

Evolution and Computation

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The Algorithm as a Lens



- It started with Alan Turing, 60 years ago
- Algorithmic thinking as a novel and productive point of view for understanding and transforming the Sciences
- Major theme of the Simons Institute
- ...and of this symposium
- This talk: Evolution

Evolution before Darwin

- Erasmus Darwin



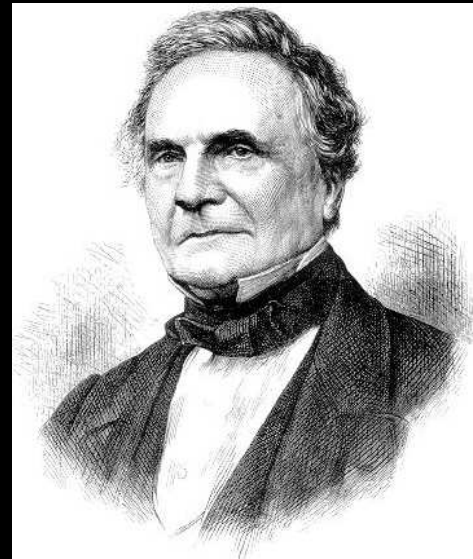
Before Darwin

- J.-B. Lamarck



Before Darwin

- Charles Babbage



[ca. 1820, paraphrased]

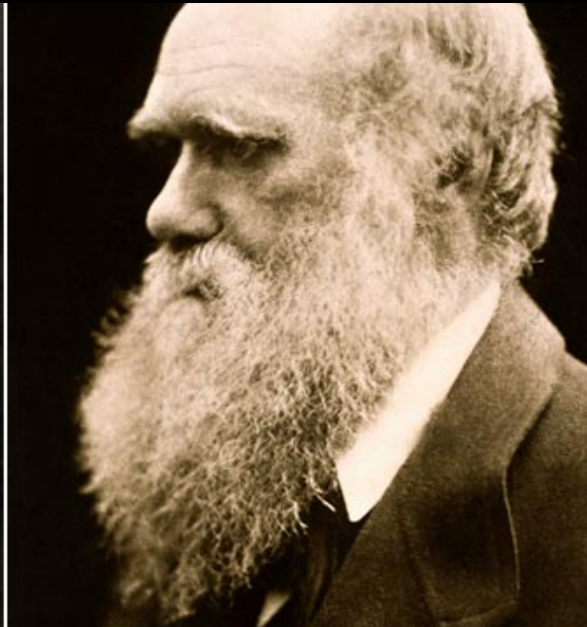
“God created not species, but the Algorithm for creating species”

The Origin of Species



- Natural Selection
- Common Ancestry
- Possibly the world's most masterfully compelling scientific argument
- The six editions: 1859, 1860, 1861, 1866, 1869, 1872

The Wallace-Darwin papers: Exponential Growth



Brilliant argument, and yet many questions left unasked, e.g.:

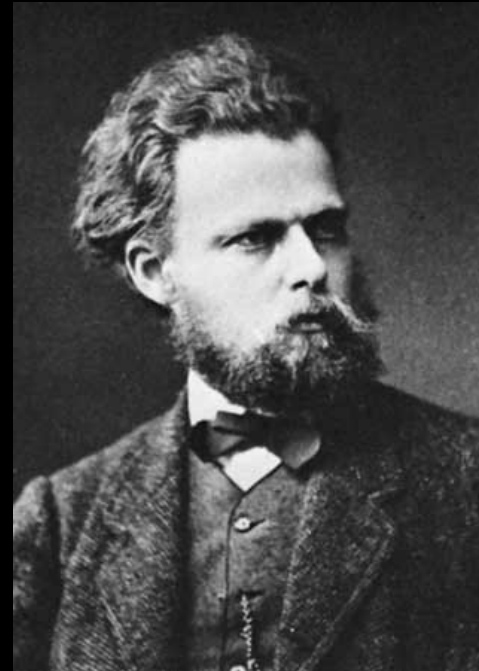
- How does novelty arise?
- What is the role of sex?

Cryptography against Lamarck

- A. Weismann

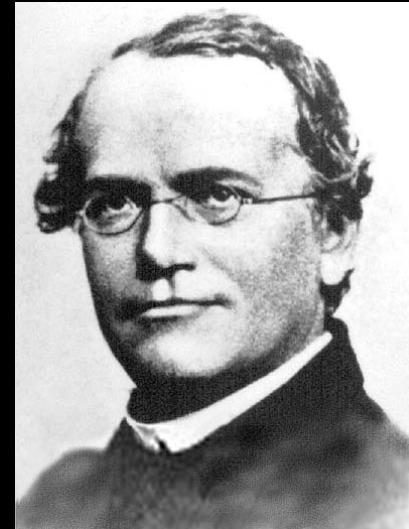
[ca. 1880, paraphrased]

“The mapping from genotype to phenotype is one-way”



Genetics

- Gregor Mendel [1866]
- Number of citations
between 1866 and 1901:



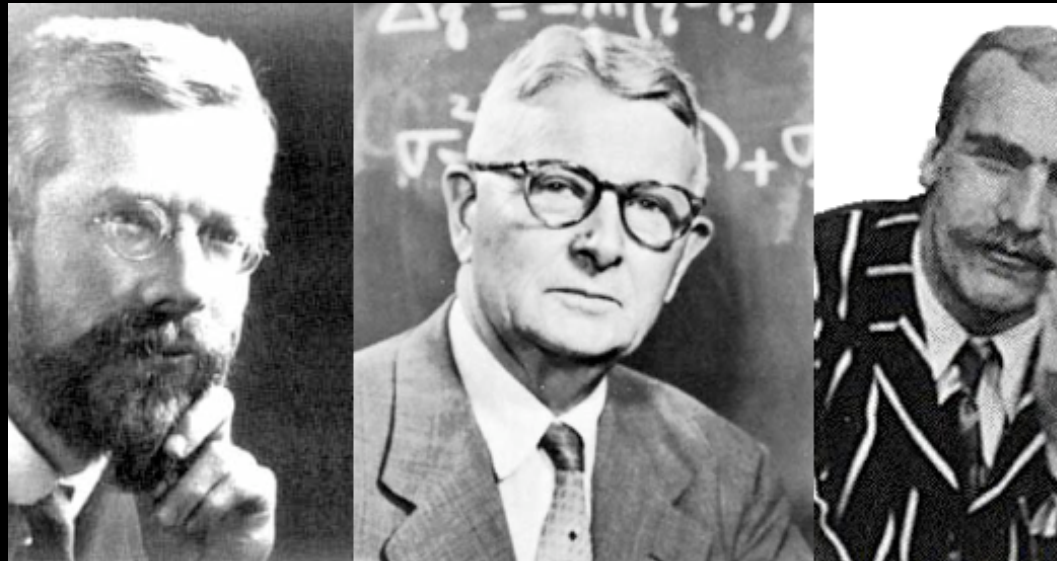
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The crisis in Evolution

1900 - 1920

- Mendelians vs. Darwinians
- Geneticists vs. Biometricists/
Gradualists

The “Modern Synthesis” 1920 - 1950



Fisher – Wright - Haldane

Big questions remain

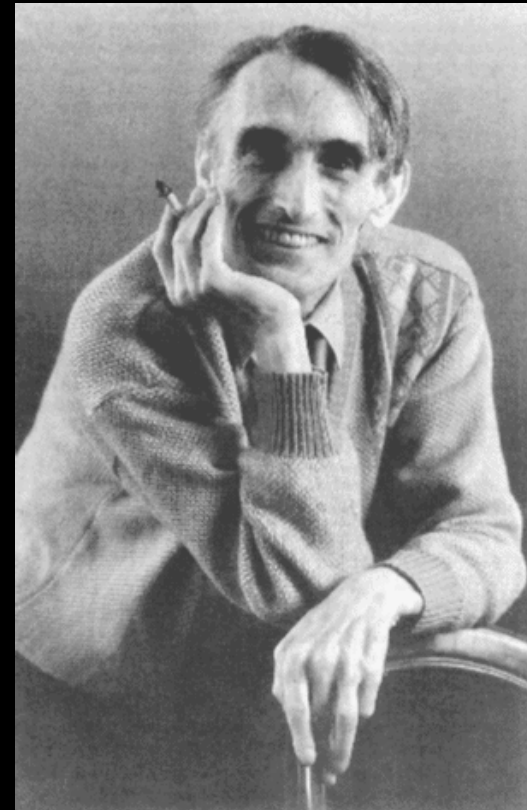
e.g.:

- How does novelty arise?
- What is the role of sex?

Disbelief

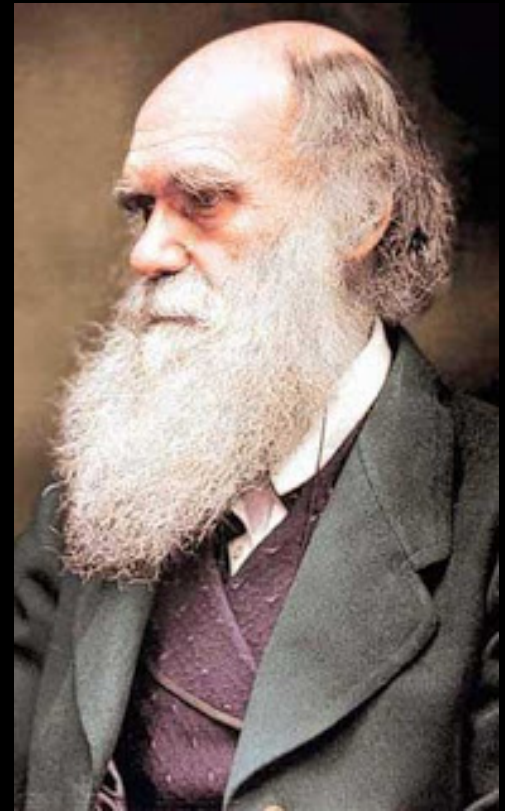
“Our thesis is that Neo-Darwinism cannot explain the basic phenomena of evolution on the basis of physico-chemistry”

Schützenberger, 1966



Disbelief at the top

“The eye to this day gives me a cold shudder.”



Disbelief, algorithmic version

“What algorithm could have achieved

all this

in a mere 10^{12} steps?”

(surprise: we have an answer...)

Valiant's Evolvability



“ How do you find a 3-billion long string in 3 billion years? ”

L. G. Valiant

Computationally-inspired model of Evolution shown to amount to a weak form of learning

Evolution and CS Practice: Genetic Algorithms [ca. 1980s]

- To solve an optimization problem...
- ...create a population of solutions/genotypes
- ...who evolve through mutations and sex...
- ...and procreate with success proportional to their objective function value
- Eventually, some very good solutions are bound to arise in the soup

And in this Corner...

Simulated Annealing

- Inspired by *asexual* reproduction
- Mutations are adopted with probability increasing with fitness/objective differential

The Mystery of Sex Deepens

- Simulated annealing (asexual reproduction) works fine
- Genetic algorithms (sexual reproduction) don't work
- In Nature, the opposite happens: Sex is successful and ubiquitous



A Radical Thought

- What if sex is a mediocre optimizer of fitness (= expectation of offspring)?
- What if sex optimizes something else?
- And what if this something else is its *raison d'être*?

Mixability!

- [Livnat et al, PNAS 2008]
- Simulations show that natural selection under **asex** optimizes fitness
- But under **sex** it optimizes *mixability*:
- = The ability of alleles (gene variants) to perform well with a broad spectrum of other alleles

Explaining Mixability: The Fisher-Wright model

- Fitness landscape of a 2-gene organism



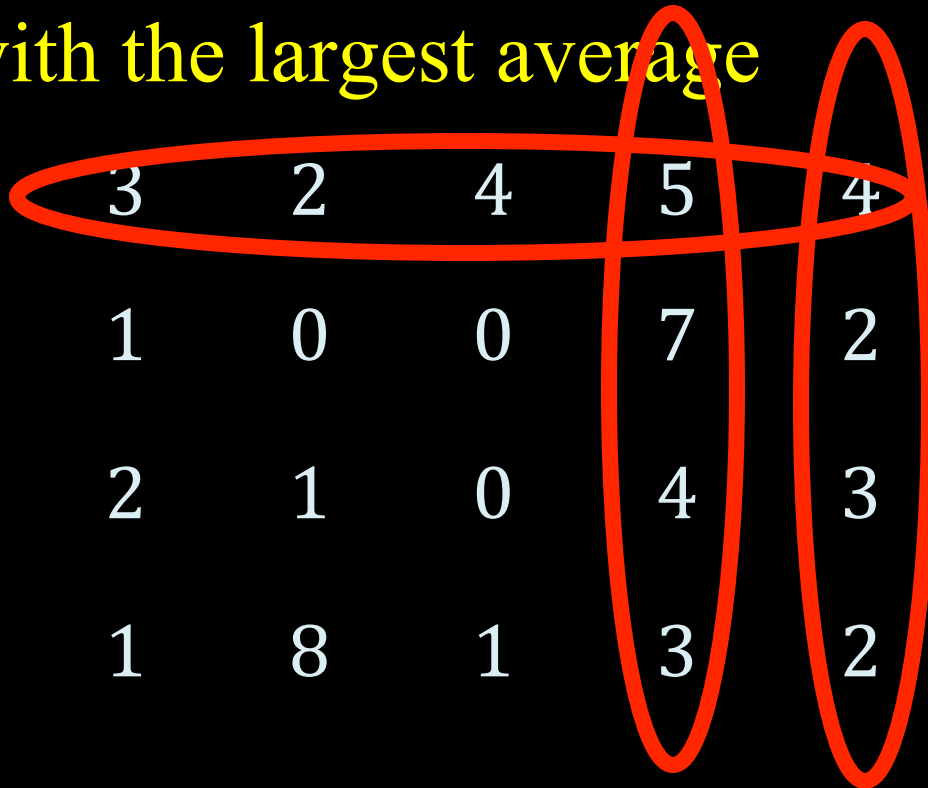
Explaining Mixability (cont)

- Asex will select the largest numbers

| | | | | |
|---|---|---|---|---|
| 3 | 2 | 4 | 5 | 4 |
| 1 | 0 | 0 | 7 | 2 |
| 2 | 1 | 0 | 4 | 3 |
| 1 | 8 | 1 | 3 | 2 |

Explaining Mixability (cont)

- But sex will select the rows and columns with the largest average



| | | | | |
|---|---|---|---|---|
| 3 | 2 | 4 | 5 | 4 |
| 1 | 0 | 0 | 7 | 2 |
| 2 | 1 | 0 | 4 | 3 |
| 1 | 8 | 1 | 3 | 2 |

Pointer Dogs



Pointer Dogs

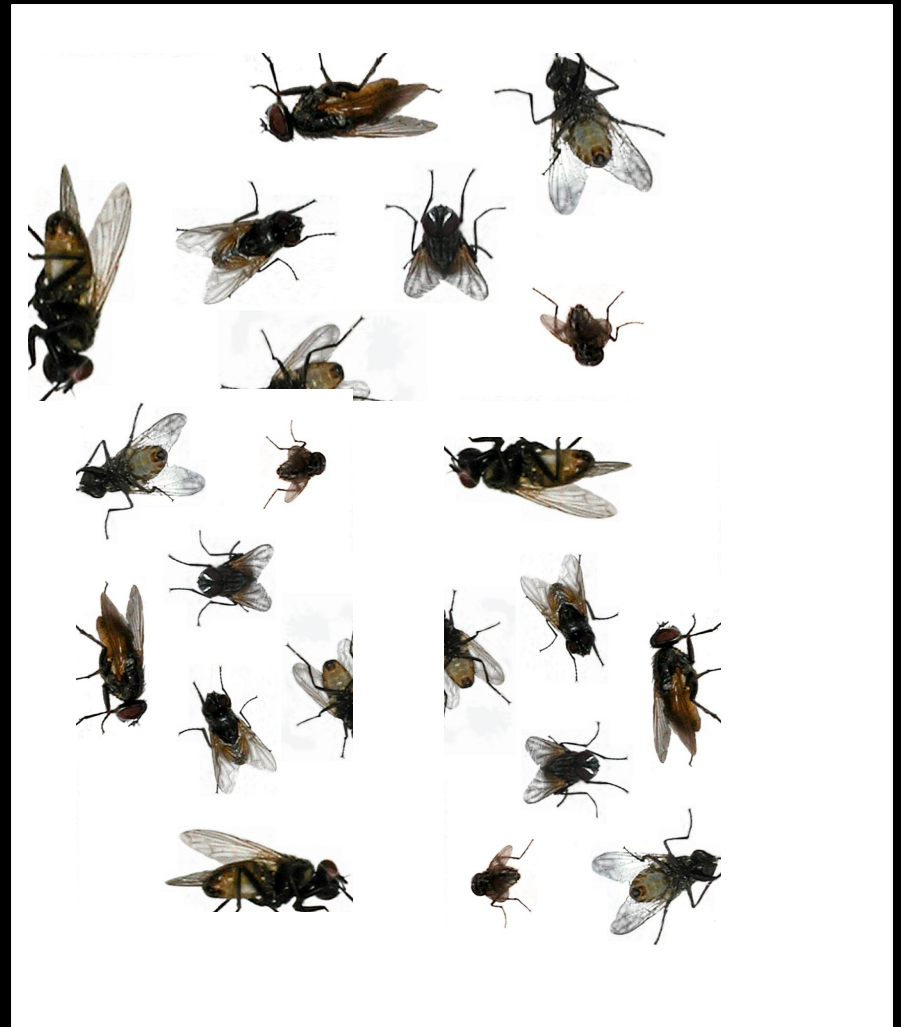


C. H. Waddington

Waddington's Experiment (1952)

Generation 1

Temp: 20° C



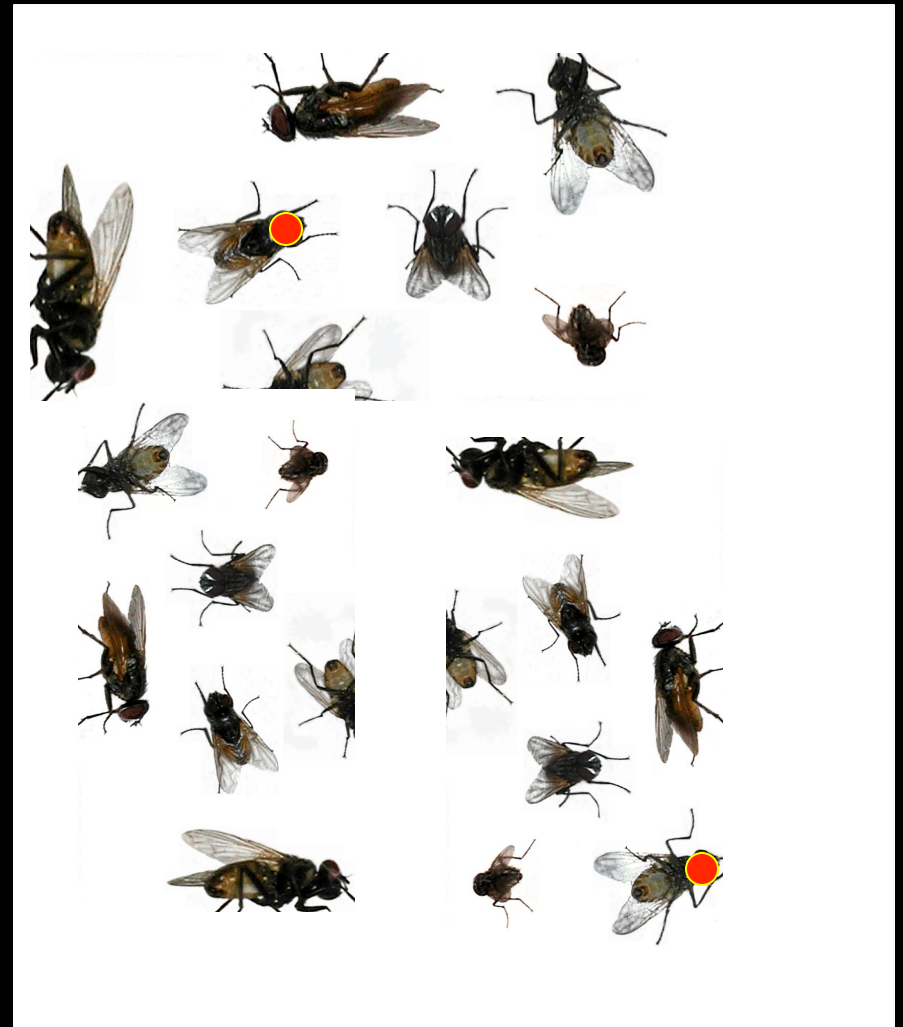
Waddington's Experiment (1952)

Generation 2-4

Temp: 40° C

~15% changed

Select and breed those



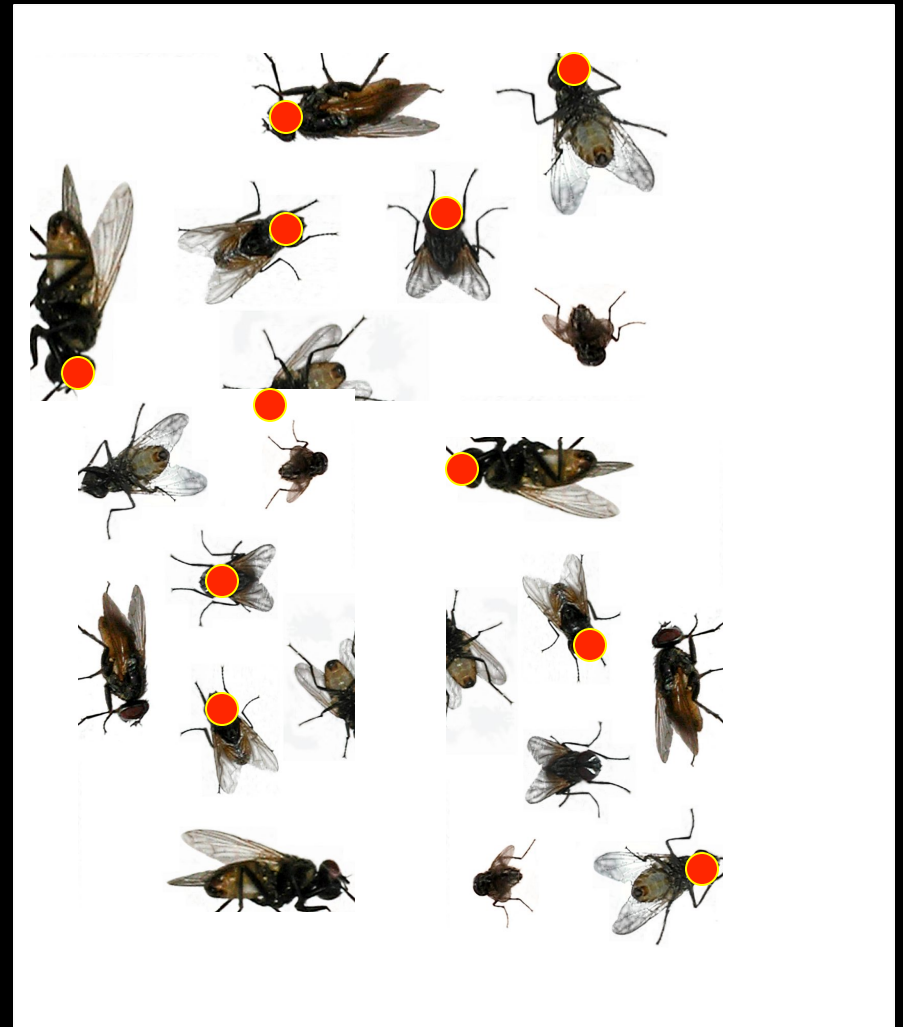
Waddington's Experiment (1952)

Generation 5

Temp: 40° C

~60% changed

Select and breed those



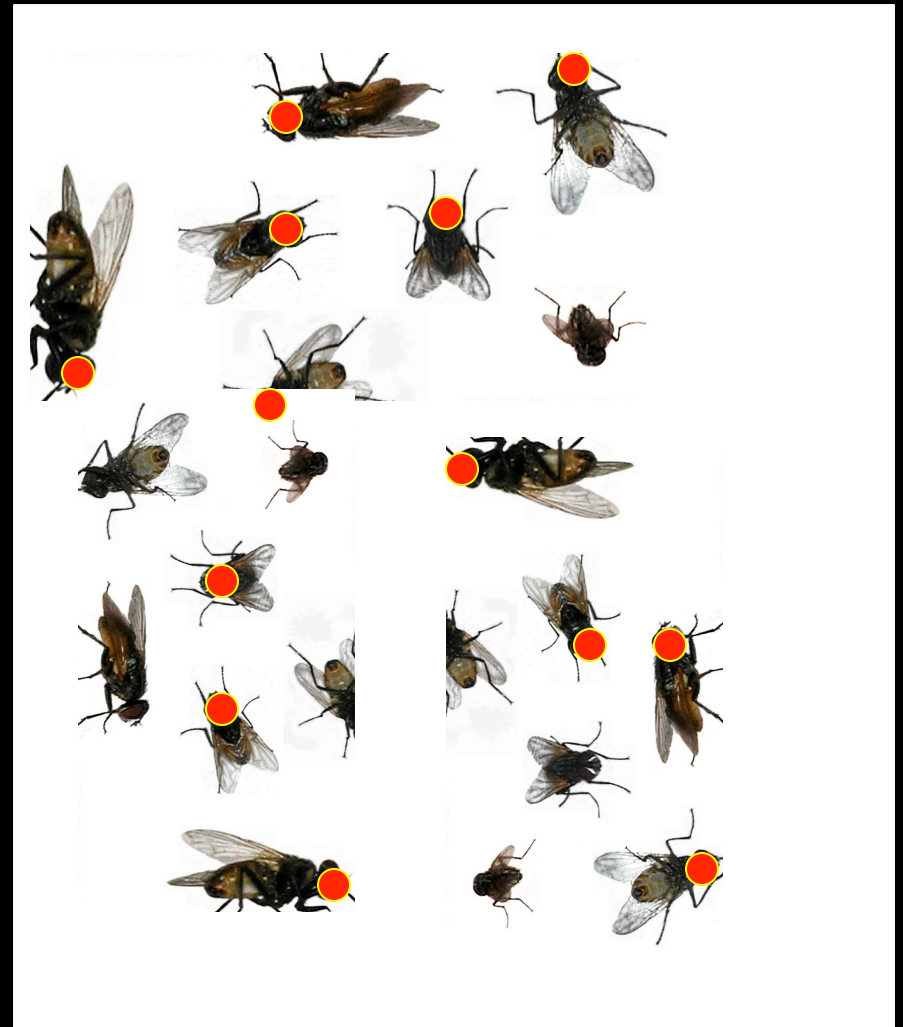
Waddington's Experiment (1952)

Generation 6

Temp: 40° C

~63% changed

Select and breed those



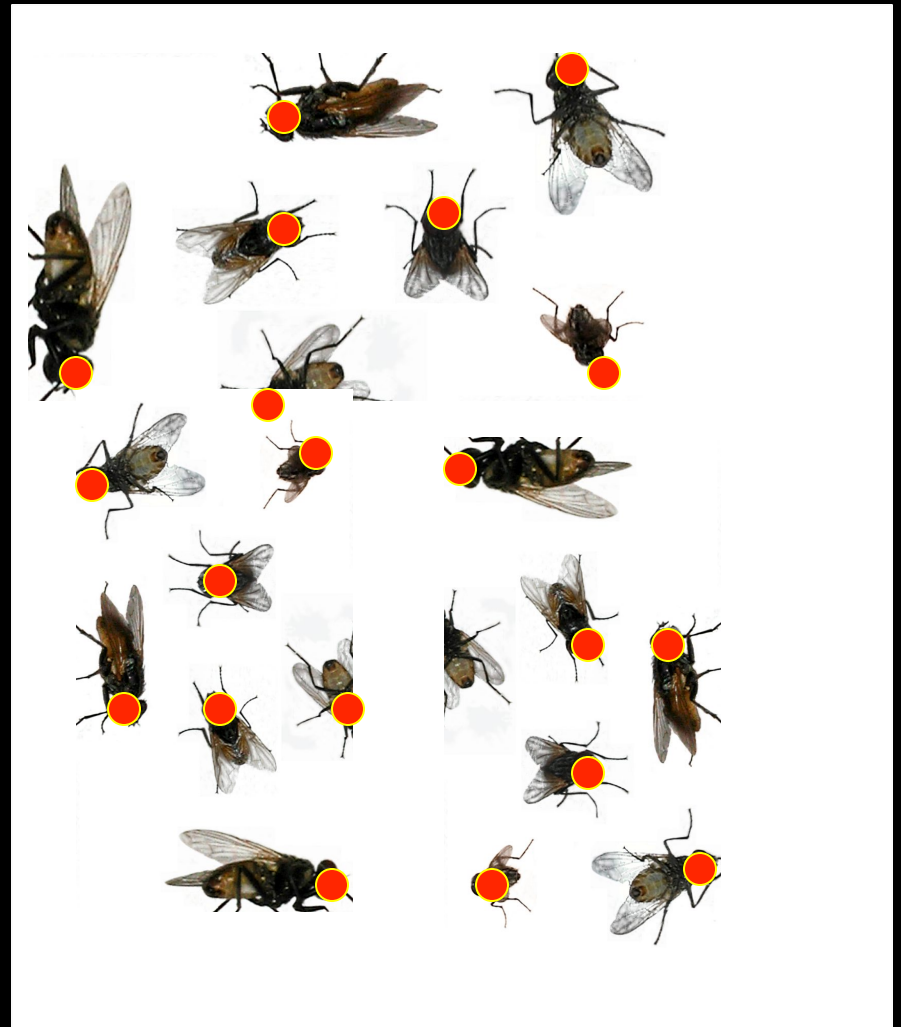
Waddington's Experiment (1952)

(...)

Generation 20

Temp: 40° C

~99% changed

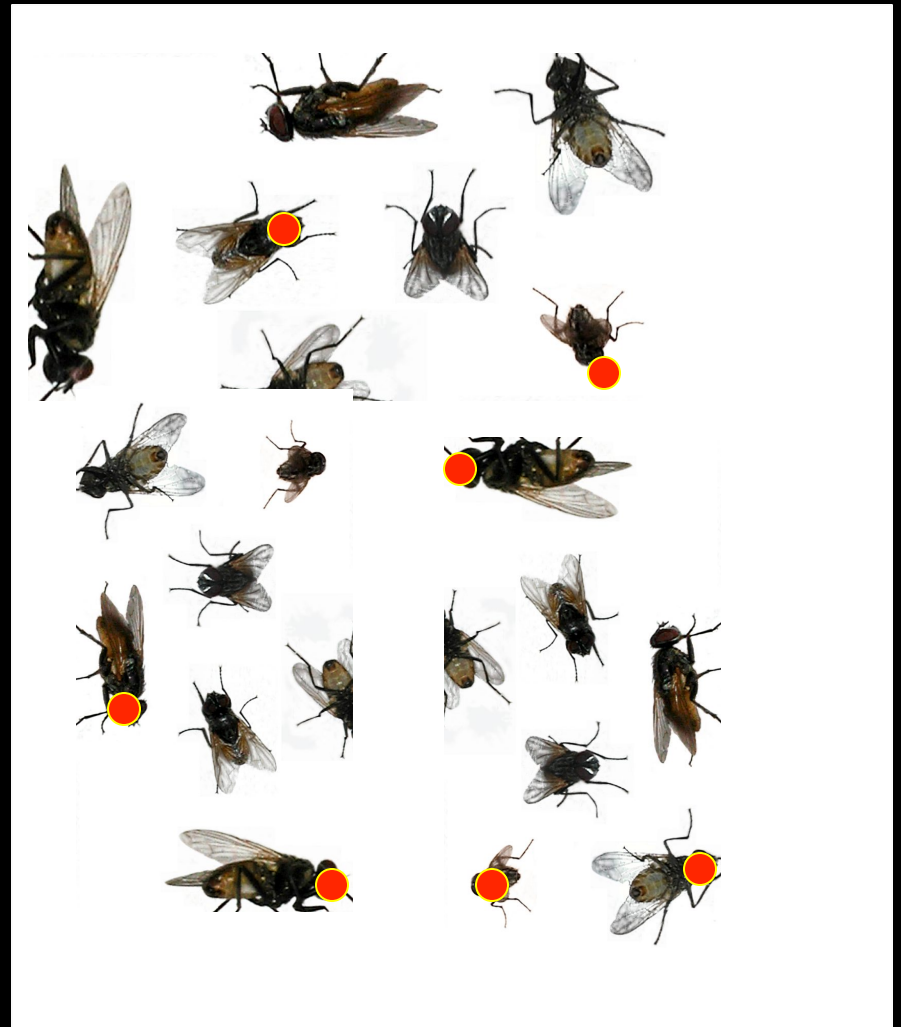


Surprise!

Generation 20

Temp: 20° C

~25% stay changed!!



- Adaptation
genetics

ome



Is There a Genetic Explanation?

Function $f(x, h)$ with these properties:

- Initially, $\text{Prob}_{x \sim p[0]} [f(x, h = 0)] \approx 0\%$
- Then $\text{Prob}_{p[0]} [f(x, 1)] \approx 15\%$
- After breeding $\text{Prob}_{p[1]} [f(x, 1)] \approx 60\%$
- Successive breedings, $\text{Prob}_{p[20]} [f(x, 1)] \approx 99\%$
- Finally, $\text{Prob}_{p[20]} [f(x, 0)] \approx 25\%$

A Genetic Explanation

- Suppose that “red head” is this Boolean function of 10 genes and “high temperature”
“red head” = “ $x_1 + x_2 + \dots + x_{10} + 3h \geq 10$ ”
- Suppose also that the genes are independent random variables, with p_i initially half, say
- All properties of the Waddington experiment satisfied
- [Stern *AN* 1958]

Arbitrary Boolean Functions

- What if we have an arbitrary function of genes (no environmental variable h)
- Suppose the satisfying genotypes have a fitness advantage ($1 + \varepsilon$ vs. 1, say)
- Will this trait be fixed eventually?

Arbitrary Functions: *Yes!*

Theorem: Any Boolean function of genes which confers an evolutionary advantage will be eventually fixed (with high probability)

(2013; with Adi Livnat, Aviad Rubinstein, Greg Valiant, Andrew Won)

Which means that...

- “With sex, all moderate-sized Boolean functions are evolvable.”
- *“Look, Ma, no mutations!”*
- Novel complex traits can emerge, through sex, in the whole population, without “Fisherian propagation”

Neutral Theory and Weak Selection

- Kimura 1970: Evolution proceeds not by leaps upwards, but mostly “horizontally,” through statistical drift
- Weak selection: the values in the fitness matrix are very close, say in $[1 - \varepsilon, 1 + \varepsilon]$

Changing the subject: The experts problem

- Every day you must choose one of n experts
- The advice of expert i on day t results in a gain $G[i, t]$ in $[-1, 1]$
- Challenge: Do as well as the best expert *in retrospect*
- Surprise: It can be done!
- [Hannan 1958, Cover 1980, Winnow, Boosting, no-regret learning, MWUA, ...]

Multiplicative weights update

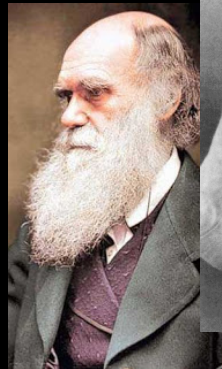
- Initially, assign all experts same weight/probability
- At each step, increase the weight of each by $(1 + \epsilon G[i, t])$ (and then normalize)
- **Theorem:** Does as well as the best expert
- MWUA solves: zero-sum games, linear programming, convex programming, network congestion,...

Disbelief



Computer scientists find it hard to believe that such a crude technique solves all these sophisticated problems

(cf: the other disbelievers)



Theorem: Under weak selection, evolution of a species *is a game*

- the players are the genes
- the strategies are the alleles
- the common utility is the fitness of the organism (*coordination game*)
- the probabilities are the allele frequencies
- game is played through multiplicative updates

(2013, with E. Chastain, A. Livnat, U. Vazirani)

Finally...

- *Variance preservation*: MWUA is known to maximize entropy
- The curious successes of Evolution and MWUA: Two mysteries united
- *This* is the role of sex in Evolution

Sooooo...

- The theory of life is deep and fascinating
- And rife with interesting technical problems that are unexpectedly computational
- How robust? (Strong selection? Complex landscapes?)
- Environment? Mutations?
- Test?

Thanks!