
OPTIMIZATION FOR AI

GLOBAL AND MULTI-OBJECTIVE OPTIMIZATION

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INTRODUCTION TO GENETIC PROGRAMMING (PART 2)

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LINEAR GP

MOTIVATIONS

- Trees are not the only way of representing programs
 - Also streams of instructions can represent programs
 - Instead of LISP-like structure we now use assembly-like commands
 - Linear GP: a linear stream of “assembly-like” instructions
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AN EXAMPLE OF LINEAR GP



Registers of a virtual
(or real!) machine

Add R1, R2, R1

Sub R3, R1, R4

Add R4, R3, R2

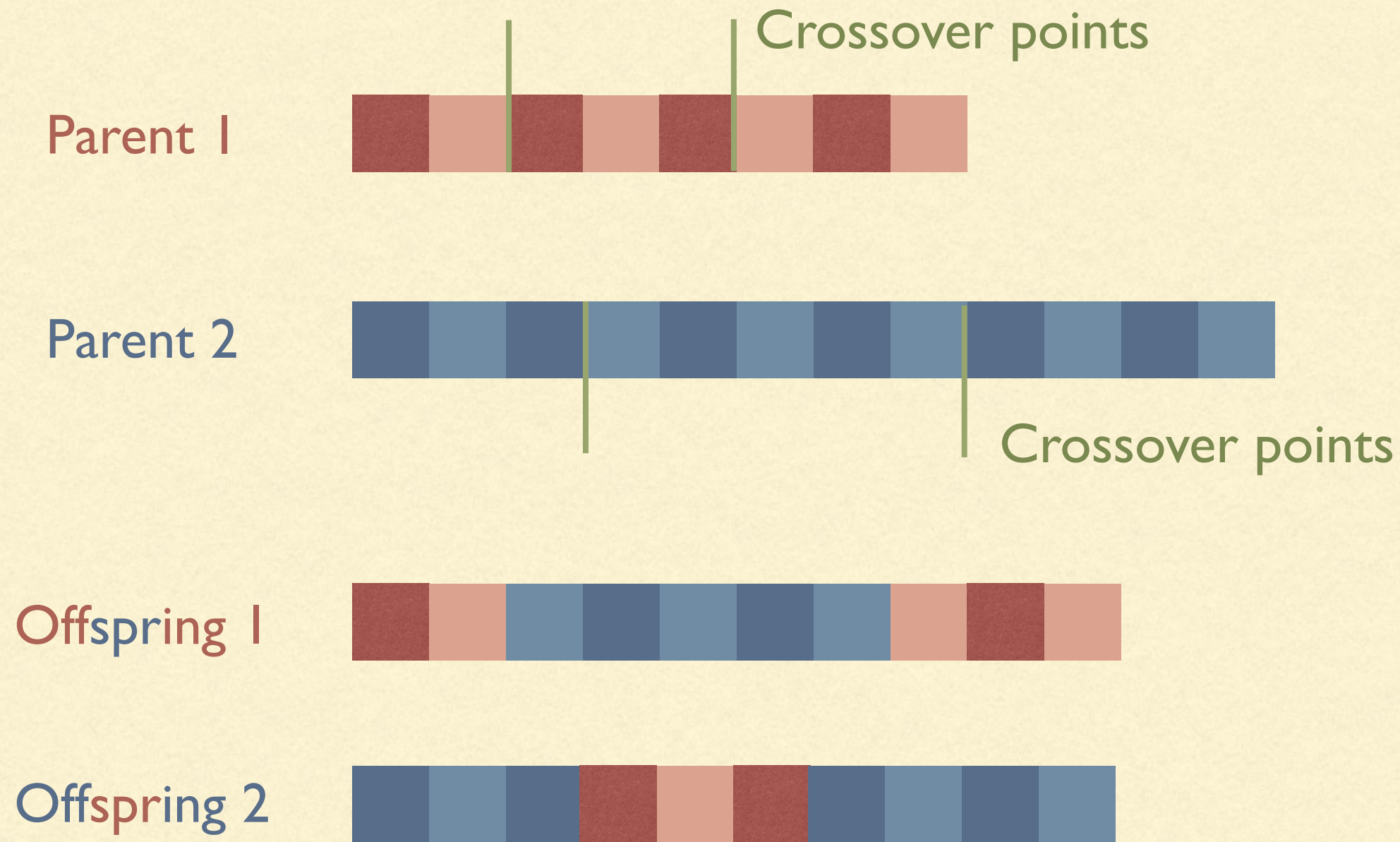
Mul R1, R1, R2

A Linear GP individual:
a list of instructions for the machine
communicating via registers

LINEAR GP AND GA

- Linear GP seems pretty similar to standard GA
 - Except that the individuals can be of non-fixed length
 - An important difference is that we are evolving programs and we “execute” the individuals
 - Most of the operators of GA can be used for Linear GP
 - Two points crossover (with possibly different crossover points between the two individuals) is usually employed
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TWO-POINTS CROSSOVER

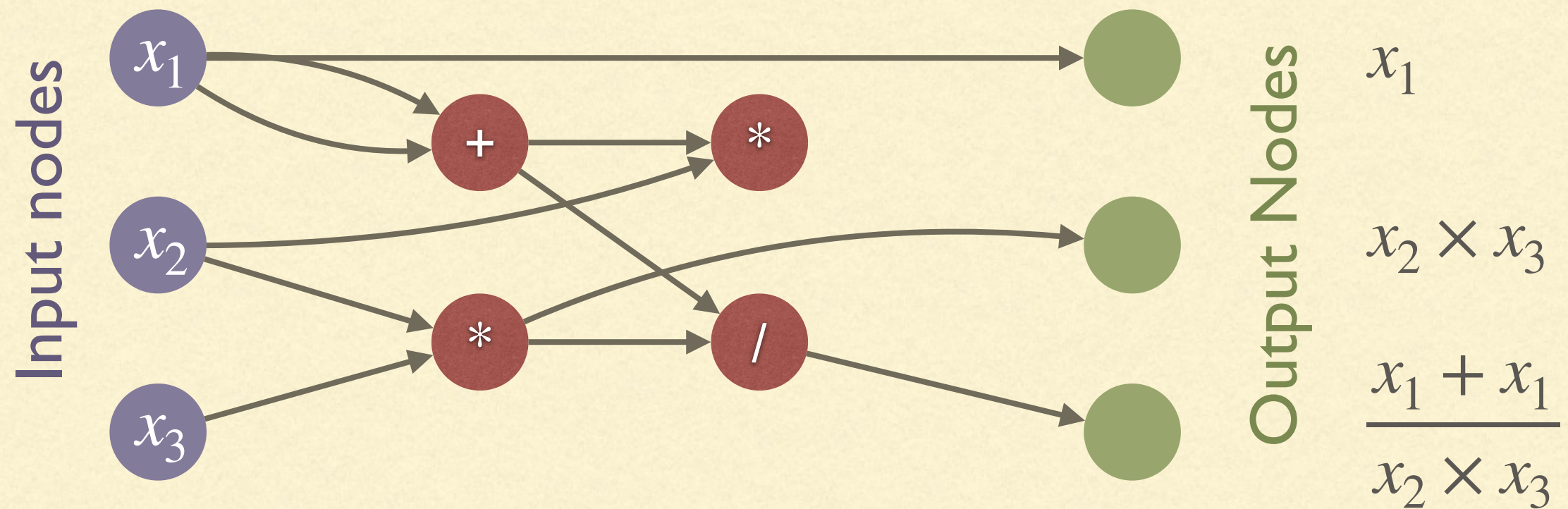


CARTESIAN GP

MOTIVATIONS

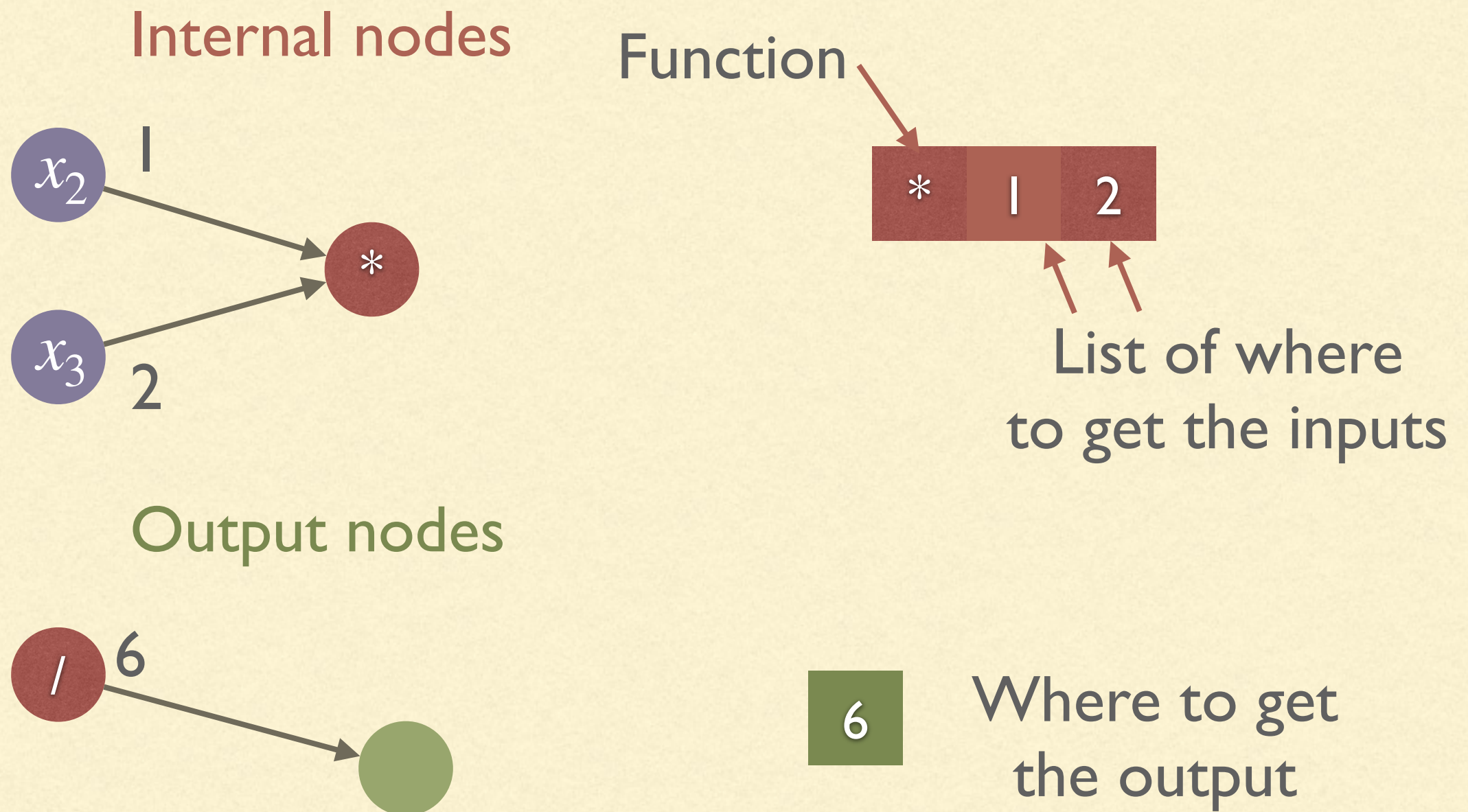
- It is possible to represent programs as circuits/graphs
 - Cartesian GP represents individuals in this way (invented by Julian F. Miller)
 - Naturally suited for problems with many inputs and many outputs (instead of using multiple trees)
 - Has some tracts in common with linear GP (the representation for the circuit/graph is encoded in a linear way)
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AN INDIVIDUAL OF CGP

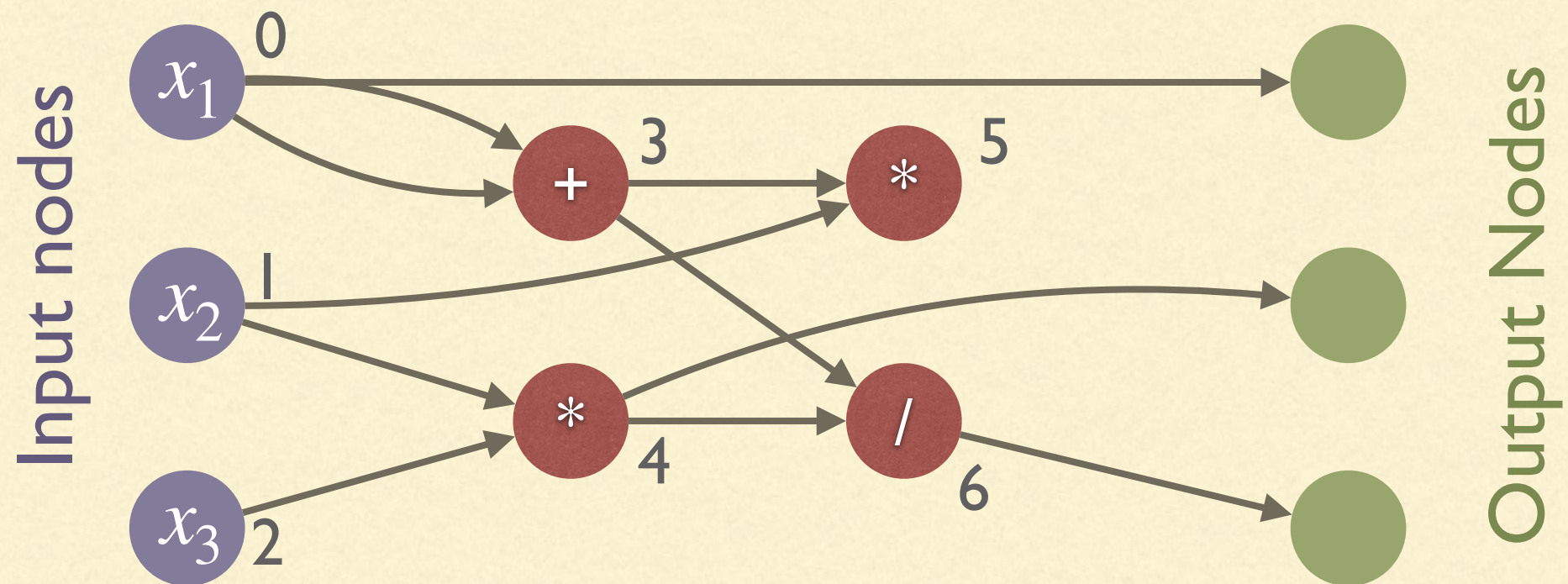


But how is the individual encoded?

ENCODING IN CGP



AN INDIVIDUAL OF CGP



CGP MUTATION



Mutation point

A connection gene \longrightarrow Random valid connection



Mutation point

A function gene \longrightarrow Random function

GRAMMATICAL EVOLUTION

WHY?

- GP trees allow to express constructs of high-level languages
 - A linear representation is easier to manage
 - Grammatical Evolution combines a linear genome with a tree-based representation (in an indirect way)
 - The idea is to define a *grammar* that is then used to interpret a linear genome as a derivation of that grammar
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BACKUS NAUR FORM (BNF)

- Invented by John Backus and refined by Peter Naur
 - Used to define context-free grammars, initially for the ALGOL programming language
 - It expresses a language as a set of production rules
 - Each rule maps a non-terminal to a string of zero or more non-terminals and terminals
-

BACKUS NAUR FORM (BNF)

- A grammar in BNF is a quadruple (T, N, P, S) where:
 - T is a set of terminal symbols
 - N is a set of non-terminal symbols
 - P is a set of production rules
 - $S \in N$ is the axiom, i.e., the start symbol
-

AN EXAMPLE OF BNF

Terminal symbols

$$T = \{+, 0, 1, 2, 3, 4, 5, 6, 7, 8, 9\}$$

Non-terminal symbols

$$N = \{S, C, D\} \quad \text{Axiom } S$$

What kind of strings
can be defined with
this grammar?

Production Rules

$$S \rightarrow S + S \mid C$$

$$C \rightarrow D \mid DC$$

$$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$$

AN EXAMPLE OF BNF

$S \rightarrow S + S \rightarrow C + S \rightarrow DC + S \rightarrow 4C + S \rightarrow 4D + S$

$\rightarrow 42 + S \rightarrow 42 + S + S \rightarrow 42 + C + S \rightarrow 42 + D + S$

$\rightarrow 42 + 8 + S \rightarrow 42 + 8 + C \rightarrow 42 + 8 + D \rightarrow 42 + 8 + 6$

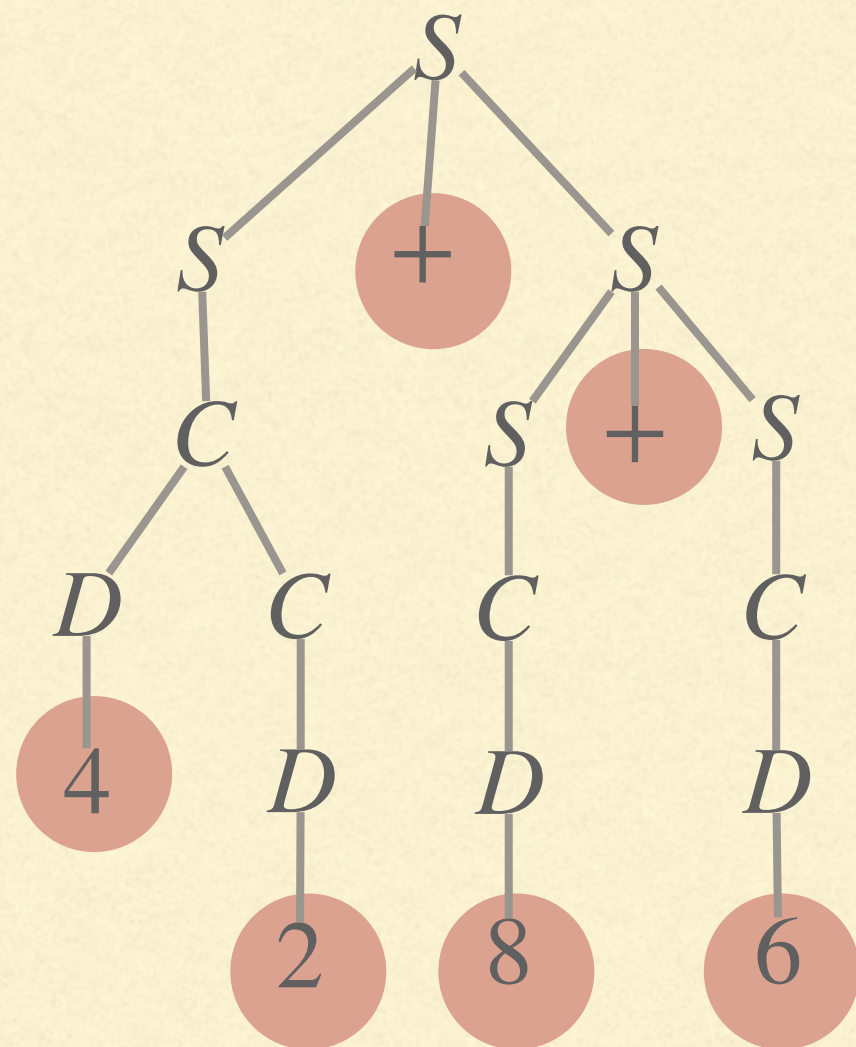
Production Rules

$S \rightarrow S + S \mid C$

$C \rightarrow D \mid DC$

$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$

AN EXAMPLE OF BNF



The expansion can also be seen as building a tree

All the non-terminal are internal nodes

The terminals are the leaves

The resulting string is obtained by reading all the leaves in order

AN EXAMPLE OF BNF

- With the previous grammar we can, depending on the expansion that we select, generate all strings that represents the sum of integers numbers
 - Notice that if we want to obtain the result of the sum we need to evaluate the string
 - This means that if we can have a grammar expressing a program (with variables) and evaluate it on a specific input
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FROM BNF TO GE

- What if we want to generate a specific string?
 - We can keep track of which expansion to select among all the possible ones for a non-terminal symbol
 - Something like:
 - Select the expansion number 1, then number 3, then number 7, then number 2, etc.
 - Seems like we can have a vector (1,3,7,2,...)
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FROM BNF TO GE

- Once we have a vector of integers we can use GA-like evolution!
 - The genome is a sequence of integers
 - To evaluate an individual we need to generate a string via a grammar...
 - ...and then evaluate the string
 - We are sure that the string is a valid program because it was generated by a grammar!
-

TOWARDS GE

$$\begin{array}{cc} 0 & 1 \\ S \rightarrow S + S & | C \end{array}$$

We number the RHS
of all production rules

$$\begin{array}{cc} 0 & 1 \\ C \rightarrow D & | DC \end{array}$$

But not all production rules
have the same number of RHS!

$$\begin{array}{cccccccccc} 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 \\ D \rightarrow 0 & | 1 & | 2 & | 3 & | 4 & | 5 & | 6 & | 7 & | 8 & | 9 \end{array}$$

Should our genome be in $\{0,1\}^n$ or in $\{0,1,\dots,9\}^n$?

TOWARDS GE

Should our genome be in $\{0,1\}^n$ or in $\{0,1,\dots,9\}^n$?

$$\begin{array}{cc} 0 & | \\ S \rightarrow S + S & | C \end{array}$$

Neither, we can use any integer!
Simply take it modulus the number of RHS

$$\begin{array}{cc} 0 & | \\ C \rightarrow D & | DC \end{array} \quad \xleftarrow{\text{Mod } 2 = 1} \quad 7548397623$$

$$\begin{array}{cccccccccc} 0 & | & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 \\ D \rightarrow 0 & | & 1 & | & 2 & | & 3 & | & 4 & | & 5 & | & 6 & | & 7 & | & 8 & | & 9 \end{array}$$

AN EXAMPLE

$$S \rightarrow (S \text{ } OP \text{ } S) \mid V \mid N$$

$$OP \rightarrow + \mid \times \mid - \mid \div$$

$$V \rightarrow x_1 \mid x_2 \mid x_3$$

$$N \rightarrow -DD \mid -D \mid D \mid DD$$

$$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$$

3	5	22	6	9	2	56	18	24	1
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AN EXAMPLE



$$S \rightarrow (S \text{ } OP \text{ } S) \mid V \mid N$$

S OP S

$$OP \rightarrow + \mid \times \mid - \mid \div$$

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3	5	22	6	9	1	56	18	24	1
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0

AN EXAMPLE



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$S \text{ } OP \text{ } S$



$N \text{ } OP \text{ } S$



3	5	22	6	9	1	56	18	24	1
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0 2

AN EXAMPLE

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$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$

$N \text{ } OP \text{ } S$



$D \text{ } OP \text{ } S$



3	5	22	6	9	1	56	18	24	1
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0 2 2

AN EXAMPLE

$S \rightarrow (S \text{ } OP \text{ } S) \mid V \mid N$

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$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$

$D \text{ } OP \text{ } S$



$5 \text{ } OP \text{ } S$



3	5	22	6	9	1	56	18	24	1
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0 2 2 6

AN EXAMPLE

$S \rightarrow (S \text{ OP } S) \mid V \mid N$

$OP \rightarrow + \mid \times \mid - \mid \div$


$V \rightarrow x_1 \mid x_2 \mid x_3$

$N \rightarrow -DD \mid -D \mid D \mid DD$

$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$

$5 \text{ OP } S$

\downarrow
 $5 \times S$



3	5	22	6	9	1	56	18	24	1
0	2	2	6	1					

AN EXAMPLE



$S \rightarrow (S \text{ } OP \text{ } S) \mid V \mid N$

$OP \rightarrow + \mid \times \mid - \mid \div$

$V \rightarrow x_1 \mid x_2 \mid x_3$

$N \rightarrow -DD \mid -D \mid D \mid DD$

$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$

$5 \times S$



$5 \times V$



3	5	22	6	9	1	56	18	24	1
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0 2 2 6 1 1

AN EXAMPLE

$S \rightarrow (S \text{ } OP \text{ } S) \mid V \mid N$

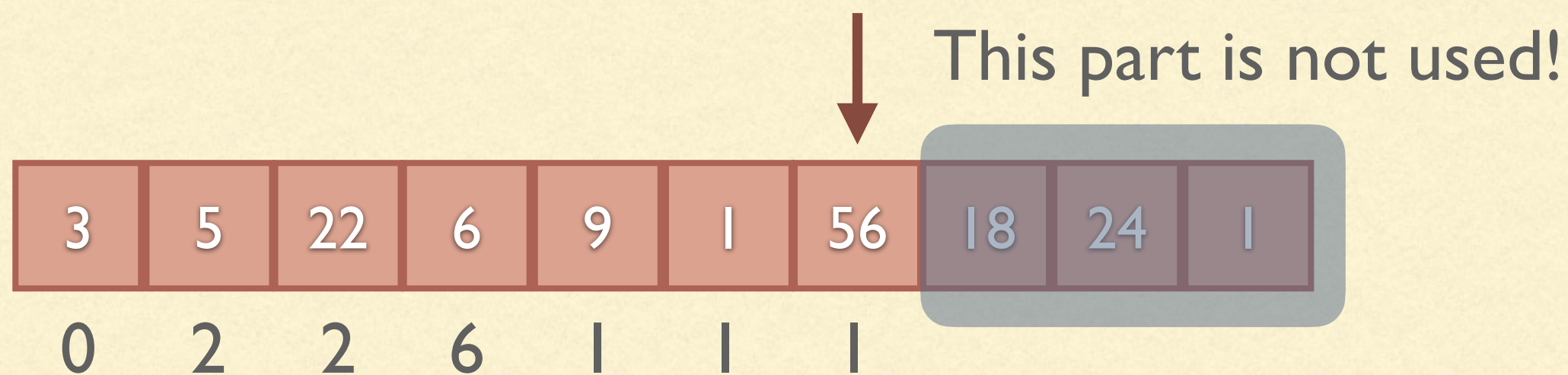
$OP \rightarrow + \mid \times \mid - \mid \div$

$V \rightarrow x_1 \mid x_2 \mid x_3$

$N \rightarrow -DD \mid -D \mid D \mid DD$

$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$

$5 \times V$
↓
 $5 \times x_2$



GE PECULIARITIES

- The individuals might not be all of the same length
 - The effective length of an individual can be shorter than the length of the genome (previous example)...
 - ...but it can also be longer (in this case we “wrap around” and continue from the beginning)
 - We must have a maximum length for the expansion!
-

AN EXAMPLE



$$S \rightarrow (S \text{ } OP \text{ } S) \mid V \mid N$$

S OP S

$$OP \rightarrow + \mid \times \mid - \mid \div$$

$$V \rightarrow x_1 \mid x_2 \mid x_3$$

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0

0

AN EXAMPLE



$$S \rightarrow (S \text{ } OP \text{ } S) \mid V \mid N$$

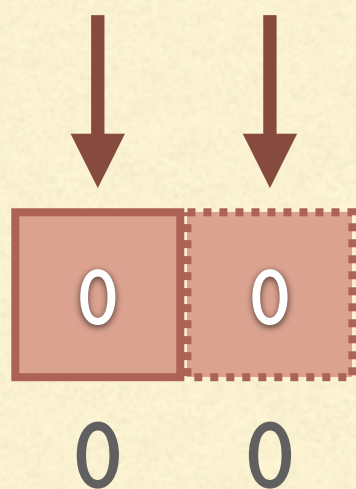
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$$N \rightarrow -DD \mid -D \mid D \mid DD$$

$$D \rightarrow 0 \mid 1 \mid 2 \mid 3 \mid 4 \mid 5 \mid 6 \mid 7 \mid 8 \mid 9$$

$$\begin{array}{c} S \text{ } OP \text{ } S \\ \downarrow \\ (S \text{ } OP \text{ } S) \text{ } OP \text{ } S \end{array}$$



The expansion will never end!

We will simply assign a very bad fitness to this individual

GE PECULIARITIES

- Notice that there is no one-to-one mapping between genotype and the phenotype
 - Multiple genotypes can be expanded to represent the same derivation tree/phenotype
 - Notice that the “meaning” of a part of an individual depends from all the genes before it (since they define in which part of the derivation we are)
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GE: OPERATORS

- We can use the same operators as GA for crossover and mutation
 - We can also use specific operators for crossover:
 - Sensible crossover is a one-point crossover that uses the actual length of an individual
 - Ripple crossover
 - Homologous crossover
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HOMOLOGOUS CROSSOVER

- In homologous crossover we align the history of rule activations in two individuals
- We perform a one-point crossover in the area where they differ



LIBRARIES

- Grammatical evolution is implemented in PonyGE2 library
 - <https://github.com/PonyGE/PonyGE2>
 - You need to clone/download the repository (you cannot use pip or anaconda for the installation)
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