

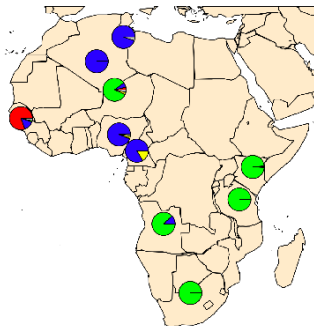
TOOLS FOR UNDERSTANDING THE GEOGRAPHY OF ADAPTATION

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May 1st, 2014
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MOTIVATION: HEMOGLOBINOPATHIES

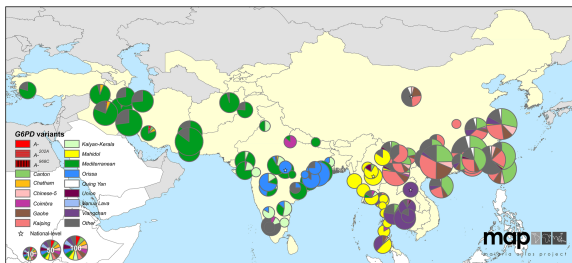


(from Flint et al 1998)

Human sickle-cell allele (HbS): (Currat et al 2002)

- ▶ Single base substitution
- ▶ provide protection against malaria (but deleterious in homozygotes)

MOTIVATION: HEMOGLOBINOPATHIES



(Howes et al 2013)

Human G6PD variants:

- ▶ over 130 G6PD deficiency alleles; 34 variants at high frequency
- ▶ provide protection against malaria but increases risk of anemia
- ▶ Estimated ages 40-400 generations (various)

MOTIVATION: A NICE STORY ABOUT MICE.



mice: AH Harris

- ▶ Dark-pigmented mammals and reptiles on volcanic outcrops in the Southwest. (Dice, Benson 1936)
- ▶ “Dark” allele beneficial on outcrops, deleterious elsewhere.
- ▶ MC1R: basis is shared **between species** but not **between populations** (Nachman, Hoekstra)

IMPLICATIONS

The geography of adaptation

i.e. how adaptations are shared or not across the landscape tells us about

- ▶ **local adaptation:** what do fitness landscapes look like?
- ▶ **constraint:** many possible solutions or not?
- ▶ **speciation:** how fast can distinct adaptations accumulate?

Main summarizing question:

What is the geographic resolution of adaptation?

Also: When is there sharing of solutions?

Every region its own solutions?

How can we tell which is happening?

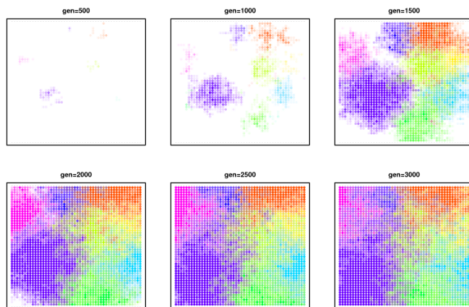
THE REST OF THE TALK

- ▶ Homogeneous landscape (quickly?)
- ▶ Patchy landscape: transients

note motivation from: Pennings & Hermisson, Soft Sweeps

FIRST: HOMOGENEOUS LANDSCAPE

- ▶ **Continuous species range** with constant population density
- ▶ Selective pressure is **geographically uniform**, changes from **deleterious** ($1 - s_d$) to **beneficial** ($1 + s_b$)
- ▶ Selected mutations are **selectively equivalent** (e.g. same base pair or on same pathway) and so *exclude* each other



DYNAMICS

Some alleles are present as **standing variation**

- ▶ in small clusters with effective density $\approx \rho\mu/s_d$.

Beneficial (s_b) alleles:

- ▶ fix locally with probability $\approx 2s_b/\xi^2$, and if they do:
- ▶ spread radially with speed $\approx \sigma\sqrt{s_b}$ (Fisher; KPP 1937)

MUTATIONS AS A POINT PROCESS

Combining these,

- ▶ Standing, and new mutations as **Poisson processes**
- ▶ **thinned** by chance of local fixation
- ▶ then spreading outward at constant speed
- ▶ and excluding further mutations.

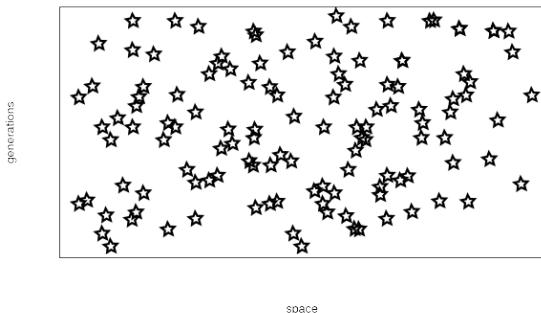


s = selection coefficient
 ρ = pop density
 μ = mutation rate
 σ = SD dispersal distance
 ξ = SD # of offspring

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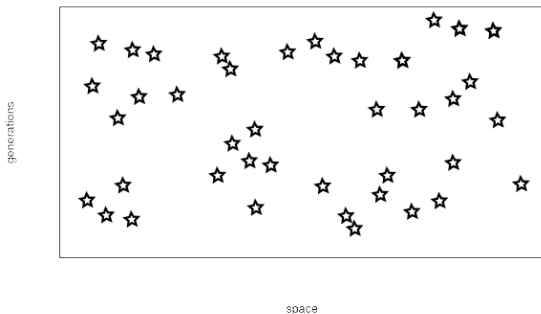


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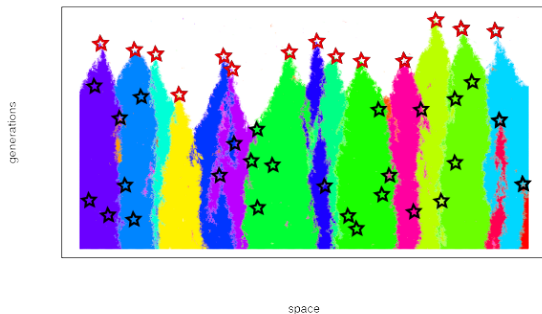


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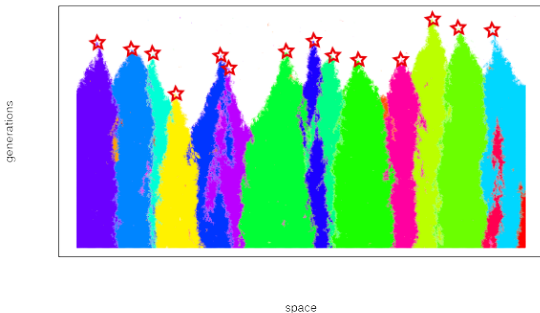


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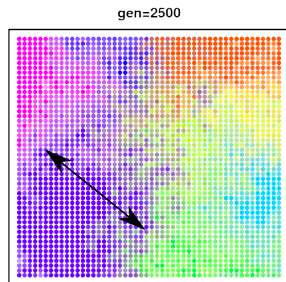
RULE OF THUMB: A CHARACTERISTIC LENGTH

Spatial properties mainly determined by a **characteristic length**, solving:

$$\lambda_0 \pi \chi^2 + \lambda \pi \chi^3 / v = 1,$$

χ is diameter of space-time cone in which expected to find one successful mutation.

λ_0 = density of standing mutations
 λ = flux of new mutations
 v = speed of wave
 s_b = fitness advantage, $t > 0$
 s_d = fitness disadvantage, $t < 0$
 ρ = pop density
 μ = mutation rate
 σ = SD dispersal distance
 ξ = SD # of offspring
 π = 3.1415 . . .

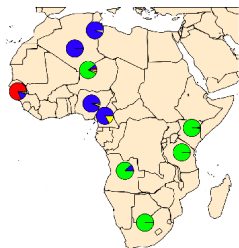


With no standing variation:

$$\chi = \left(\frac{\sigma \xi^2}{\rho \mu \sqrt{2s} \pi} \right)^{1/3}$$

Also: mean time until adaptation;
proportion from standing variation; size of
sampled cluster, etc.

SICKLE-CELL ALLELE (HbS)



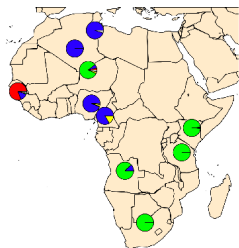
Human sickle-cell allele (HbS): (Currat et al 2002)

- ▶ Single base substitution: $\mu = 10^{-8}$
- ▶ Balancing selection: no problem
- ▶ Say, $\sigma = 50$ km and $\rho = 2$ people/ km^2 and $s_d = .05$ or $= .5$.

(from Flint et al 1998)

	estimated	weaker	stronger	
s_b	0.15	0.15	0.15	
s_d	—	0.05	0.5	
σ	10–100	50	50	km
μ	10^{-8}	10^{-8}	10^{-8}	
ρ	2	2	2	km^{-2}
χ	1000	980	1120	km
age	10–70	63	77	gen
z_0	—	52%	6%	

SICKLE-CELL ALLELE (HBS)



Observed pattern:

- ▶ Haplotype pattern on scale of 1000 km
- ▶ Estimated age 10-70 generations

(Currat et al; Modiano et al)

(from Flint et al 1998)

	estimated	weaker	stronger	
s_b	0.15	0.15	0.15	
s_d	—	0.05	0.5	
σ	10–100	50	50	km
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SICKLE-CELL ALLELE (HBS)

s_b = fitness advantage, $t > 0$

s_d = fitness disadvantage, $t < 0$

ρ = pop density

μ = mutation rate

σ = SD dispersal distance

χ = characteristic length

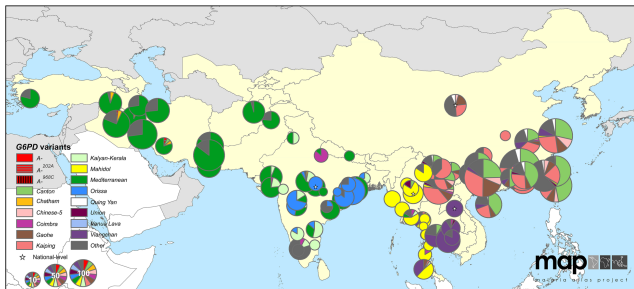
z_0 = proportion from standing

We compute:

- ▶ characteristic length $\chi \approx 1000$ km
- ▶ mean adaptation time
 $\mathbb{E}[\tau] \approx 70$ generations
- ▶ proportion from standing variation z_0 significant
- ▶ time until pattern is erased
 $R^2/\sigma^2 \approx 6,400$ generations

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EXAMPLE: G6PD

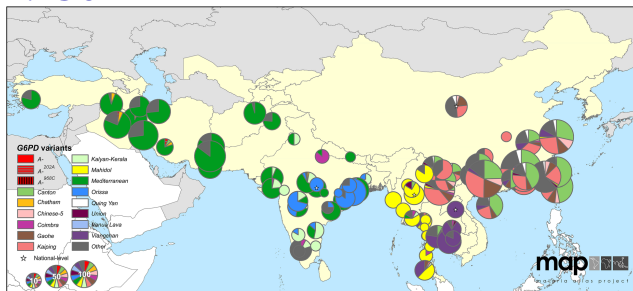


Human G6PD variants:

(Howes et al 2013)

- ▶ 34 variants across 4,000,000 km²: $\chi \approx 350$ km
- ▶ Estimated ages 40-400 generations (various)
- ▶ Estimated $s = .25$ (Slatkin et al 2008)
or $s = .04$ (Tishkoff et al 2001)
- ▶ 150 coding bases: $\mu = 150 \times 10^{-8}$
- ▶ Say, $\sigma = 50$ km and $\rho = 2$ people/km²

EXAMPLE: G6PD

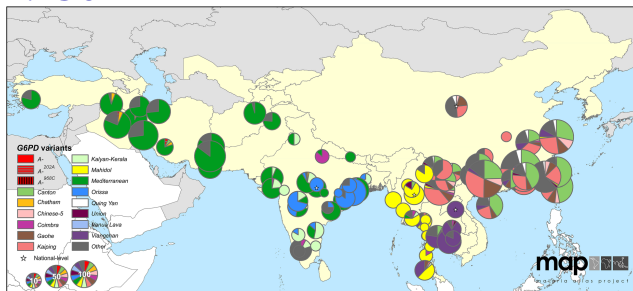


Human G6PD variants:

(Howes et al 2013)

	estimated	weaker	stronger	
S_b	0.25 or 0.04	0.25	0.04	
S_d	0.1	—	—	
σ	50	—	—	km
μ	1.5×10^{-6}	—	—	
ρ	2	—	—	km^{-2}
χ	350	144	254	km
age or $\mathbb{E}[\tau]$	40–400	7	32	gen
Z_0	—	83%	51%	

EXAMPLE: G6PD



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SECOND CASE: PATCHY ENVIRONMENT

Focal allele is **beneficial** in patches; **deleterious** between:



- A = patch size
- R = patch separation
- s_D = on-patch selective advantage
- $-s_I$ = between-patch selective cost
- ρ = pop density
- μ = mutation rate
- σ = SD dispersal distance
- ξ = SD # of offspring

What is the time scale of:

- ▶ appearance of new mutations in unadapted patches?
- ▶ transit of mutations between patches?

What does the latter look like?

ADAPTATION BY NEW MUTATION

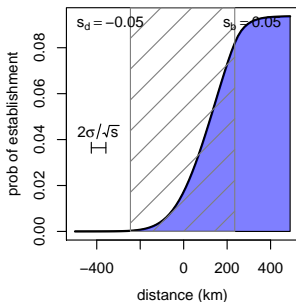
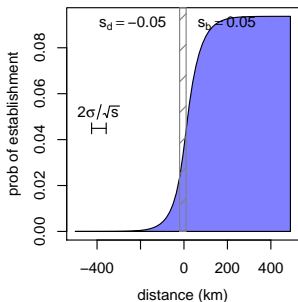
Naively: a new mutant in a patch establishes with probability

$$p_{\text{estab}} = \frac{2s_b}{\xi^2}$$

so rate of influx is

$$\lambda_{\text{mut}} = A\rho\mu \frac{2s_b}{\xi^2} \text{ per generation.}$$

Actually, probability of establishment at distance r is $\simeq \exp(-r\sqrt{s_d}/\sigma)$
(Barton 1987) ... naive calculation **does pretty good**:



MIGRATION: TRANSITING FAMILIES

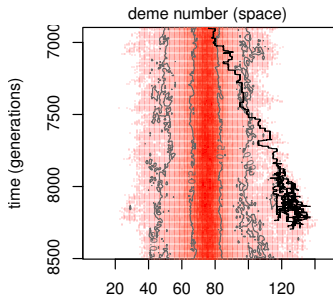
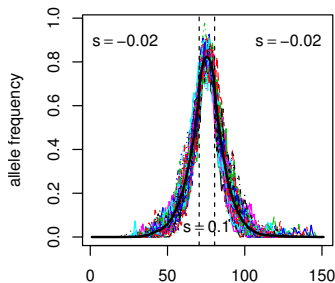
Suppose a mutation has fixed locally in one patch.

How long until it reaches another at distance R ?

At migration-selection balance: frequency at distance r is

$$q(r) \approx \frac{1}{2} \left(\frac{r\sqrt{2s_d}}{\sigma} \right)^{-\frac{d-1}{2}} \exp \left\{ -\frac{r\sqrt{2s_d}}{\sigma} \right\}.$$

This deterministic “equilibrium” is composed of **rare long-distance migrant families**.



FAMILY DECOMPOSITION

Draw a circle at distance r_0 from the original patch.

Definition: Any two individuals **outside the patch** that share an ancestor who **lived outside r_0** are **in the same “family”**.

In the continuum limit, these families are **subcritical branching processes**, killed on hitting a patch, with inhomogeneous branching rate.

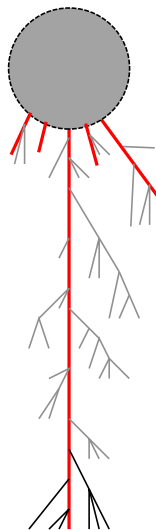
For large r_0 , branching rate (nearly) homogeneous $1 - s_d$.

(at least those making it to the new patch)

MODELING TRANSITING FAMILIES

General idea:

- ▶ Between patches, transiting families die out:
growth rate $-s_d < 0$ (“subcritical”).
- ▶ Chance that one lives for t generations is $\sim e^{-s_d t}$
- ▶ In the (rare) event it does, looks like a single “trunk” with transient “branches” (Geiger 1999)
- ▶ Trunk moves as a random walk.



THE EFFECTIVE MIGRATION RATE

Rate at which mutations **transit between patches** is

$$\lambda_{\text{mig}}(R) = (\text{outflux of families}) \\ \times (\text{prob family establishes in patch at } R).$$

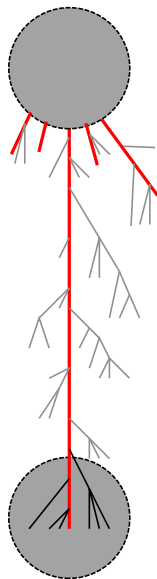
On the other hand, without the new patch,

$$q(R) = (\text{outflux of families}) \\ \times (\text{occupation time of a family near } R).$$

Each is the probability that the family **hits** the patch, multiplied by

- ▶ the probability it establishes, or
- ▶ its occupation time

given it gets there.



MIGRATION AND MUTATION COMPARED

Constant is $\approx \rho \times 2s_b/\xi^2 \times 1/s_d$, so

$$\lambda_{\text{mig}} \approx \frac{\rho s_b}{\xi^2 s_d} \left(\frac{R\sqrt{2s_d}}{\sigma} \right)^{-\frac{d-1}{2}} \exp\left(-\frac{R\sqrt{2s_d}}{\sigma}\right).$$

The relevant time scales of each are $T_{\text{mig}} = 1/\lambda_{\text{mig}}$ and $T_{\text{mut}} = 1/\lambda_{\text{mut}}$.

$$\lambda_{\text{mut}} = A\rho\mu 2s_b/\xi^2, \quad \text{so}$$

$\lambda_{\text{mut}} \approx \lambda_{\text{mig}}$ if (in 1D)

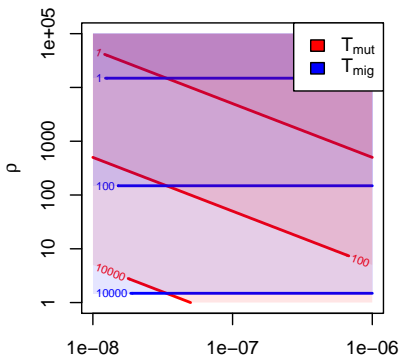
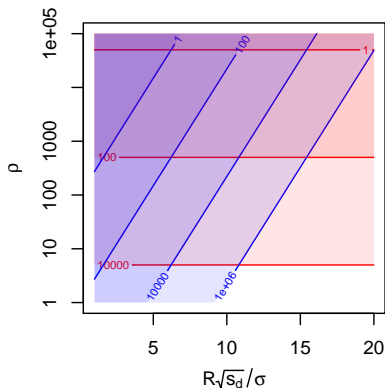
$$A\mu \approx (1/s_d) \exp(-R\sqrt{2s_d}/\sigma).$$

MIGRATION AND MUTATION COMPARED

Constant is $\approx \rho \times 2s_b/\xi^2 \times 1/s_d$, so

$$\lambda_{\text{mig}} \approx \frac{\rho s_b}{\xi^2 s_d} \left(\frac{R\sqrt{2s_d}}{\sigma} \right)^{-\frac{d-1}{2}} \exp\left(-\frac{R\sqrt{2s_d}}{\sigma}\right).$$

The relevant time scales of each are $T_{\text{mig}} = 1/\lambda_{\text{mig}}$ and $T_{\text{mut}} = 1/\lambda_{\text{mut}}$.



CAN WE DETECT ALLELES SHARED BY MIGRATION?

Shared alleles will share a haplotype; how long?

- ▶ Haplotype whittled down by recombination between patches:
- ▶ if transit between patches takes t generations,
- ▶ shared haplotype will be \approx Exponential with mean $1/t$.

The trunk: \approx Brownian, killed at rate s_d .

Let τ be the hitting time of the new patch, at distance r .

If length of shared haplotype is L , then in 1-D:

$$\mathbb{P}\{L > \ell\} = \mathbb{E} [e^{-\ell\tau} | \tau < \infty] = \exp\left(-\frac{R}{\sigma} \left(\sqrt{2(\ell + s_d)} - \sqrt{2s_d}\right)\right).$$

LENGTH OF SHARED HAPLOTYPE

The probability the haplotype is of length $> \ell$ is

$$\mathbb{P}\{L > \ell\} = \mathbb{E} [e^{-\ell\tau}] = \exp \left\{ -(R/\sigma) \left(\sqrt{2(\ell + s_d)} - \sqrt{2s_d} \right) \right\}$$

For large R :

- ▶ Mean transit time of allele is $\mathbb{E}[\tau] \approx R/(\sigma\sqrt{2s_d})$.
- ▶ Haplotype length is $\mathbb{E}[L] \approx \sigma\sqrt{2s_d}/R$.
- ▶ $L \stackrel{d}{=} (Y + \sqrt{2s_d})^2 - 2s_d$, with $Y \sim \text{Exponential}(R\sqrt{2}/\sigma)$.
- ▶ CLT for τ as $R \rightarrow \infty$:

$$\tau \approx \frac{R}{\sigma\sqrt{2s_d}} + \sqrt{\frac{R}{\sigma(2s_d)^{3/2}}} Z, \quad Z \sim N(0, 1).$$

APPLICATION TO ROCK POCKET MICE

Back to *Chaetodipus intermedius* (Dice, Benson, Nachman, Hoekstra, etc)



APPLICATION TO ROCK POCKET MICE

Back to *Chaetodipus intermedius* (Dice, Benson, Nachman, Hoekstra, etc)

Black Tank (BLK)



White Hills (WHT)



Tinajas Altas (TIN)



Tule Well (TUL)



Pinacate Lava (PIN)



Carrizozo (CAR)



Armendaris (ARM)



Fra Cristobol (FRA)



Kenzin (KNZ)



Afton (AFT)



Mexico
(MEX)



Avra Valley
(AVR)



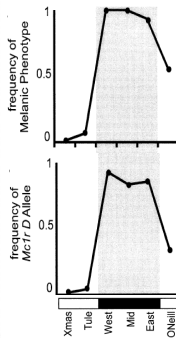
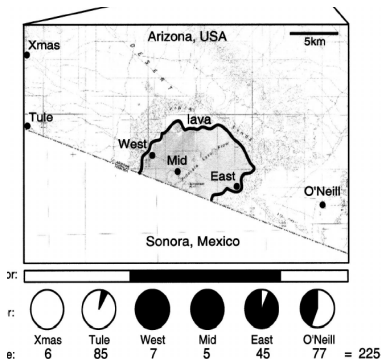
Tumamoc Hill
(TUM)



Portal
(POR)

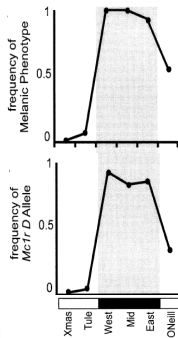
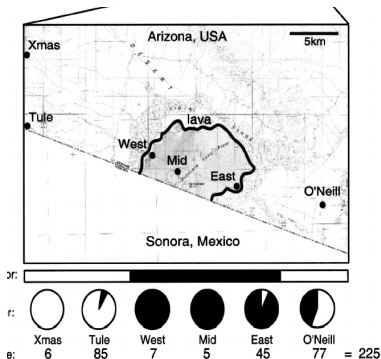
APPLICATION TO ROCK POCKET MICE

On one flow (Pinacate) an allele of MC1R is responsible for much of the change to a dark pelage (Nachman et al 2003).
Hoekstra et al (2004): further study of nearby flows.



APPLICATION TO ROCK POCKET MICE

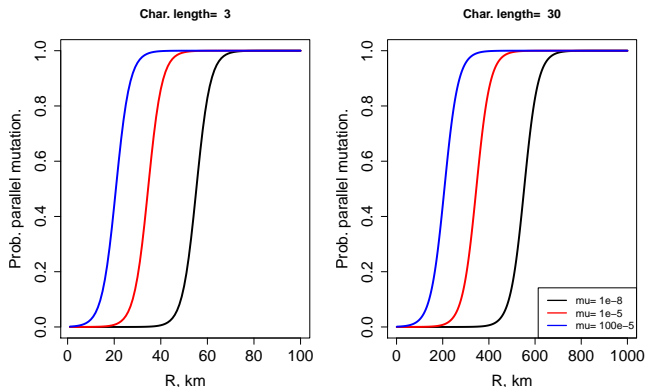
- ▶ Hoekstra et al (2004) estimated $0.03 < s_b < 0.03$.
- ▶ Cline width suggests $3\text{km} < \sigma/\sqrt{s_d} < 30\text{km}$
- ▶ Another species suggests $\sigma = 0.28\text{km}$
- ▶ ... so maybe $10^{-4} < s_d < 0.01$?



APPLICATION TO ROCK POCKET MICE

Probability of parallel adaptation, i.e. adaptation by mutation before migration, is

$$\frac{\lambda_{\text{mut}}}{\lambda_{\text{mut}} + \lambda_{\text{mig}}} = \frac{2A\mu s_b}{2A\mu s_b + \exp(-\sqrt{2s_d R}/\sigma)}$$



... and size of shared haplotype $\approx 56\text{Kb}$.

CONCLUSIONS

- ▶ Order-of-magnitude estimates and Poisson process calculations get **not unreasonable** answers
- ▶ Some aspects of geometry are **insensitive** to parameters.
- ▶ We can describe well the initial dynamics; what happens next?
- ▶ Need to better model **interactions** (e.g. G6PD).

thanks



gideon bradbud



graham coop



yaniv brandvain

Michael Nachman

WITHIN A PATCH

In a discrete deme, large population model, lineages of the adapted allele moves from x to y at rate

$$q(x \leftarrow y) = \rho p(y)(1+s(y))m(y \rightarrow x).$$

If migration is nearest-neighbor, the stationary distribution is

$$\pi(x) = p(x)(1 + s(x)).$$

Diffusion limit for motion of the lineage is

$$dX_t = \pi'(X_t)dt + dB_t.$$