#### On hybrid zones and the effect of barriers

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Agencia Nacional de Investigación y Desarrollo

Ministerio de Ciencia, Tecnología, Conocimiento e Innovación

Gobierno de Chile

#### Thanks goes to



Work under the supervision of Alison Etheridge

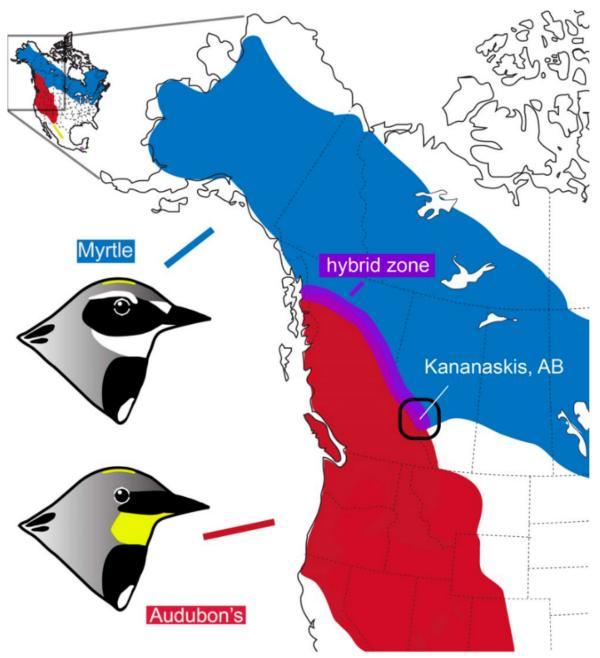


For many meaningful discussions

Kim Becker



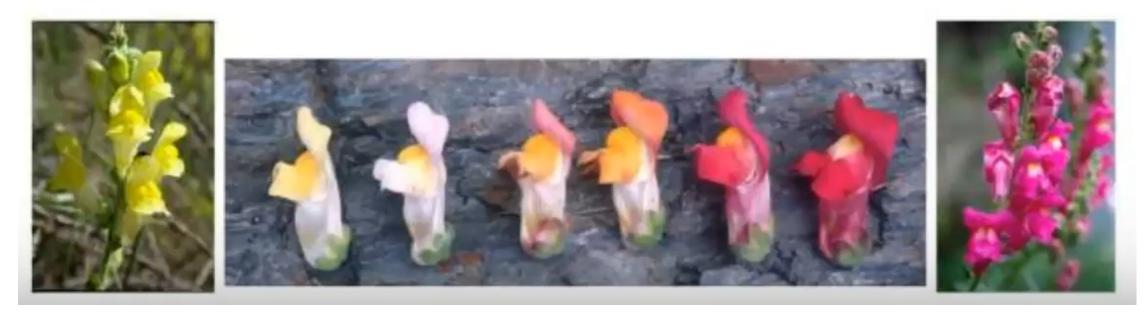
For pedagogical reasons technical details will be handwaved.



Hybrid zone: narrow genetically geographic region where two genetically distinct populations are found and hybridise to produce mixed offsprings.

Example:

Warblers in the United States [1]



Antirrhinum (Flowers) in Europe [2]. Sparrows in the Italic Peninsula [3]







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How can we understand the evolution of hybrid zones under hypothesis 2?

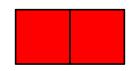
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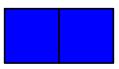
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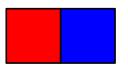
Hardy-Weinberg equilibrium: there is a proportion of  $u(x,t)^2$  individuals of type AA,  $(1-u(x,t))^2$  individuals of type BB, 2u(x,t)(1-u(x,t)) individuals of type AB.

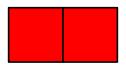
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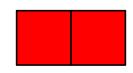




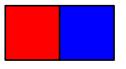


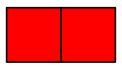
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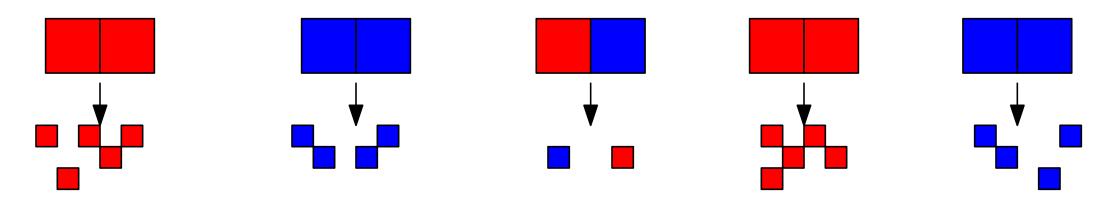




For reproduction: a sample of N individuals of type AB produces K gametes, BB produces K(1+s) and AA produces  $K(1+s+\delta)$ .

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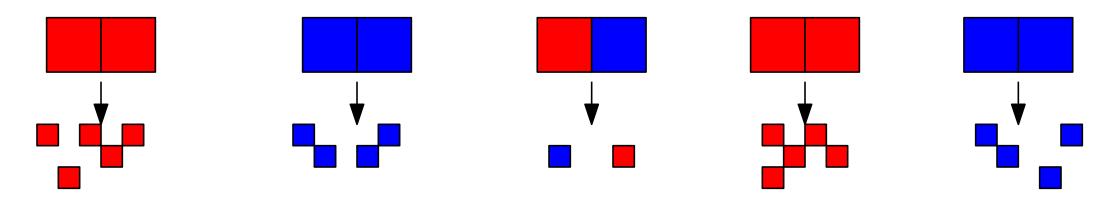
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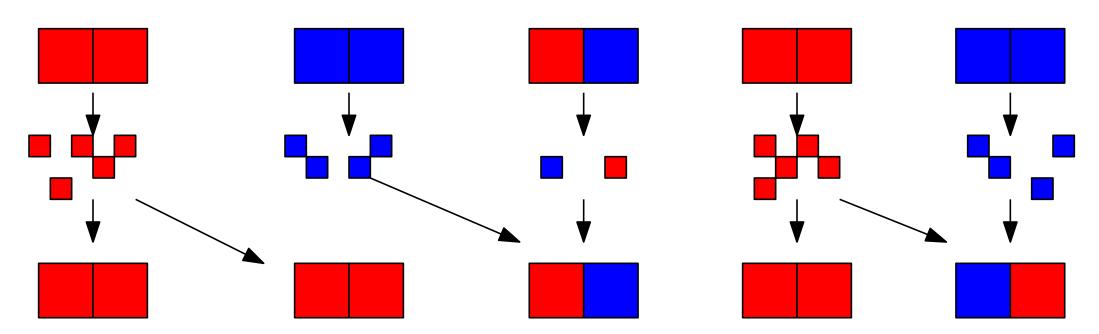
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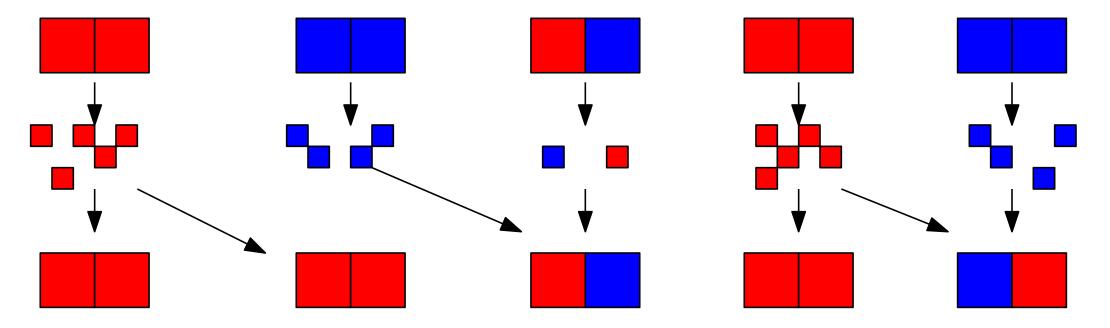
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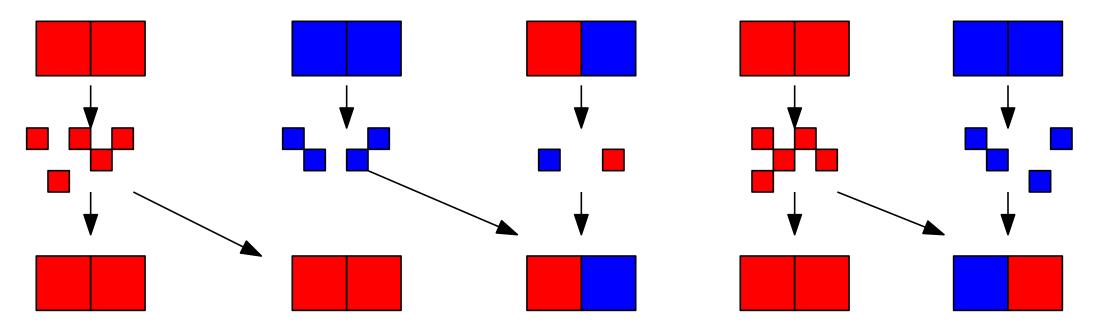
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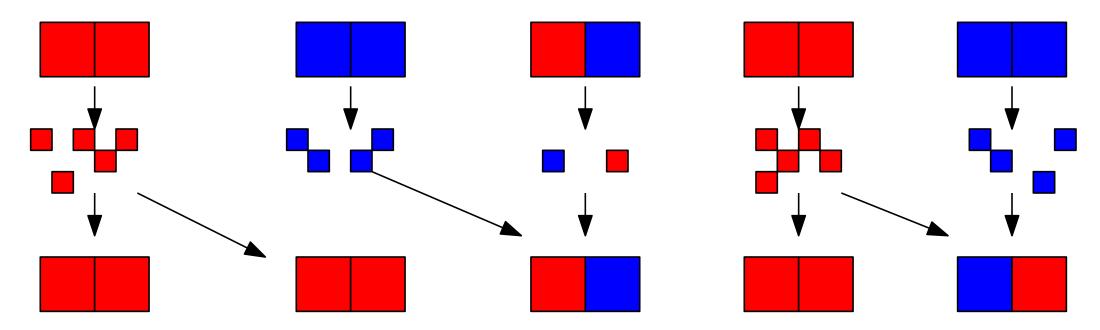
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The local change of the allele A per reproductive event is

 $\mathbb{P}(\text{sampling allele }A\text{ at }x)-u(x,t)\sim Mu(x,t)(1-u(x,t))(2u(x,t)-(1-\gamma))$  for some  $M(s,\delta,K)>0$  and  $1>\gamma(\delta)>0$ .



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Normalising M=1, supposing finite variance movement of individuals and ignoring genetic drift:

$$\partial_t u = \text{change by movement} + \text{change by reproduction},$$
  
=  $\Delta u + u(1-u)(2u - (1-\gamma)).$ 

Assuming **space homogeneity** we get:

$$(AC) \begin{cases} \partial_t u = \Delta u + u(1-u)(2u - (1-\gamma)) & t > 0, x \in \mathbb{R}^d, \\ u(0,x) = u_0(x) & x \in \mathbb{R}^d. \end{cases}$$

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#### Video!

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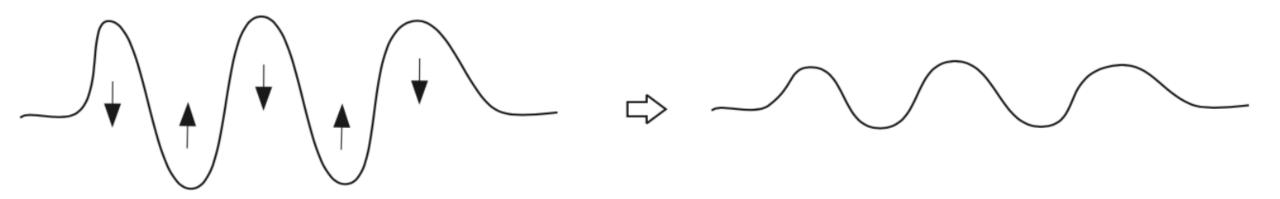
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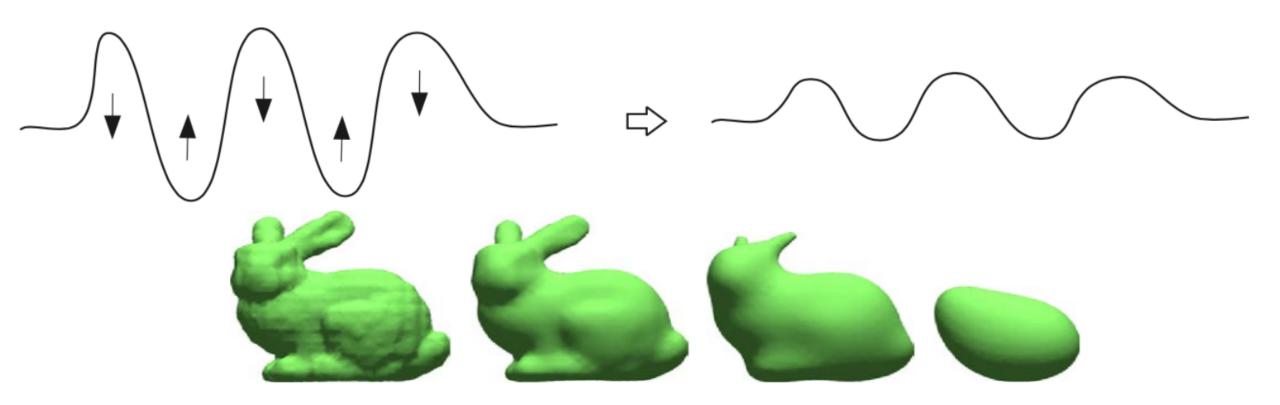
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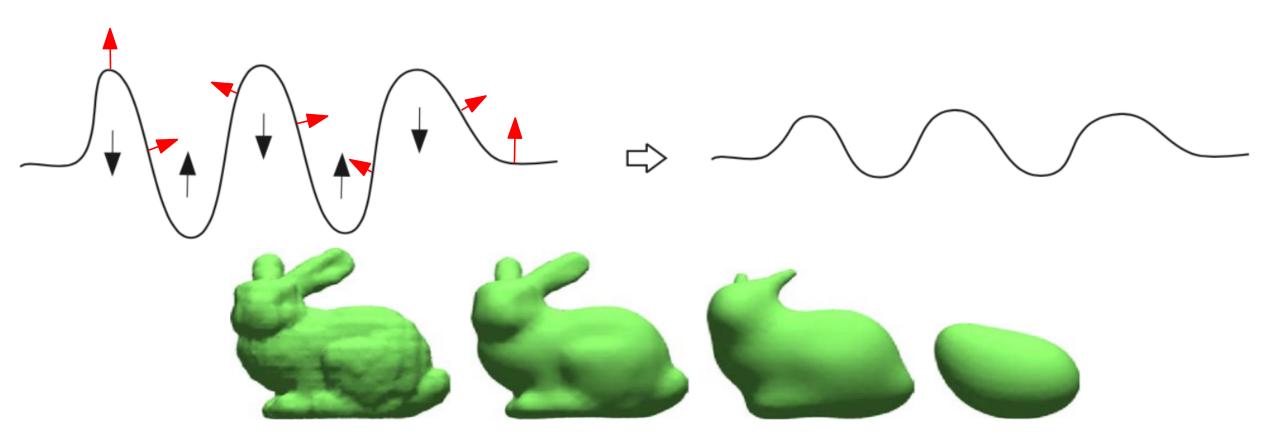
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If  $\nu \neq 0$ , constant flow adds a velocity  $\nu$  going *outside* the surface.



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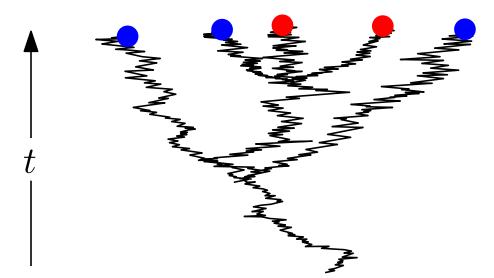
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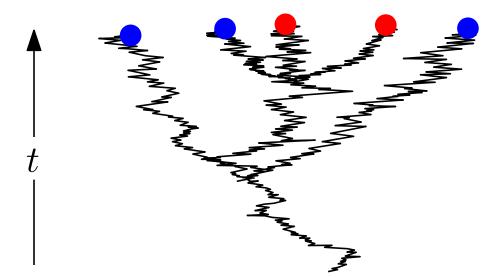
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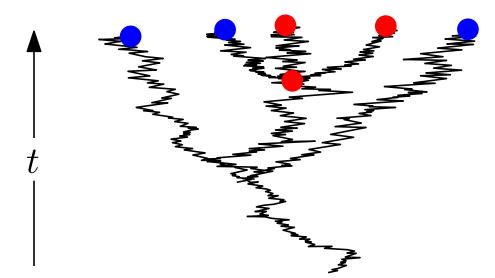
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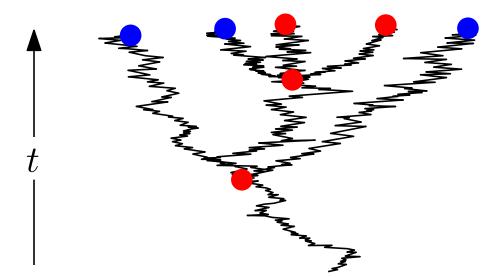
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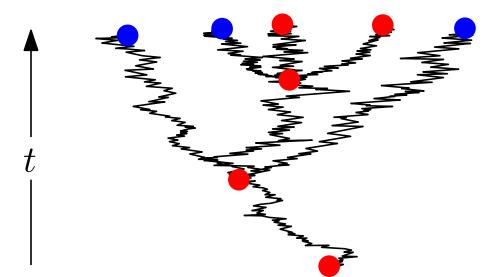
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Let  $\mathbb{P}_x^{\epsilon}$  be the law where the BBM starts at x and  $V=(1+\epsilon\nu)\epsilon^{-2}$ .

Let  $\mathbb{V}_{u_0}(\mathcal{W}(t))$  be the vote of the root of the voting system, with the vote of the leaves given by  $u_0$ .

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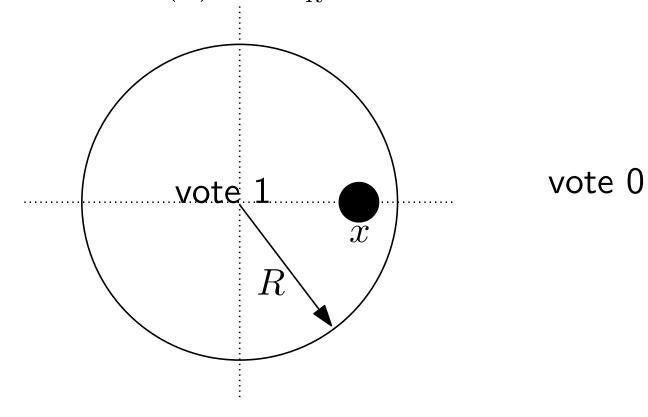
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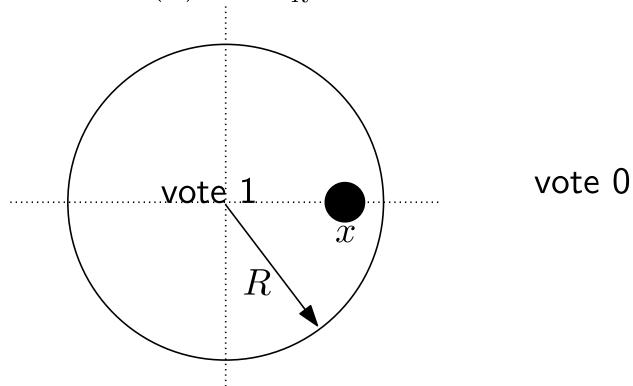
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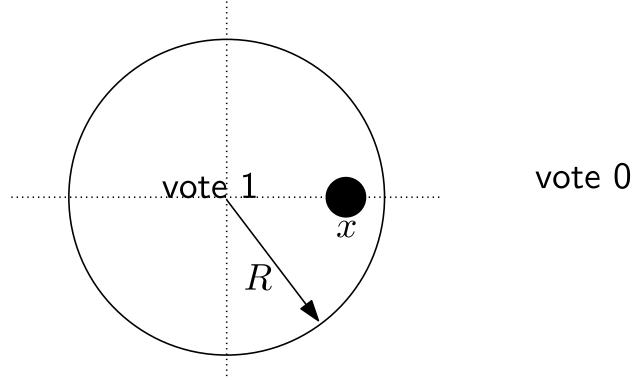


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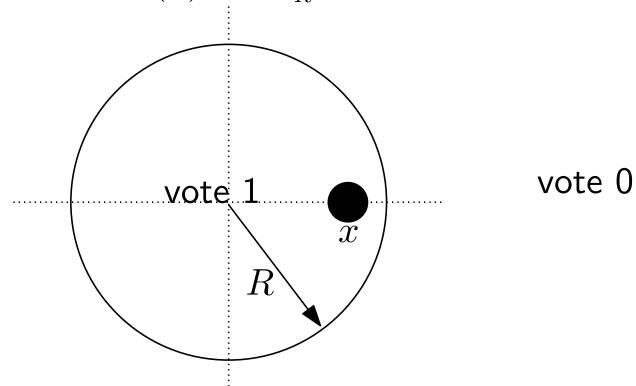
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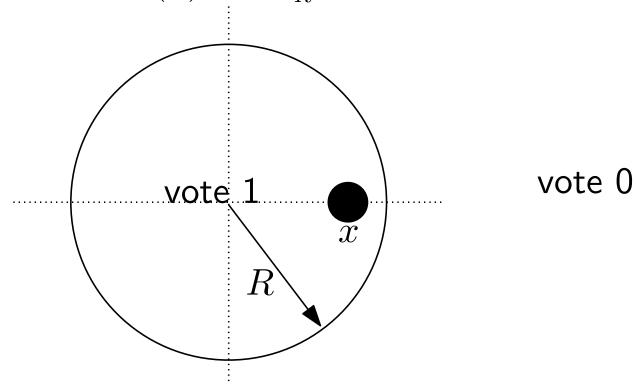


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 $\Rightarrow$  the hybrid zone should move as curvature flow for small times. For big times use Markov property + iterate carefully.

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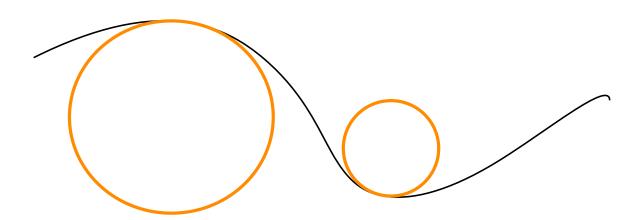
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And  $\partial\Gamma_0$  looks *locally* like a circle, we apply the first case.



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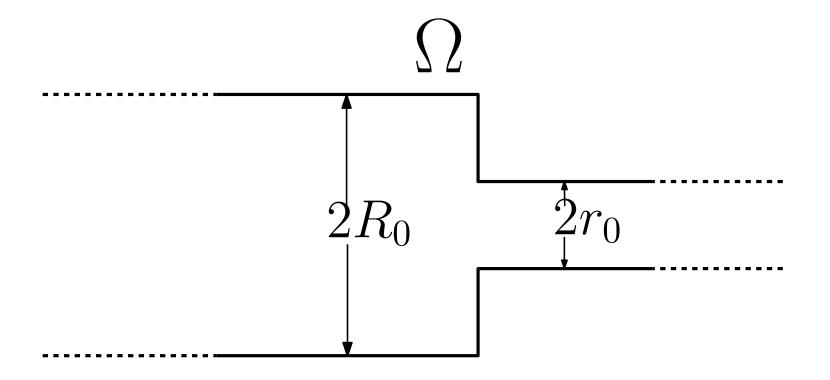
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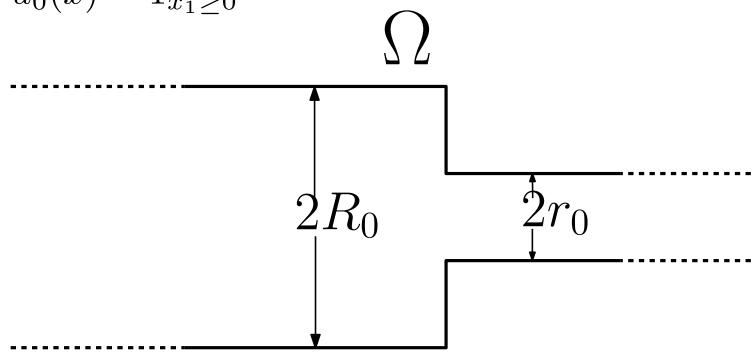
What can we say about the evolution of hybrid zones?

Mean curvature flow does not behave well with boundaries.

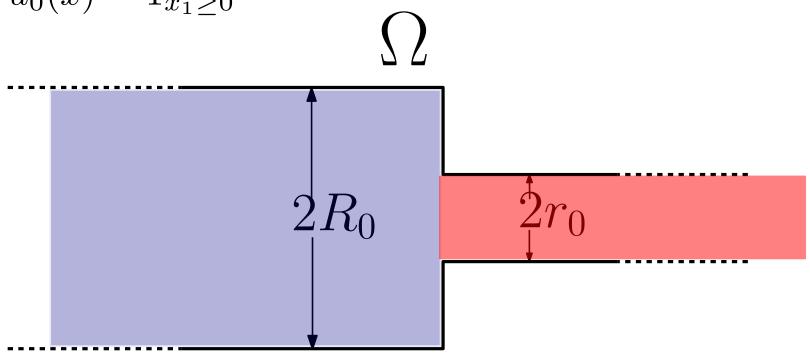
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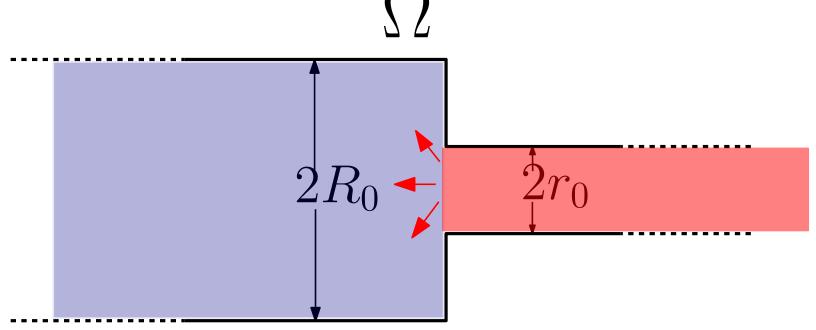
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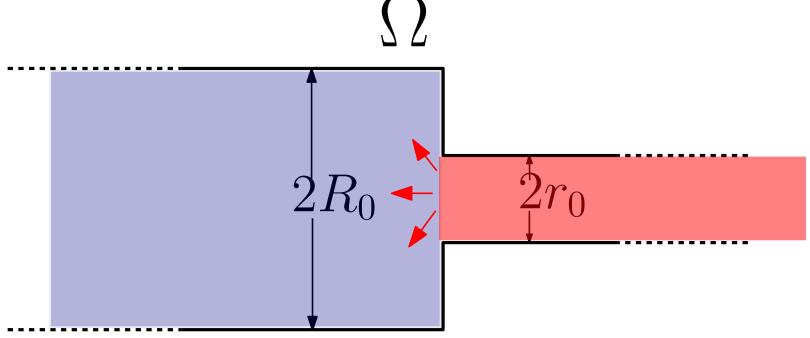
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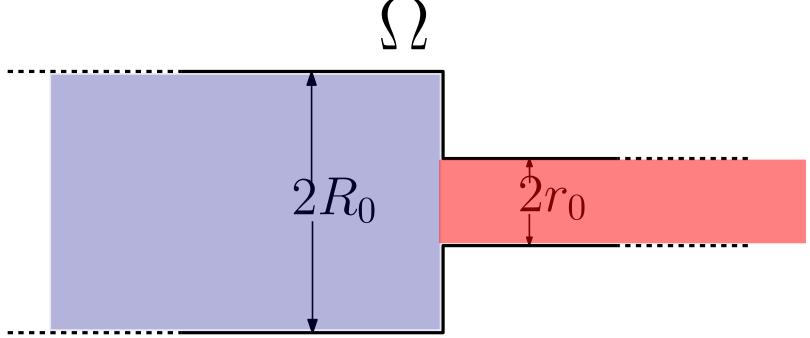


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Who wins? Dispersal or selection?

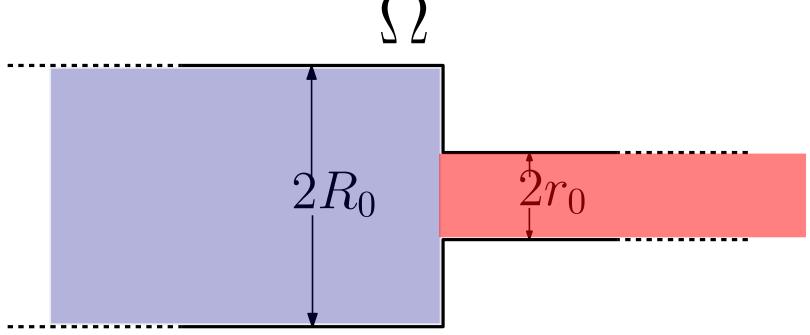
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Theorem (Beresticky et al. 2017 really short version)

If  $r_0$  is big enough in comparison to  $\nu$  then invasion occurs. If  $r_0$  is small enough in comparison to  $\nu$  then blocking occurs.

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Can we say more? Maybe using BBM and the voting system?

Theorem 3 (Etheridge, L.)

Let W(t) be the historical paths of a Branching **reflected** B.M.

Let  $\mathbb{V}_{u_0}(\mathcal{W}(t))$  be the vote defined in the same way as in Theorem 2.

Then 
$$u(x,t) := \mathbb{P}_x^{\epsilon}[\mathbb{V}_{u_0}(\mathcal{W}(t)) = 1]$$
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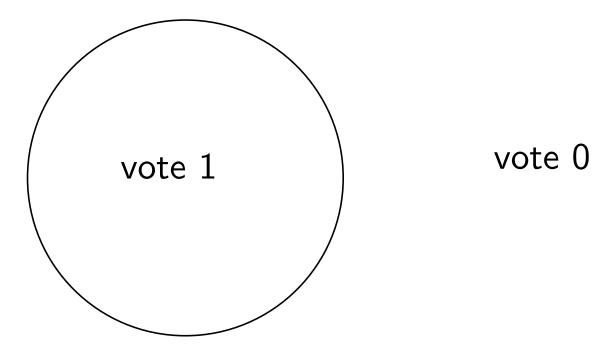
#### Theorem 5 (Etheridge, L.)

If  $r_0 > \frac{d-1}{\nu}$ , for small  $\epsilon$ , the solution of  $(N - AC_{\epsilon})$  presents invasion.

Ideas for Theorem  $3 \Rightarrow$  Theorem 4, Theorem 5,

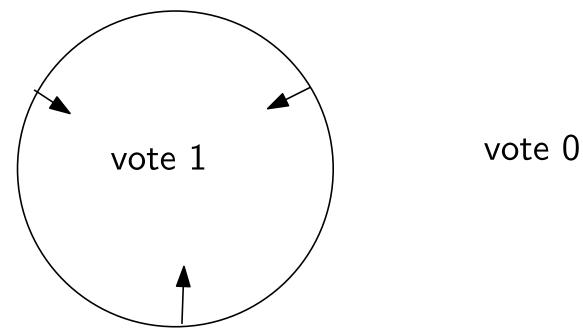
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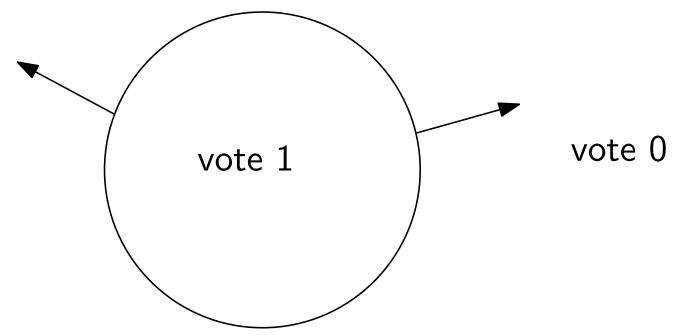
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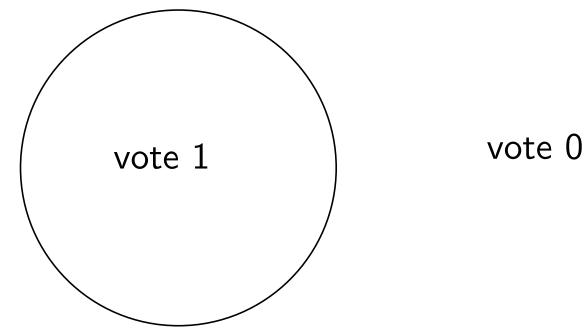


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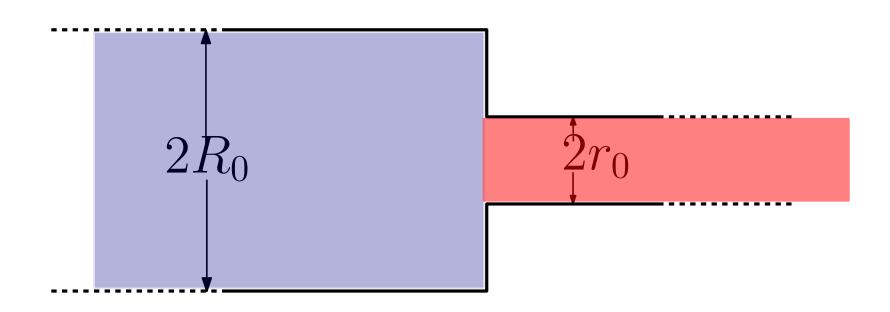
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Equilibrium if  $R=(d-1)/\nu$ . External push if  $R>(d-1)/\nu$ .

Theorem  $3 \Rightarrow$  Theorem 4 (Blocking)

Now we think about the barrier

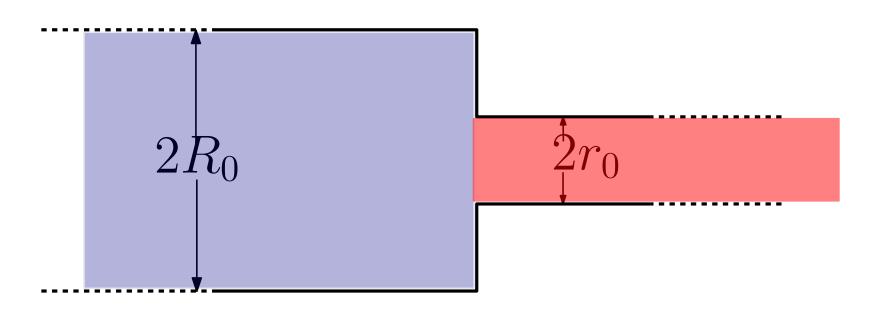
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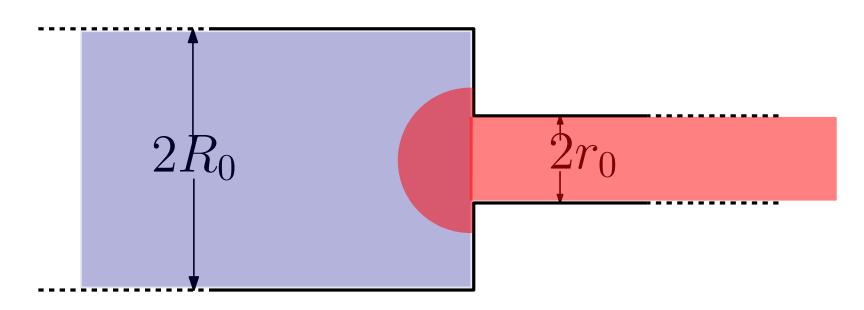


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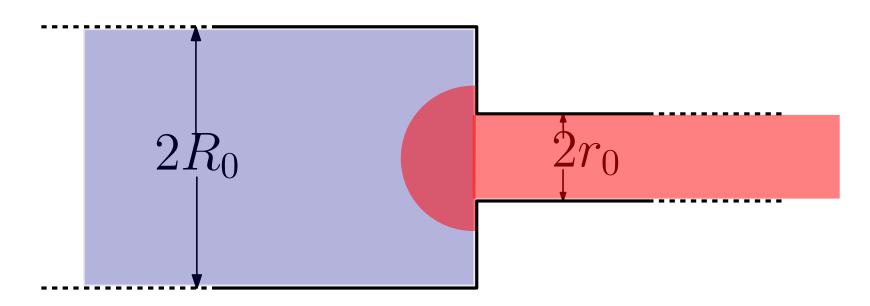


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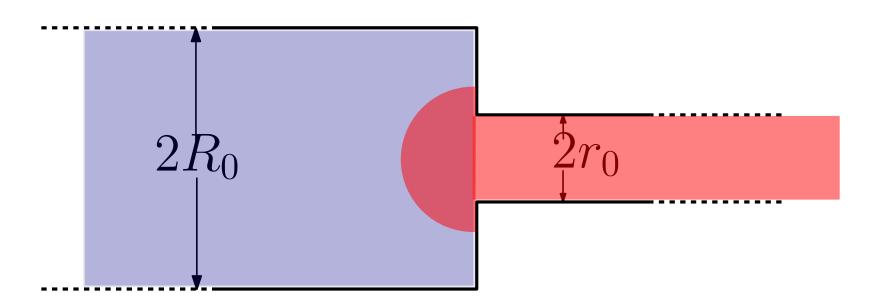


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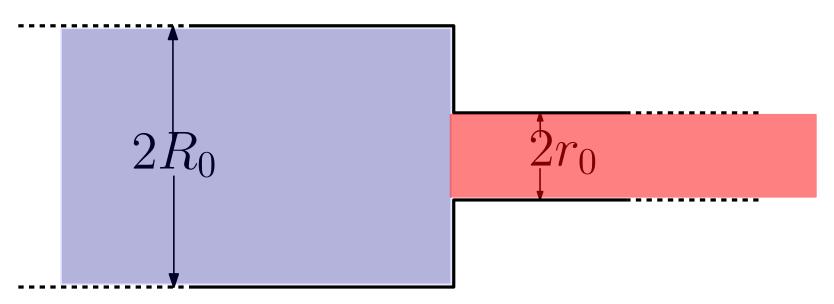
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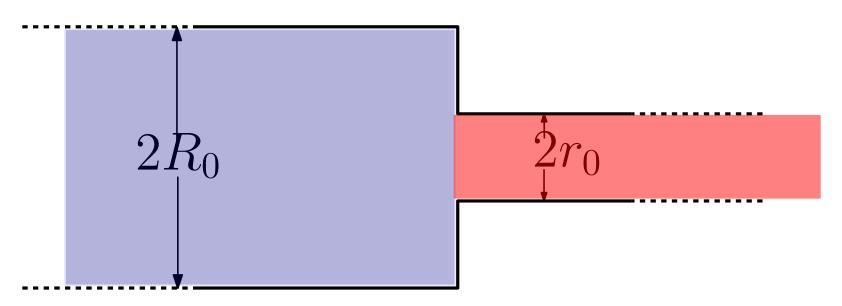
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$$r_0 < r \le (d-1)/\nu \Rightarrow \mathsf{Blocking}$$

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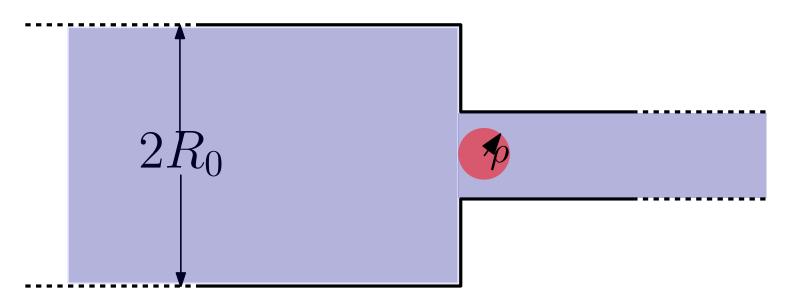


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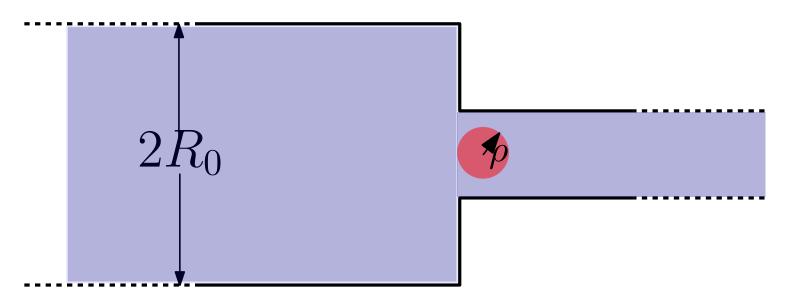
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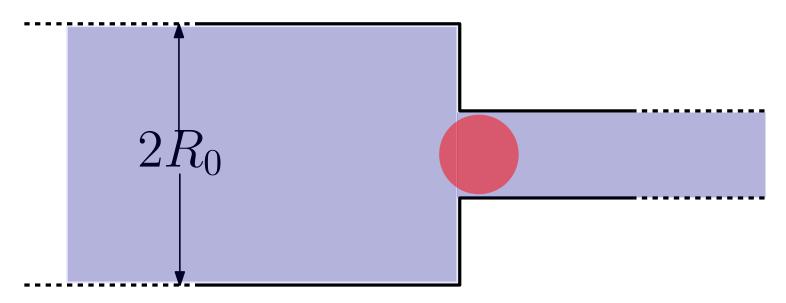


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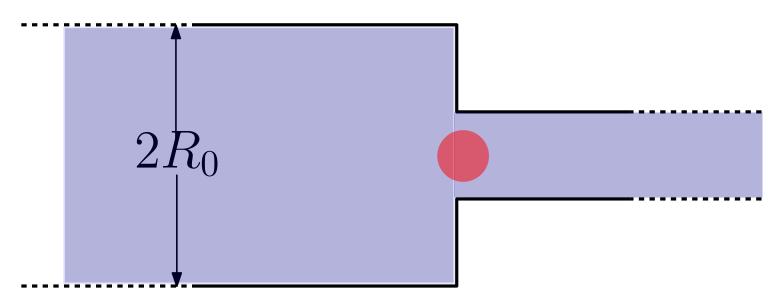
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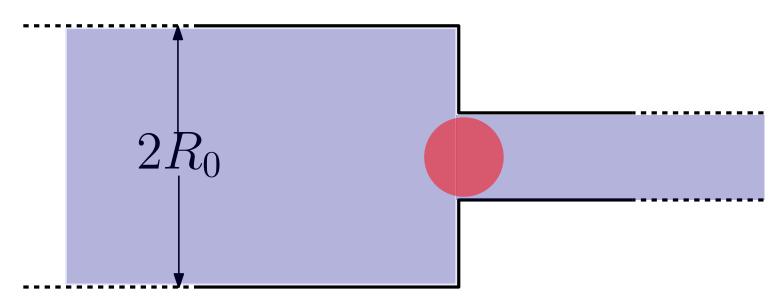
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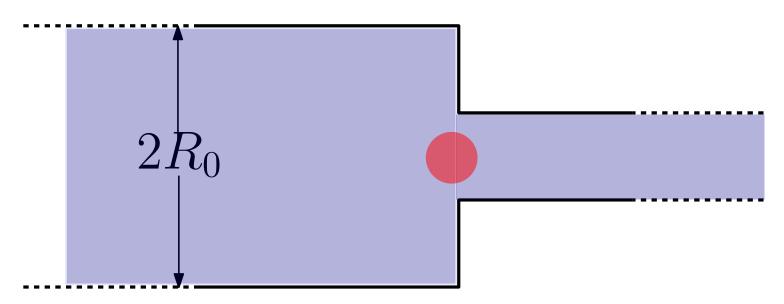
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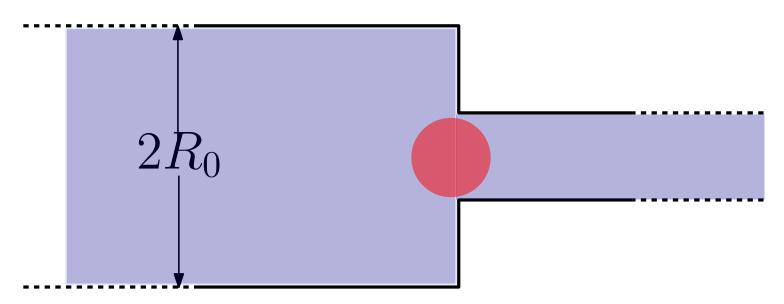
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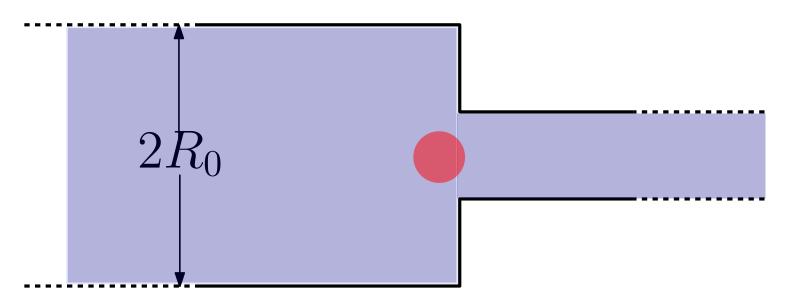
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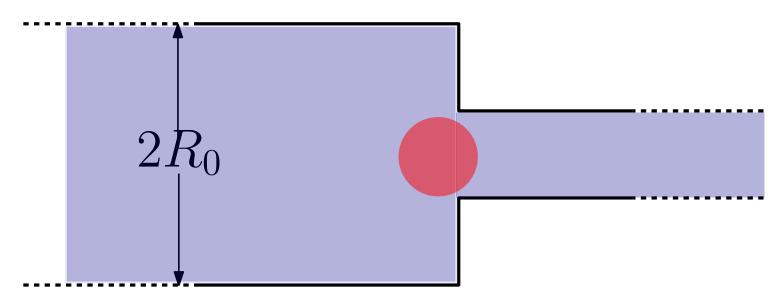
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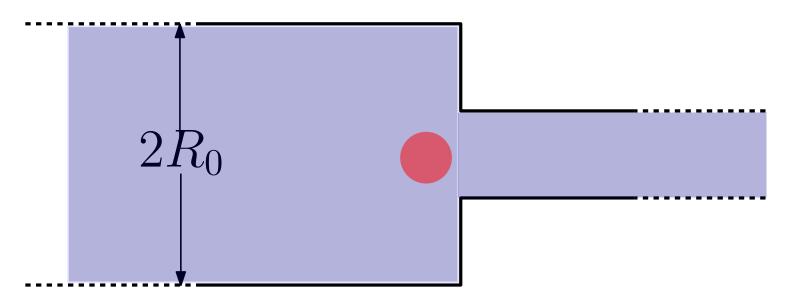
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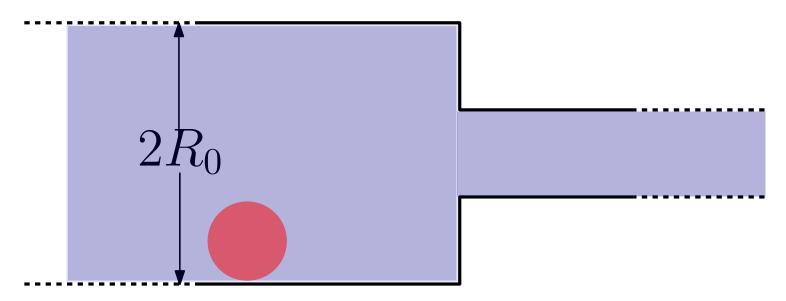
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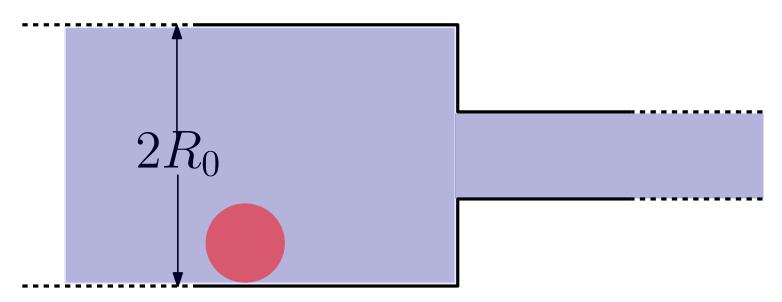
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We eventually can reach any point in  $\Omega$ . Any point of  $(AC-\epsilon)$ , for big time, is bounded below by  $\sim 1$ .

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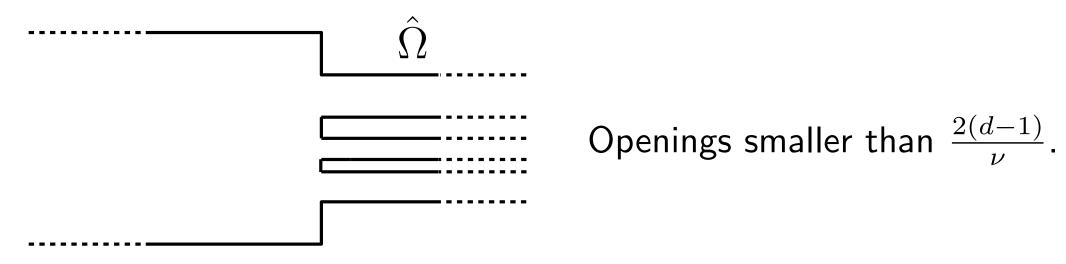
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For blocking cover openings with small balls perpendicular to the boundary.

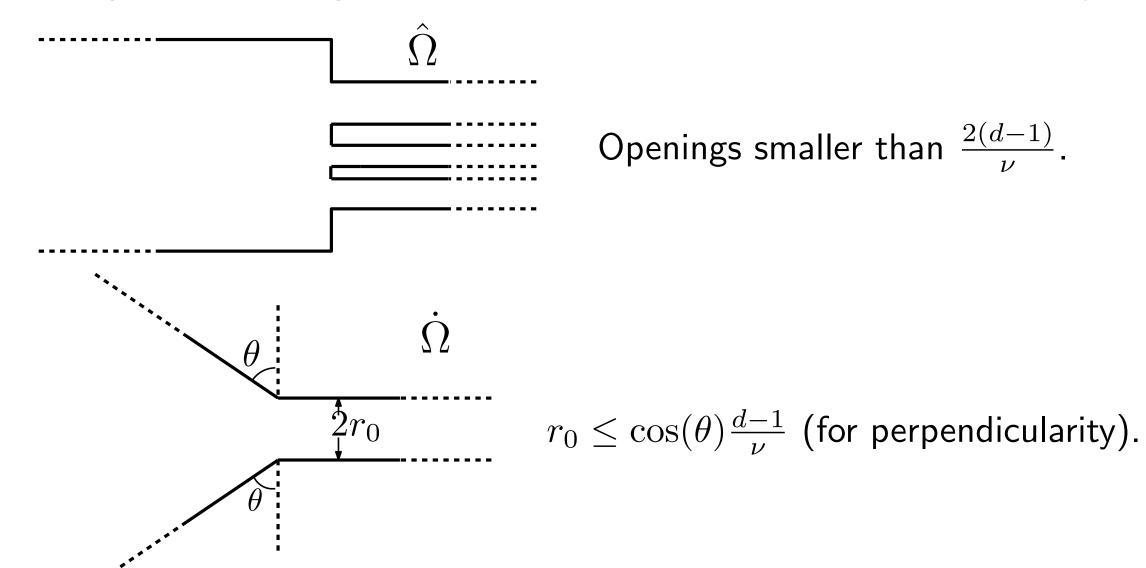
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**Basically**, we add coalescence to the B.M.M. of Theorem 1/3.

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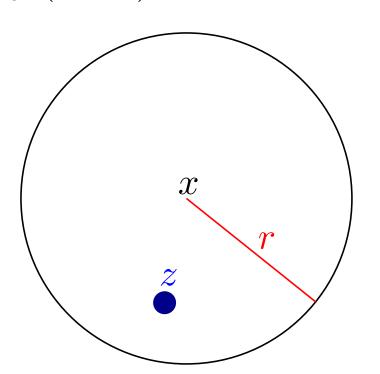
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Choose  $z \sim U(B_r(x))$ 

Then take  $K \sim Ber(w(t^-, z))$ .

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$$w(t,y) = (1-u)w(t^-,y) + uK.$$



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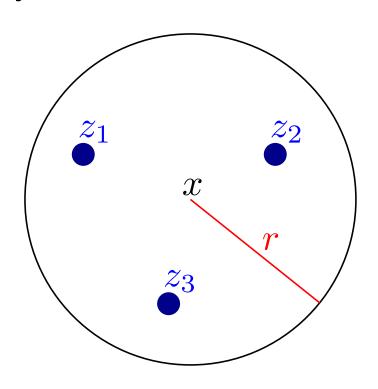
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type of individual at x at time  $t \sim \mathbb{V}_{w(0,x)}(\mathcal{W}(t))$ ,

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**Remark:** one can define the S $\Lambda$ FVS with reflection in the boundary of poligonal domains by the method of images.

Scale  $x \to n^{\beta}x$  and  $t \to nt$  for  $0 < \beta < \frac{1}{4}$ . And, for some  $\epsilon_n \to 0$  fast enough,

$$u_n = \frac{u}{n^{1-2\beta}}, \qquad s_n = \frac{1+\epsilon_n \nu}{\epsilon_n^2 n^{2\beta}}, \qquad \gamma_n = \nu \epsilon_n.$$

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Individuals move like Brownian motion

Selective event in an ancestral lineage happen at rate  $s_n u_n n \sim \frac{1+\epsilon_n \nu}{\epsilon_n^2}$ 

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In conclusion the type of an ancestral lineage  $\sim \mathbb{V}_{w_0}(\mathcal{W}(t))$ 

 $\Rightarrow w^n(x,t) \sim u(x,t)$  with u the solution of  $(N-AC_{\epsilon})$ .

Scale  $x \to n^{\beta}x$  and  $t \to nt$  for  $0 < \beta < \frac{1}{4}$ . And, for some  $\epsilon_n \to 0$  fast enough,

$$u_n = \frac{u}{n^{1-2\beta}}, \qquad s_n = \frac{1+\epsilon_n \nu}{\epsilon_n^2 n^{2\beta}}, \qquad \gamma_n = \nu \epsilon_n.$$

Theorem 4 (Etheridge, L.)(Scaling for small noise regime)

Let  $\Omega = (\mathbb{R}_- \times B_{d-1}(x, R_0)) \cup (\mathbb{R}_+ \times B_{d-1}(x, r_0))$ , for  $r_0 < R_0$ .

If  $r_0 < \frac{d-1}{\nu}$ , for big enough n,  $\mathbb{E}_{1_{x>0}}(w^n(t,x))$  is blocked.

If  $r_0 > \frac{d-1}{\nu}$ , for big enough n,  $\mathbb{E}_{1_{x>0}}(w^n(t,x))$  presents invasion.

What if the noise is strong? Suppose d=2. Let  $(u_n)_{n\in\mathbb{N}}\subseteq(0,1)$  with

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Individuals still move like Brownian motion.

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So 
$$(\xi_t^n) \sim W(t)$$

 $\Rightarrow w^n(x,t) \sim \text{solution to the heat equation.}$ 

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Theorem 5 (Etheridge, L.)(Scaling for big noise regime)

As n goes to infinity we have  $\mathbb{E}_{w(0,x)}(w^n(t,x))$ 

converges to the solution of the heat equation.

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As n goes to infinity we have  $\mathbb{E}_{w(0,x)}(w^n(t,x))$ 

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**Remark**: No blocking in this case! (The heat equation converges to a constant).

The branching rate of potential ancestor,  $s_n n^{2\beta}$ , could go to infinity.

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So we see BBM but with less particles. Harder to get blocking.

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Scale  $x \to n^{\beta}x$  and  $t \to \frac{n^{2\beta}}{u_n}t$ . For  $(s_n)_{n \in \mathbb{N}}$  suppose  $\frac{s_n n^{2\beta}}{u_n \log(n)} \to K$ .

Conjecture (Etheridge, L.)(Scaling for mild noise regime)

Let 
$$\Omega = (\mathbb{R}_- \times B_{d-1}(x, R_0)) \cup (\mathbb{R}_+ \times B_{d-1}(x, r_0))$$
, for  $r_0 < R_0$ .

There is some  $r_* << \frac{d-1}{\nu}$  such that

If  $r_0 < r_*$ , for big enough n,  $\mathbb{E}_{1_{x \ge 0}}(w^n(t,x))$  presents blocking.

If  $r_0 > r_*$ , for big enough n,  $\mathbb{E}_{1_{x>0}}(w^n(t,x))$  presents invasion.

#### Extra references

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- [2] Nick Barton and his group.
- [3] Cassandra N. Trier, Jo S. Hermansen, Glenn-Peter Sætre, Richard I. Bailey. Evidence for Mito-Nuclear and Sex-Linked Reproductive Barriers between the Hybrid Italian Sparrow and Its Parent Species.
- [4] N H Barton and G M Hewitt.

  Adaptation, speciation and hybrid zones.
- (Pictures of MCF) Curve by Grayson (1987), Bunny by C.Stocker (found in Voigt 2007).

#### On hybrid zones and the effect of barriers

Ian Patrick Letter Restuccia

15.11.2021 · Oxford probability seminar





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