



# Evolvability: What is It and How Do We Get It?

Coolidge Otis Chapman Honors Program

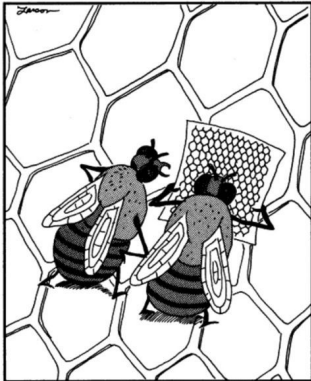
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# Evolutionary Algorithm

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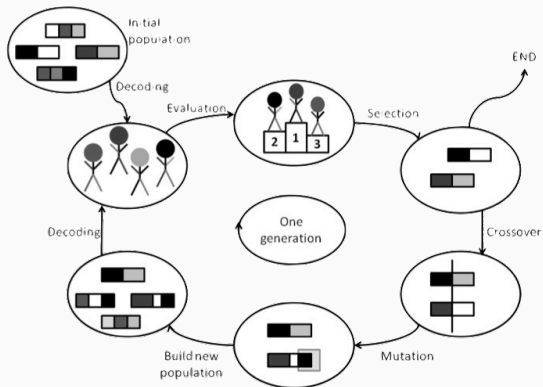


"Face it, Fred—you're lost!"

- **common scenario:** you can recognize a good solution, but you don't know how to find one
- encountered by computer scientists (and everyone else, too)
- **common approach:** try different options, evaluate outcomes to help choose next options to try
- this is called **search**

# Evolutionary Algorithm

- individual
- population
- fitness function
- selection
- mutation
- genotype
- phenotype



**Figure 1:** A schematic illustration of the evolutionary algorithm [Prothmann et al., 2009, Figure 1].

Figure 2: Evolution in Action [Cheney et al., 2013]

# Evolutionary Algorithm: Problem Statement

What makes an evolutionary algorithm work?

# Defining Evolvability

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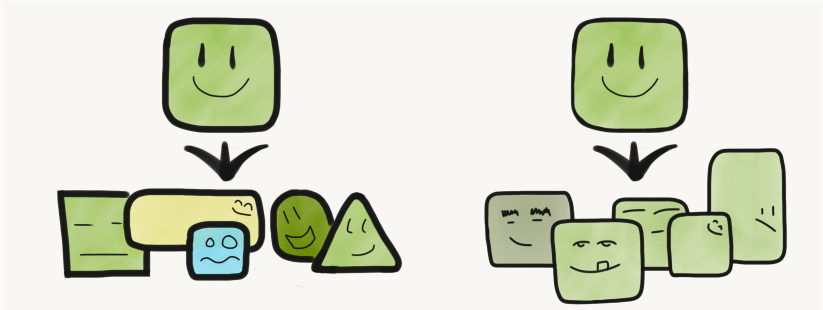
# Defining Evolvability

consensus: the amount of **useful variation** generated by the evolutionary process

- evolvability as the amount of **novel variation** generated
- evolvability the proportion of variation that is **useful**



# Evolvability as Novel Variation

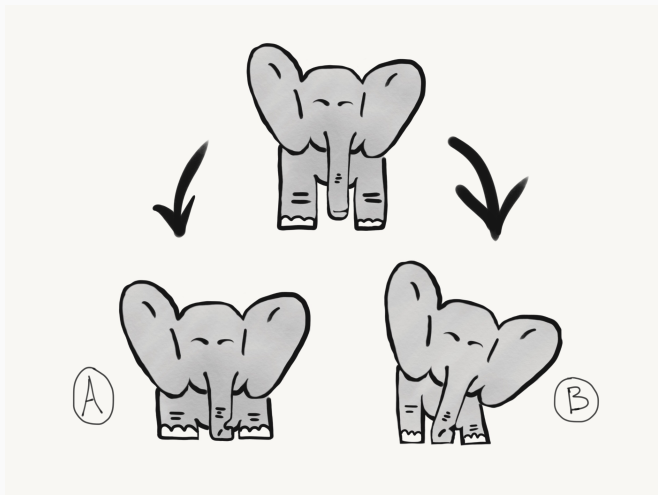


(a) high individual evolvability

(b) low individual evolvability

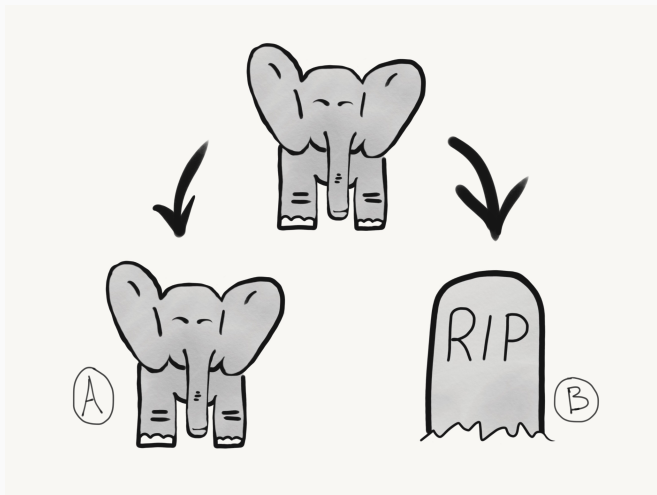
**Figure 3:** An illustration of individual evolvability, considering evolvability as heritable variation [Wilder and Stanley, 2015].

## Evolvability as Bias towards Useful Variation



**Figure 4:** Illustration of developmental constraint; high evolvability left and low evolvability right [Smith et al., 1985, Tuinstra et al., 1990].

## Evolvability as Bias towards Useful Variation



**Figure 5:** Illustration of robustness; high evolvability left and low evolvability right [Downing, 2015].

## Organizing and Analyzing Factors that Promote Evolvability

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# Proximate / Ultimate Thinking



**Figure 6:** Why is the flower purple? Proximate and ultimate explanations differ [Wilson, 2007].

# Proximate/Intermediate/Ultimate Organization

## Proximate Causality

- describes specific organismal processes or structures

## Intermediate Causality

- describes characteristics of an organism as a whole

## Ultimate Causality

- describes the relation between the individual and its environment

# Proximate/Intermediate/Ultimate Organization

## Proximate Causality

- duplication and divergence
- developmental constraint
- hidden genetic variation
- exploratory growth
- weak linkage
- indirect encodings

## Intermediate Causality

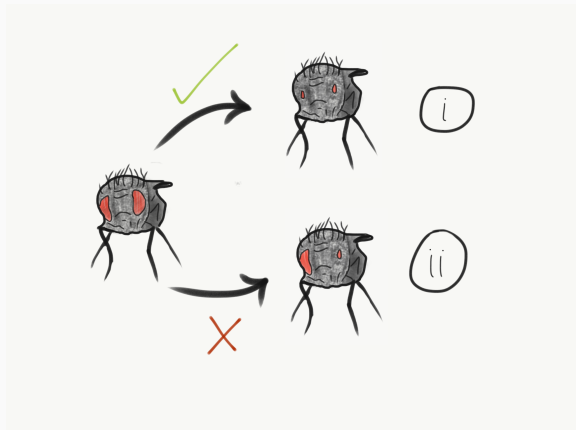
- modularity
- robustness
- canalization
- plasticity
- intraindividual degeneracy
- interindividual degeneracy
- regularity

## Ultimate Causality

- temporally varying goals
- environmental influence on phenotype
- fitness degeneracy

# Proximal Causality: Developmental Constraint

idea: the phenotype results from development, so developmental processes influence the phenotypic outcomes of mutation

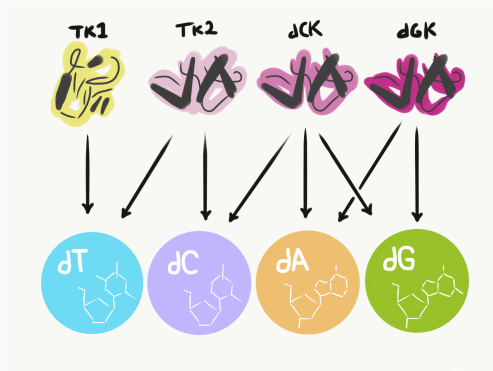


**Figure 7:** Illustration of Canalization Against Bilateral Asymmetry in *Drosophila melanogaster* [Tuinstra et al., 1990].



# Intermediate Causality: Intraindividual Degeneracy

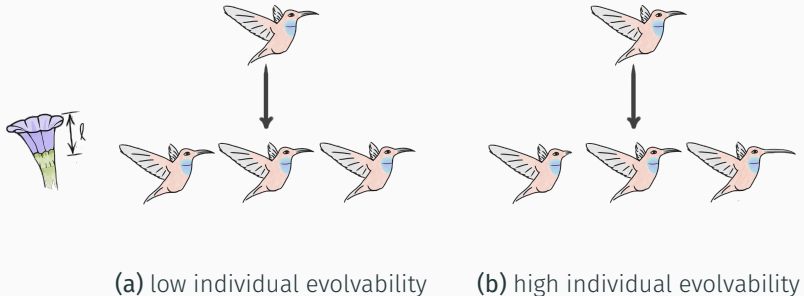
idea: employing a diverse collection substructures that provide identical or near-identical functionality promote robustness through redundancy while providing many jumping off points for variation through repurposing or elaboration



**Figure 8:** Mammalian deoxyribonucleoside kinases exhibit degeneracy [Sandrini and Piskur, 2005].

# Ultimate Causality: Modularly Varying Fitness Function

idea: if evolution sets a moving target, organisms that produce variable offspring will be selected for



**Figure 9:** An hypothetical illustration of a modularly varying fitness function [Kashtan and Alon, 2005].

## observations:

- proximal and intermediate causality relates to the genotype-phenotype mapping
- ultimate causality is related to interaction with the environment to determine fitness



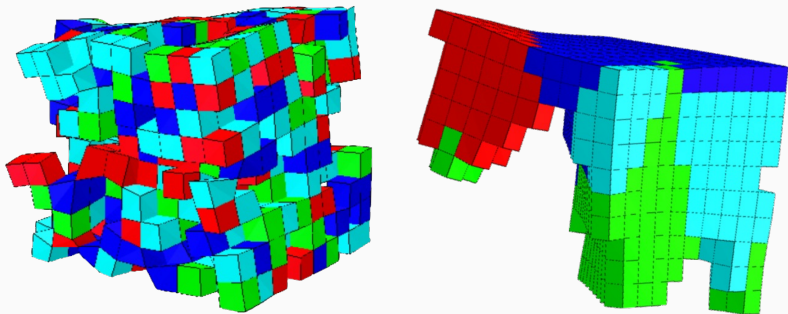
## paths forward:

- taking a broader view of fitness and selection
- taking a more nuanced view of the developmental process

# Evolvability in Action

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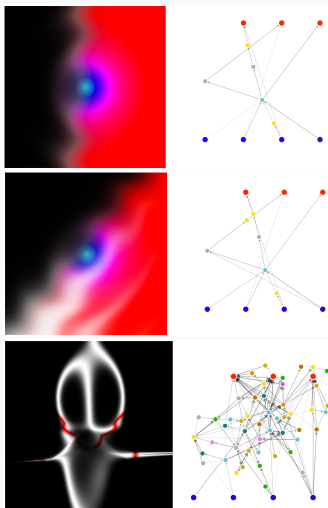
## Promoting Evolvability: Indirect Encoding



(a) direct encoding (low regularity)    (b) indirect encoding (high regularity)

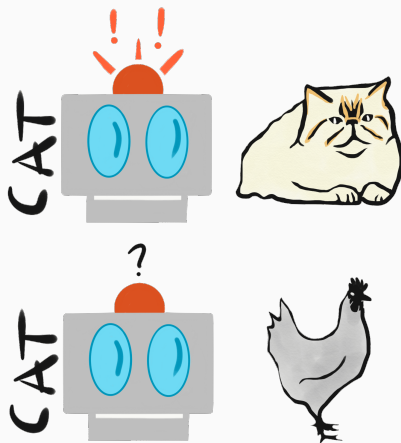
**Figure 10:** Representative examples of soft robots evolved with direct and indirect representations [Cheney et al., 2013, Figures 6, 7]

# Promoting Evolvability: Fitness Niches



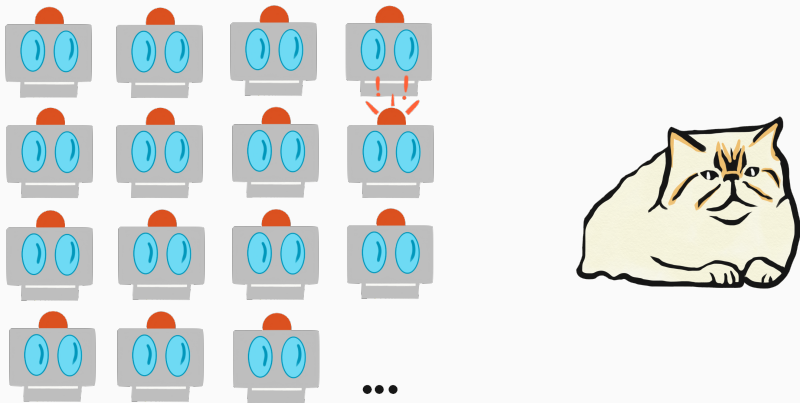
**Figure 11:** Illustration of compositional pattern producing networks (right) and their output images (left) generated via [Ha, 2015].

## Promoting Evolvability: Fitness Niches



**Figure 12:** A deep neural network (DNN) is trained to recognize a specific category of images.

## Promoting Evolvability: Fitness Niches



**Figure 13:** Several hundred fitness niches are defined using DNNs each trained to recognize different categories [Nguyen et al., 2015].



# Promoting Evolvability: Fitness Niches

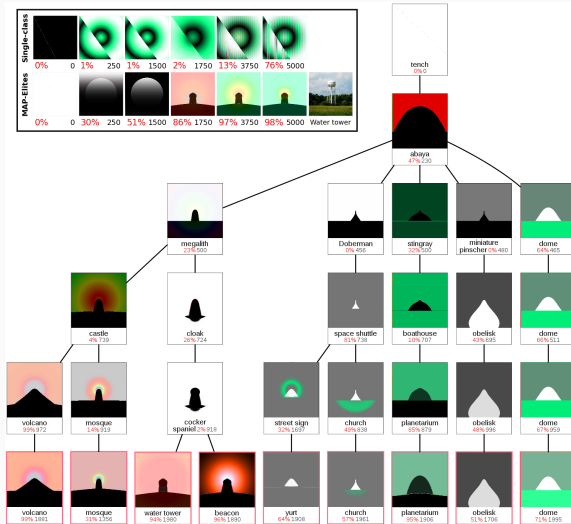


Figure 14: An illustration of goal-switching, where offspring from a parent that occupies one niche invade another [Nguyen et al., 2015, Figure 9]. Individuals that promote phenotypically variable offspring are rewarded [Mengistu et al., 2016].

# Promoting Evolvability: Fitness Niches

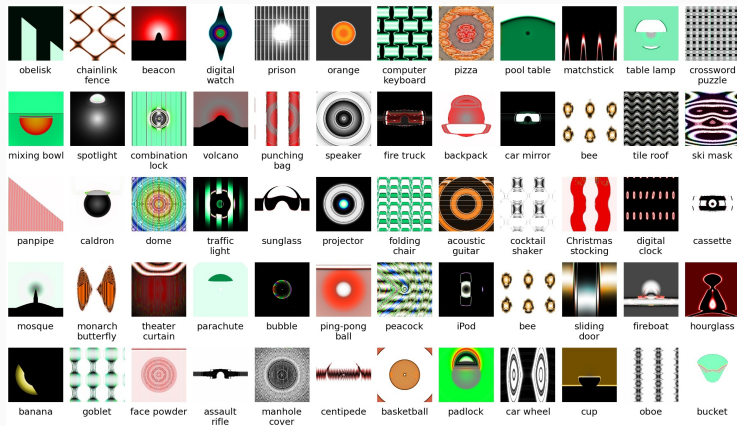
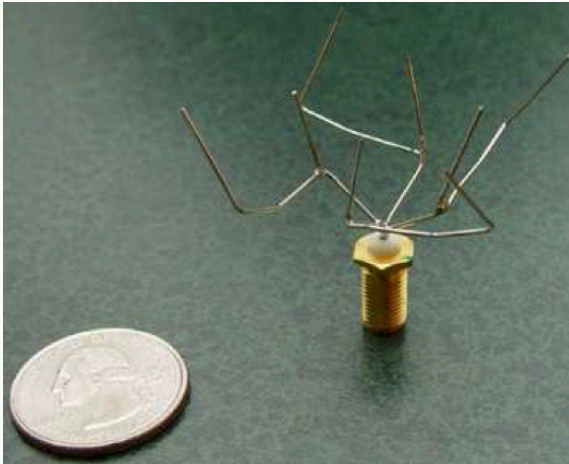


Figure 15: Selected champion individuals from a sample of environmental niches [Nguyen et al., 2015, Figure 7].

## Closing Thoughts

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**Figure 16:** A spacecraft antenna design generated using evolutionary methods [Hornby et al., 2006, Figure 2(a)].

# Scientific Questions

- at what level of abstraction can the power of biological evolution be harnessed in a computational model?
- what are the fundamental mechanisms at play in evolution?



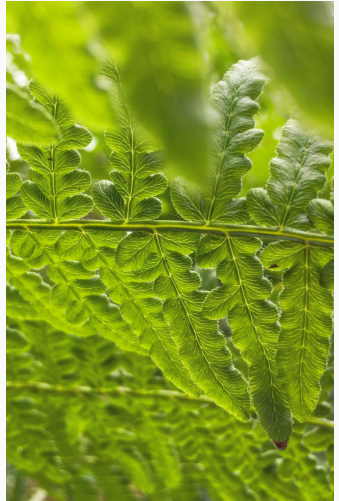
# Scientific Questions

- evolutionary biology provides continuing inspiration for new techniques in evolutionary computing
- evolutionary models move theory evaluation from a qualitative endeavor towards a quantitative endeavor



# Acknowledgements





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Questions?



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