

# Automating Algorithm Design through Genetic Programming Hyper-Heuristics

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Morris, MN

# What does the title mean?

- Reducing the human component in algorithm design



<https://scratch.mit.edu/discuss/m/topic/200574/>

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- Reducing the human component in algorithm design
- More work at the beginning, more possibilities
- Genetic programming hyper-heuristics as a method to the madness



<https://scratch.mit.edu/discuss/m/topic/200574/>

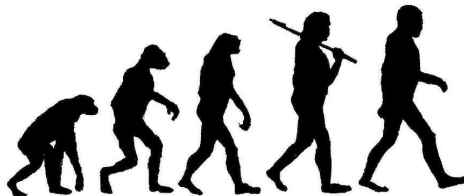
# Outline

- 1 Background
- 2 Hyper-heuristics
- 3 Genetic Programming Variants
- 4 Autoconstruction
- 5 Summary

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- 1 **Background**
  - Evolutionary Computation
  - Genetic Programming
- 2 Hyper-heuristics
- 3 Genetic Programming Variants
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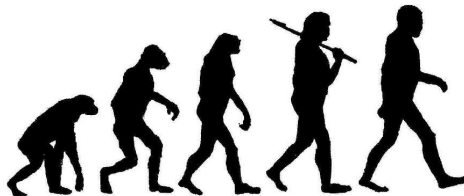
# Evolutionary Computation



<https://www.spigotmc.org/attachments/evolution-jpg.137048/>

- Subfield of Artificial Intelligence

# Evolutionary Computation

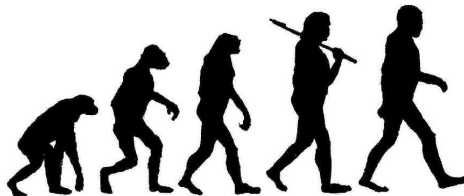


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- Subfield of Artificial Intelligence
- Algorithms based on biological evolution



# Evolutionary Computation



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- Subfield of Artificial Intelligence
- Algorithms based on biological evolution
- Uses lots of terminology from biology, doesn't always mean same thing as term means in biology.

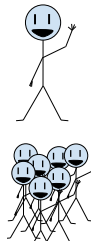
# Evolutionary Computation – Terminology

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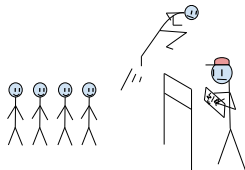
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- **Individual** – a potential solution to a problem (or set of problems)
- **Population** – a group of individuals
- **Fit** – how well suited an individual is at solving a problem
- **Fitness Test** – a set of tests to determine how fit an individual is.



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- **Global optima** – best solution (or solutions) possible

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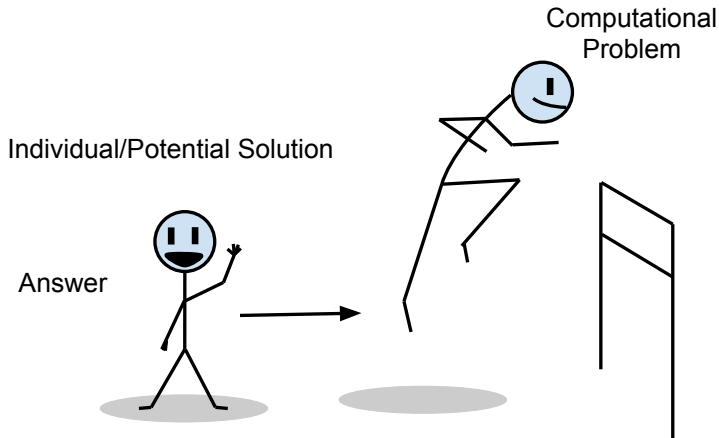
- **Parent** – Individual A
- **Child** – Individual B



# Genetic Programming

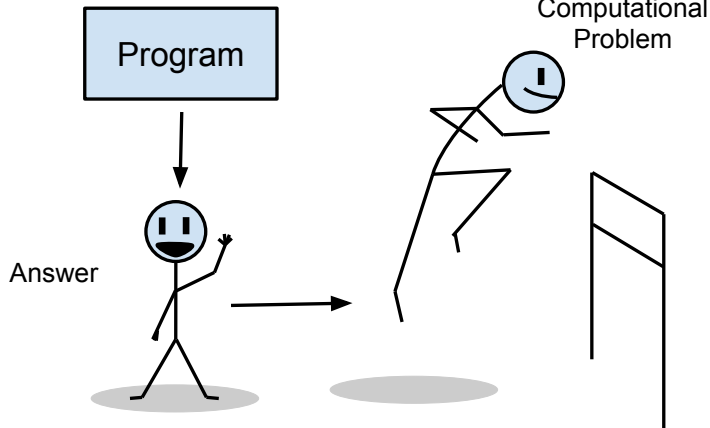
A family of algorithms in Evolutionary Computation that uses biological techniques to create programs to solve computational problems.

# Genetic Programming – structure for evolution



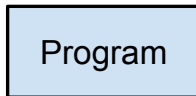
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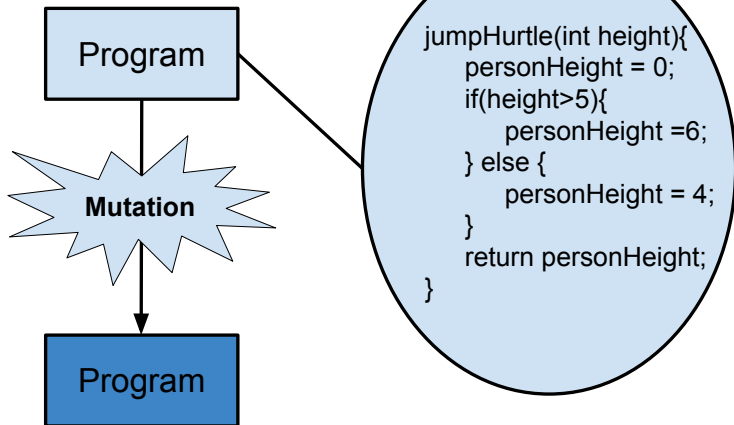


```
jumpHurtle(int height){  
    personHeight = 0;  
    if(height>5){  
        personHeight =6;  
    } else {  
        personHeight = 4;  
    }  
    return personHeight;  
}
```



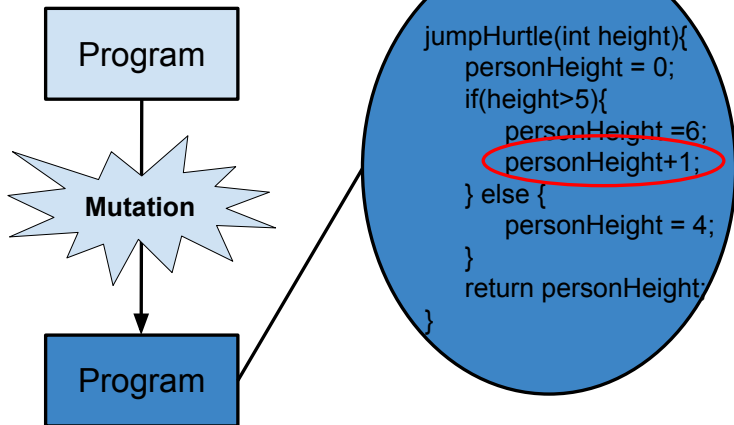
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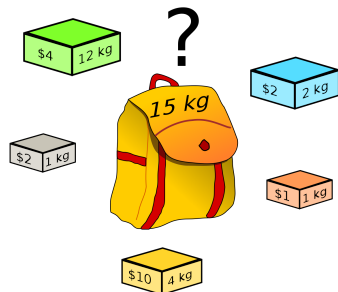
- 1 Background
- 2 Hyper-heuristics**
  - What they are
  - How they work
- 3 Genetic Programming Variants
- 4 Autoconstruction
- 5 Summary

# Heuristic

*Heuristic* – a function that ranks alternatives in a search algorithm at each branching step and uses that information to choose which branch to follow.

# Heuristic

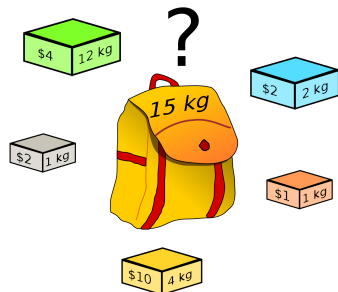
*Heuristic* – a function that ranks alternatives in a search algorithm at each branching step and uses that information to choose which branch to follow.



`https://upload.wikimedia.org/wikipedia/commons/thumb/f/fd/Knapsack.svg/1200px-Knapsack.svg.png`

# Heuristic

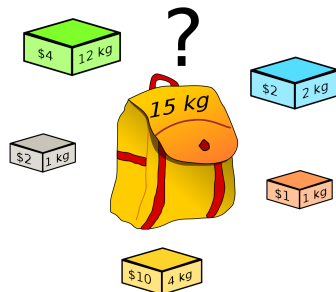
"if knapsack is not full, put largest valued item into knapsack. If this item would cause knapsack to be overweight, take the next highest valued item and put it into the knapsack. Repeat until all items are gone"



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# Heuristic

“if knapsack is not full, put largest valued item into knapsack. If this item would cause knapsack to be overweight, take the next highest valued item and put it into knapsack. Repeat until all items are gone or knapsack is full”



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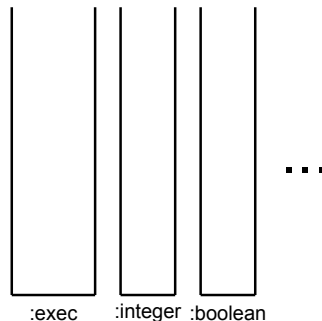
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  - Why they matter
  - Stack-based genetic programming
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# Stack-based genetic programming

Data-stacks are used for managing input and output of operations.

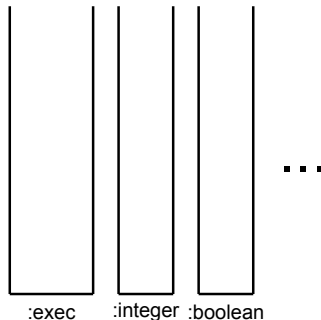


# Stack-based genetic programming

Data-stacks are used for managing input and output of operations.

Programs are represented as linear sequences of literals and instructions. Below is an example of a simple Push program:

```
(1 2 integer_equal)
```

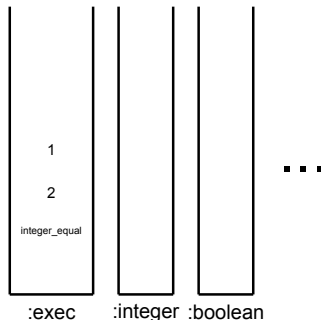


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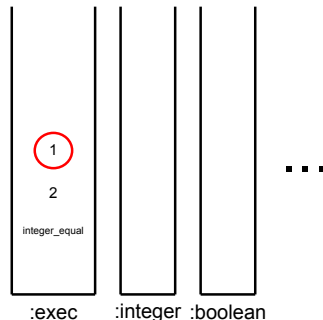


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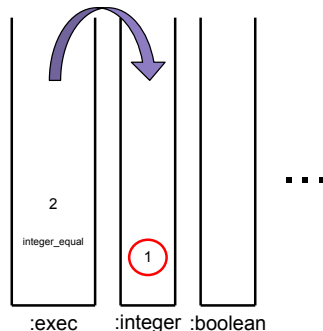


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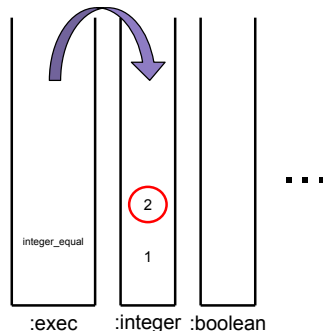


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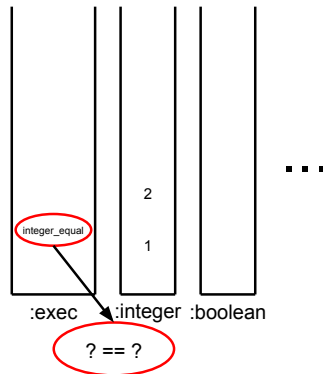


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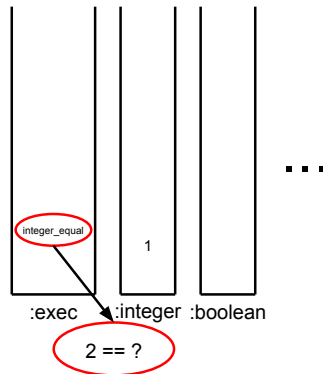


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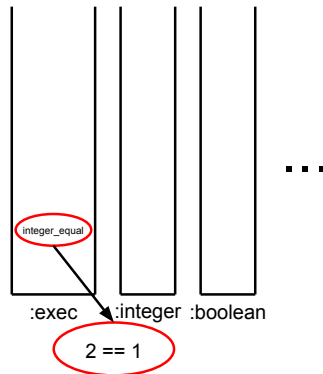


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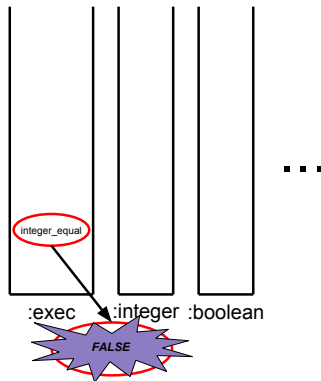


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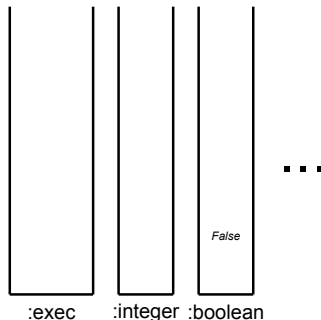


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- 2 Hyper-heuristics
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  - What is it?
  - AutoDoG
  - Results
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# What is Autoconstruction?

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- In most GPHH, the individual programs are evolving, but everything else is specified by the engineer; in autoconstruction, evolution is evolving as well.
- Programs are responsible for evolving solutions *and* responsible for evolving their offspring.



# AutoDoG

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