

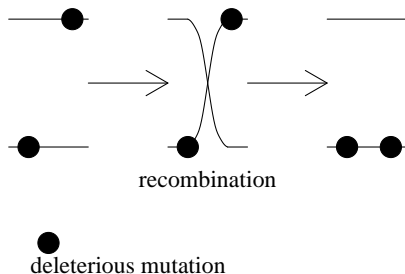
On the rate of Muller's ratchet

facts, heuristics, asymptotics

Joint with Alison Etheridge (Oxford) and Anton Wakolbinger (Frankfurt)

Asexual versus sexual reproduction

- ▶ Difference between asexually and sexually reproduction: recombination
- ▶ Most mutations **slightly deleterious**



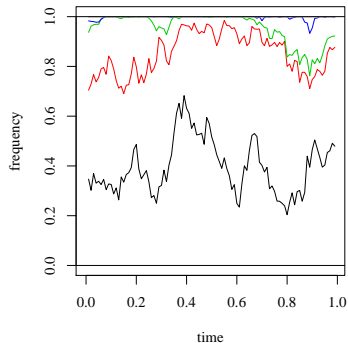
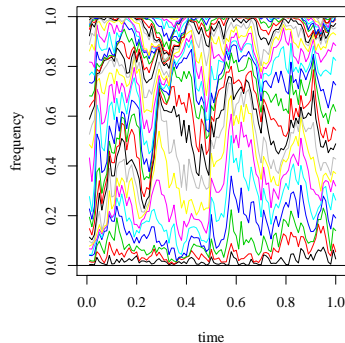
Recombination:
production of fit
genotypes

Muller's ratchet

- ▶ Asexually reproducing organism
 - ▶ Y_k : frequency of individuals carrying k mutations
 - ▶ Individual has k mutations: **fitness** = $(1 - s)^k$
 - ▶ Poisson(λ) **new mutations** for each individual
 - ▶ If $Y_0 = 0$ there will never again be an individual with 0 mutations
- the **ratchet has clicked**

Muller's ratchet

- Frequent and rare clicks depending on parameters



Muller's ratchet

- ▶ **Simple model:** parameters N, s, λ
- ▶ **Simple question:** average time between clicks
- ▶ **No exact answer** until today!

Haigh (1978)

- ▶ $Y_k(t)$: frequency of individuals with k mutations at time t
- ▶ **Selection:** $\rightarrow Y_k(t)(1 - s)^k$
- ▶ **Mutation:** $\rightarrow Y_k(t)(1 - s)^k + H, H \sim \text{Poisson}(\lambda)$
- ▶ $Y_k(t) = \text{Poisson}(\theta)$:
Selection and mutation: $\rightarrow \text{Poisson}(\theta(1 - s) + \lambda)$
- ▶ **Fixed point:** $\theta = \frac{\lambda}{s}$

Diffusion approximation

- ▶ For large N , small s, λ , approximately:

$$dY_k = \left(\sum_j s(j-k) Y_j Y_k + \lambda(Y_{k-1} - Y_k) \right) dt + \sum_{j \neq k} \sqrt{\frac{1}{N} Y_j Y_k} dW_{jk}$$

where $Y_{-1} := 0$, and $(W_{jk})_{j>k}$, $W_{jk} = -W_{kj}$ are independent Brownian motions

- ▶ Especially, with $\mathbf{M}_1 = \sum \mathbf{j} Y_j$,

$$d\mathbf{Y}_0 = \mathbf{Y}_0 (s\mathbf{M}_1 - \lambda) + \sqrt{\frac{1}{N} Y_0 (1 - Y_0)} dW$$

Moment equations

- ▶ Rate of the ratchet = speed of M_1
- **find equation for dM_1**
- ▶ Speed of M_1 determined by variance:
- ▶ With $M_2 = \sum_j (j - M_1)^2 Y_j$,

$$dM_1 = (\lambda - sM_2)dt + \sqrt{\frac{1}{N}M_2}dW$$

- ▶ Without noise, this is seen from:

$$\begin{aligned}\frac{d \sum_k k Y_k}{dt} &= \sum_{k,j} sk(j-k)Y_j Y_k + \lambda \sum_k k(Y_{k-1} - Y_k) \\ &= \lambda - sM_2\end{aligned}$$

Including stochastic effects

- ▶ Similarly,

$$dM_2 = \left(-\frac{1}{N}M_2 + (\lambda - sM_3)\right)dt + \sqrt{\frac{1}{N}M_3}dW$$

$$dM_3 = \left(-\frac{3}{N}M_3 + (\lambda - s(M_4 - 3M_{2,2}))\right)dt + \sqrt{\frac{1}{N}M_6 + \dots}dW$$

etc.

- ▶ **No closed system of equations!**

Cumulants

- ▶ Recall: **Equilibrium is Poisson**
Only the Poisson distributions has all cumulants equal
- ▶ **Cumulants** $\kappa_1, \kappa_2, \dots$ satisfy

$$\log \sum_{k=0}^{\infty} x_k e^{-\xi k} = \sum_{k=1}^{\infty} \kappa_k \frac{(-\xi)^k}{k!}.$$

- ▶ $\kappa_1, \kappa_2, \kappa_3$ are the first three centered moments
- ▶ Ignore random effects and compute

$$\frac{d\kappa_k}{dt} = \lambda - s\kappa_{k+1}.$$

⇒ **Linear System!**

Cumulants

- ▶ The solution can be computed.
- ▶ Especially,

$$x_0(t) = e^{-\kappa_0(t)} = x_0(0) \frac{\exp\left(-\frac{\lambda}{s}(1 - e^{-st})\right)}{\left(\sum_{k=0}^{\infty} x_k(0)e^{-stk}\right)}$$

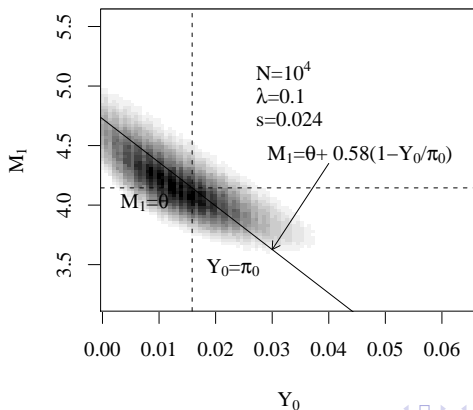
and

$$\kappa_1(t) = -\frac{\partial}{\partial \xi} \log \sum_{k=0}^{\infty} x_k(0)e^{-\xi k} \Big|_{\xi=st} + \frac{\lambda}{s}(1 - e^{-st}).$$

- ▶ **Still no solution including random effects...**

One-dimensional diffusion heuristics

- ▶ Equation for Y_0 : **prediction of M_1 given Y_0** necessary
- ▶ Simulations show **correlation** between M_1 and Y_0 :



One-dimensional diffusion heuristics

- ▶ Idea from **Haigh (1978)**:
By random effects, $Y_0 - \pi_0$ is **distributed on all classes**
- ▶ \Rightarrow observed states are of the form

$$\Pi(Y_0) = \left(Y_0, \frac{1 - Y_0}{1 - \pi_0} (\pi_1, \pi_2, \dots) \right)$$

π_k Poisson weight for parameter $\theta := \frac{\lambda}{s}$

Poisson profile approximation

- ▶ In particular **$M_1(Y_0) = M_1(\Pi(Y_0))$ can be computed**

One-dimensional diffusion heuristics

- ▶ Haigh: observed states are of the form $\Pi(Y_0)$
- ▶ However: **Random effects and dynamical system interact**
- ▶ Our idea: **observed states are of the form**

$$\Pi(Y_0)S_\tau$$

for some τ (S : semigroup of dynamical system)

One-dimensional diffusion heuristics

- ▶ Use **explicit solution** of dynamical system: observed states have

$$M_1(\tau) = \theta + \frac{\eta}{e^\eta - 1} \left(1 - \frac{y_0(\tau)}{\pi_0} \right).$$

for $\tau := \frac{A}{s} \log \theta$ and $\eta := \theta^{1-A}$

► **Three parameter regimes:**

$$A \text{ small,} \quad \eta \approx \theta, \quad M_1 \approx \frac{\theta}{1-\pi_0}(1 - Y_0),$$

$$A = 1, \quad \eta = 1, \quad M_1 \approx \theta + 0.58 \left(1 - \frac{Y_0}{\pi_0}\right),$$

$$A \text{ big,} \quad \eta \approx 0, \quad M_1 \approx \theta + \left(1 - \frac{Y_0}{\pi_0}\right)$$

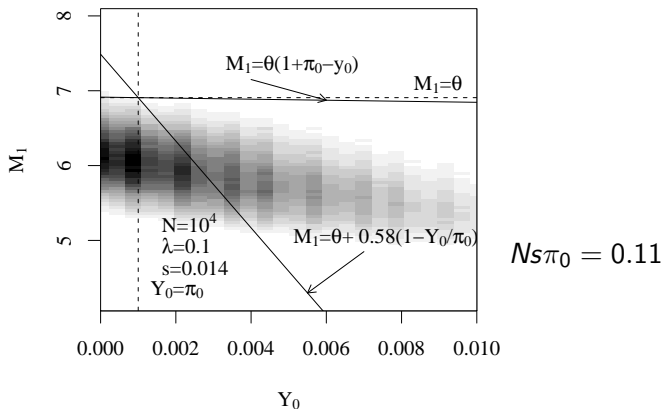
► Corresponding one-dimensional diffusions:

$$A \text{ small,} \quad dY_0 = \lambda(\pi_0 - Y_0)Y_0 dt + \sqrt{\frac{1}{N}Y_0}dW,$$

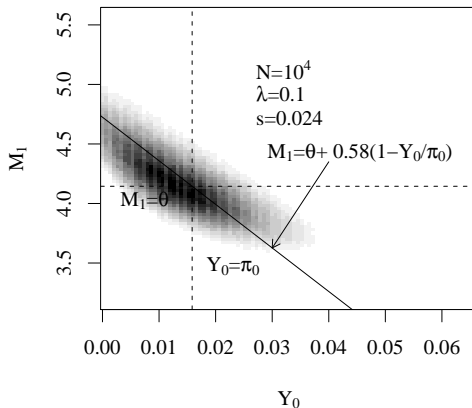
$$A = 1, \quad dY_0 = 0.58s \left(1 - \frac{Y_0}{\pi_0}\right) Y_0 dt + \sqrt{\frac{1}{N}Y_0}dW_0,$$

$$A \text{ big,} \quad dY_0 = s \left(1 - \frac{Y_0}{\pi_0}\right) Y_0 dt + \sqrt{\frac{1}{N}Y_0}dW_0,$$

- **A small:** no time for the dynamical system to relax to equilibrium \Leftrightarrow **frequent clicks**

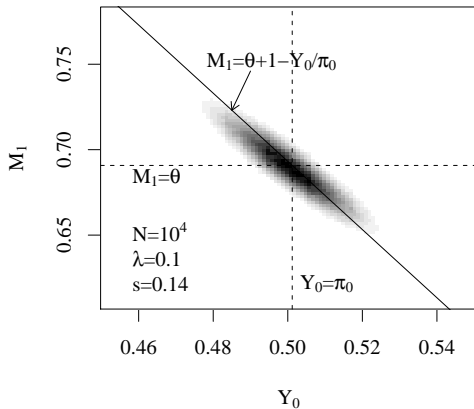


- ▶ **A = 1** : speed for relaxation equal to speed of noise
- ▶ See also by Stephan et al. and Gordo and Charlesworth



$$Ns\pi_0 = 3.7$$

- **A big:** system cannot exit equilibrium \Leftrightarrow **rare clicks**



$$Ns\pi_0 = 685$$

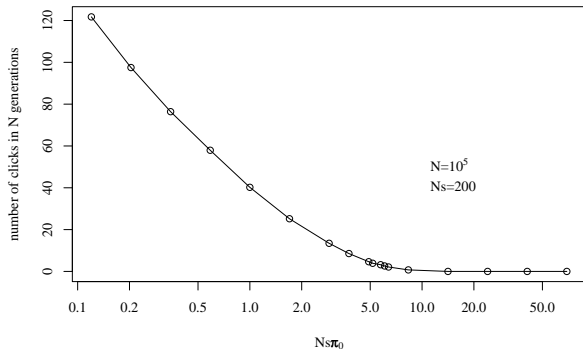
- ▶ Use **rescaling**

$$Z(t) = \frac{1}{\pi_0} Y_0\left(\frac{t}{N\pi_0}\right)$$

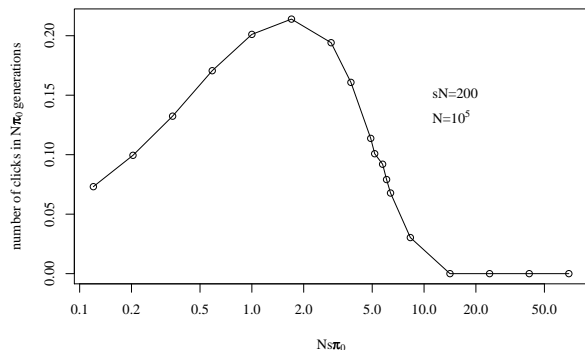
- ▶ Consider the intermediate regime $A = 1$

$$A = 1: dZ = 0.58Ns\pi_0(1 - Z)Zd\tau + \sqrt{Z}dW.$$

- ▶ Consider $\lambda, s \rightarrow 0$, $N \rightarrow \infty$
- ▶ Clicks only for small $Ns\pi_0$



- ▶ Consider $\lambda, s \rightarrow 0$, $N \rightarrow \infty$
- ▶ In case of clicks, interclick time is of order $N\pi_0$



Conclusion

- ▶ **Exact rate of the ratchet still not obtained**, but
- ▶ **Conjecture:**

$$Ns\pi_0 = \mathcal{O}(1) \implies \text{Interclick time } \mathcal{O}(N\pi_0)$$

$$Ns\pi_0 \gg 1 \implies \text{Interclick time } \gg N\pi_0$$