

Adaptive evolution of molecular quantitative traits

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Thanks

Theory of quantitative traits:

Torsten Held (Cologne)
Armita Nourmohammad (Princeton)
Stephan Schiffels (Sanger)

Drosophila gene expression:

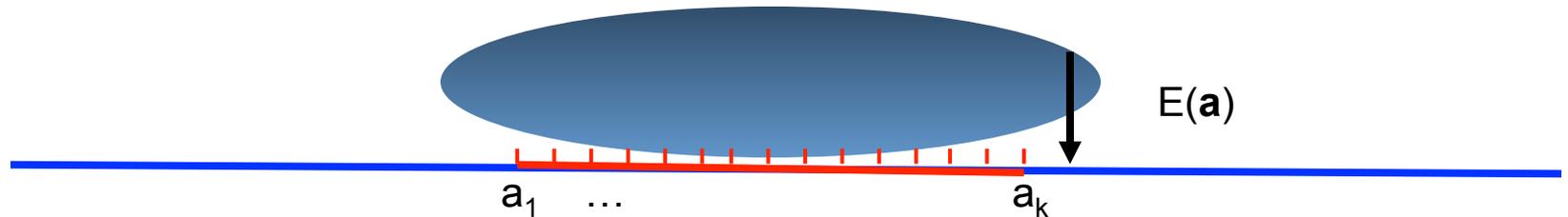
Armita Nourmohammad (Princeton)
Joachim Rambaut (Cologne, Montpellier)
Torsten Held (Cologne)
Johannes Berg (Cologne)



Examples of phenotypes

- **Molecular traits:**

Molecular binding interactions, e.g., between transcription factors and regulatory DNA



- **Cellular traits:**

metabolic, regulatory, and signaling pathways

- **Organismic traits:**

body weight and size, shapes, metabolic rates
fitness, longevity

➔ **Quantitative traits occur in coarse-grained descriptions of increasing level.**

Evolutionary theories

- **Sequence level: population genetics**
- **Cellular level: systems biology**
- **Organismic level: quantitative genetics**

➔ **Evolutionary systems biology**

- **We need to rethink the quantitative genetics of molecular traits.**

1. Evolutionary statistics of quantitative traits

Population genetics

- **Stochastic equation of motion for population frequencies:**

$$\frac{dx}{dt} = \underbrace{s(x)g(x)}_{\text{selection}} + \underbrace{mx}_{\text{mutations}} + \underbrace{\eta(x, t)}_{\substack{\langle \eta(x, t)\eta(x, t') \rangle = \frac{g(x)}{N} \delta(t - t') \\ \text{reproductive fluctuations} \\ \text{(genetic drift)}}$$

- **Equation of motion for the frequency distribution:**

$$\frac{\partial}{\partial t} P(x, t) = \nabla \left[\frac{1}{N} \nabla g(x) - s(x)g(x) - mx \right] P(x, t)$$

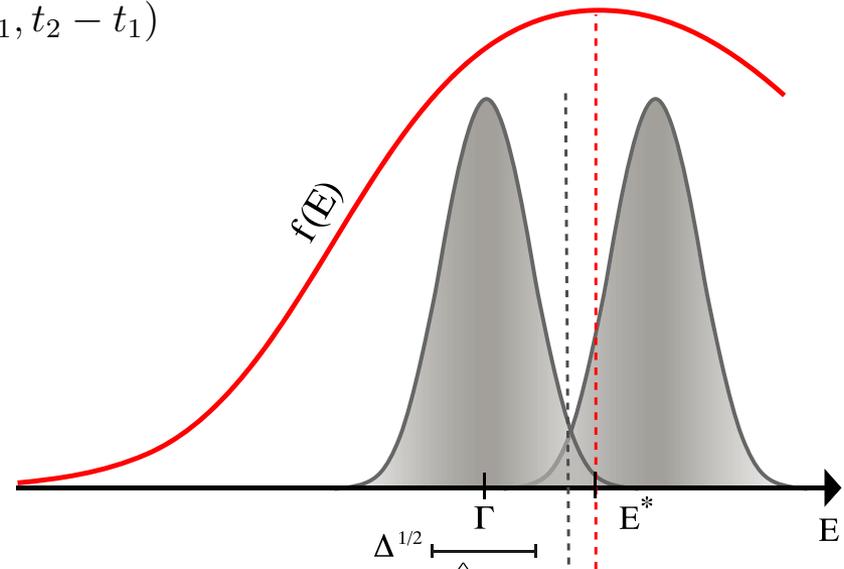
Evolutionary statistics of phenotypes

- **Trait mean and diversity** within one population:

$$\Gamma \equiv \bar{E} = \int E \mathcal{W}(E) dE, \quad \Delta \equiv \overline{(E - \Gamma)^2} = \int (E - \Gamma)^2 \mathcal{W}(E) dE,$$

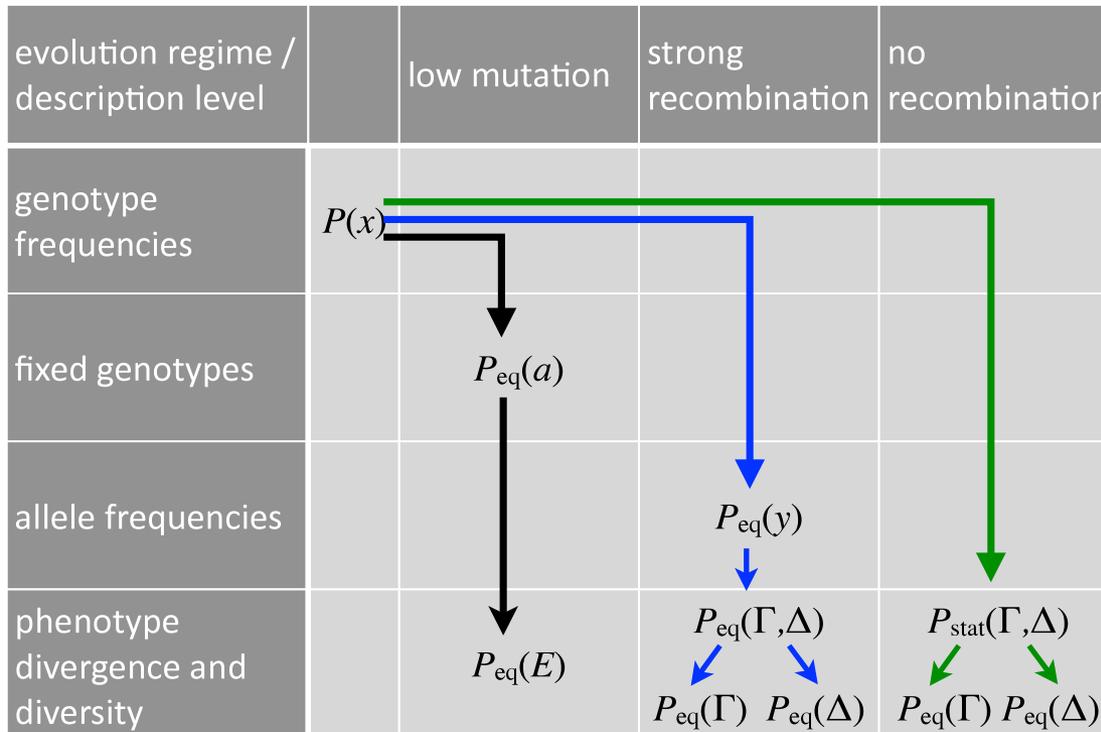
- **Trait divergence** between populations:

$$\langle (\Gamma(t_1) - \Gamma(t_2))^2 \rangle = \int d\Gamma_1 d\Gamma_2 Q(\Gamma_2; \Gamma_1, t_2 - t_1)$$



Coarse-graining to phenotypes

- Genotype-phenotype mapping with and without recombination:



[Berg, Willman, ML, BMC Evol. Biol. 2004; Sella and Hirsh, PNAS 2005]

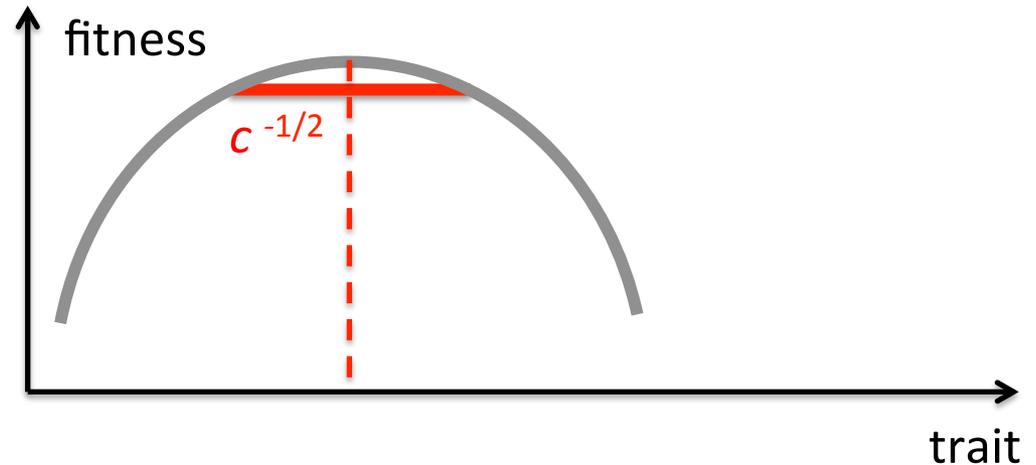
[Quantitative Genetics]

[Nourmohammad, Schiffels, ML, JSTAT 2013]

[Held, Nourmohammad, ML, JSTAT 2014]

Evolution under stabilizing selection

- **Fitness landscape model:**
Minimal dynamics under stabilizing selection



$$\frac{d\Gamma}{dt} = -\langle\Delta\rangle 2c_0(\Gamma - E^*) - 2\mu(\Gamma - \Gamma_0) + \eta_\Gamma(t)$$

with noise generated by genetic drift: $\langle\eta_\Gamma(t)\eta_\Gamma(t')\rangle = \frac{\langle\Delta\rangle}{2N}$

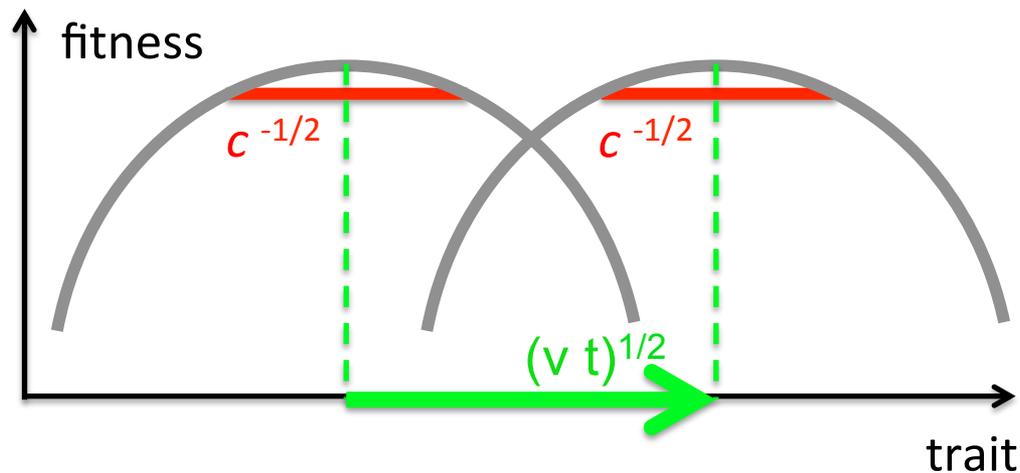
+ similar equation for Δ

- **The landscape model generates equilibria of trait mean and variance:**

$$Q_{\text{eq}}(\Gamma) = \frac{1}{Z} \tilde{Q}_0(\Gamma) \exp[2N\tilde{F}(\Gamma)], \quad Q_{\text{eq}}(\Delta) = \frac{1}{Z} Q_0(\Delta) \exp[2N\tilde{F}(\Delta)],$$

Adaptive evolution

- **Fitness seascape model:**
Minimal adaptive dynamics



$$\frac{d\Gamma}{dt} = -\langle \Delta \rangle 2c_0(\Gamma - E^*(t)) - 2\mu(\Gamma - \Gamma_0) + \eta_\Gamma(t)$$

$$\frac{dE^*}{dt} = \frac{v}{r^2}(E^* - \mathcal{E}) + \eta_{E^*}$$

with noise generated by genetic drift and fitness peak displacements:

$$\langle \eta_\Gamma(t) \eta_\Gamma(t') \rangle = \frac{\langle \Delta \rangle}{2N} \quad \langle \eta_{E^*}(t) \eta_{E^*}(t') \rangle = v E_0^2 \delta(t - t')$$

+ similar equation for Δ .

- **The seascape model generates non-equilibrium adaptive evolution.**

Quantifying adaptation

- A **population history** is a sequence of frequency measurements

$$\mathbf{x} = (x_0, \dots, x_n) \quad \text{at times} \quad (t_0, \dots, t_n).$$

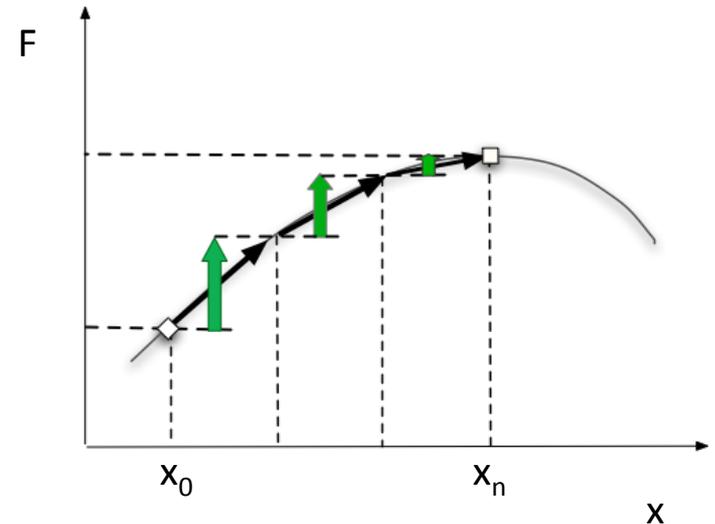
- The **fitness flux** of a population history is the cumulative **selective effect of frequency changes**:

$$\Phi(\mathbf{x}) \equiv \sum_{i=1}^n \Delta x_i s(x_i, t_i).$$

- Flux in a **fitness landscape**:

$$s(x) = \nabla F(x)$$

$$\begin{aligned} \Phi(\mathbf{x}) &= \sum_{i=1}^n \Delta x_i \nabla F(x_i) \\ &= F(x_n) - F(x_0). \end{aligned}$$



Quantifying adaptation

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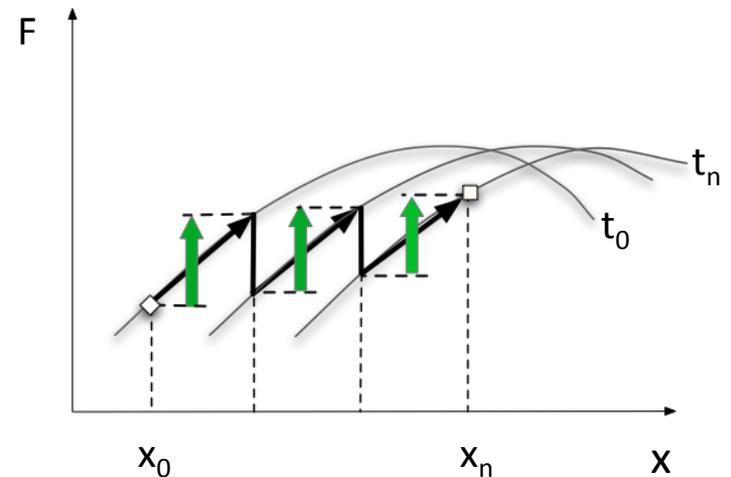
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- Flux in a **fitness seascape**:

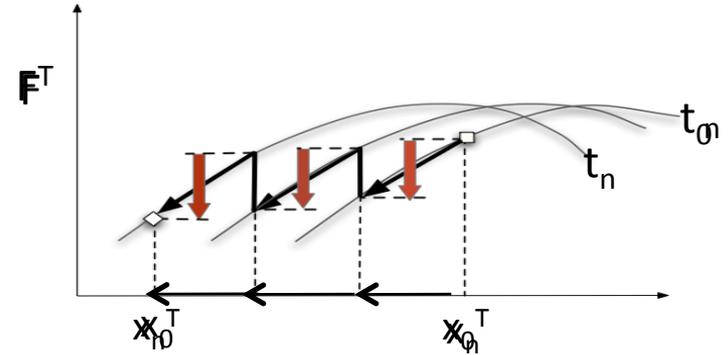
$$s(x, t) = \nabla F(x, t)$$

$$\begin{aligned} \Phi(\mathbf{x}) &= \sum_{i=1}^n \Delta x_i \nabla F(x_i, t_i) \\ &\neq F(x_n, t_n) - F(x_0, t_0). \end{aligned}$$



Quantifying adaptation

- Each population history \mathbf{x} has a **reverse history** \mathbf{x}^T , in which all frequency transitions have opposite fitness effects:



- Fitness flux** measures the amount of non-equilibrium / adaptation:

$$\mathcal{P}(\mathbf{x}^T) = \mathcal{P}(\mathbf{x}) e^{-N\Phi(\mathbf{x}) + \Delta\mathcal{H}(\mathbf{x})}$$

| fitness flux
| entropy difference of initial conditions

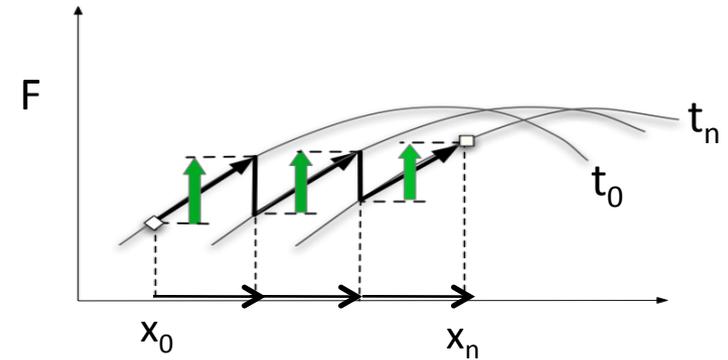
$$N\langle\Phi\rangle = H(\mathcal{P}|\mathcal{P}^T) - H(P(t_n)|P_0) + H(P(t_0)|P_0)$$

| KL entropy
| boundary terms

Quantifying adaptation

- **Phenotypic fitness flux**

$$\Phi(\tau) = \int_0^\tau \frac{\partial F(\Gamma, t)}{\partial t} \frac{d\Gamma}{dt} dt$$



- **Average fitness flux in the seascape model:**

$$\langle 2N\Phi(\tau) \rangle = c v \tau$$

Inferring adaptation

- Scaled divergence-diversity ratio:

$$\Omega(\tau) = 2\mu N \frac{\langle D(\tau) \rangle}{\langle \Delta \rangle}$$

- **Ω test:** The time-dependence of Ω measures:

- stabilizing strength

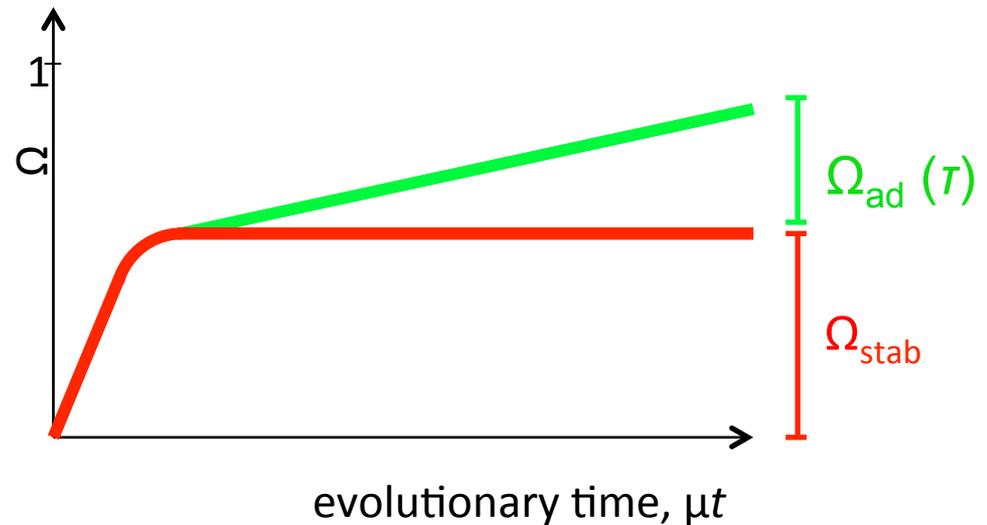
$$c \approx \frac{1}{\Omega_{\text{stab}}}$$

- driving rate

$$v \approx \frac{2\Omega_{\text{ad}}(\tau)}{\tau}$$

- fitness flux

$$\langle 2N\Phi(\tau) \rangle \approx \frac{2\Omega_{\text{ad}}(\tau)}{\Omega(\tau) - \Omega_{\text{ad}}(\tau)}$$



Inferring adaptation

Comparison with other methods

1. Q_{st}/F_{st} analysis:

infers directional selection if it is the dominant part of selection on short time scales ($\Omega > \Omega_0$).

2. Ornstein-Uhlenbeck model:

heuristics for equilibrium under stabilizing selection.

3. McDonald-Kreitman test:

requires divergence data from query sequence vs. neutral sequence at one divergence time.

Ω test:

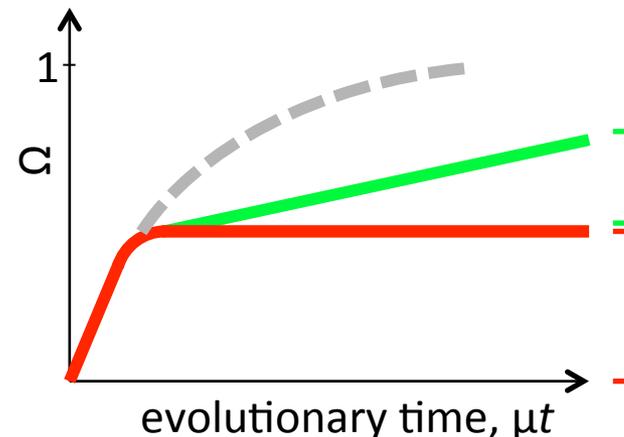
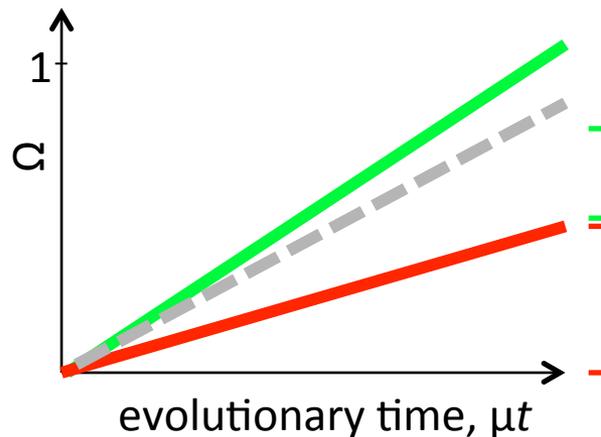
joint inference of directional and stabilizing selection (also for $\Omega > \Omega_0$).

Landscape model:

determines Ornstein-Uhlenbeck parameters in terms of quantitative genetics

Ω test:

requires trait divergence data for at least two sufficiently large divergence times.



Conclusions (1)

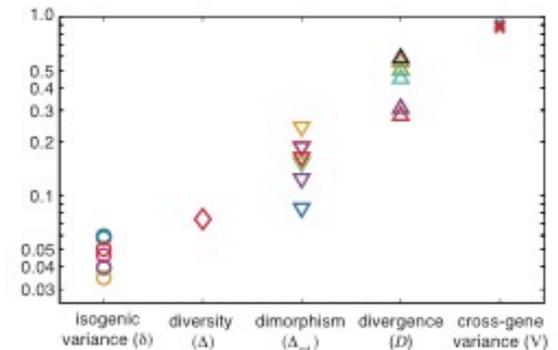
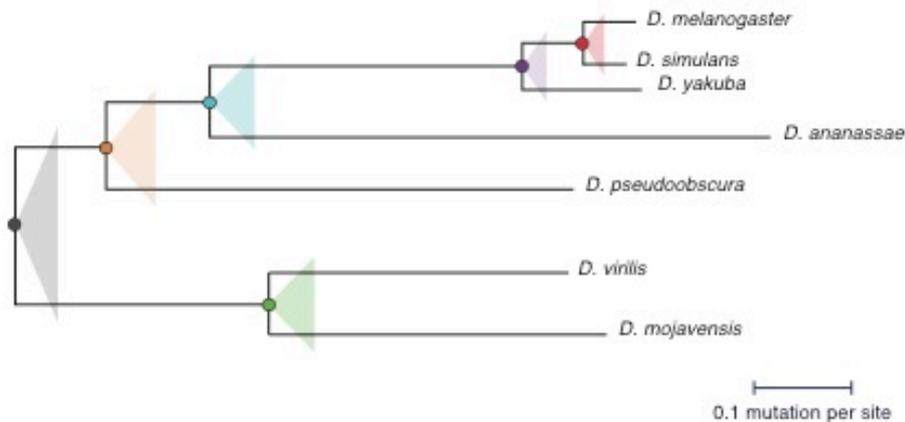
- **The evolutionary statistics of quantitative traits shows *universal* characteristics.**
- **These can be used to infer phenotypic fitness land- and seascapes.**
- **The single-peak seascape - unifies stabilizing and directional selection.
- provides a minimal model for adaptive evolution.**

2. Evolution of gene expression levels in *Drosophila*

Variation of gene expression across 7 Drosophila species

Comparative assay of genome-wide expression levels [Zhang et al, Nature 2007]

- RNA levels of > 6000 genes from inbred lines,
- 7 Drosophila species, 2 populations of *D. simulans*,
- data from males and females.

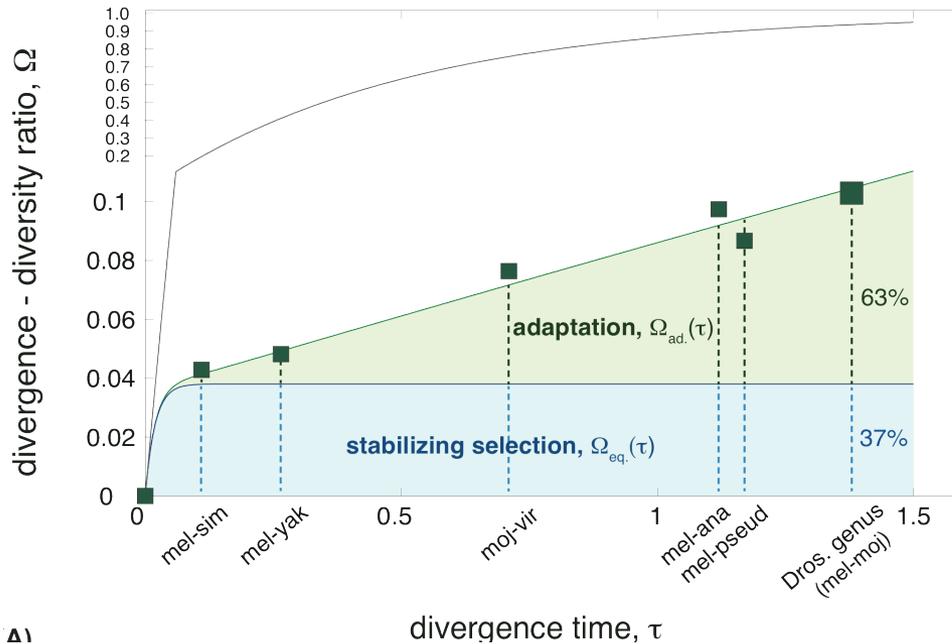


clade	mel - sim	mel - yak	moj - vir	mel - ana	mel - pse	mel - moj (Dros.)
time, τ_0	0.11	0.27	0.71	1.12	1.17	1.38
species	<i>D. melanogaster</i> <i>D. simulans</i>	<i>D. melanogaster</i> <i>D. simulans</i> <i>D. yakuba</i>	<i>D. mojavensis</i> <i>D. virilis</i>	<i>D. melanogaster</i> <i>D. simulans</i> <i>D. yakuba</i> <i>D. ananassae</i>	<i>D. melanogaster</i> <i>D. simulans</i> <i>D. yakuba</i> <i>D. ananassae</i> <i>D. pseudoobscura</i>	<i>D. melanogaster</i> <i>D. simulans</i> <i>D. yakuba</i> <i>D. ananassae</i> <i>D. pseudoobscura</i> <i>D. mojavensis</i> <i>D. virilis</i>

— *D. mel*
— *D. sim*
— *D. yak*
— *D. ana*
— *D. pse*
— *D. vir*
— *D. moj*

The pattern of gene expression divergence

▪ Ω ratio for clades of different divergence times



Δ

- The Ω data are
- incompatible with neutral evolution.
 - incompatible with evolution under time-independent stabilizing selection.
 - **consistent with evolution in a single-peak fitness seascape.**

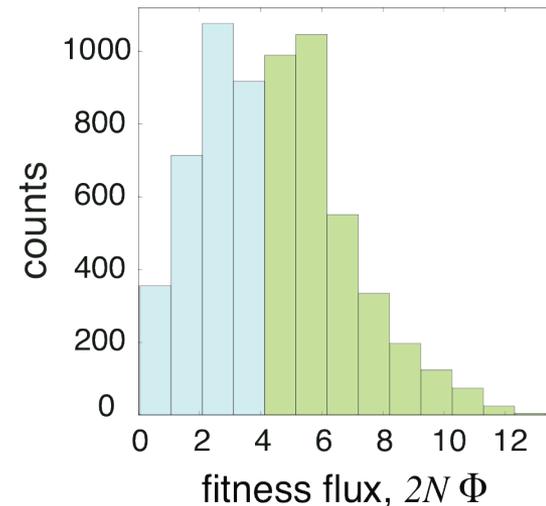
Inferring adaptive evolution

▪ Aggregate analysis

Between *mel* and *sim*, about 8% of the observed expression divergence is adaptive.
Between *mel* and *pse*, 64%

▪ Probabilistic inference for individual genes:

- 54% of all genes are significantly adaptive ($\Phi_{ML} > 4$).
- Less adaptation in genes with specific codon usage and in genes with high expression level.
- More adaptation in genes with broad codon usage, genes with specific functions (sensory perception, regulation, neural maturation, regulation of growth, aging and morphology), genes with male-biased expression.



Inferring adaptive evolution

- **Testing alternative evolutionary scenarios:**
 - *Lineage-specific demography.*
 - *Lineage- and gene-specific relaxation of stabilizing selection: stochastic gene loss.*
 - *Punctuated fitness seascape: large peak shifts with small rate.*

Conclusions (2)

- **Adaptive evolution of gene expression in *Drosophila* is pervasive.**
- **The adaptive pattern of expression has two molecular clocks.**
- **Quantitative genetics can be put to use for evolutionary systems biology.**