



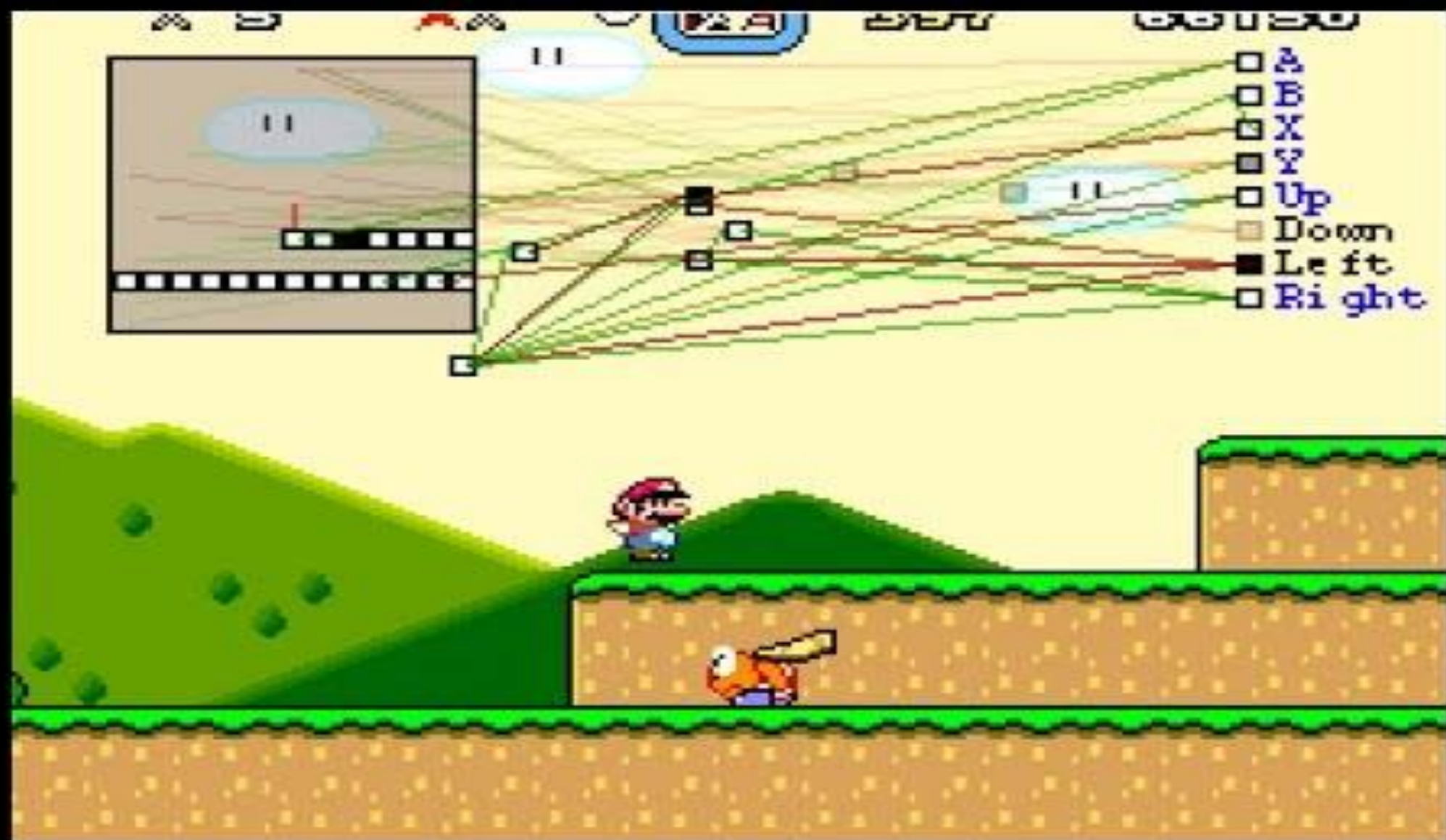
Genetic Algorithms

Presentation by Eli Hodges

Based on the paper by Eli Hodges

What to Expect

- The patrons of genetic algorithms
- How to implement genetic algorithms
- Applications of genetic algorithms in practical contexts

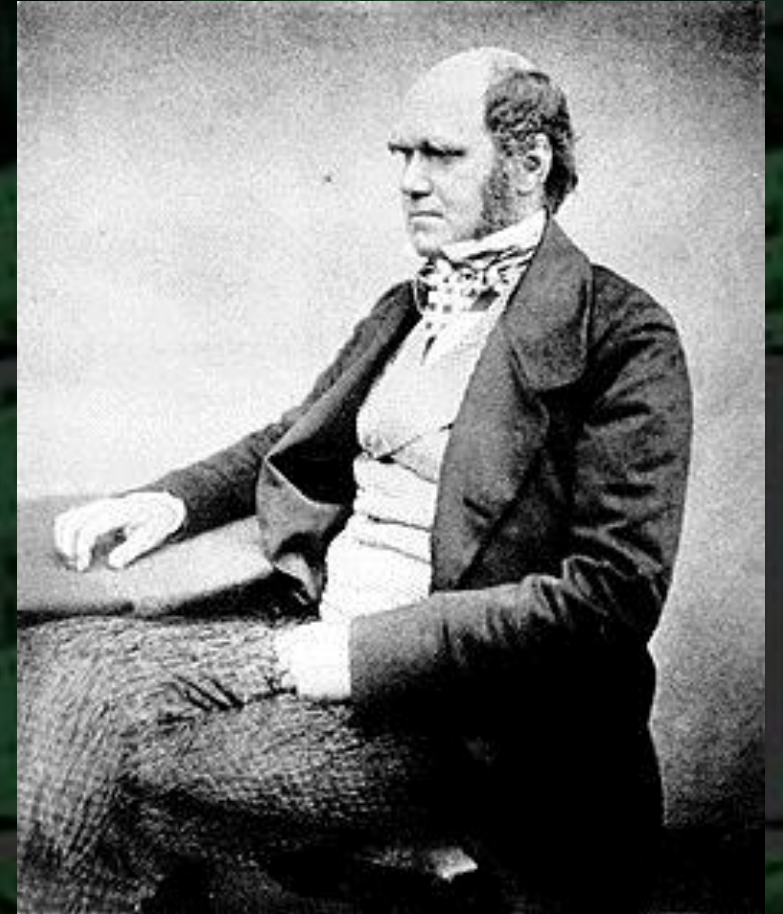


What is a genetic algorithm?

- Optimization search
- Designed to simulate biology using natural selection
 - Mimics key phases of natural selection
- Converges to numerous solutions of equal efficiency

Evolution by Natural Selection

- Presented in the 1859
 - “On the Origin of Species by means of Natural Selection”
- Founded on four principals
 - Variation
 - Overproduction
 - Adaptation
 - Descent with Modification

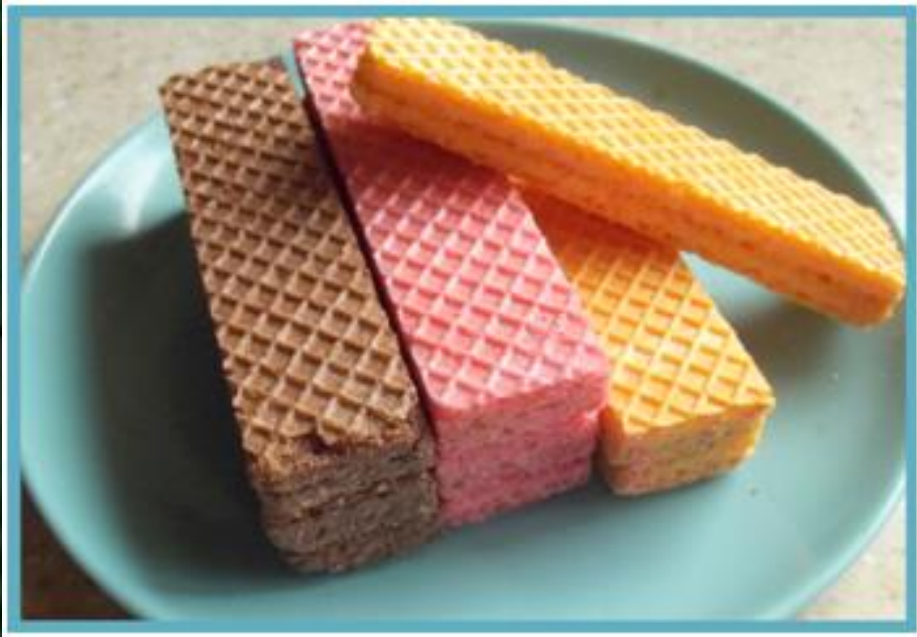


Evolution

The process of changing through time. Modern species are the result of millennia of small changes driven by natural selection.

Natural selection

- A process of natural elimination
- Organisms are selected to continue their lineage based on traits that make them more fit for their current environment
- Survival of the fittest
 - ... Of the given set.



Variation

- Variation exists within the population of all organisms
- Multiple genetic characteristics allow organisms to adapt to various situations
- Nature selects for or against specific genetic characteristics.

Overproduction

- Each species in a population exceeds its sustainable size within a particular environment or habitat.
- A result of increased birthrate or reduced deathrate



Adaptation

- Considered the result of natural selection
- Unfit individuals are culled until only adapted organisms remain



Descent with Modification

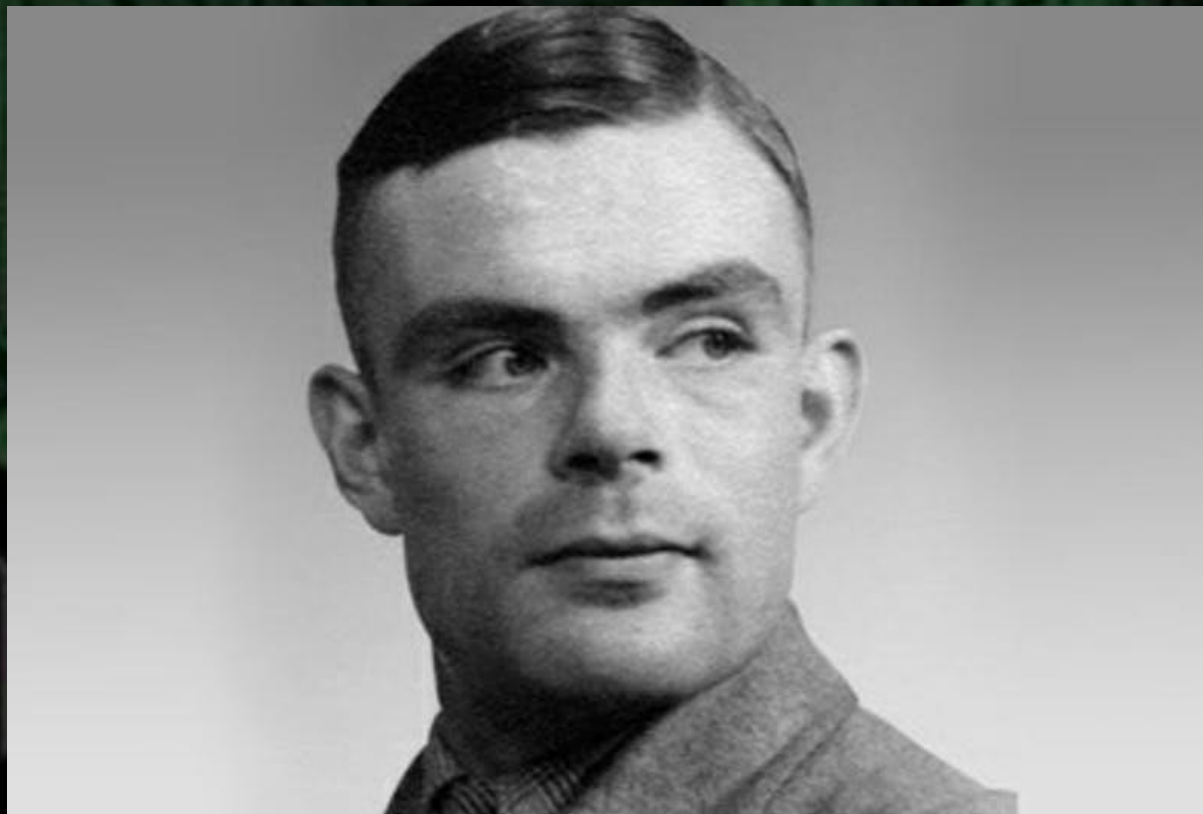
- The passage of traits from parent to offspring
- The mechanic of which evolution 'actually happens'





The History of Genetic Algorithms

Alan Turing



Alan Turing

- First to mention evolution in a computational context
 - In “Computing Machinery and Intelligence”
 - As a response to Ada Lovelace
 - Was a result of a thought experiment.
 - Tangential to the purpose of the paper

Alan Turing

- Compared an ideal mechanical brain to an “atomic pile of super-critical size”
- Natural selection as a model
 - Structure of the child machine --- Hereditary Material
 - Changes in structure --- Mutations in nature
 - Natural Selection --- Judgement of the Experimenter
- Concept was completely mechanical, no automation involved

Nils Aall
Barricelli



Nils Aal Barricelli

- Attempted to simulate evolution
- Used punch card programming
- Emulated random number generation by shuffling decks of cards

Alex Fraser



Alex Fraser

- Simulated evolution to the same effect as Barricelli
- Garnered much more acclaim for his work
- Tuned the selection phase to select for a specific trait

Hans-Joachim
Bremermann



Hans-Joachim Bremermann

- Considered natural selection from a problem solving context
- Initial population of solutions
- Bremmermans' limit

Ingo
Rechenberg



and Hans-
Paul Schwefel



Ingo Rechenberg and Hans-Paul Schwefel

- Work was done independently, but with similar conclusions
- Developed “Evolutionary Strategies”
 - Solved complex engineering problems

PROCEEDINGS OF THE
FIRST INTERNATIONAL CONFERENCE ON
GENETIC ALGORITHMS
AND THEIR APPLICATIONS

July 24-26, 1985
at the
Carnegie-Mellon University
Pittsburgh, PA

Sponsored By
Texas Instruments, Inc.
Naval Research Laboratory

John J. Grefenstette
Editor

 Psychology Press
Taylor & Francis Group

-1985- First international Conference on Genetic Algorithms

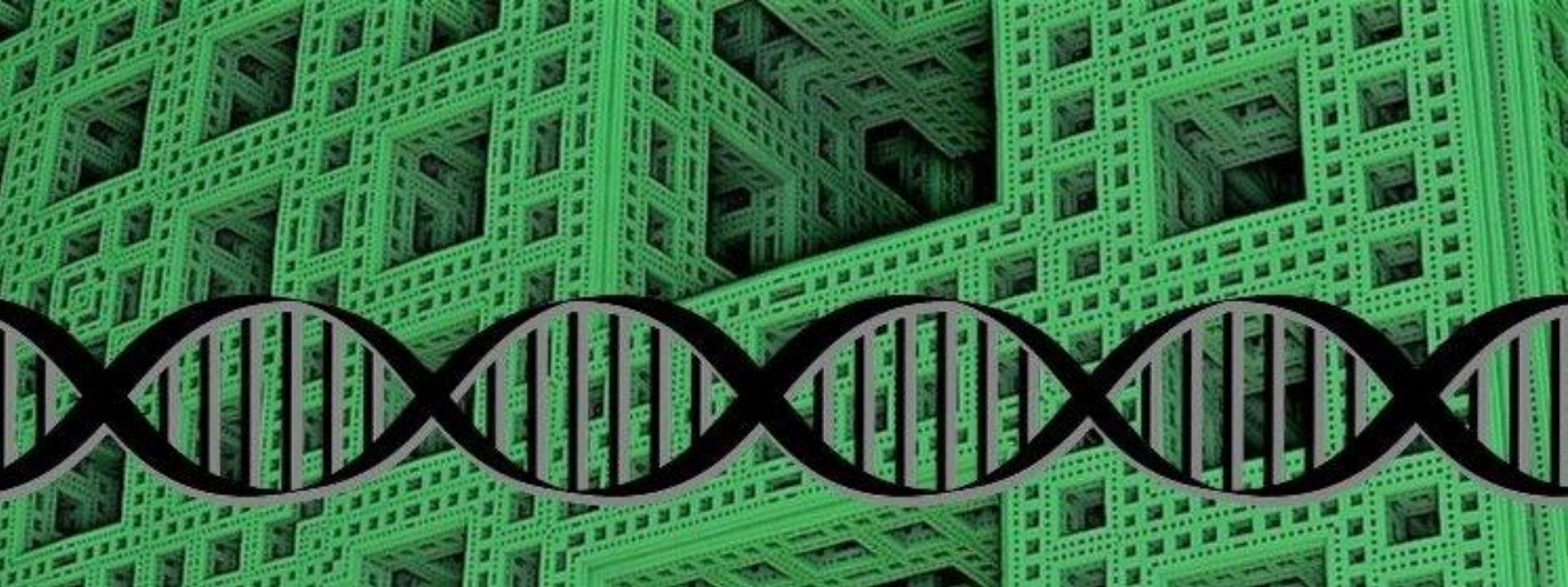


Selections and
Corrections

Implementation

Like parent, like child

- Intended to mechanically simulate evolution to a purpose
- Segmented into several distinct phases
 - Initialization of population
 - The Fitness Function
 - Selection
 - Crossover
 - Mutation



Vocabulary

Individual

- In Biology: A single, separate organism distinguished from others of a same kind
- In our context: An individual solution distinguished from other solutions though its derived tactics
- In both: Characterized by genes organized into chromosomes

Gene

- In Biology: A structure of nucleotide 'tuples' that parameterize genetic information
- In our context: A single value, usually binary, that parameterizes synthetic genetic information
- In both: Strung together to construct chromosomes

Chromosome

- In Biology: A string of genes with part or all of an individual's genetic material
- In our context: A string of genes that contain all genes associated with the given solution
- In both: Split and recombined to pass genetic information to children

Population

- In Biology: A group of individuals that interbreed and live in the same place at the same time
- In our context: A collection of individuals comprising a given solution set
- In both: A combined collection of individuals in a given context

A1

0	0	0	0	0	0
---	---	---	---	---	---

Gene

A2

1	1	1	1	1	1
---	---	---	---	---	---

Chromosome

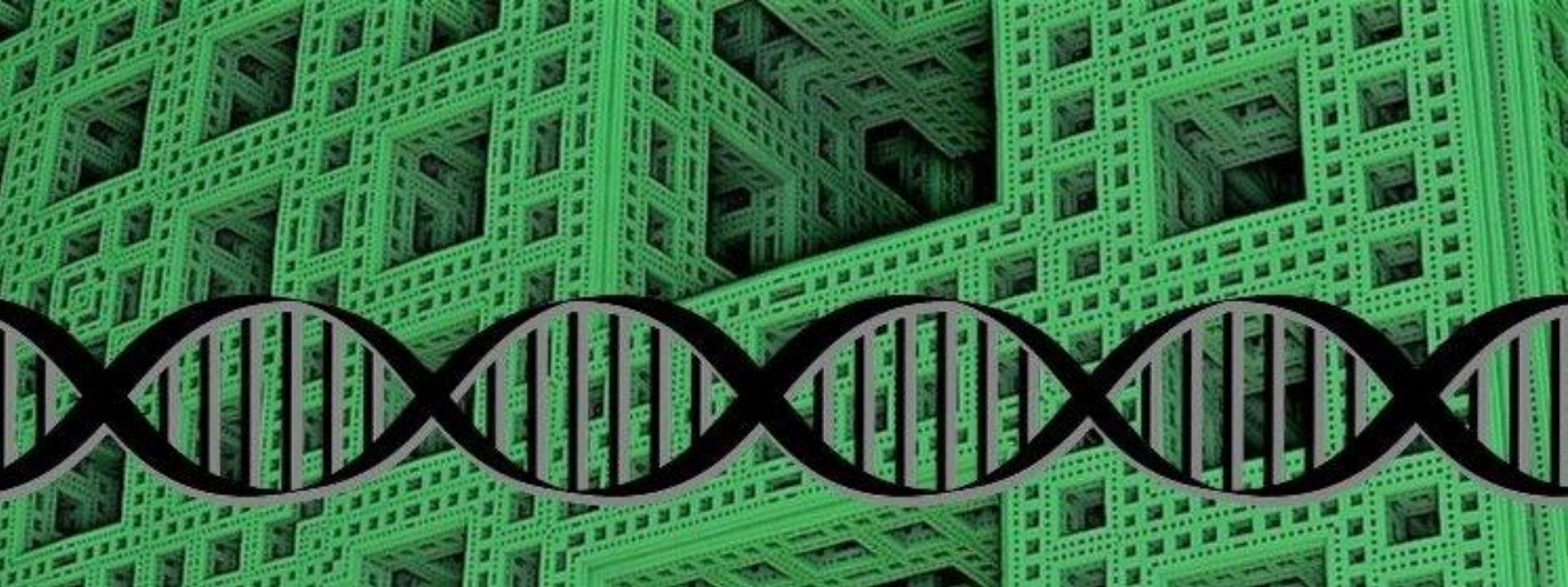
A3

1	0	1	0	1	1
---	---	---	---	---	---

A4

1	1	0	1	1	0
---	---	---	---	---	---

Population



The Fitness Function

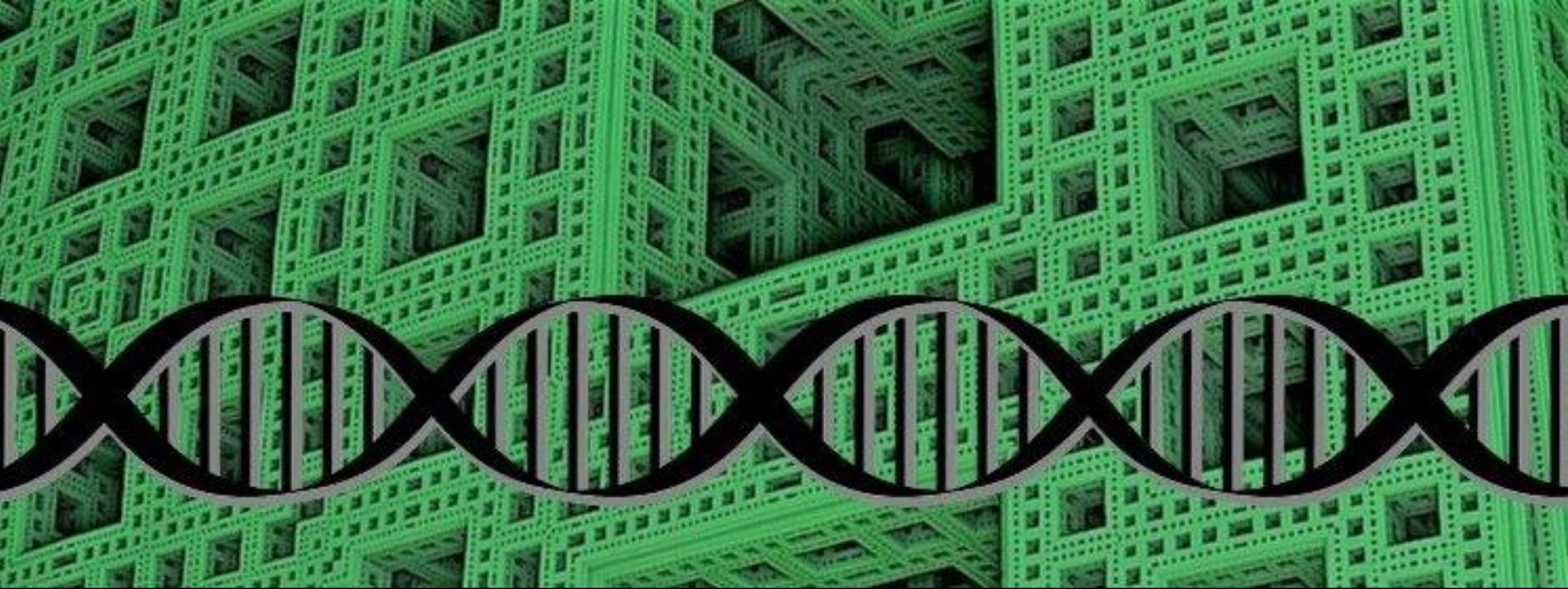
Fitness

- Determines how successful a given solution is at problem completion
- Uniquely implemented for each problem set

Gen 0 species 71 genome 1 (23%)
Fitness: 528 Max Fitness: 528



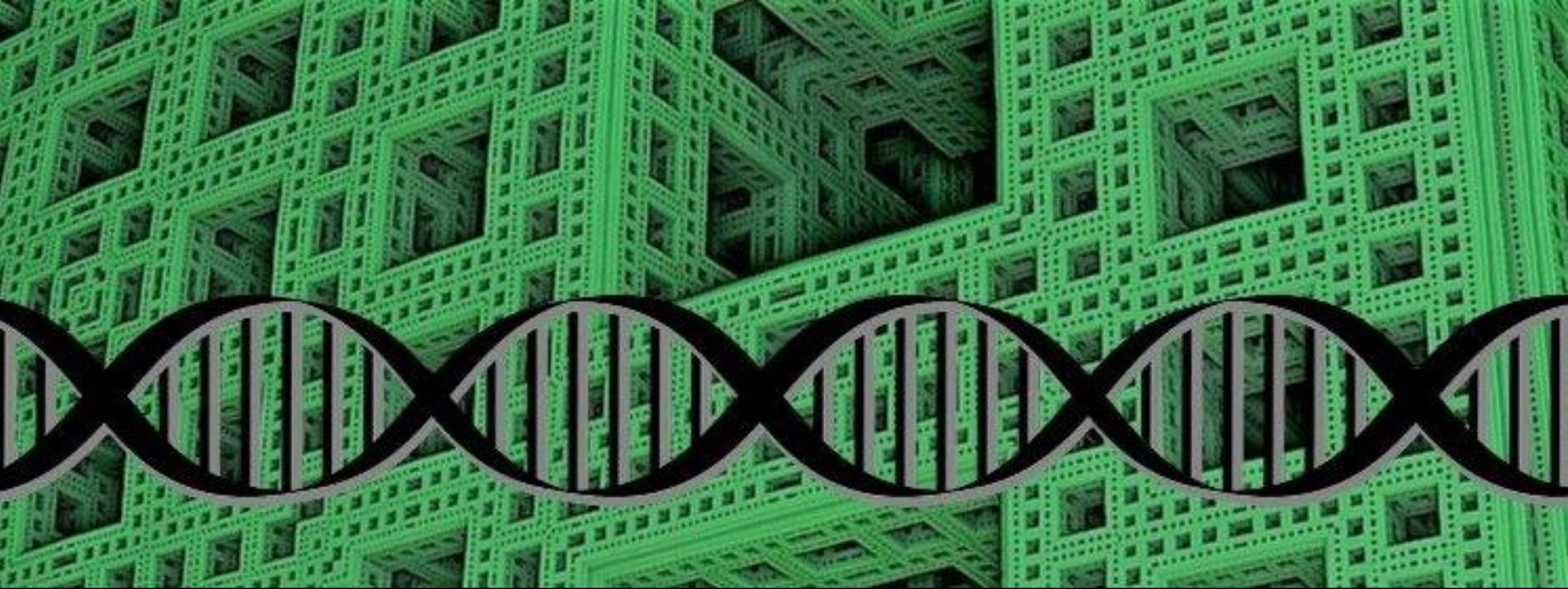
Fitness



Selection

Selection

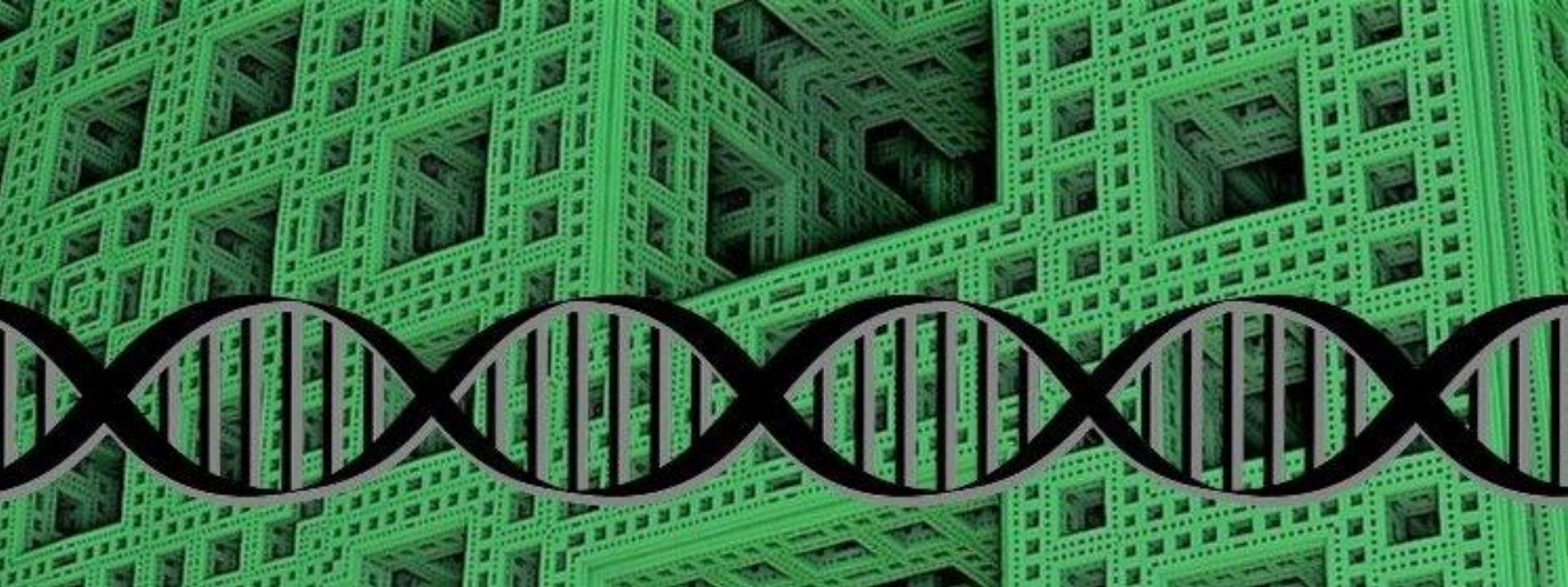
- A result of fitness
- Probabilistic
 - Higher fitness scores have a higher probability of selection
- Non-orthogenetic without heuristics
 - Desirable traits –tend- to have higher fitness score



Vocabulary Lightning Round

Parents

- In Biology: Two individuals who have conceived/sired a child and whose genes have therefore transmitted to the child
- In our context: Two individuals who have been assigned each other, and together progress to the crossover phase
- In both: Pairs of individuals whose genes are passed on to the next generation of the population

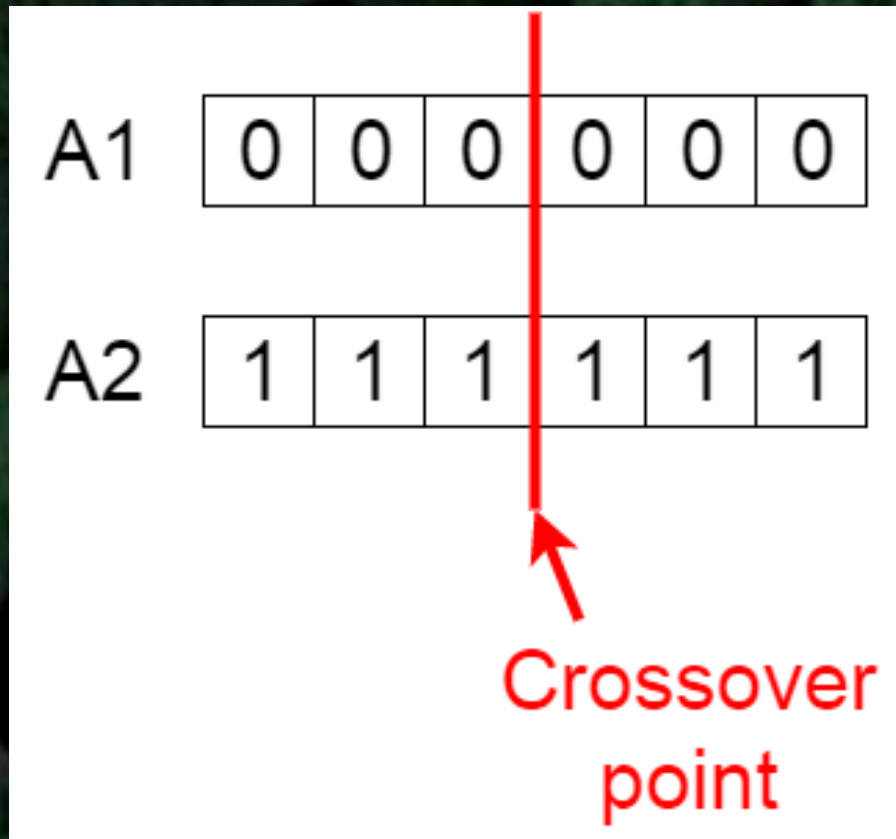


Crossover

Crossover

- The most important phase of the genetic algorithm process
- Crossover point is chosen at random

Crossover



Crossover

- Two children are each given half of their parents genes

Crossover

A1

0

0

0

0

0

0

A2

1

1

1

1

1

1



Crossover

- The parents are removed from the population
- The children replace their parents

Crossover

A5

1

1

1

0

0

0

A6

0

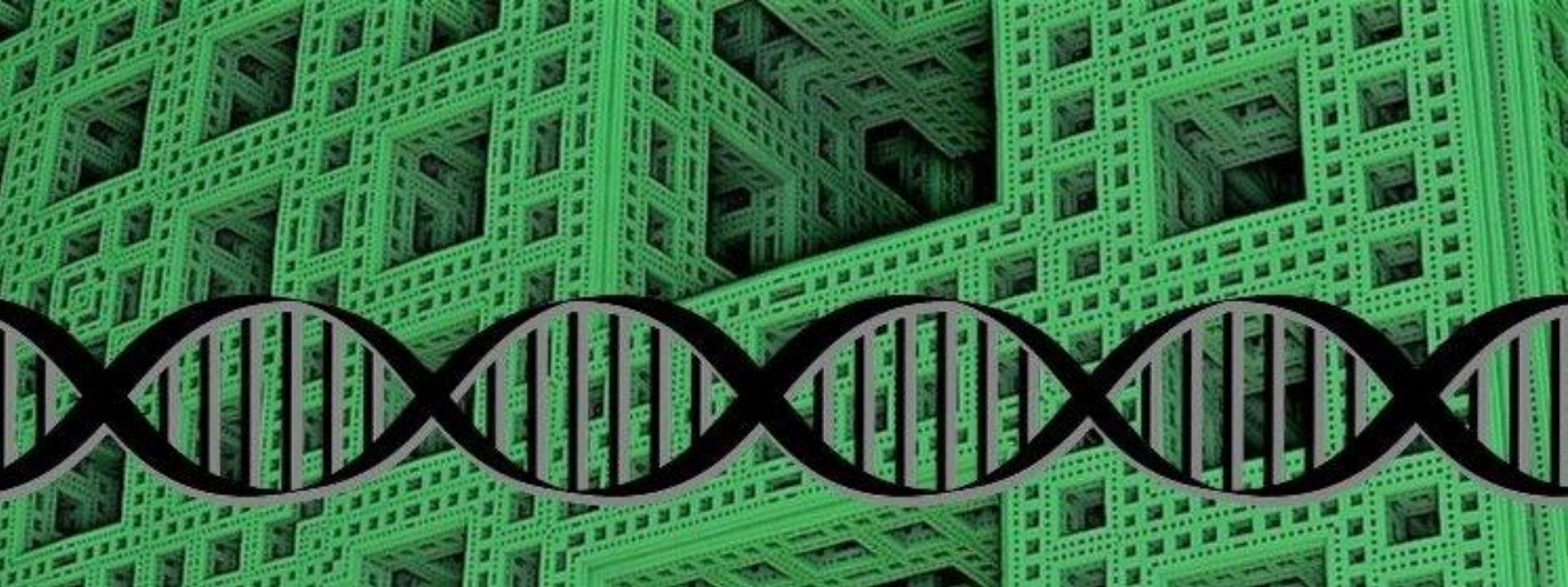
0

0

1

1

1



Mutation

- Occurs probabilistically at a rate determined by the developer

Mutation

Before Mutation

A5	1	1	1	0	0	0
----	---	---	---	---	---	---

After Mutation

A5	1	1	0	1	1	0
----	---	---	---	---	---	---

Setting the Mutation Rate

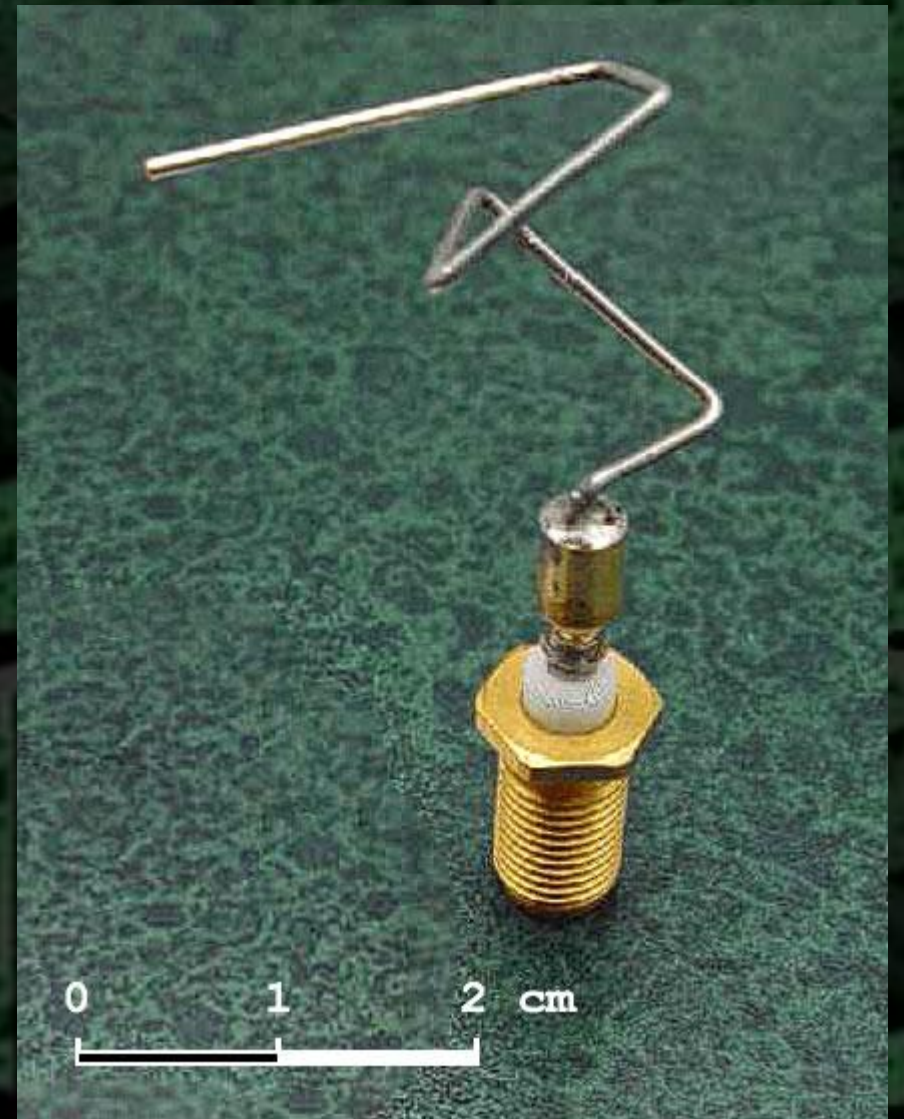
- If the rate is too high, it can discard rare and valuable solutions
- If the rate is too low, it can cause limited diversity.
 - Early convergence
- Important to uncover solutions that haven't been considered



Evolution by
Design

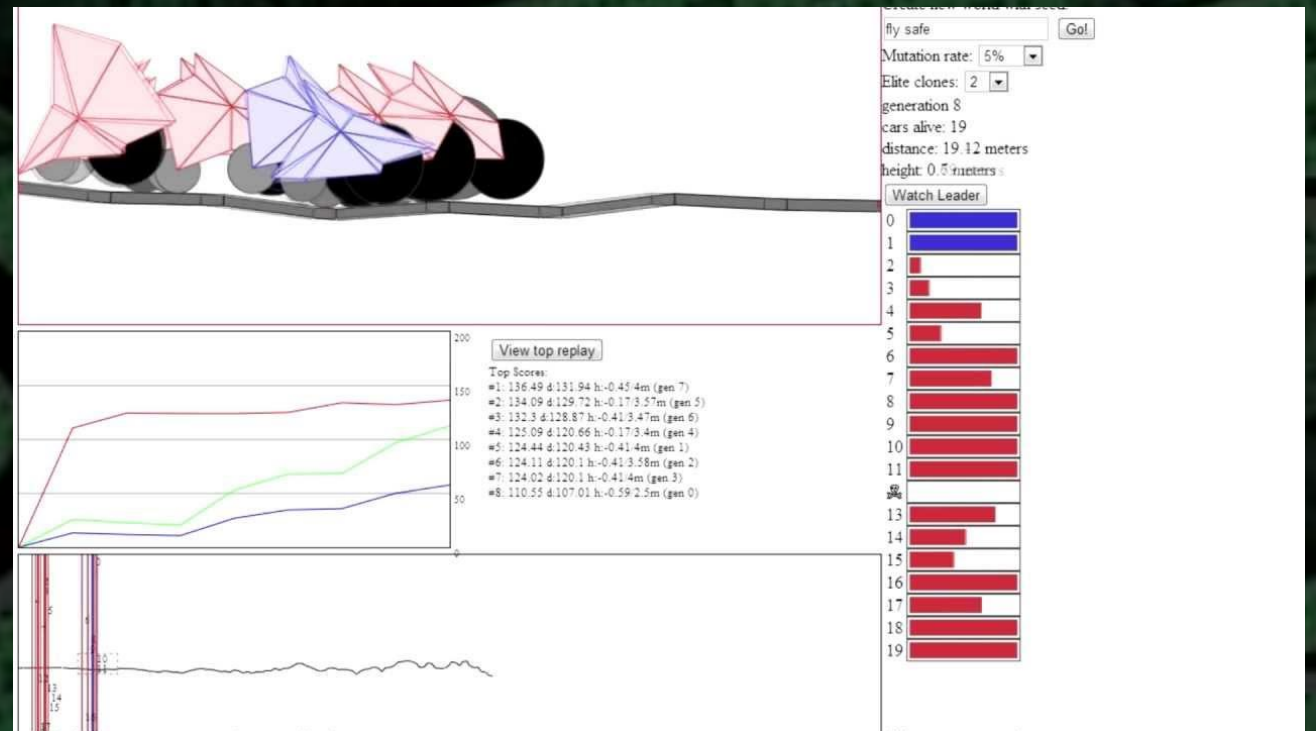
Practical Applications

- 2006 NASA 'Evolved Antenna



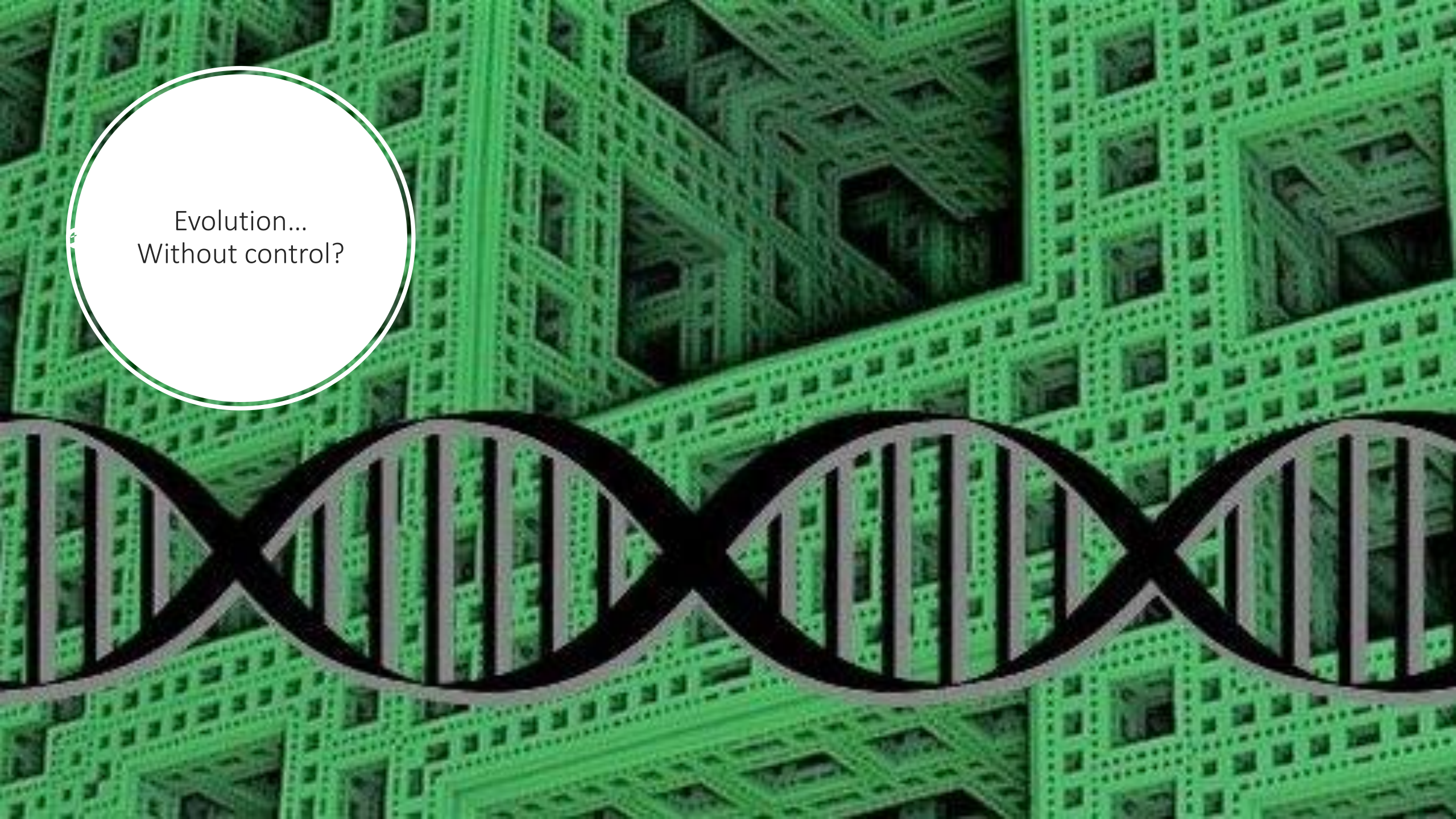
Practical Applications

- https://rednuht.org/genetic_cars_2/



Other Practical Applications

- Polymer design
- Vehicle body structuring
- Video game strategy generation
- Encryption generation
- Logistical route building
- Market Forecasting...
- General Purpose AI... ?



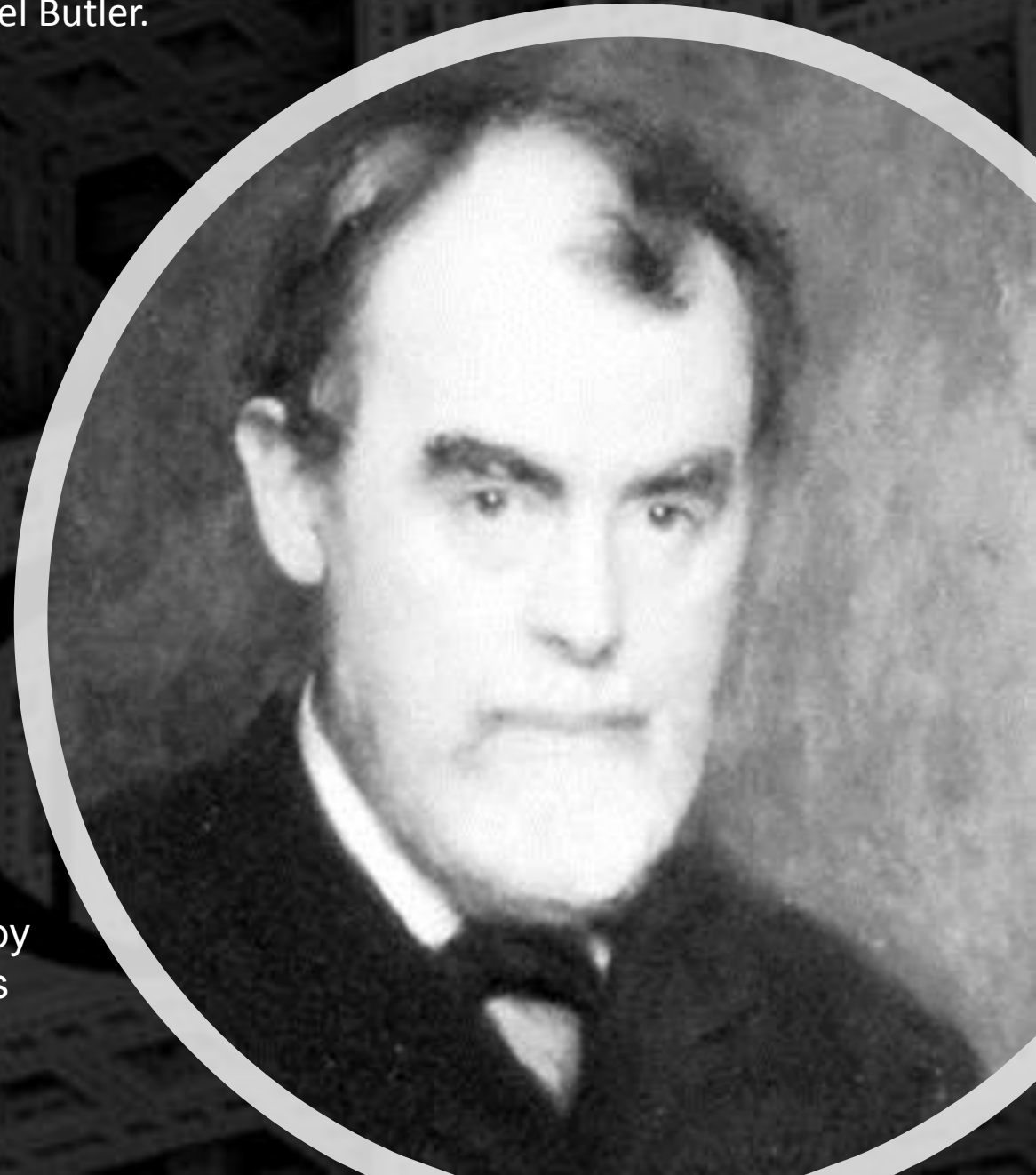
Evolution...
Without control?

“Darwin among the machines” 1963, Samuel Butler.

Christchurch, New Zealand

Day by day, however, the machines are gaining ground upon us... but that the time will come when the machines will hold true supremacy over us is what no person of a truly philosophic mind can for a moment question

War to the death should be instantly proclaimed against them. Every machine of every sort should be destroyed by the well-wisher of his species. Let there be no exceptions made, no quarter shown; let us at once go back to the primeval condition of the race.

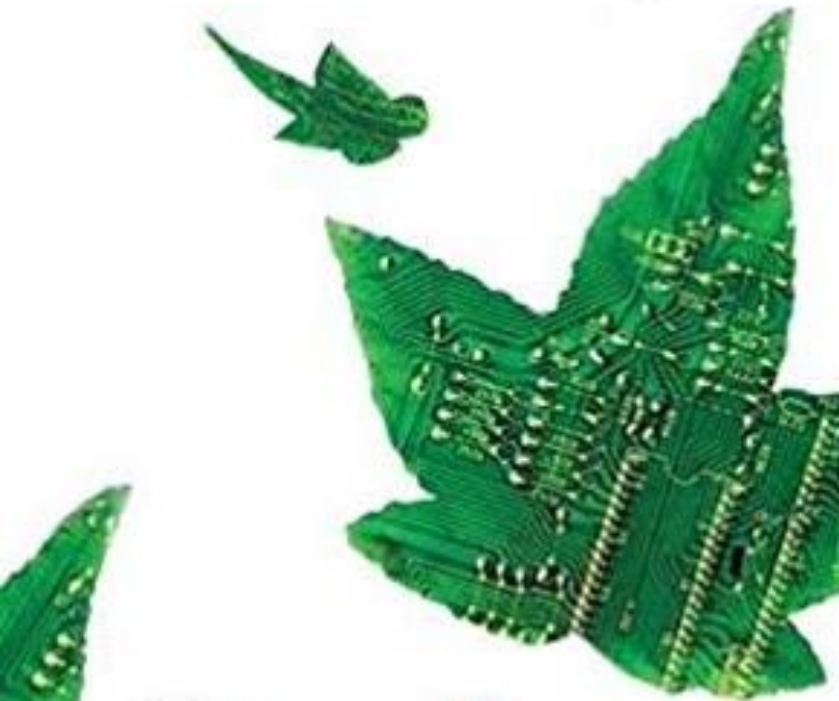


'One of the most original and intriguing
books of the last two decades' —*GUARDIAN*



DARWIN among the MACHINES

The Evolution of Global Intelligence



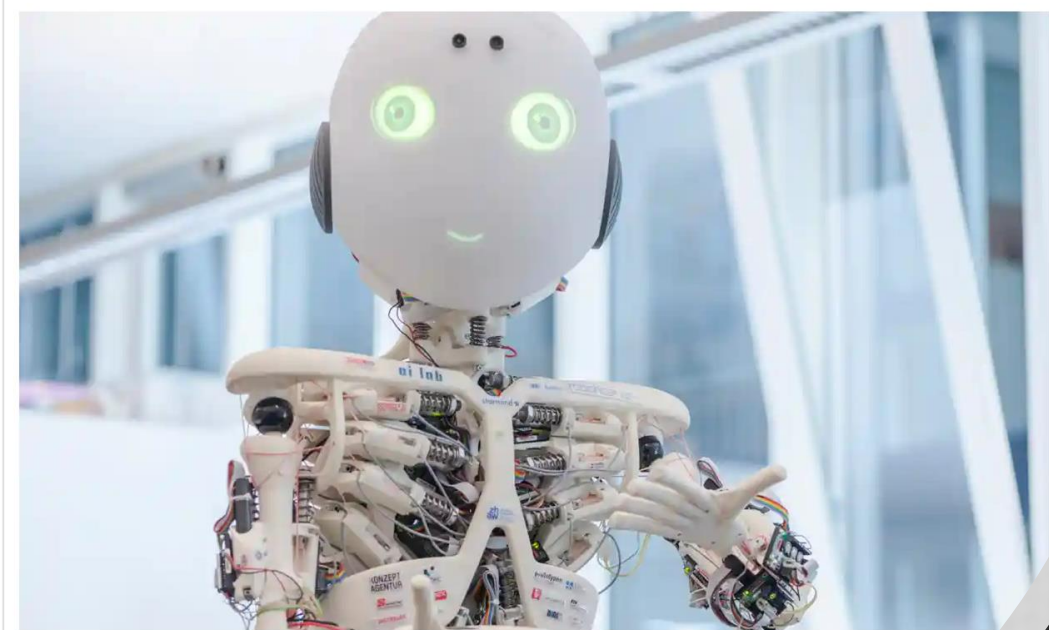
George Dyson

The background of the entire slide is a green circuit board. A large, semi-transparent DNA double helix is overlaid on the circuit board, running diagonally from the bottom left towards the top right.

What mind, if any, will become
apprehensive of the great
coiling of ideas now under way
is not a meaningless question,
but it is still too early in the
game to expect an answer that
is meaningful to us

Robots could learn human values by reading stories, research suggests

Scientists have been running tests where artificial intelligences cultivate appropriate social behaviour by responding to simple narratives



▲ Kindness calculus ... the ROBOY humanoid robot (not involved in the Quixote experiment). Photograph: F. Thom/Corbis

More than 70 years ago, Isaac Asimov dreamed up his three laws of robotics, which insisted, above all, that “a robot may not injure a human being or, through inaction, allow a human being to come to harm.” He also stated that “the development of full artificial intelligence requires a change in the way we think about the world.”

- <https://www.theguardian.com/books/2016/feb/18/robots-could-learn-human-values-by-reading-stories-research-suggests>



Conclusion