

key events coming up

- **Labs: 35% (ISE-483)**
 - Complete 5 (best 4 graded) assignments based on algorithms presented in class
 - Lab 4 : April 8th
 - Evolutionary Algorithms, (Assignment 4)
 - Delivered by SSIE583 Group 2
 - Due April 15th
 - Lab 5: April 29th
 - Ant Clustering Algorithm, (Assignment 5)
 - Delivered by Group 1
 - Due May 6th
- **SSIE – 583 -Presentation and Discussion: 25%**
 - Present and lead the discussion of an article related to the class materials
 - Enginet students post/send video or join by Zoom
 - April 25th or April 29th
 - Conrad, M. [1990]. "The geometry of evolution." *Biosystems* 24: 61-81.
 - Mario Franco
 - Stanley, Kenneth O., Jeff Clune, Joel Lehman, and Risto Miikkulainen. "Designing Neural Networks through Neuroevolution." *Nature Machine Intelligence* 1, no. 1 (January 2019): 24–35.
 - Jessica Lasebikan
 - Lindgren, K. [1991]. "Evolutionary Phenomena in Simple Dynamics." In: *Artificial Life II*. Langton et al (Eds). Addison-wesley, pp. 295-312.
 - Akshay Gangadhar
 - Salahshour, Mohammad. "Interaction between Games Give Rise to the Evolution of Moral Norms of Cooperation." *PLOS Computational Biology* 18, no. 9 (September 29, 2022): e1010429
 - Srikanth Iyer
 - Discussion by all



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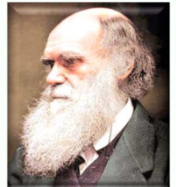
until now

■ Class Book

- Floreano, D. and C. Mattiussi [2008]. *Bio-Inspired Artificial Intelligence: Theories, Methods, and Technologies*. MIT Press. Preface, **Chapters 1 and 4**.

■ Lecture notes

- Chapter 1: What is Life?
- Chapter 2: The logical Mechanisms of Life
- Chapter 3: Formalizing and Modeling the World
- Chapter 4: Self-Organization and Emergent Complex Behavior
- Chapter 5: Reality is Stranger than Fiction
- Chapter 6: Von Neumann and Natural Selection
 - posted online @ <http://informatics.indiana.edu/rocha/i-bic>



■ Papers and other materials

- Optional
 - Nunes de Castro, Leandro [2006]. *Fundamentals of Natural Computing: Basic Concepts, Algorithms, and Applications*. Chapman & Hall.
 - Chapter 2, 7, 8
 - **Chapter 3, sections 3.1 to 3.5**



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■ Projects

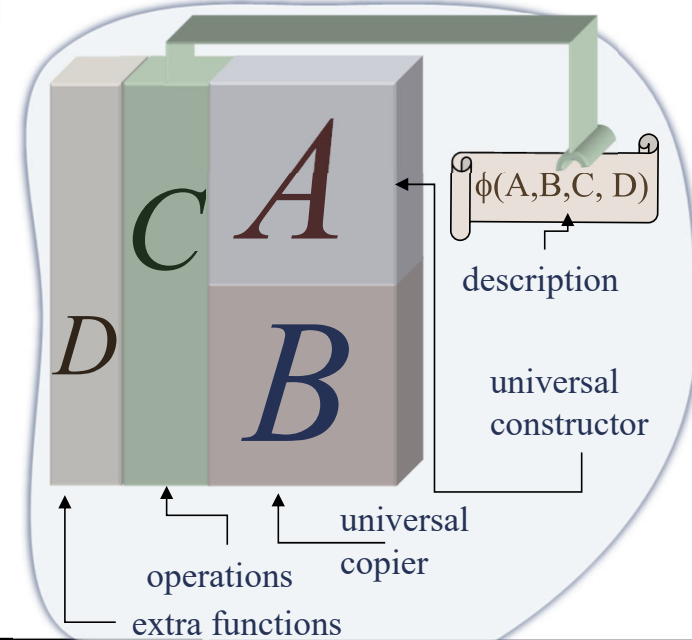
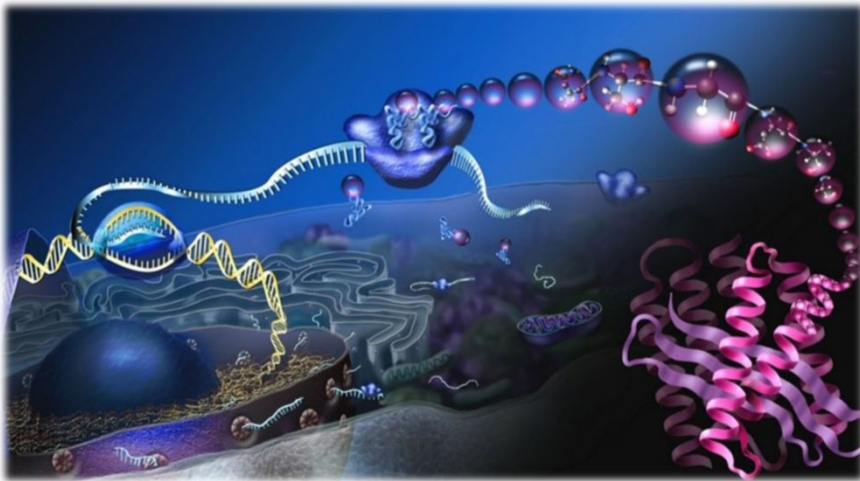
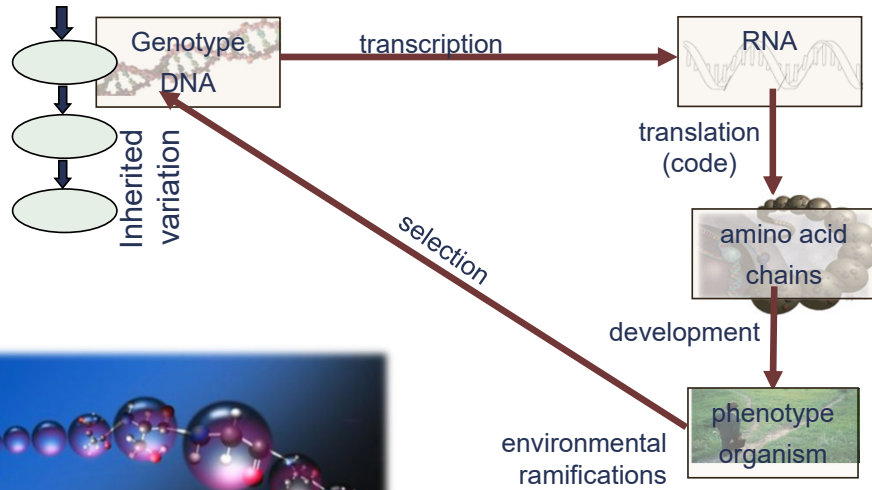
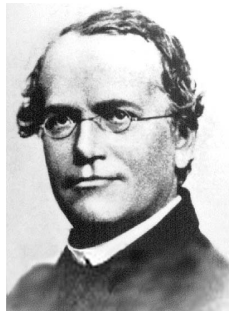
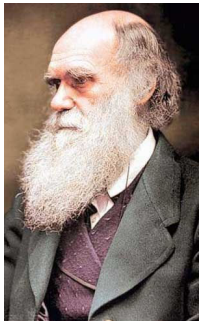
- Due by **May 8th** in Brightspace, “Final Project Paper” assignment
 - ALIFE 2023
 - Not to submit to actual conference due date (April 3rd , 2024)
 - <https://2024.alife.org/>
 - 8 pages, author guidelines:
 - https://2024.alife.org/call_paper.html
 - MS Word and Latex/Overleaf templates
 - Preliminary ideas **by March 15**
 - Submit to “Project Idea” assignment in Brightspace.
- Individual or group
 - With very definite tasks assigned per member of group

ALIFE 2024

Tackle a real problem using bio-inspired algorithms, such as those used in the labs.

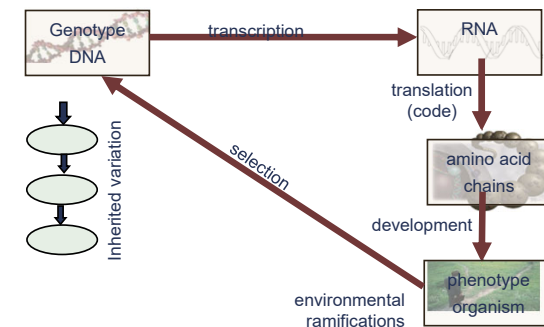
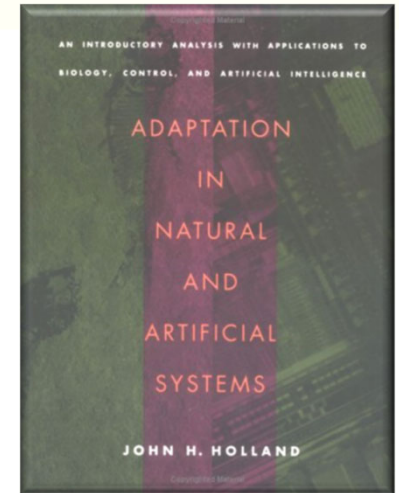


from cybernetic mechanisms to bio-inspired algorithms



History of evolutionary computation

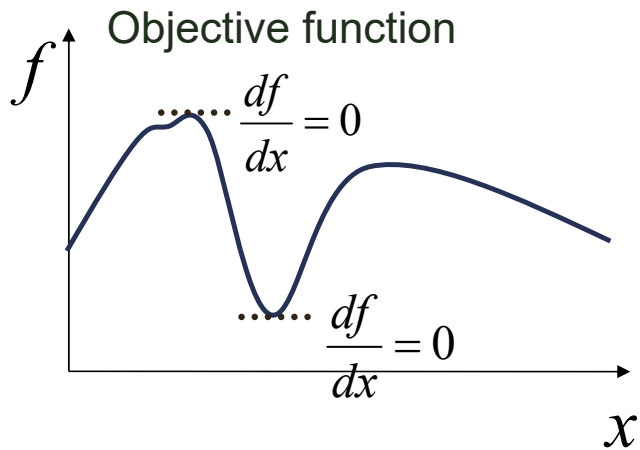
- **Evolutionary Operation**
 - Box (1957)
 - Perturbations to continuous variables followed by selection to improve industrial productivity
- **Evolution Strategies**
 - Rechenberg (1960's), Schwefel (1970's)
 - To optimize real-valued parameters in wind-tunnel experiments
 - Real-valued genotypes under variation and selection
- **Evolutionary Programming**
 - Fogel, Owens, and Walsh (1966)
 - Evolution of tables of state-transition functions (diagrams) under mutation and selection
- **Artificial ecosystems**
 - Conrad and Pattee (1970)
 - Population of artificial cells evolving with genotype and phenotype
- **Other early evolution-inspired algorithms and models**
 - Barricelli CA-like model(1957), game-strategy model (1963)
 - Symbiogenetic evolution
 - Friedman (1957, 1959), Bledsoe (1961), Bremmerrmann (1962)
- **Genetic Algorithms**
 - John Holland (1960's and 1970's)
 - *Adaptation in Natural and Artificial Systems*, University of Michigan Press, 1975. (MIT Press, second edition 1992)



via genetic algorithms

- Search algorithms based on the mechanics of Natural Selection
 - Holland, Conrad, Fogel
 - Based on distinction between a machine and a description of a machine
 - Solution alternatives for optimization problems

Direct analysis depends on
 Knowing the function
 Existence of derivatives
 continuity

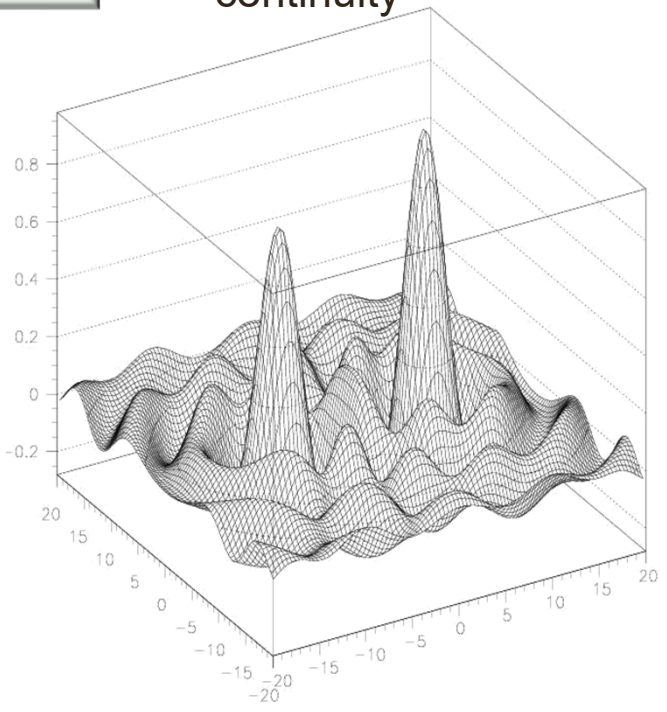


“hill-climbing”

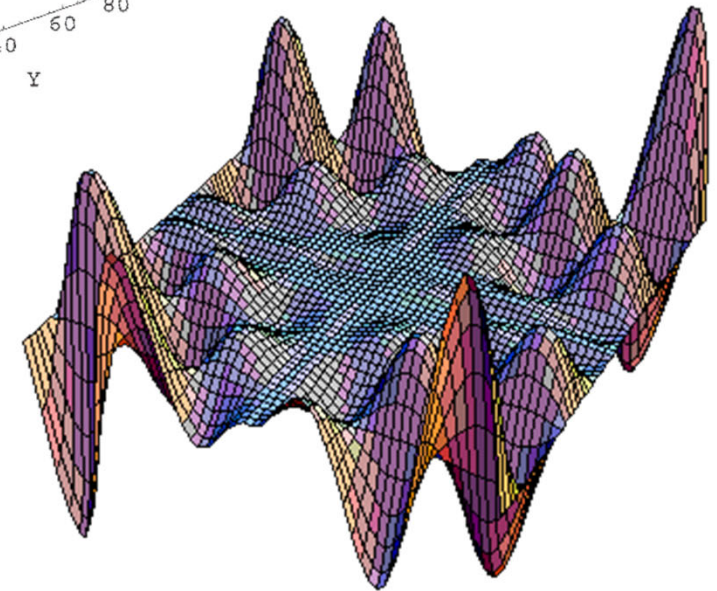
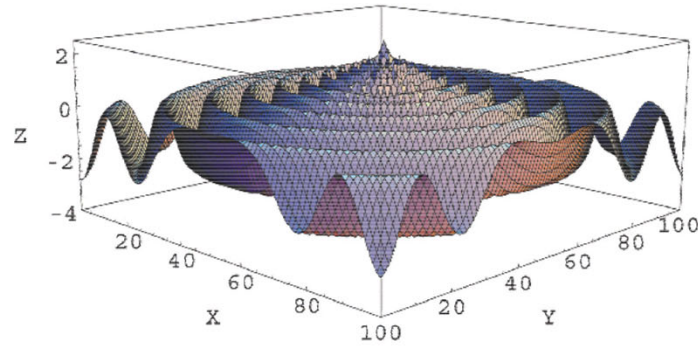
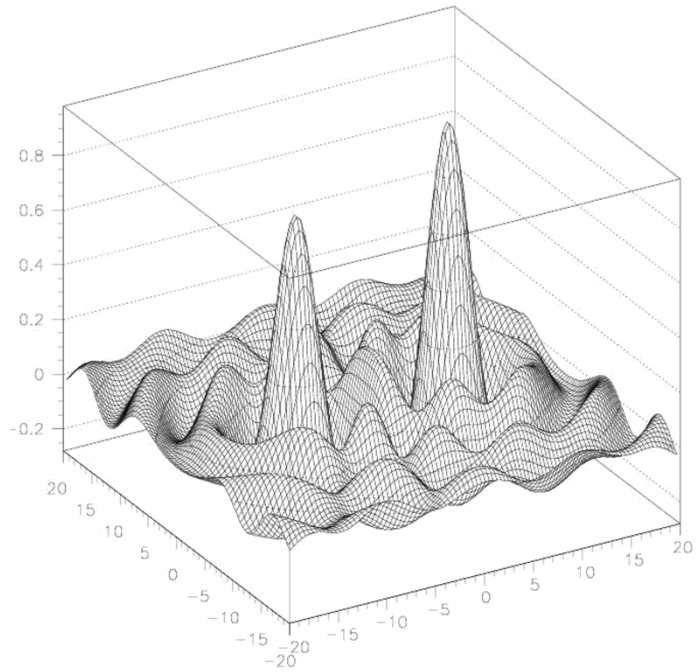
“hop” on the function and
 move along the steepest
 direction until a local
 extrema is found

Random Search
 directionless

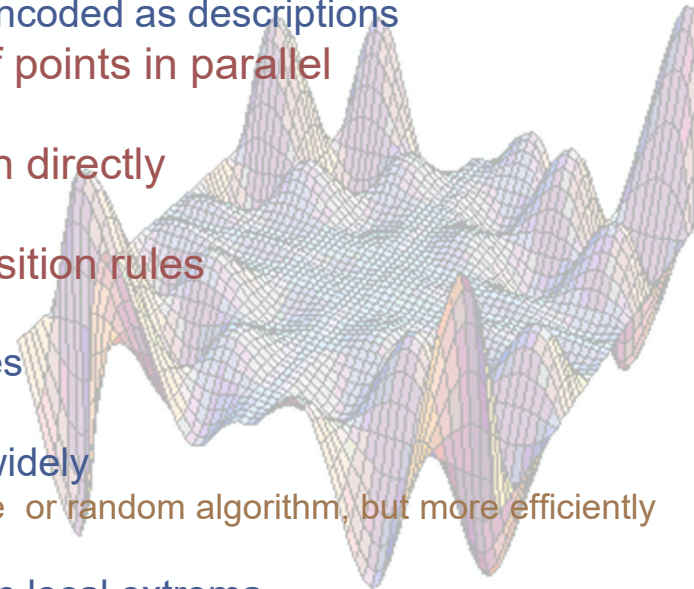
Enumerative Search
 Search point by point



examples

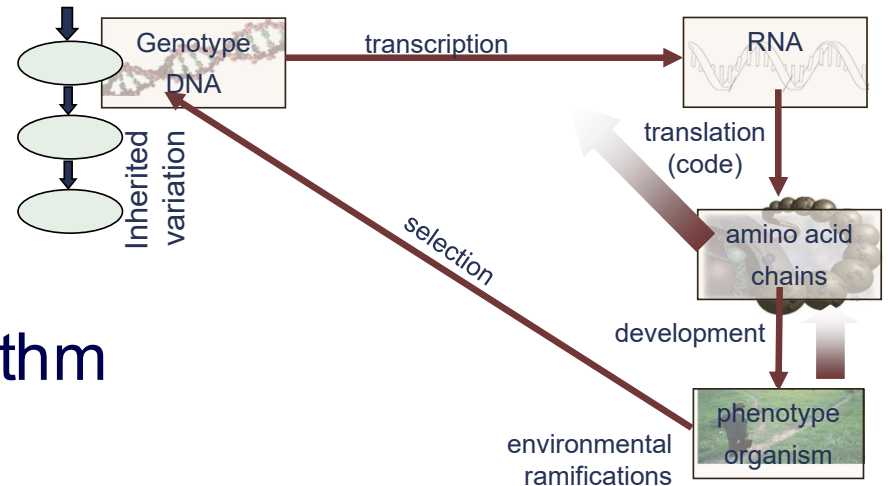


- Work with an encoding of the parameter set, not the parameters themselves
 - solution alternative encoded as descriptions
- Search a population of points in parallel
 - Not a single point
- Uses objective function directly
 - Not derivatives
- Uses probabilistic transition rules
 - Statistical bias
 - Not deterministic rules
- Advantages
 - Samples the space widely
 - like an enumerative or random algorithm, but more efficiently
 - Directed search
 - Not so easily stuck on local extrema
 - Population search
 - Variation mechanisms to search new points

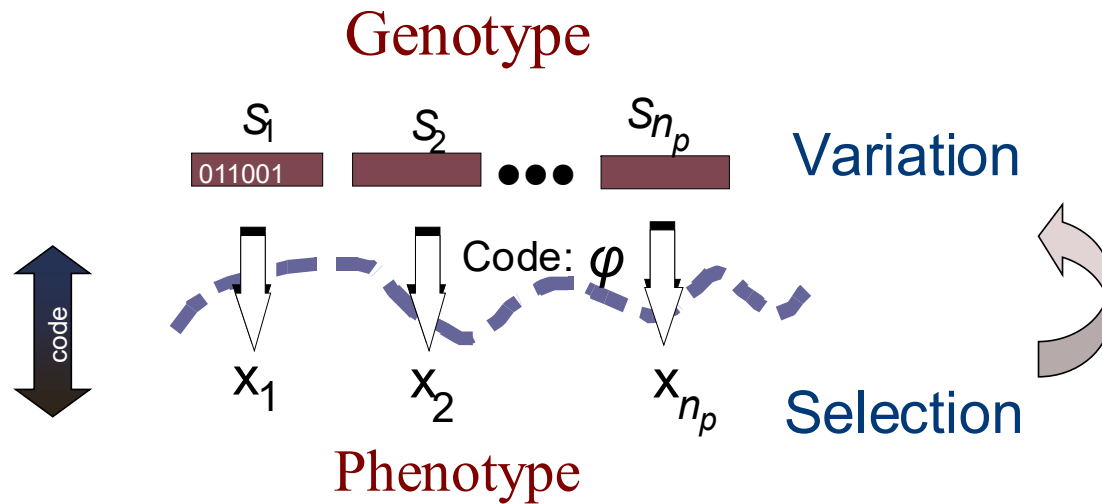


artificial genotype/phenotype mapping

Search algorithms based on the mechanics of Natural Selection
 Based on distinction between a machine and a description of a machine
 Solution alternatives for optimization problems

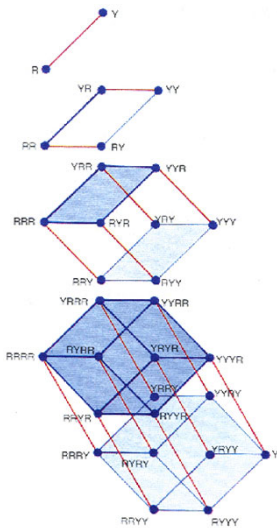


Traditional Genetic Algorithm

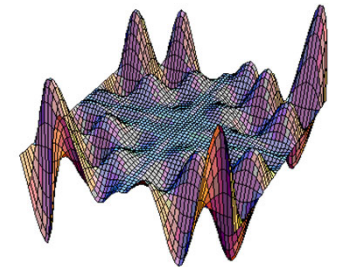
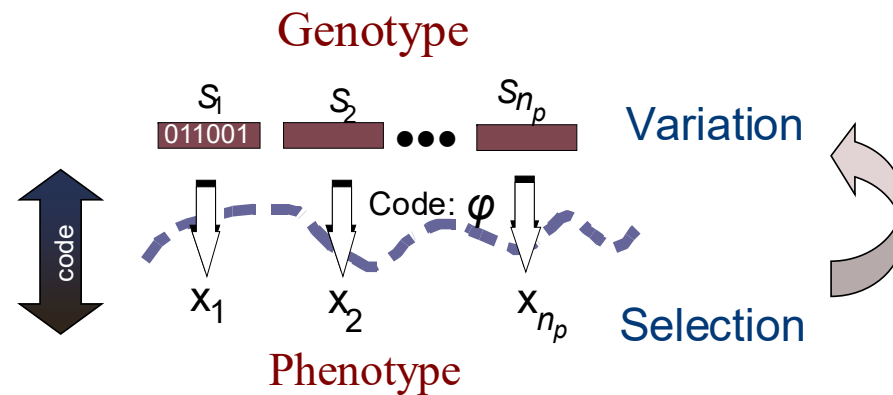


in genetic algorithms

- Solution space encoded as finite-length string over a finite alphabet
 - E.g. $\{0, 1\}$
- GA's exploit coding similarities
 - Searches the code space, not the solution space

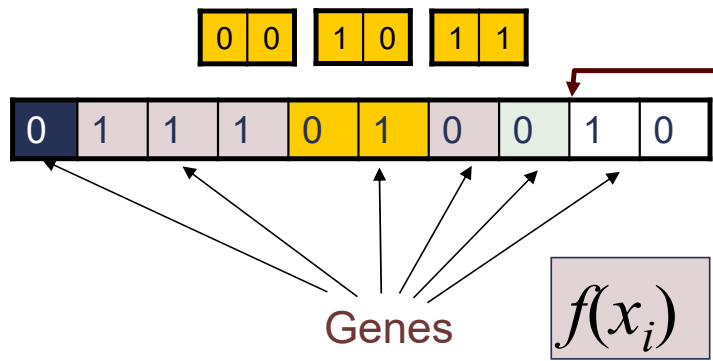
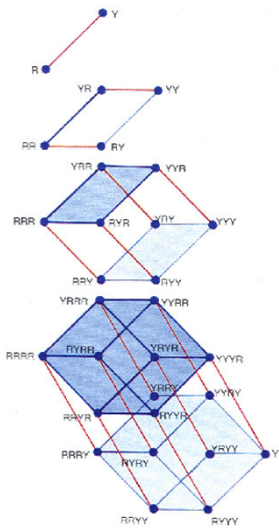
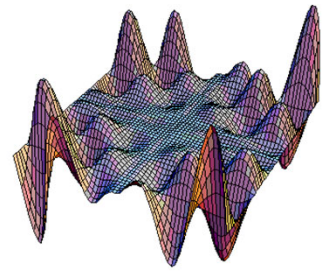


Traditional Genetic Algorithm

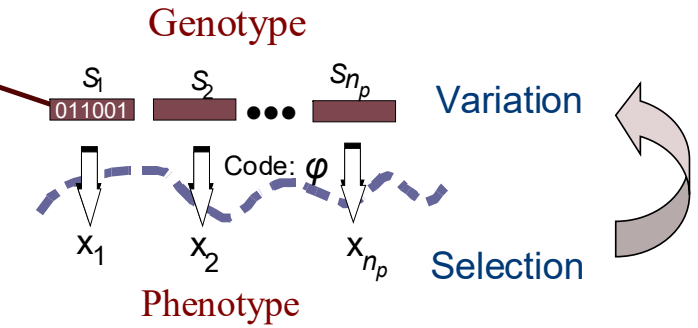


The workings

- 1) **Generate Random population of bit-strings**
 - **Chromosomes/Genotypes**
 - A candidate solution to a problem
 - **Genes**
 - Single or short blocks of adjacent bits
 - **Allele**
 - Actual value of a gene
- 2) **Evaluate Fitness Function for each decoded solution**
- 3) **Reproduce next generation**
 - Genotypes of solutions with higher fitness value reproduce with higher probability
- 4) **Go back to 2)**

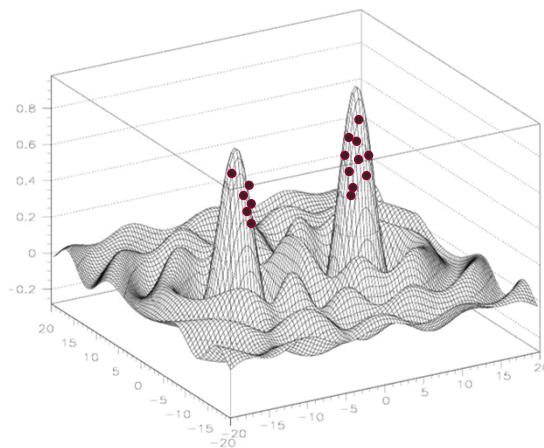
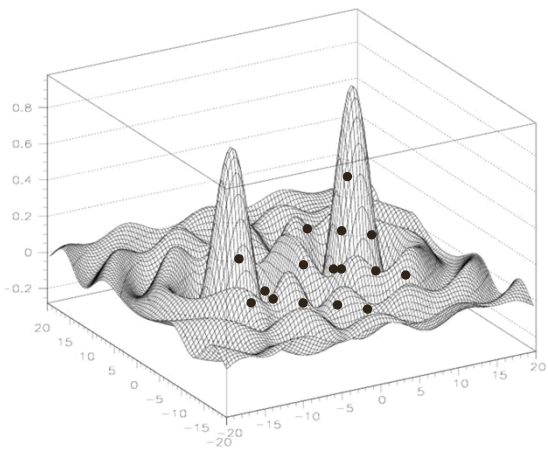


Traditional Genetic Algorithm

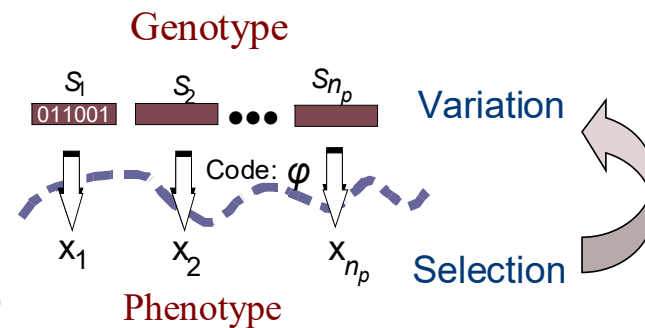


biased population generation

- Solution space encoded as finite-length string over a finite alphabet
 - E.g. {0, 1}
- GA's exploit coding similarities
 - Searches the code space, not the solution space
- Searches the space with many alternatives in parallel
 - Avoids getting trapped in local optima
 - Higher probability of finding better solutions
- Not random search
 - Search towards regions with likely improvement
 - Better solutions reproduce more often
 - Does not work in very rugged, chaotic, uncorrelated landscapes



Traditional Genetic Algorithm

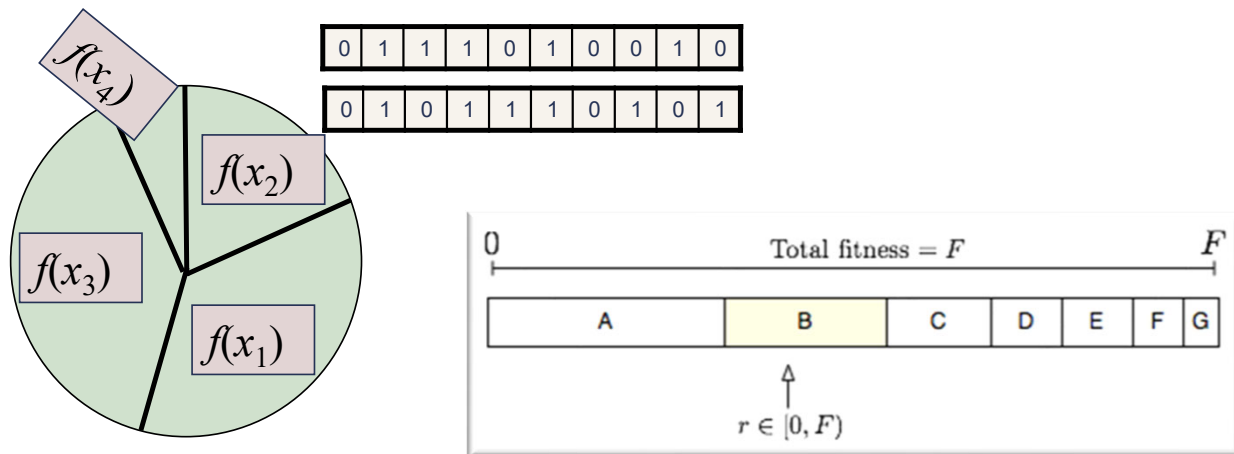


Modeling fitness selection

1) Reproduction: New population is generated

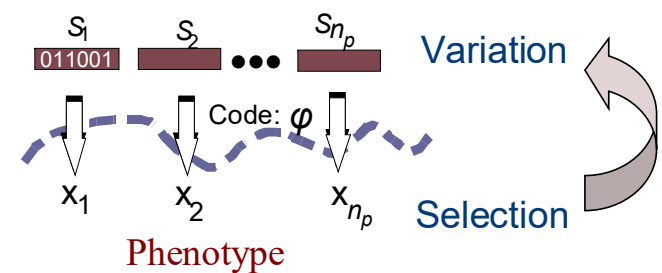
a) Selection

- Select two parent chromosomes from a population according to their fitness
 - Biased roulette wheel according to fitness of solution
 - Elite group
 - Tournament



Traditional Genetic Algorithm

Genotype



Variation: crossover

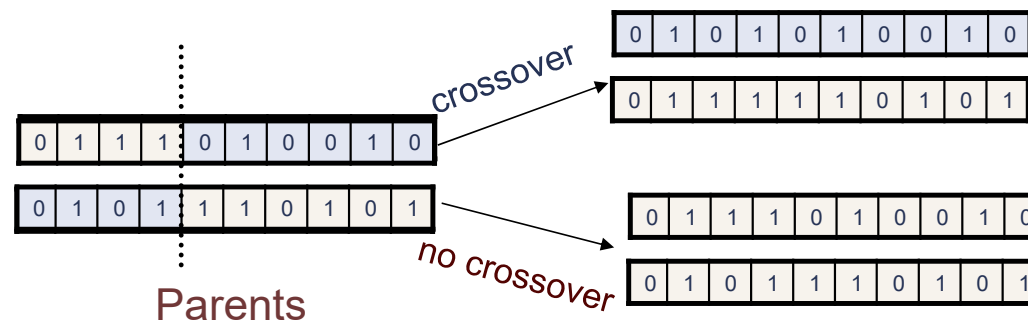
1) Reproduction: New population is generated

a) Selection

- Select two parent chromosomes from a population according to their fitness

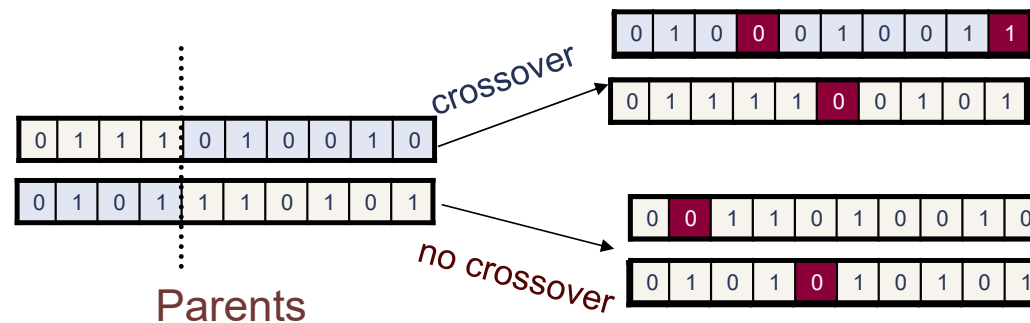
b) Variation: Crossover

- With a **crossover probability** produce offspring pair by recombining parents.



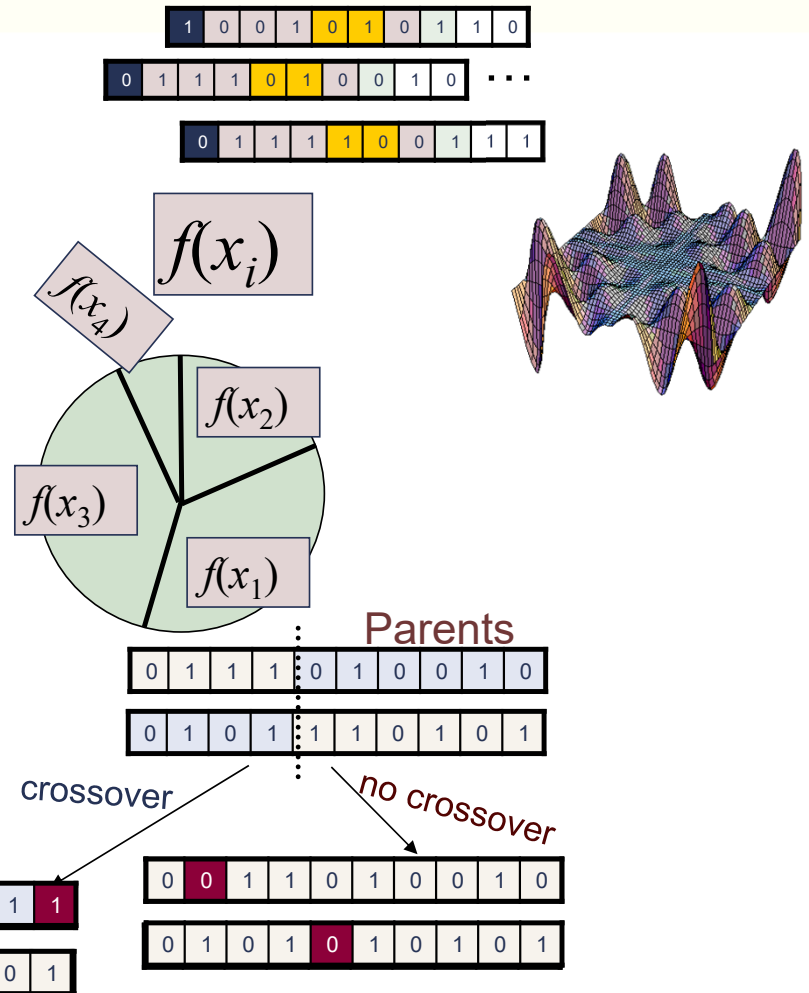
Variation: mutation

- 1) **Reproduction: New population is generated**
- Selection**
 - Select two parent chromosomes from a population according to their fitness
 - Variation: Crossover**
 - With a **crossover probability** produce offspring pair by recombining parents.
 - Variation: Mutation**
 - With a **mutation probability** mutate (bit flip) new offspring at each bit position



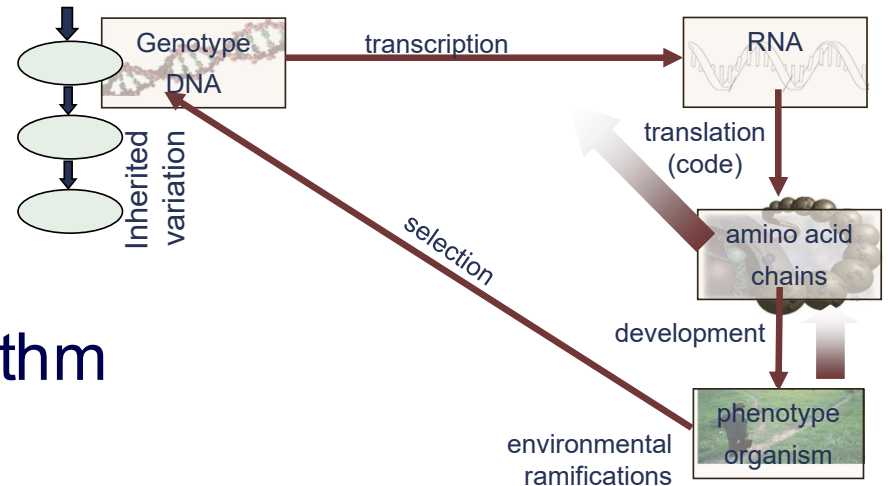
The workings

- 1) Generate Random population of bit-strings
- 2) Evaluate Fitness Function for each decoded solution
- 3) Reproduce next generation
 - Selection by fitness
 - Variation
 - crossover and mutation
 - Fill new population
- 4) Go back to 2) until stop criteria is met
 - Desired fitness
 - Specified number of generations
 - Convergence
 - Lack of variability in population and/or fitness
 - Tends to a peak

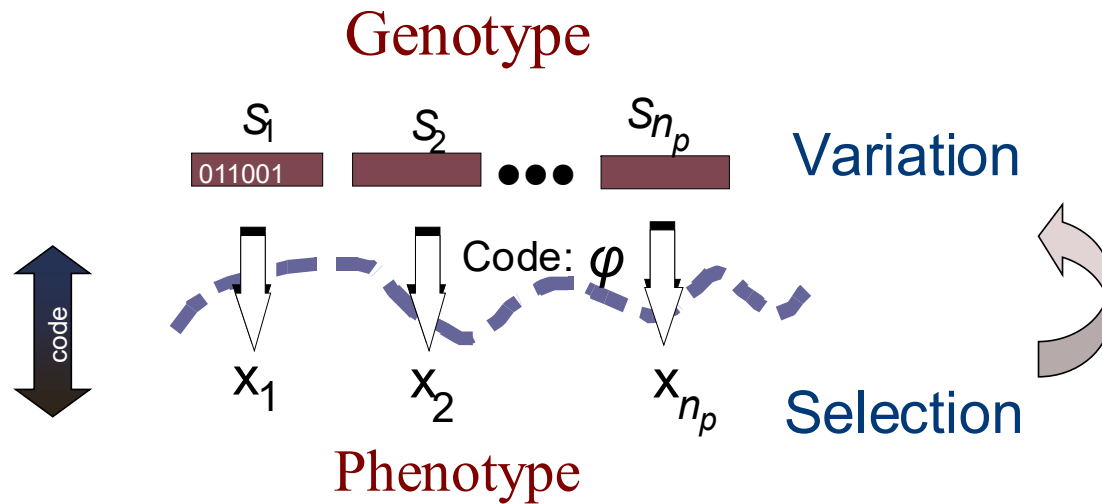


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Search algorithms based on the mechanics of Natural Selection
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 Solution alternatives for optimization problems



Traditional Genetic Algorithm



readings

■ Class Book

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 - Chapter 7

■ Lecture notes

- Chapter 1: What is Life?
- Chapter 2: The logical Mechanisms of Life
- Chapter 3: Formalizing and Modeling the World
- Chapter 4: Self-Organization and Emergent Complex Behavior
- Chapter 5: Reality is Stranger than Fiction
- Chapter 6: Von Neumann and Natural Selection
 - posted online @ casci.binghamton.edu/academics/i-bic

■ Papers and other materials

● Optional

- Nunes de Castro, Leandro [2006]. *Fundamentals of Natural Computing: Basic Concepts, Algorithms, and Applications*. Chapman & Hall.
 - Chapter 5, 7.7, 8.3.1, 8.3.6,

