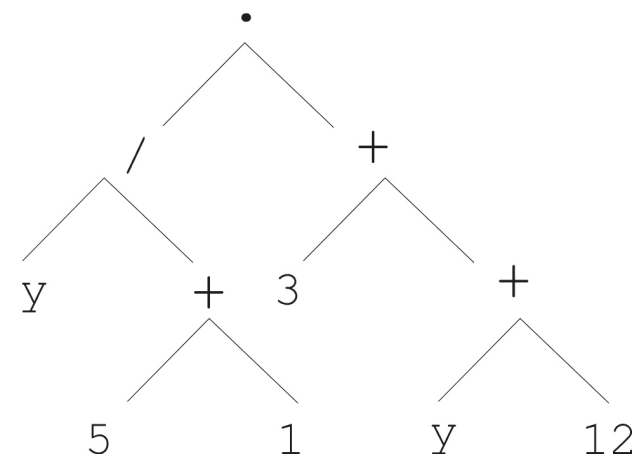
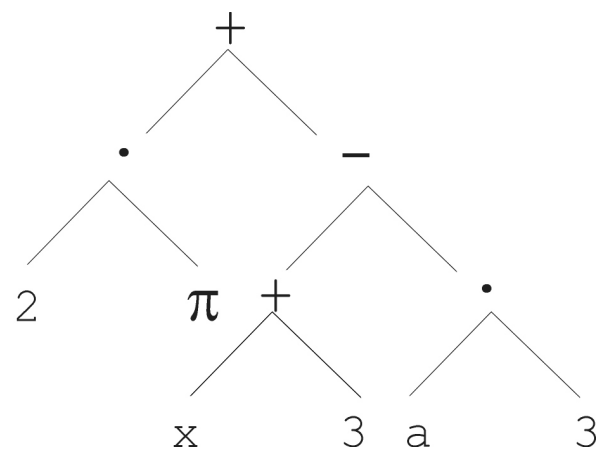
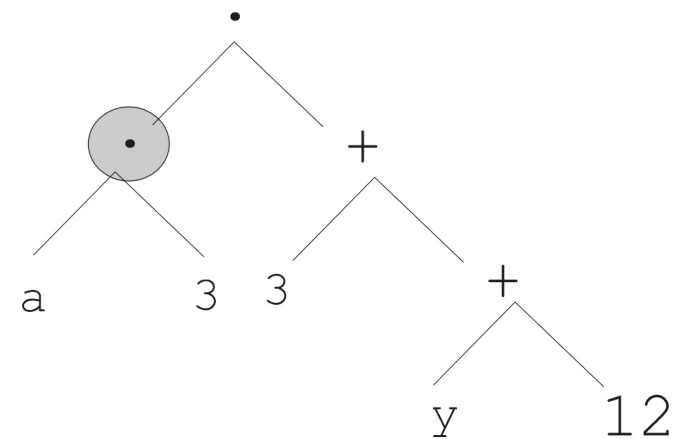
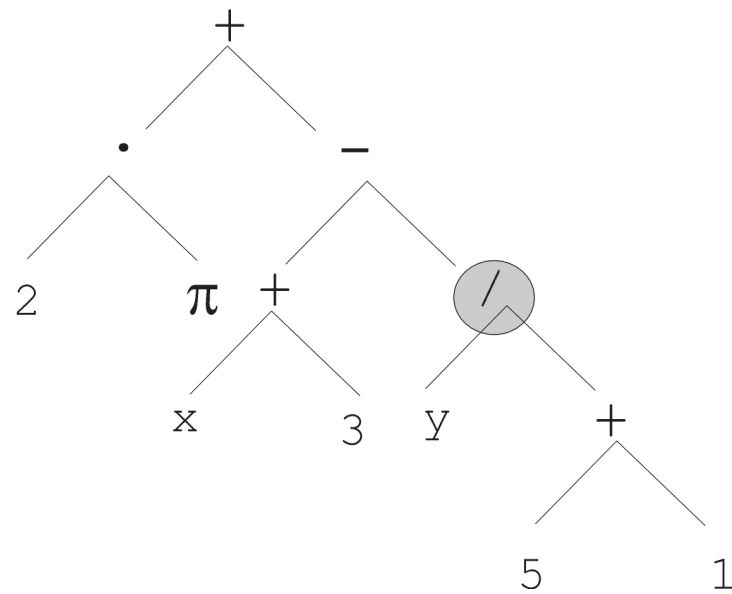


# Tree Representation: Mutation

- *most common mutation operator*: replace a randomly chosen subtree with a randomly generated tree
- in this case, it's simply the variable 'y' from the terminal set
- but it could be larger



# Tree Representation: Recombination





# Reading & References

- slides based on and adapted from, Chapter 4 (and slides) of Eiben & Smith's *Introduction to Evolutionary Computing*
- W.M. Spears: Evolutionary Algorithms: The Role of Mutation and Recombination, Springer 2000
- K. Deb: Representations. Part 4 of T. Bäck, D. Fogel and Z. Michalewicz (editors) Evolutionary Computation I: Basic Algorithms and Operators, Institute of Physics Press
- *note that above link leads directly to a .pdf download*