

Non-Adaptive Evolvability

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Evolution Fails to Optimize Mutation Rates
(though it would improve evolvability)

Evolution Fails to Produce Modularity For Adaptive Reasons
(though it would improve evolvability)

Part I: Mutation Rates

2008

PLoS COMPUTATIONAL BIOLOGY

Natural Selection Fails to Optimize Mutation Rates for Long-Term Adaptation on Rugged Fitness Landscapes

Jeff Clune^{1,2*}, Dusan Misevic³, Charles Ofria¹, Richard E. Lenski⁴, Santiago F. Elena^{2,5}, Rafael Sanjuán^{2,6}

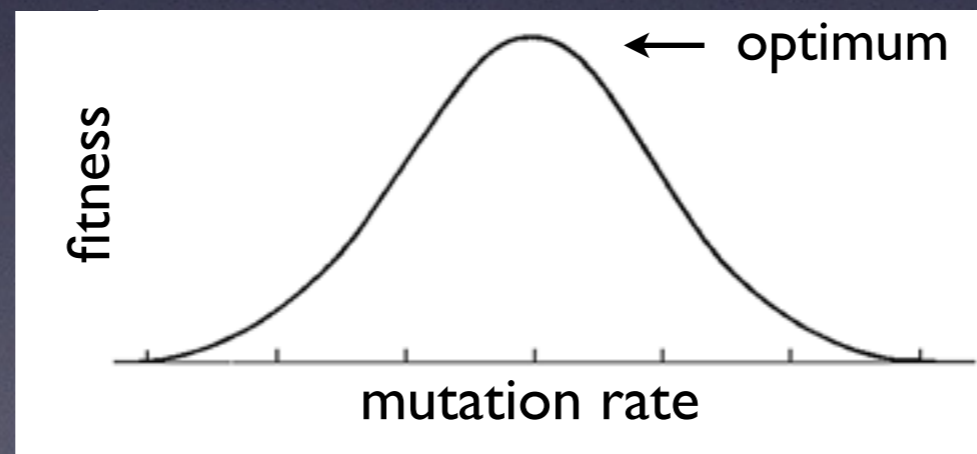
- natural selection is short-sighted
- a non-low mutation rate
 - good in the long-term
 - bad in the short term

Mutation Rates

- Key driver of evolvability
- Optimized?
 - (for long-term adaptation)

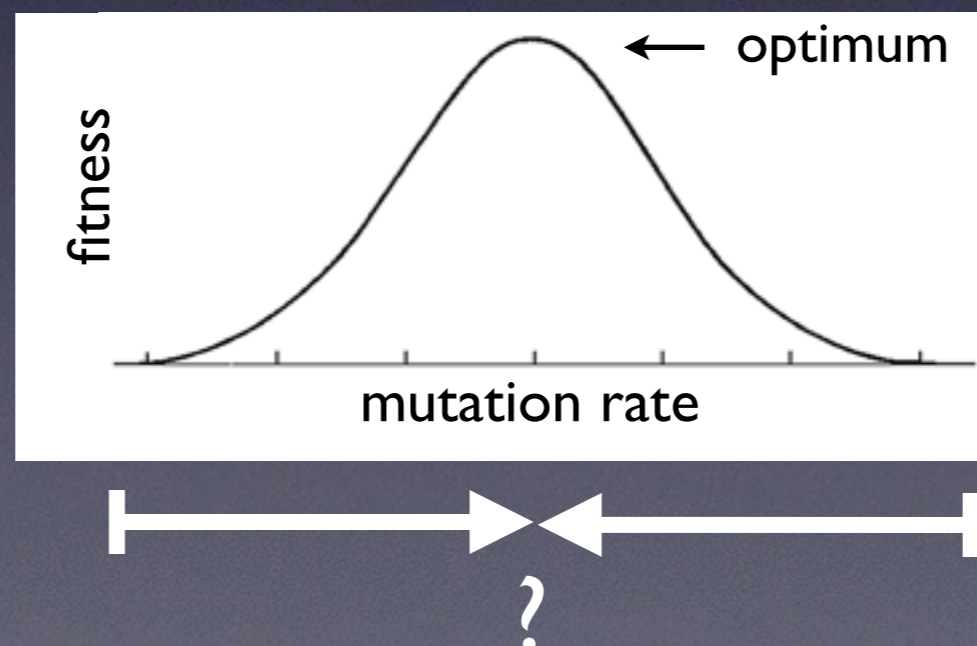
Experimental Design

- Identify the optimum
 - evolve organisms with different, fixed (non-evolving) mutation rates in new environment



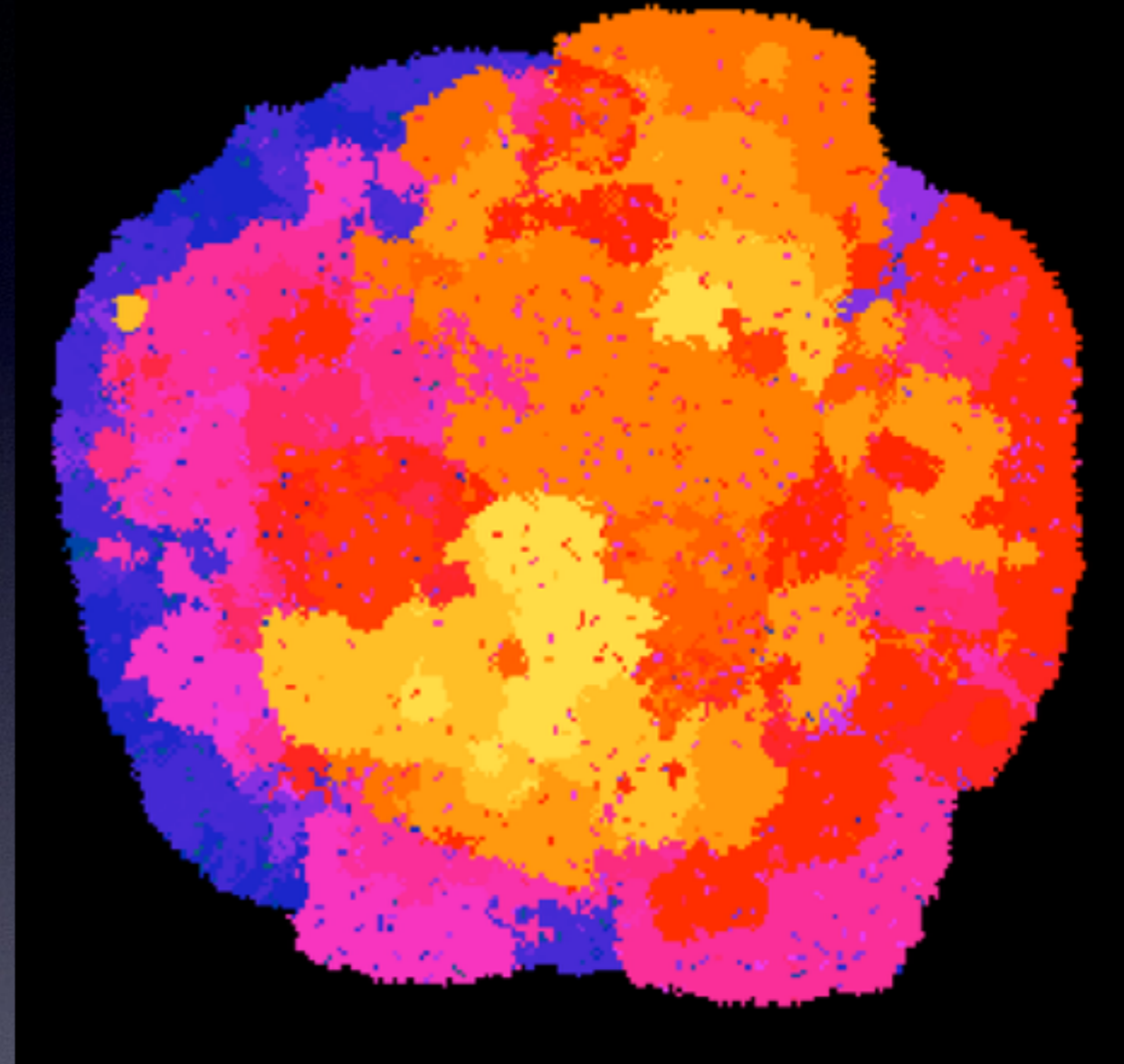
Experimental Design

- Identify the optimum
 - evolve organisms with different, fixed (non-evolving) mutation rates in new environment
- Does evolution produce the optimum?
 - allow mutation rates to evolve
 - start well below and well above the optimum



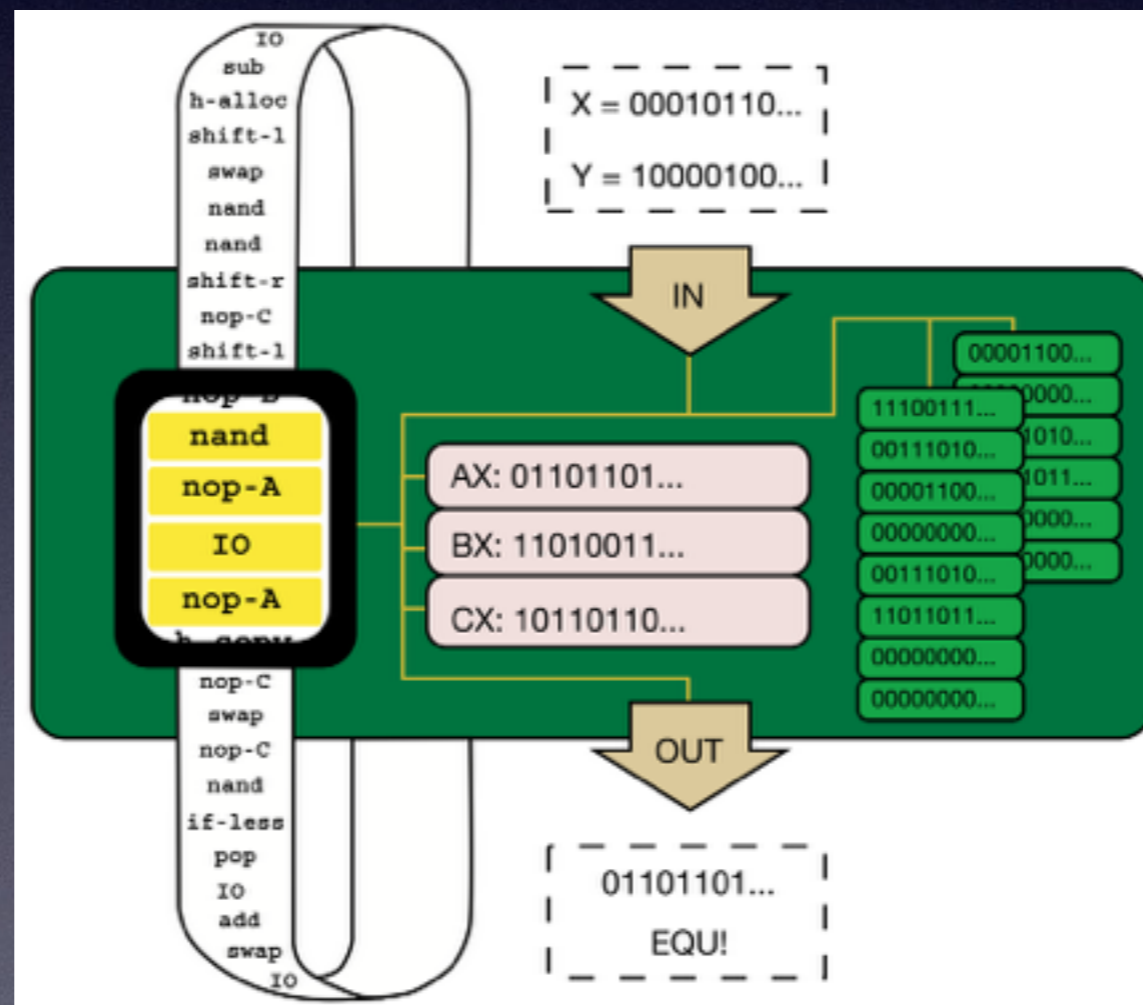
System

- computational evolution
- Avida
 - well-studied
 - Lenski et al. Nature 2003
 - Lenski et al. Nature 1999
 - Adami et al. PNAS 2000
 - Wilke et al. Nature 2001
 - Chow et al. Science 2004
 - etc.
 - population of self-replicating digital organisms



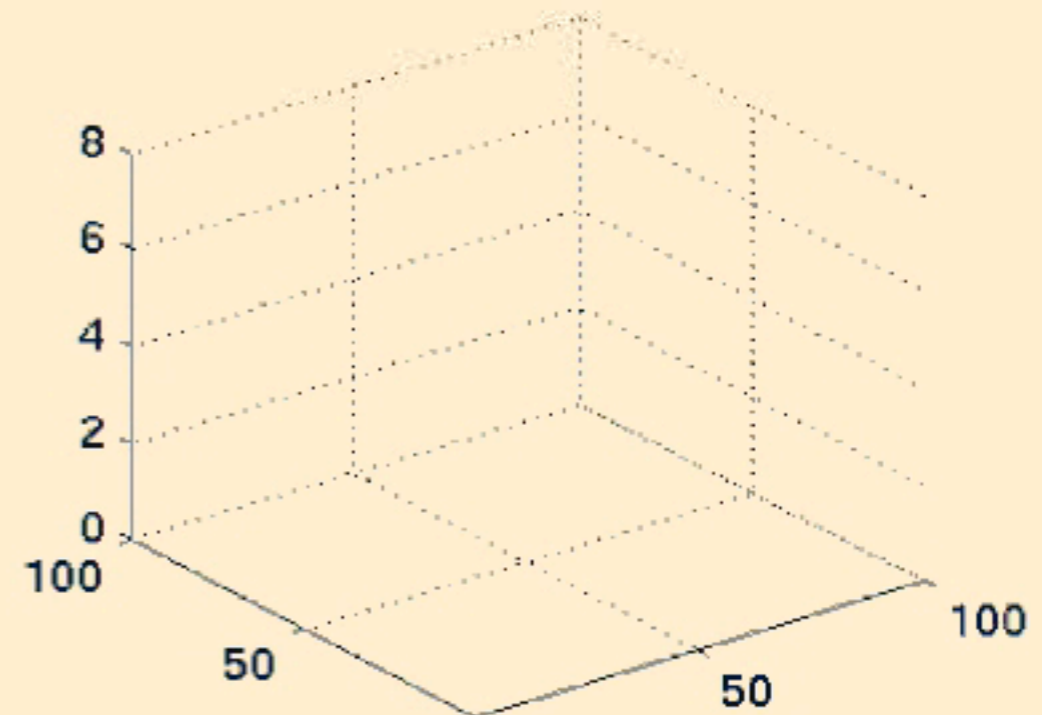
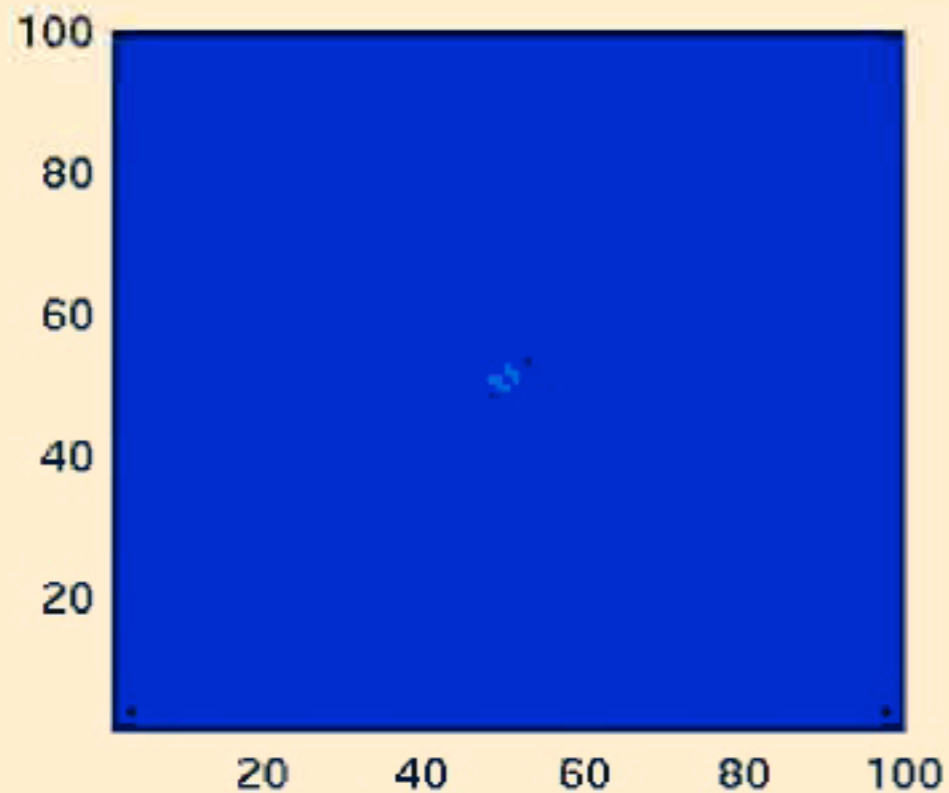
Avida Organisms

- genome: list of computer instructions
- phenotype: execution of instructions with virtual hardware



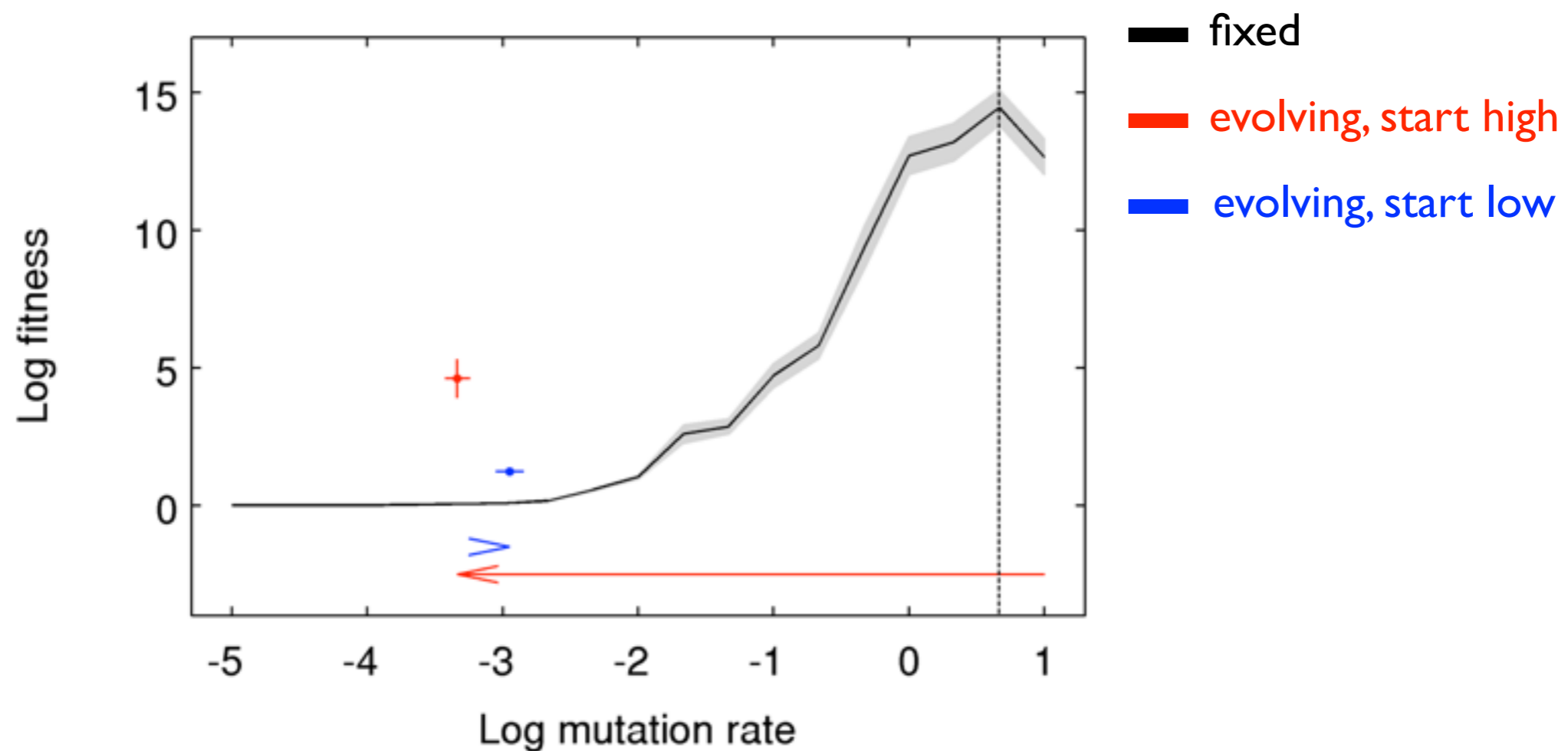
Fitness

- limited space (overwrite neighbors)
- faster replication = more offspring
- extra energy = faster replication
 - traditional: 9 logic tasks (Lenski et al. 2003)

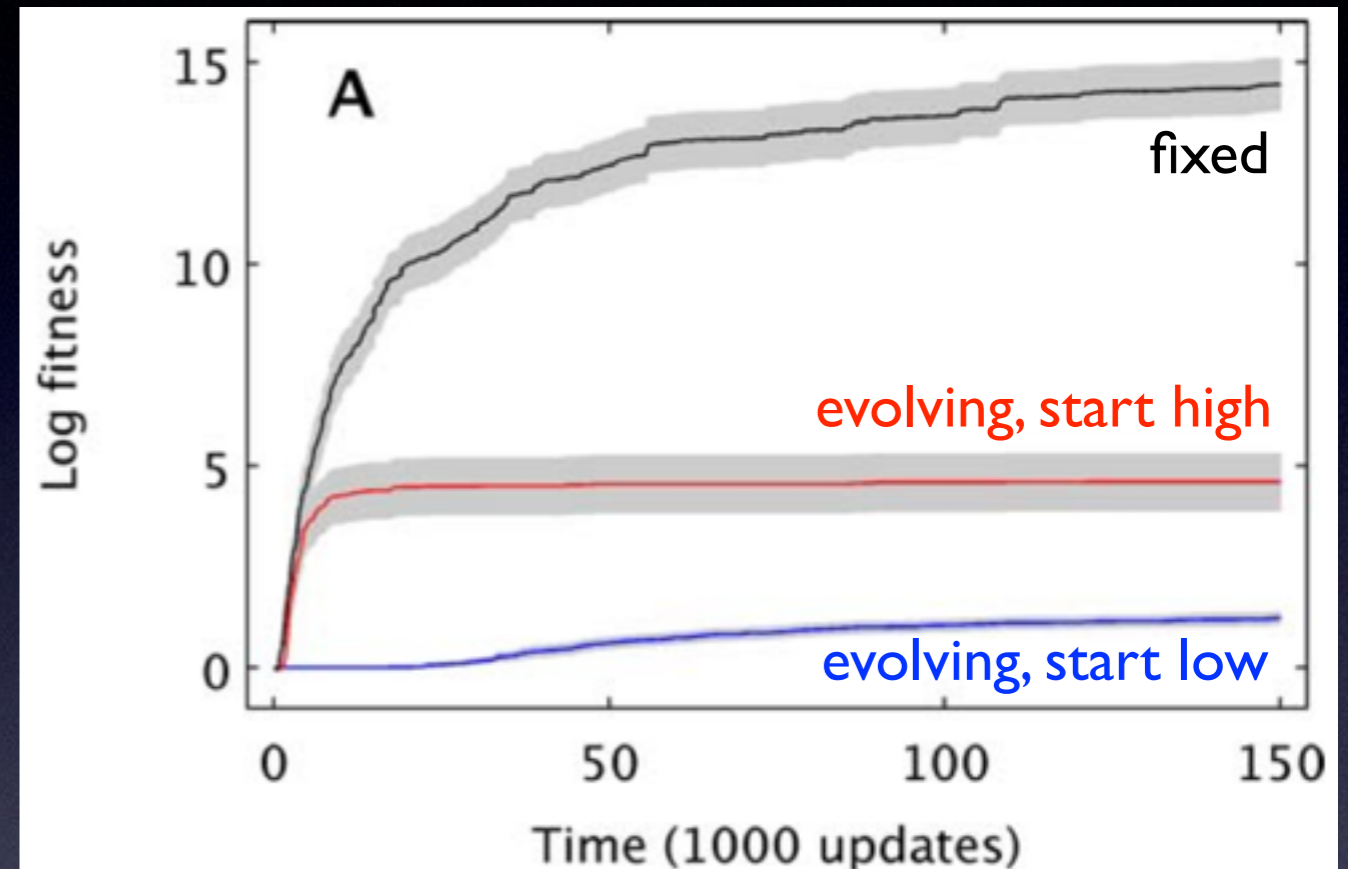
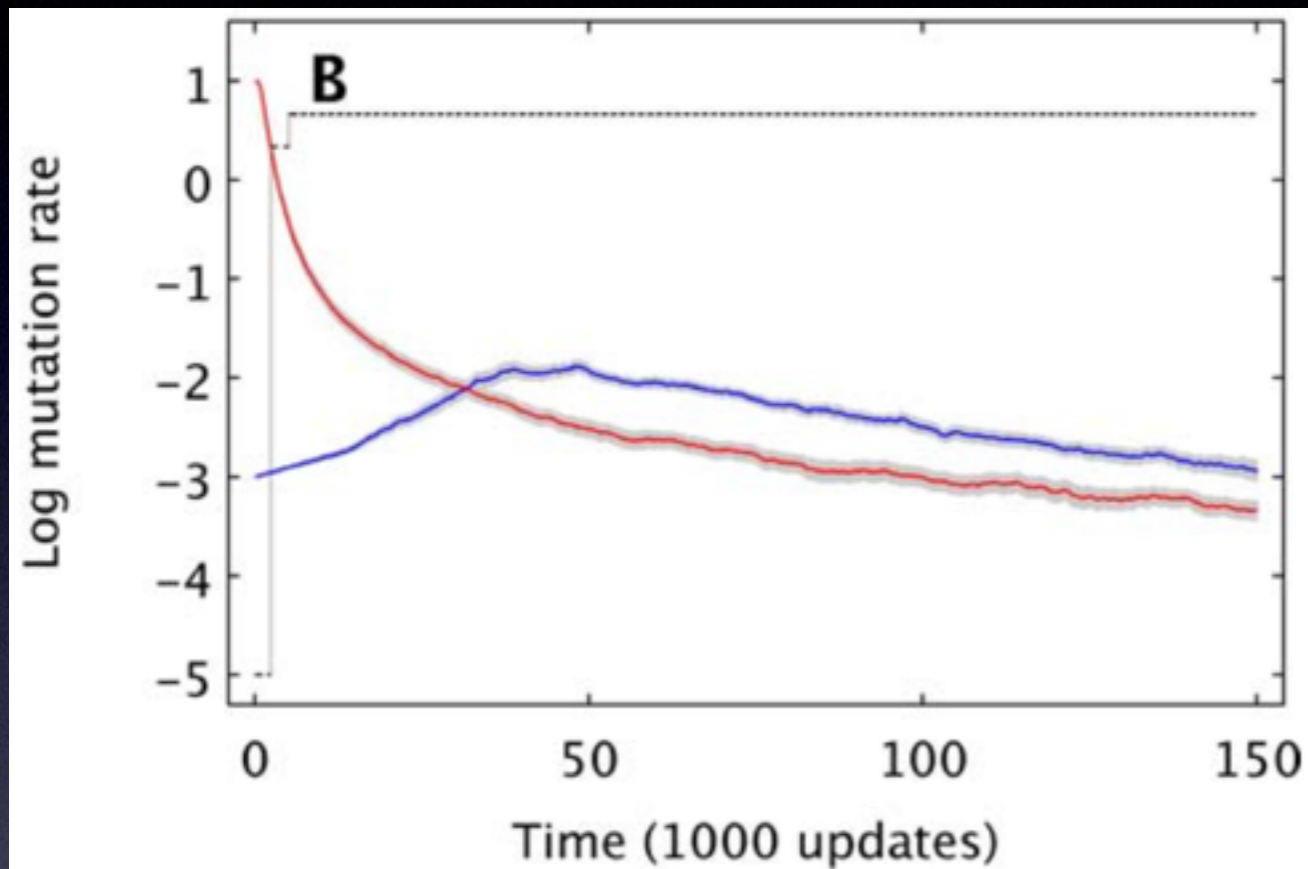


Experiments

- sweep range of fixed mutation rates
- allow mutation rates to evolve



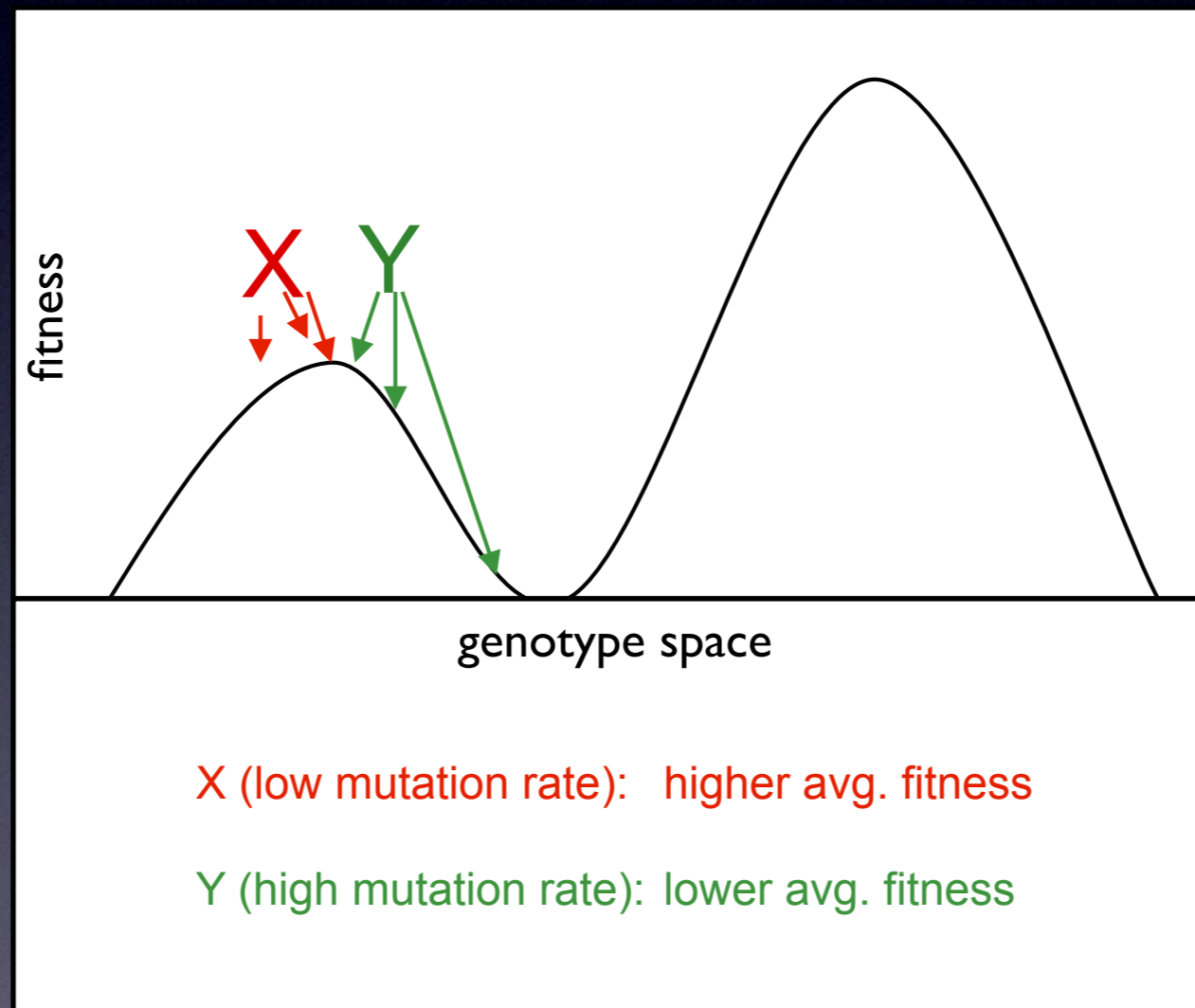
Evolved Mutation Rates Less Fit



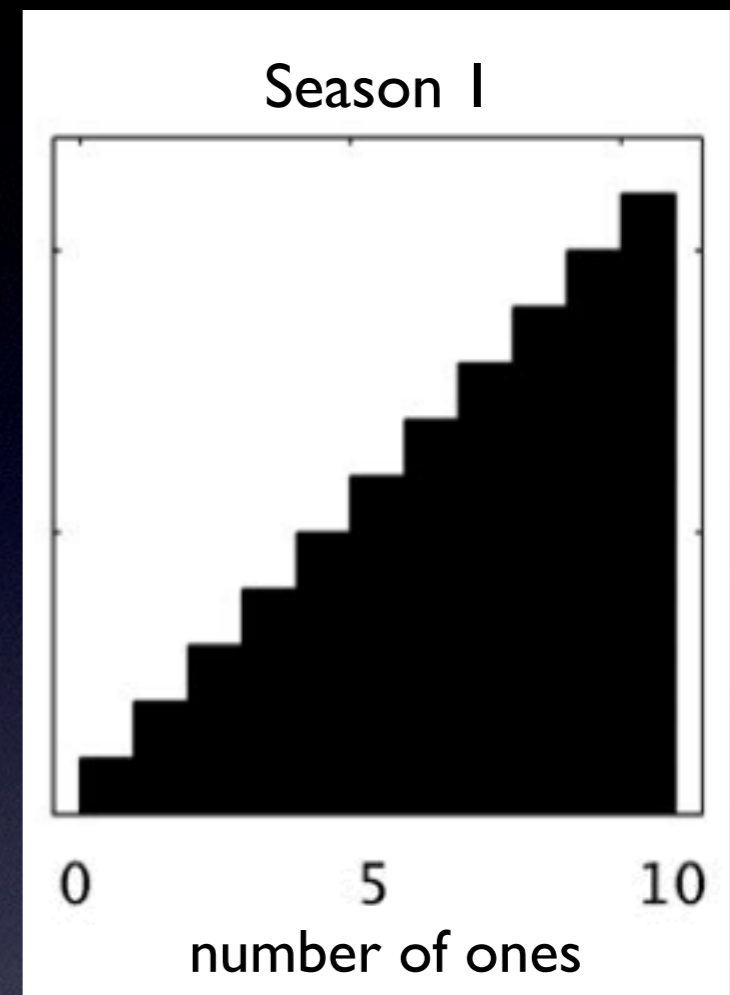
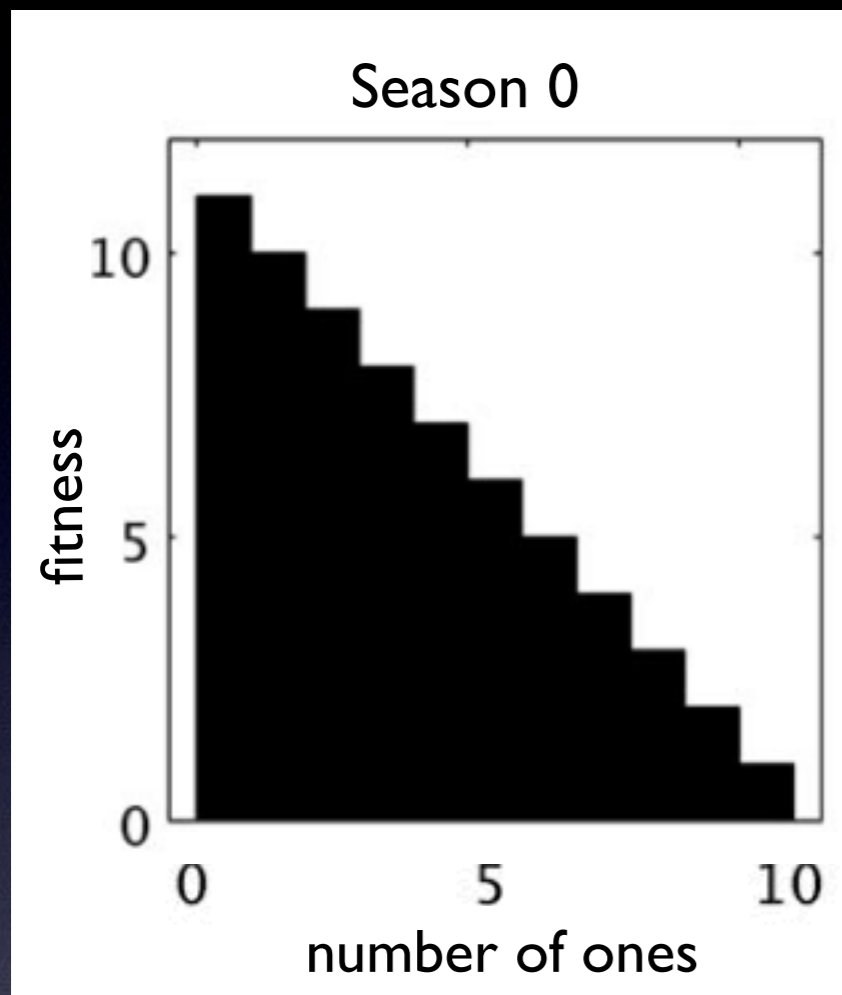
- natural selection fails to optimize for long-term
- ...in a complex fitness landscape (Avida default)

Hypothesis

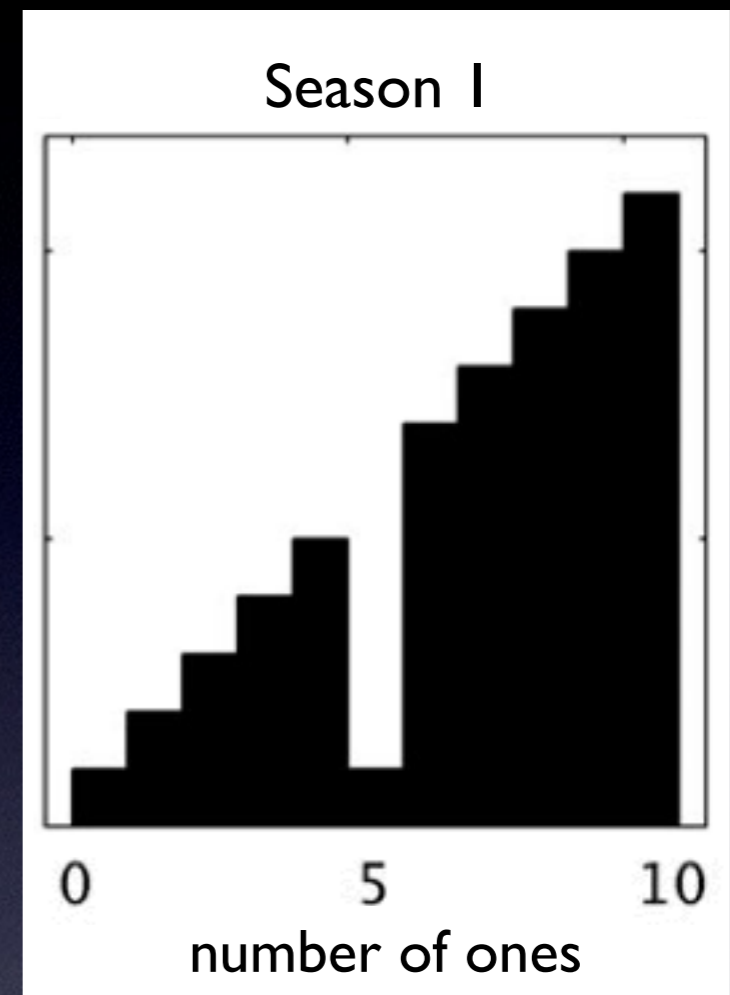
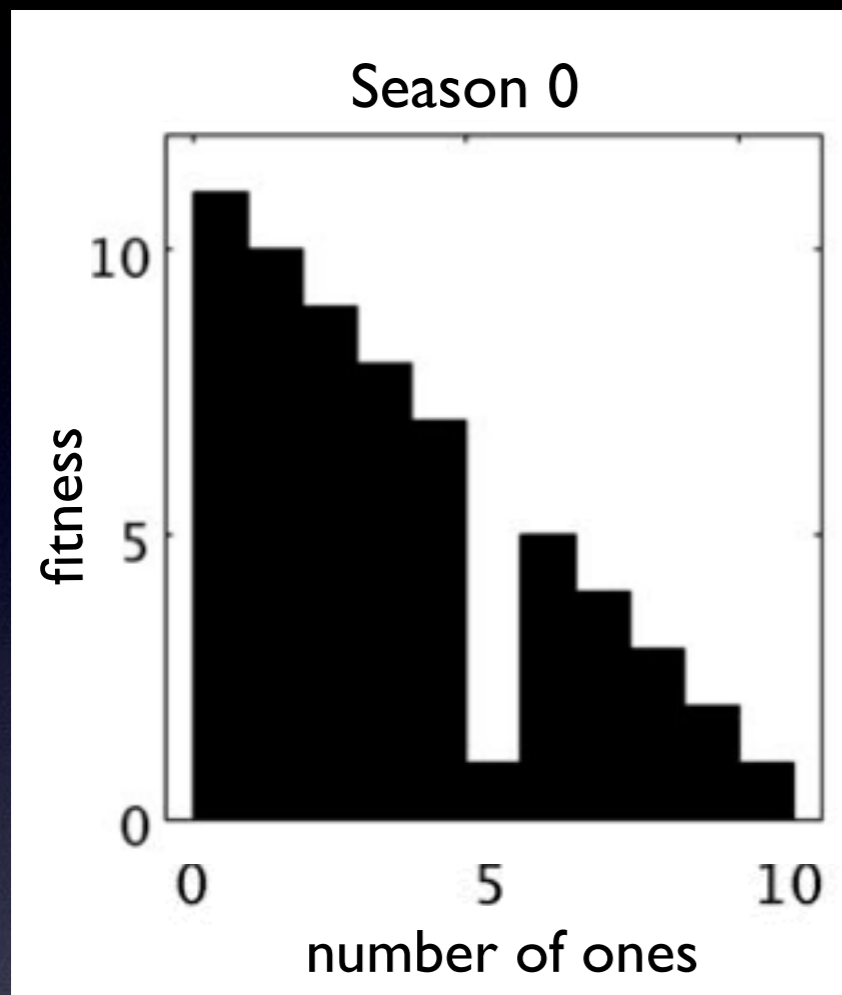
Ruggedness of fitness landscape?

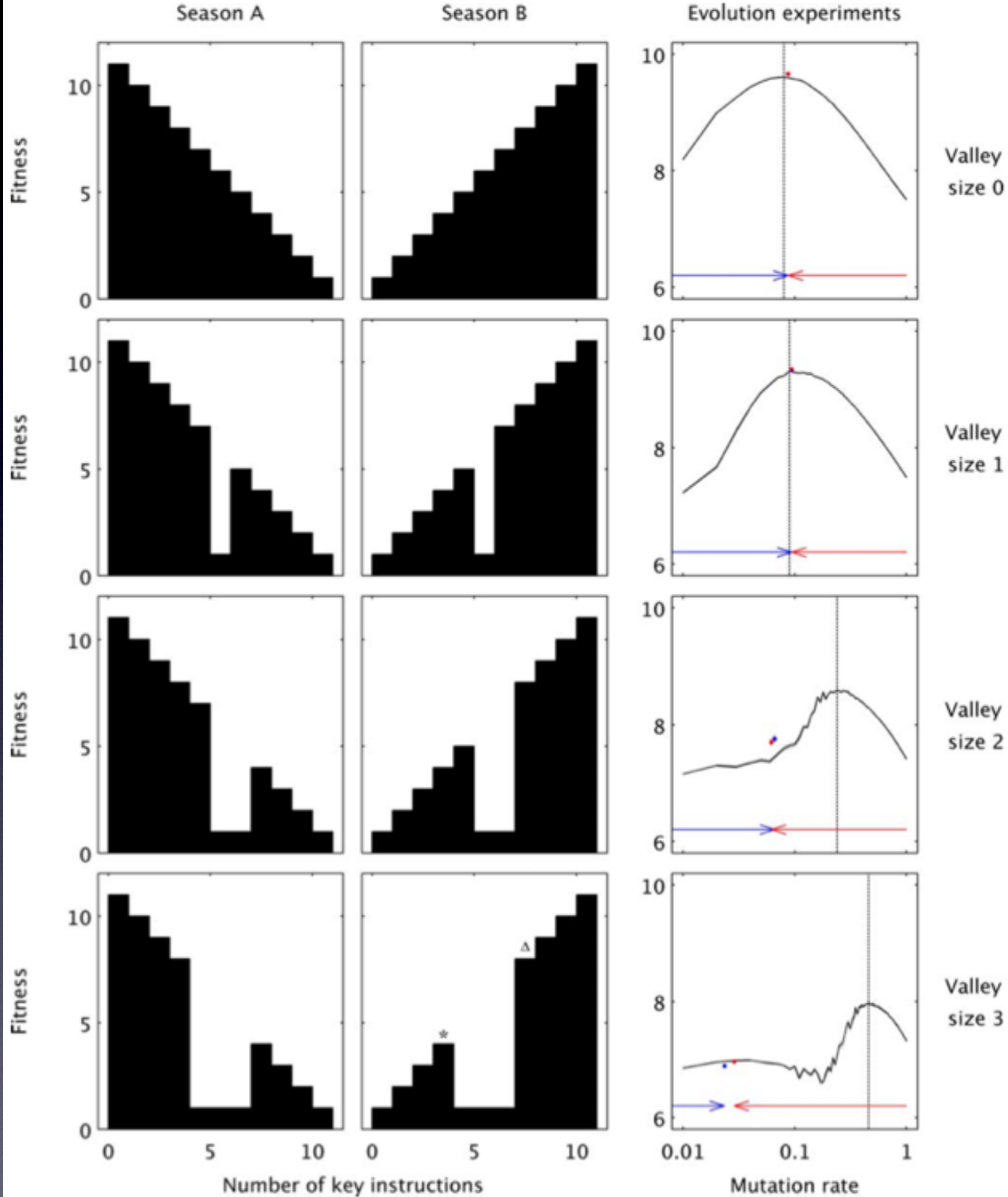


Simplified Avida Environment



Simplified Avida Environment





- Optimized on smooth landscapes
- Not optimized when ruggedness above threshold
- Valley is crossed many times, but any delay = self-reinforcement

Same Results with Different...

- implementations of mutation rate evolution
 - size of changes
 - frequency of changes
 - increases more likely
 - self-reflexive
- environments
 - complexity
 - static vs. changing
 - rate of change
- ancestors

Part I: Evolvability

Conclusions

- natural selection fails to optimize mutation rates for long-term adaptation on rugged fitness landscapes
- i.e. natural selection is short-sighted
 - sounds obvious, but many disagree!

Part II

Evolutionary Origins of Modularity

PROCEEDINGS OF THE ROYAL SOCIETY

2013



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Marie Curie University
Paris, France



Hod Lipson

Cornell
University

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Modularity

- Localization of function in an encapsulated unit (Lipson 2007)
 - Car (spark plug, muffler, wheel), bodies (organs), software, etc.
- Enables increased
 - Complexity
 - Adaptability

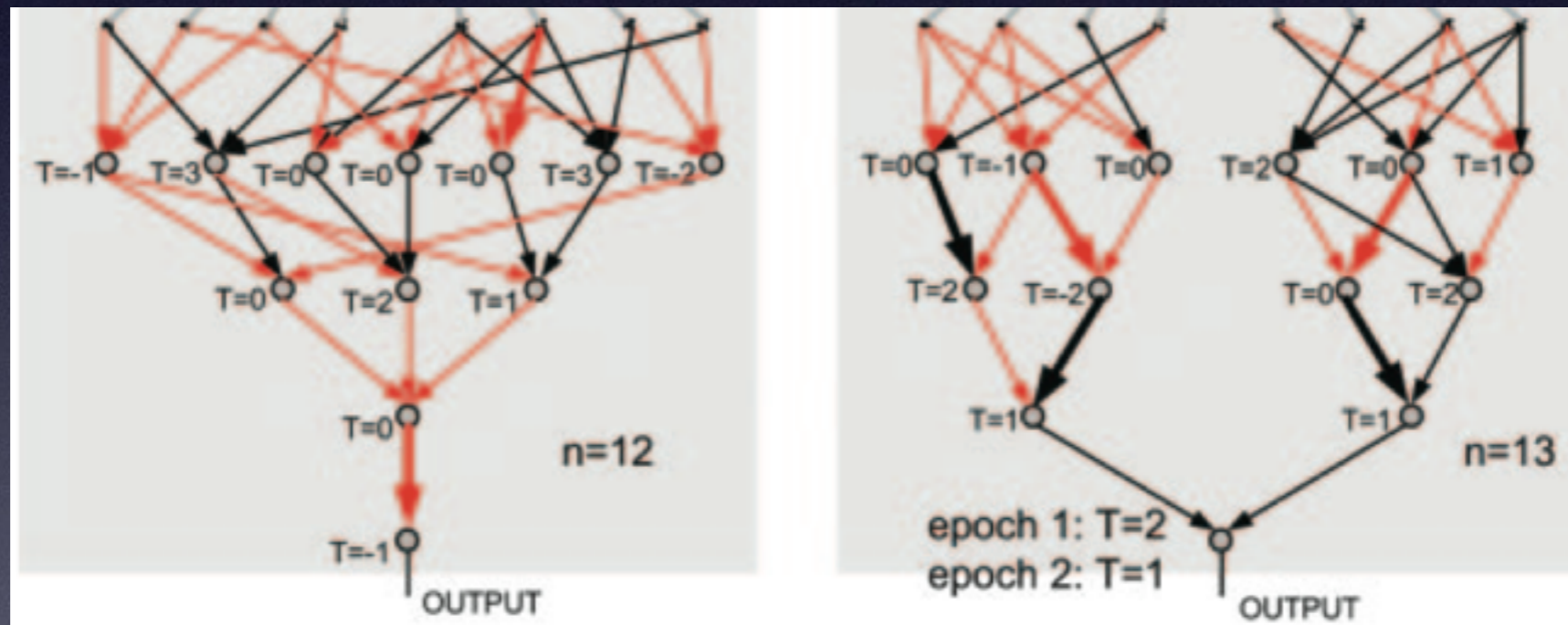


Modularity: Major driver of Evolvability

- For the same reasons as in engineering
 - reuse building blocks in new combinations
 - tinker with one module without affecting everything

Modularity

- Rare in current neuroevolution
- Suggests selection on performance alone does not produce modularity



Kashtan and Alon 2005

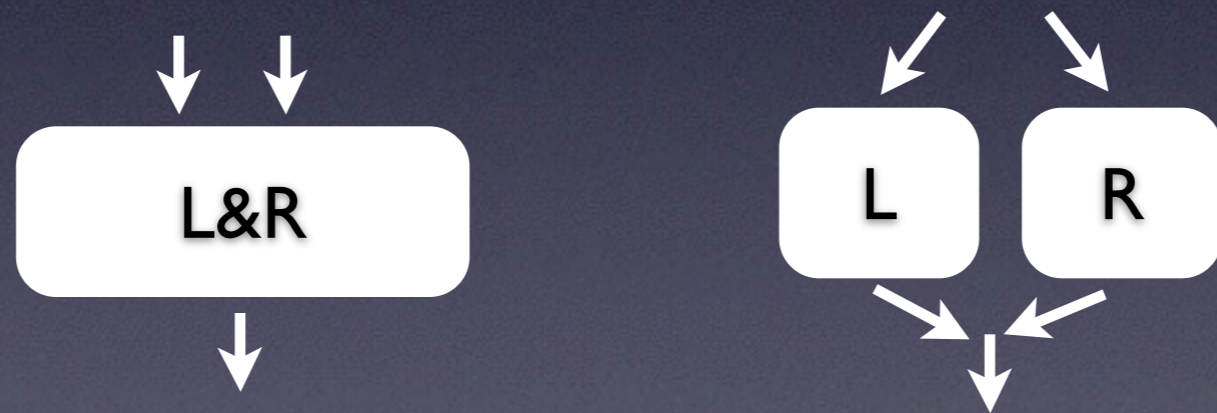
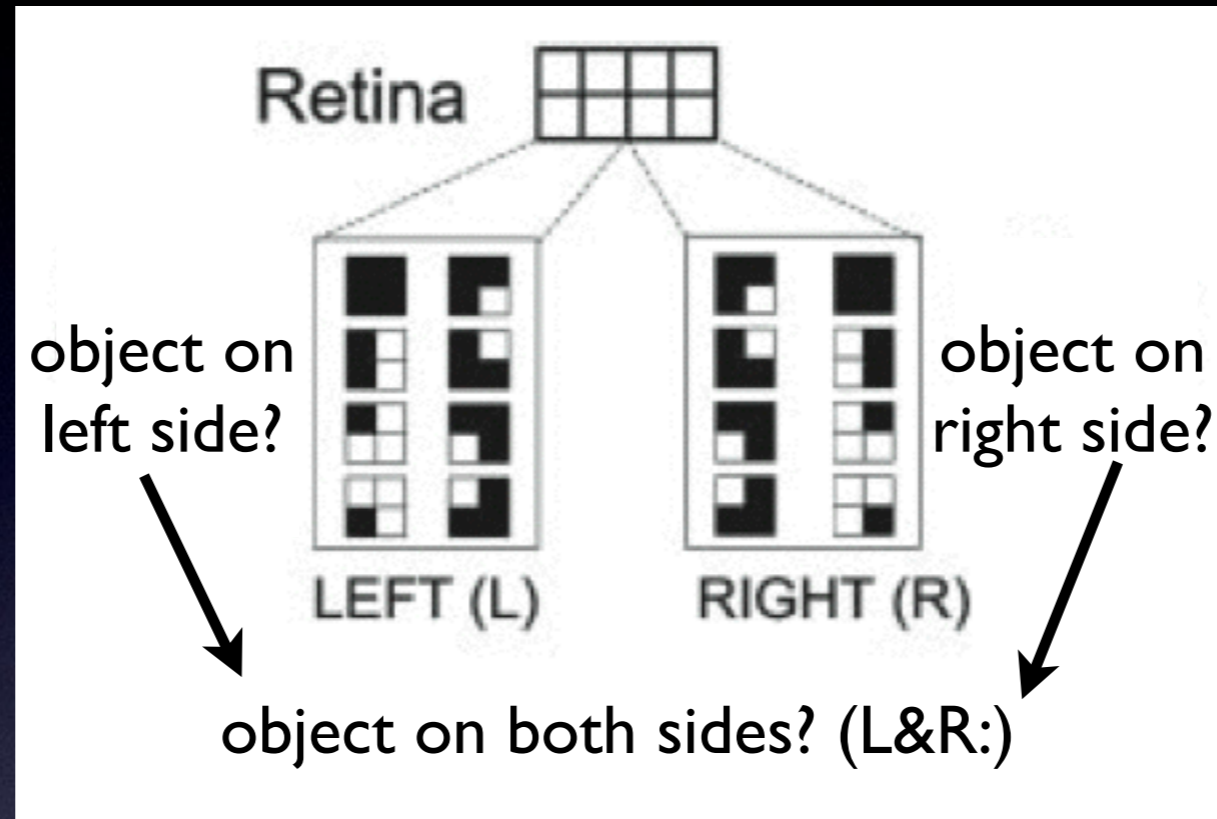
Why did modularity evolve?

- Leading Hypothesis: Selection for evolvability
- We provide evidence for a new force:
 - Selection to minimize connection costs

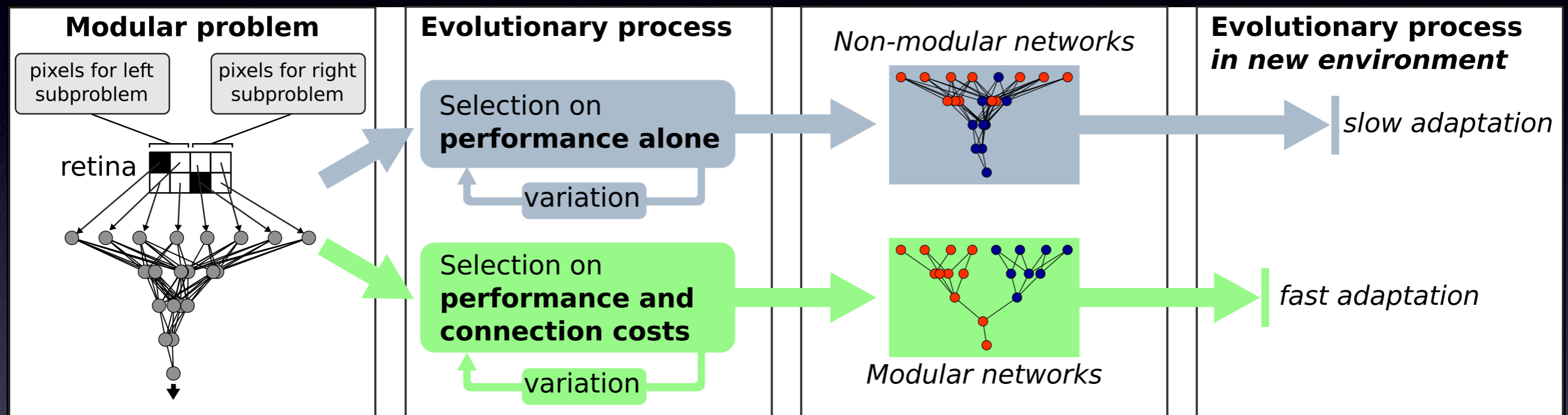
Minimizing Connection Costs

- Hypothesis from founding neuroscientist (Ramón y Cajal 1899)
 - There are costs in biological networks
 - Evidence that selection acts to minimize costs
- Test by evolving neural networks
- Why?
 - answer longstanding, fundamental biological question
 - harness for artificial intelligence

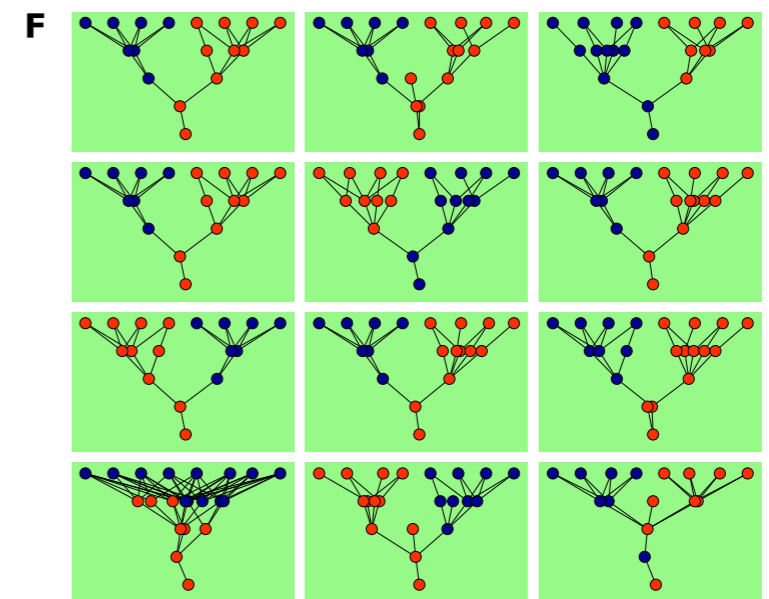
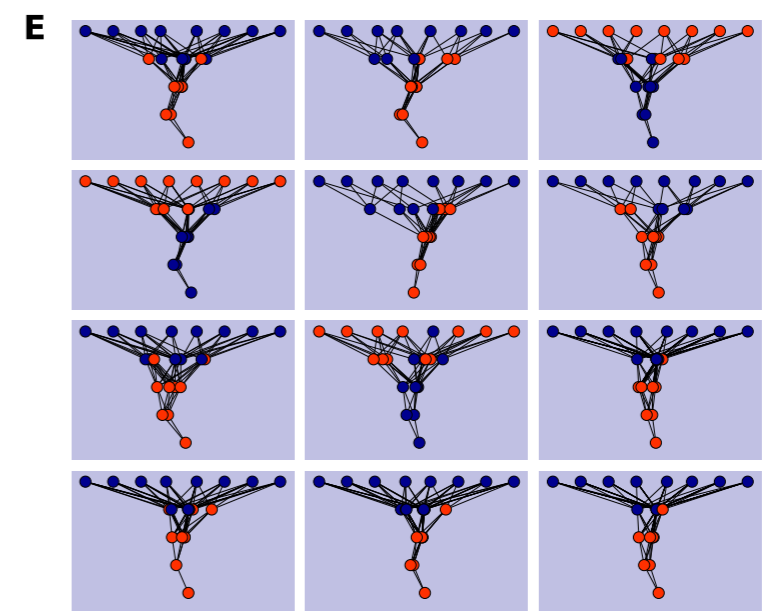
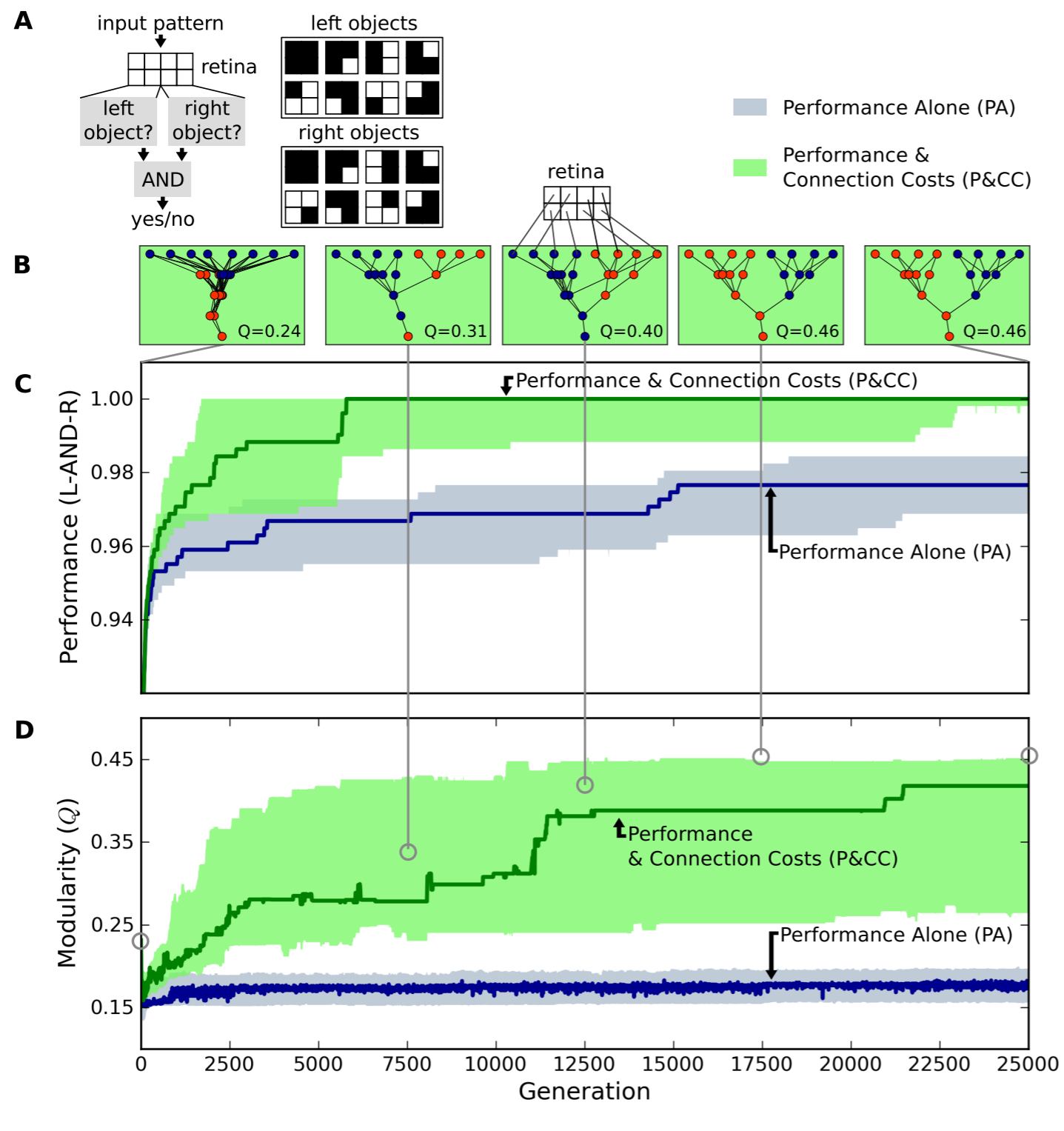
Retina Problem



Summary

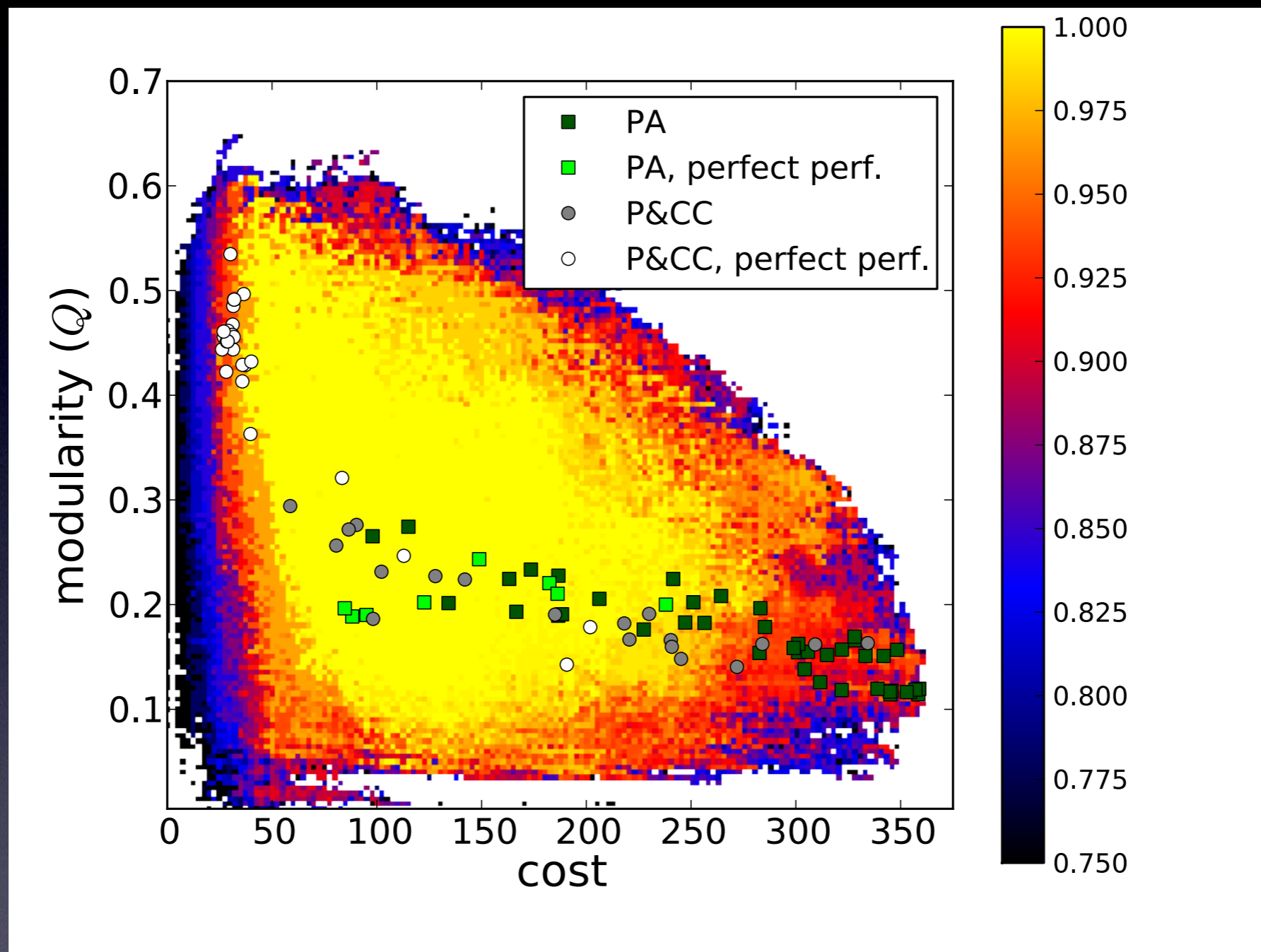


- Performance Alone (PA)
- Performance & Connection Costs (P&CC)



- P&CC significantly more modular, higher-performing ($p < 0.0001$)
- Perfect decomposition in 56% of P&CC, never for PA ($p < 0.0001$)

Why?

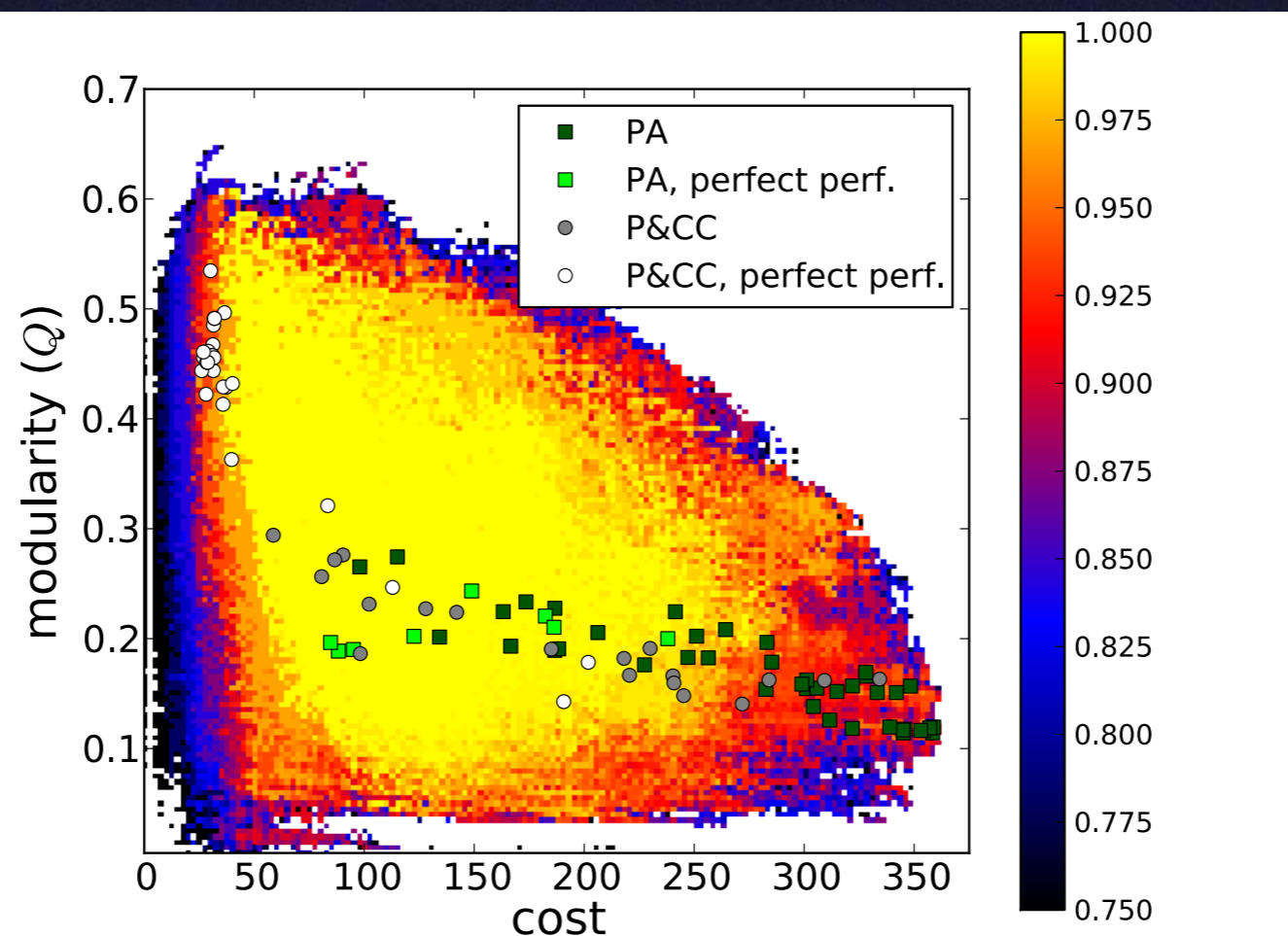
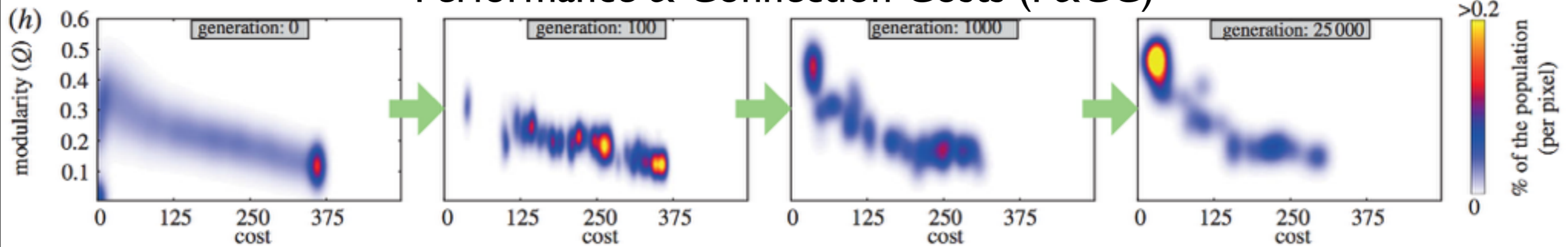


- New technique: MOLE map
 - Multi-Objective Landscape Exploration

Performance Alone (PA)



Performance & Connection Costs (P&CC)

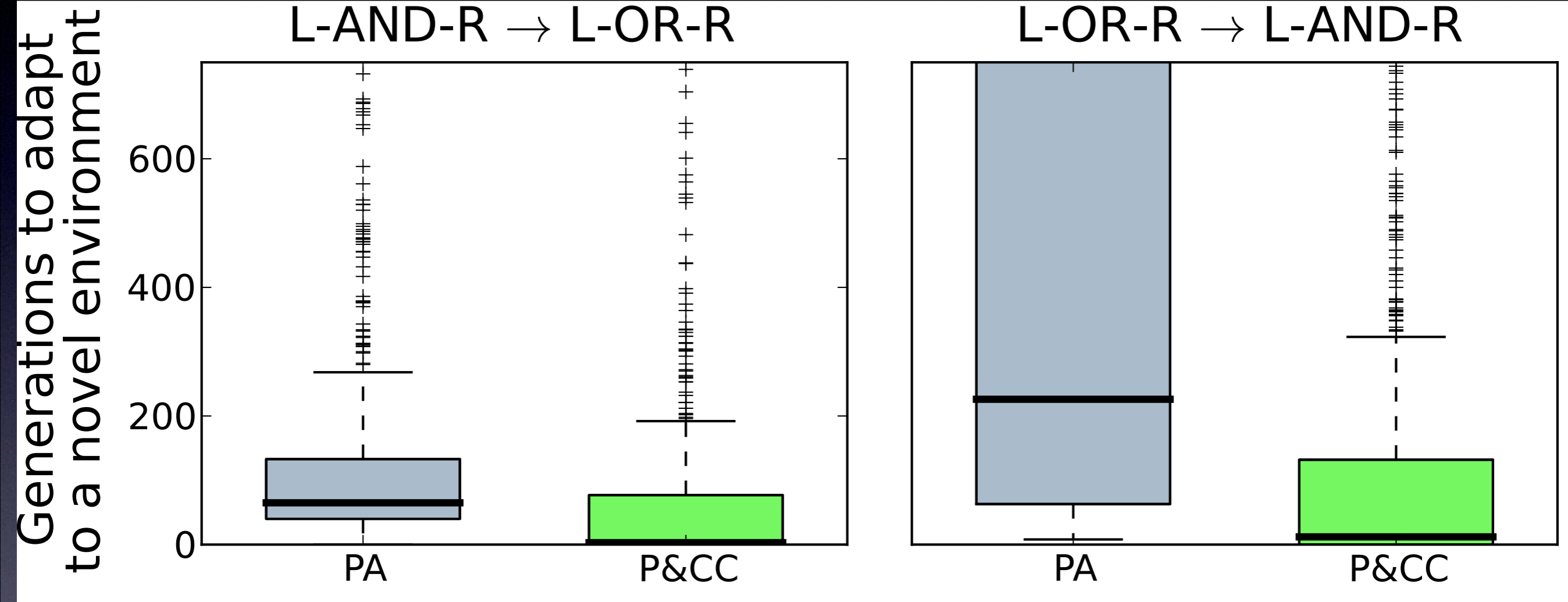


More evolvable?

- Evolved in one environment, transfer to another
 - L-AND-R \rightarrow L-OR-R
 - L-OR-R \rightarrow L-AND-R
- Ran extra trials until 50 had perfect networks

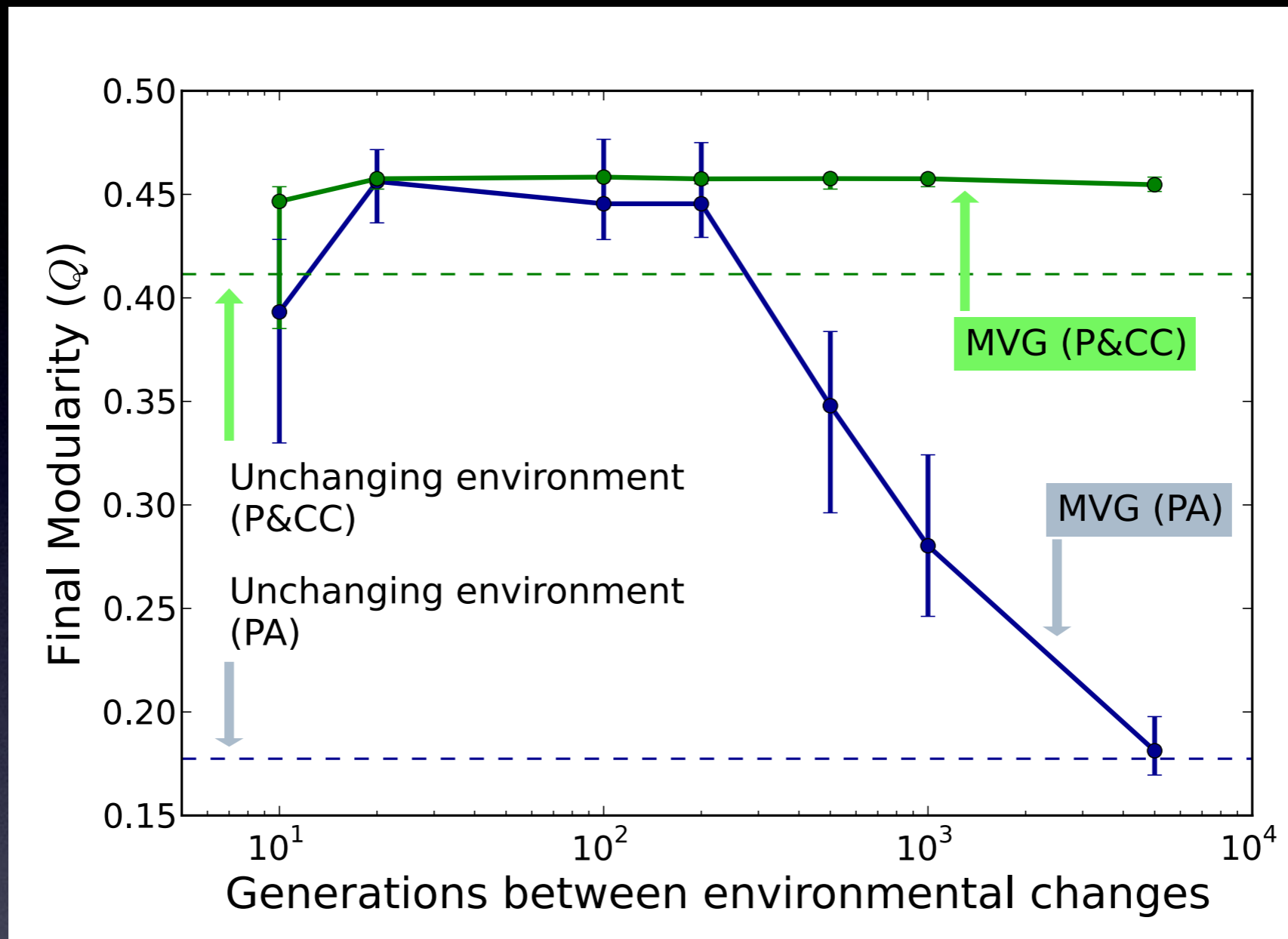
P&CC significantly more evolvable

$P\&CC < PA$ ($p < 0.0001$)



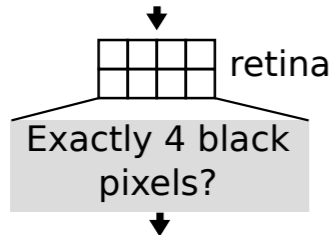
Evolve modularity to reduce connection costs, which happens to help because of problem-modularity

Generality

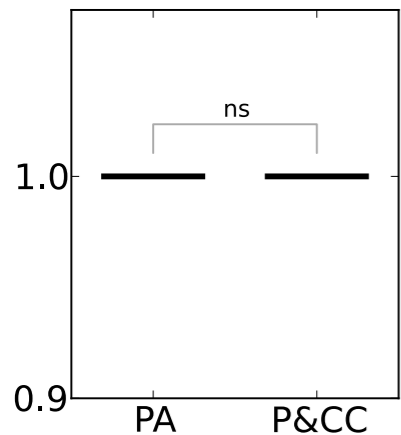


- Modularity forces can combine
- P&CC less sensitive to rate of environmental change
 - P&CC \geq MVG at its strongest

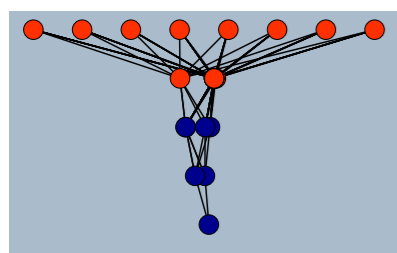
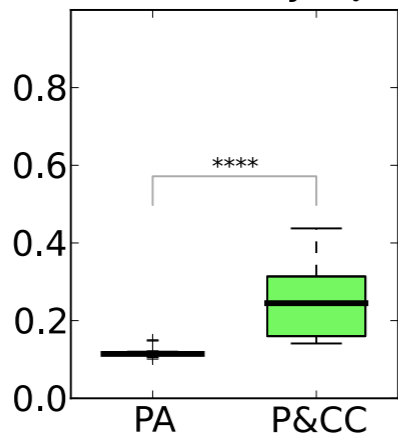
A Non-Modular Problem



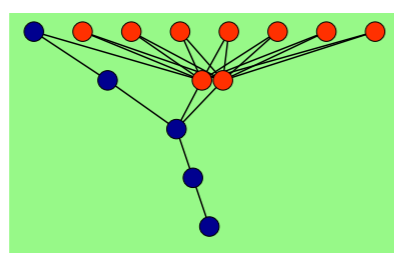
Performance



Modularity (Q)

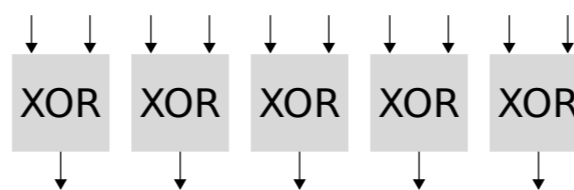


PA

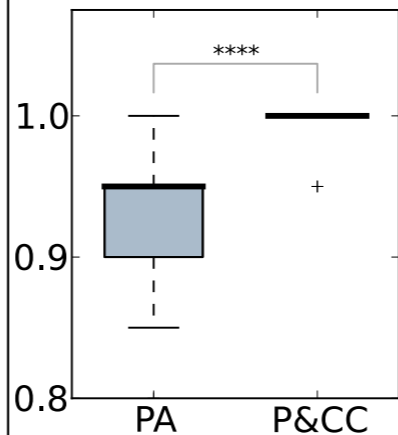


P&CC

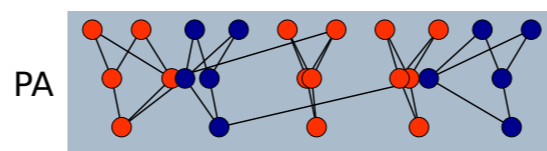
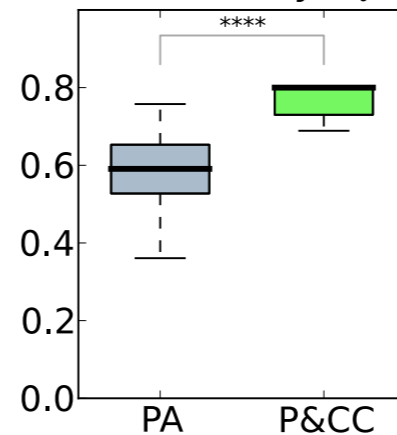
B Multiple, Separable Problems



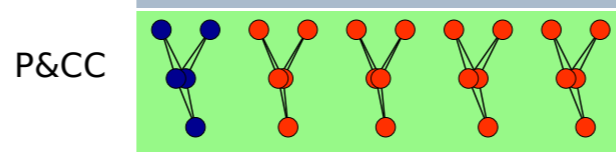
Performance



Modularity (Q)

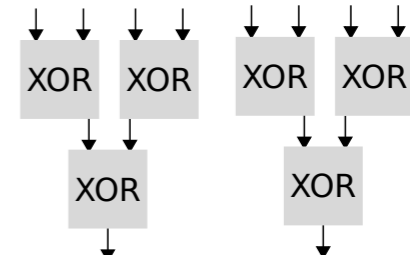


PA

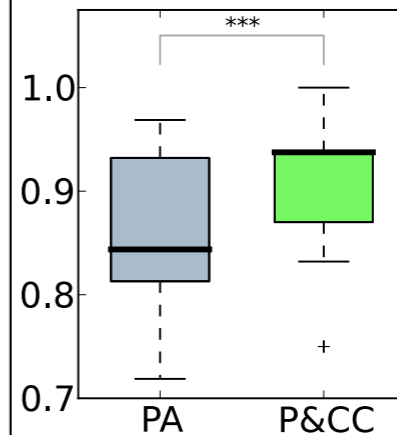


P&CC

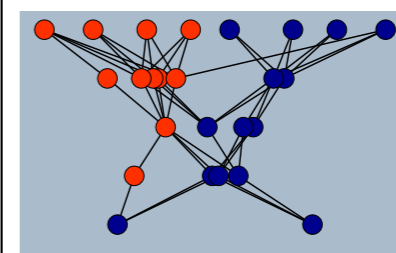
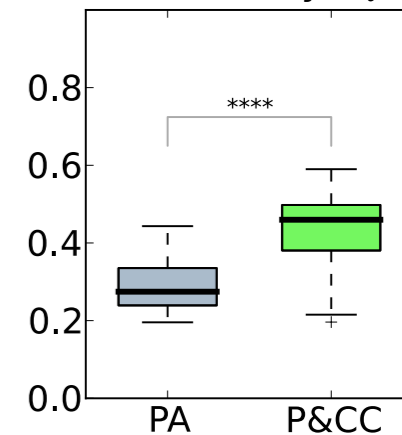
C Hierarchical, Separable Problems



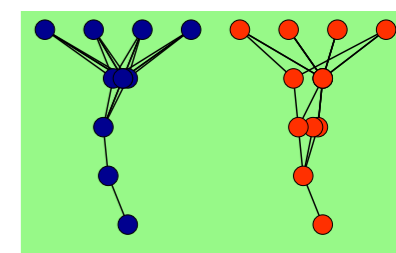
Performance



Modularity (Q)

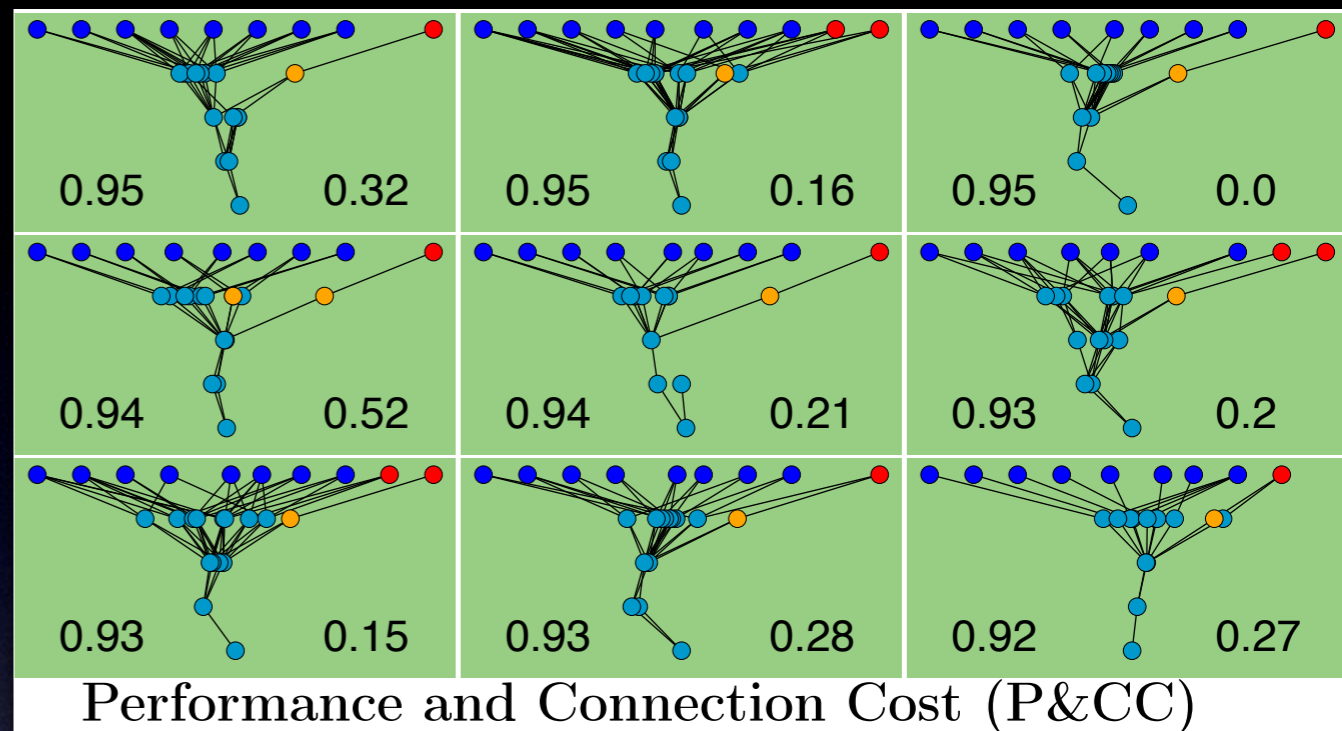
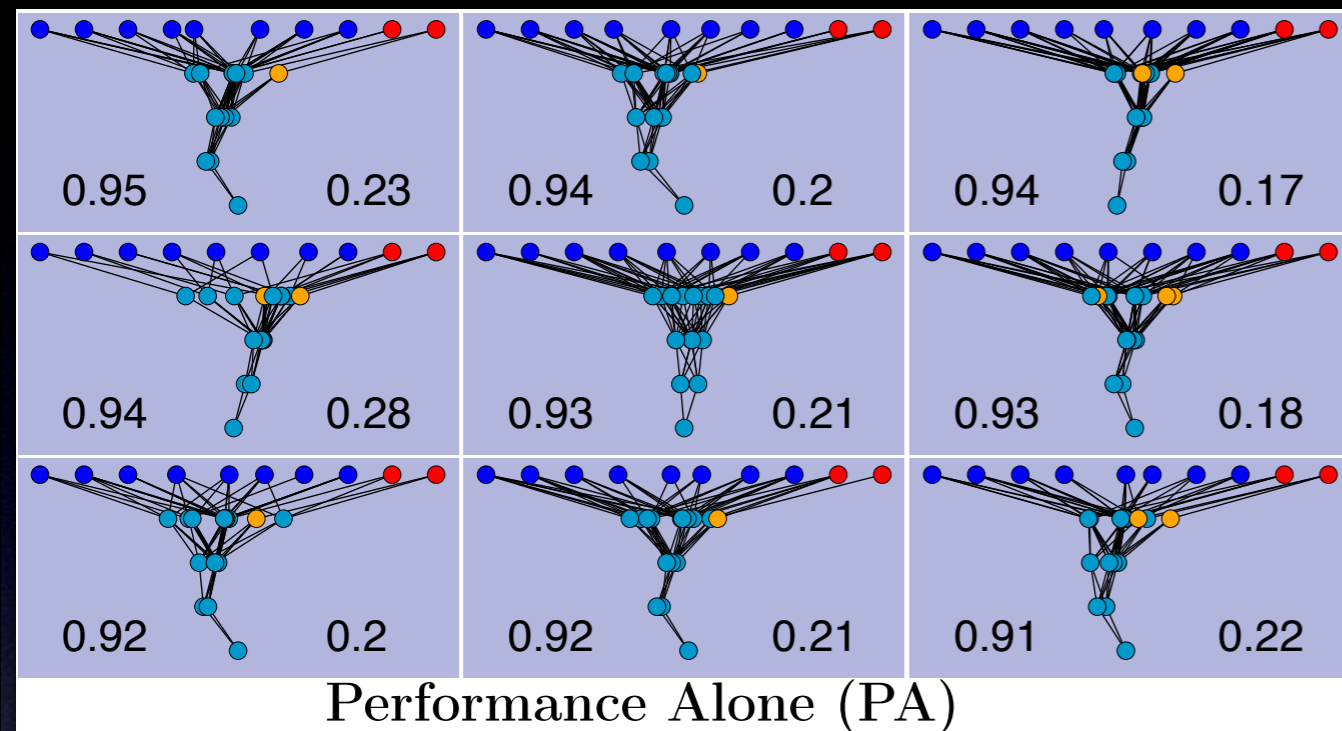


PA

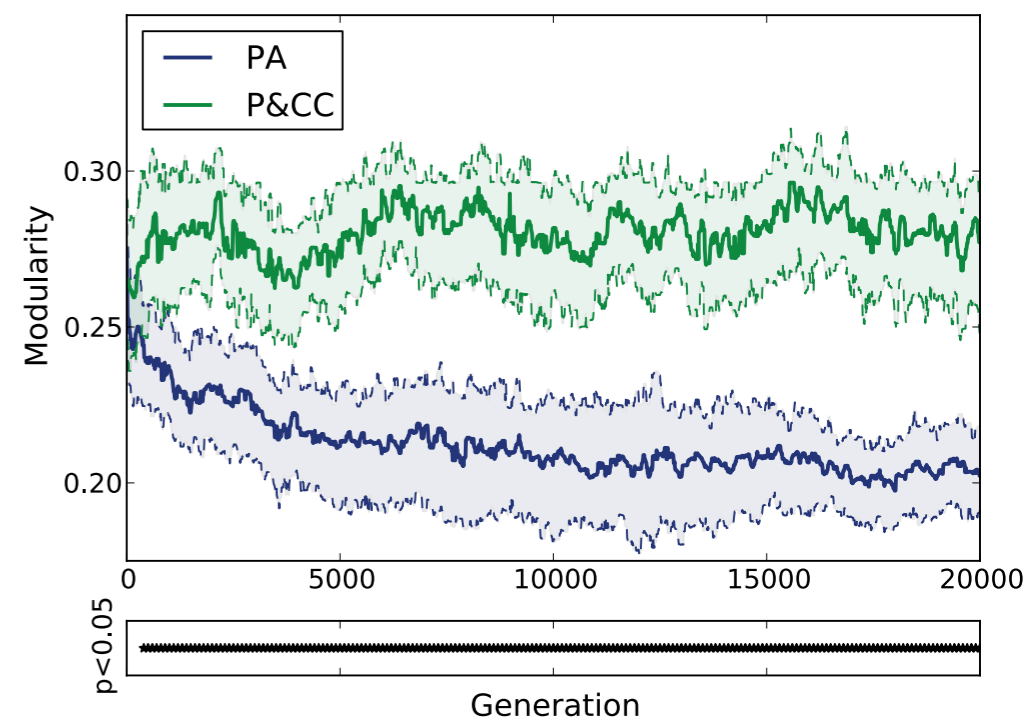
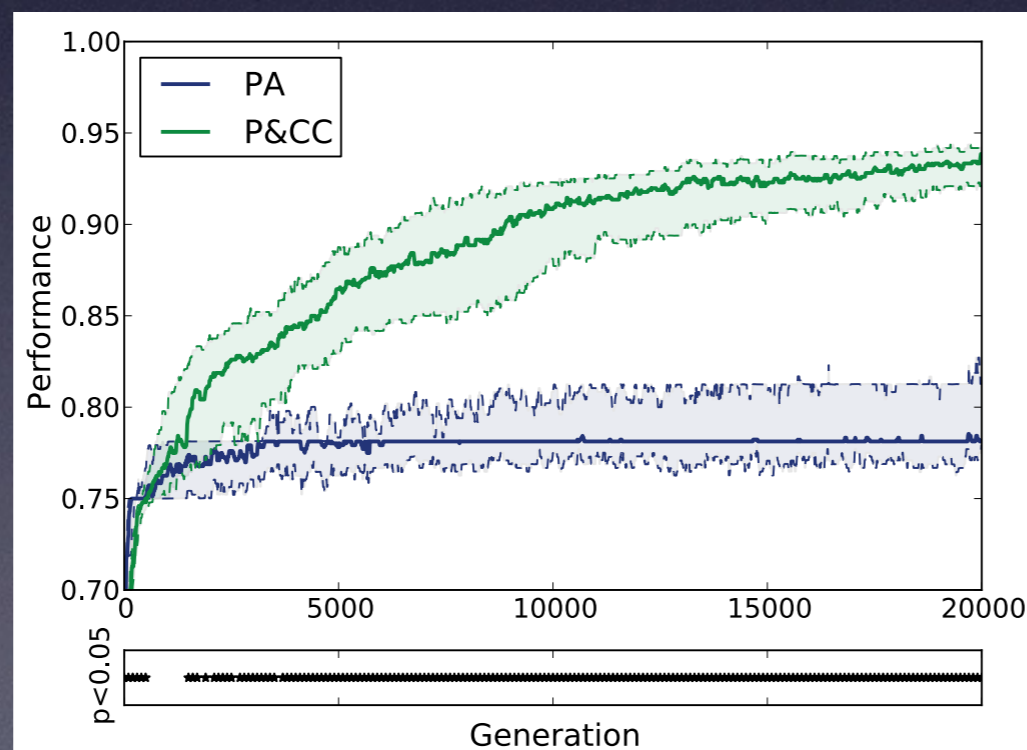


P&CC

Modularity Improves Learning



CC minimizes
“catastrophic
forgetting”



Biological Implications

- May be a major explanatory force behind evolved modularity
- May bootstrap evolvability explanations
 - initial modularity due to connection costs
 - indirect selection for evolvability takes over

Neuroevolution Implications

- Adding a cost increased
 - performance
 - modularity
 - evolvability
- Could be powerful technique for evolutionary algorithms

Non-Adaptive Evolvability

- Evolution fails to evolve optimal mutation rates
 - any evolvability likely due to cost of fidelity
- Evolution fails to evolve modularity
 - any evolvability likely due to connection costs
- How many other cases of evolvability are non-adaptive?
 - converse: how many examples of evolvability do we know are adaptive?

Non-Adaptive Evolvability

Thanks!

- Evolution fails to evolve optimal mutation rates
 - any evolvability likely due to cost of fidelity
- Evolution fails to evolve modularity
 - any evolvability likely due to connection costs
- How many other cases of evolvability are non-adaptive?
 - converse: how many examples of evolvability do we know are adaptive?



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Computer Science



Minimizing Connection Costs

- Many studies suggest overall wire length in brains and nervous systems are minimized
 - Most connections in brains are *short*
 - Most nodes are not connected
 - Neuron placement optimized to reduce wire length
- Primary reason may be *connection costs*
 - clear in networks with physical connections (neural)
 - building, maintenance, energy to use, signal delays, weight, etc.
 - may also exist in other networks (e.g. genetic regulatory)
 - slow replication, slow regulation, added constraints