

# Taking advantage of pathogen diversity and plant immunity to minimize disease prevalence

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ModStatSAP  
November 24 2020

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*Inria*

# French context

- Pesticide use impacts public health and biodiversity
- Target: halve pesticide use by 2025
- Constraints:
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  - Breakdown and durability of resistance



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## Wanted New agro-ecological methods

- Host mixtures remain to be optimized
- Plant immunity is key but absent from mathematical models so far



# The Yunnan province experimentation (2000)

In mixtures, the prevalence of Rice blast was reduced from 20% to 1% on susceptible varieties compared to susceptible monocultures (dilution effect)

## letters to nature

### Genetic diversity and disease control in rice

Yuyong Zhu<sup>1</sup>, Hairu Chen<sup>1</sup>, Jinghua Fan<sup>1</sup>, Yanyue Wang<sup>1</sup>, Yan Li<sup>1</sup>, Jianbing Chen<sup>1</sup>, Jinxiang Fan<sup>1</sup>, Shisheng Yang<sup>1</sup>, Lingping Hu<sup>2</sup>, Hei Leung<sup>3</sup>, Tom W. Mew<sup>4</sup>, Paul S. Teng<sup>4</sup>, Zonghua Wang<sup>1</sup> & Christopher C. Mundt<sup>4</sup>

<sup>1</sup> The Phytopathology Laboratory of Yunnan Province, Yunnan Agricultural University, Kunming, Yunnan 650201, China

<sup>2</sup> Honghe Prefecture Plant Protection Station of Yunnan Province, Kaiyuan 661400, China

<sup>3</sup> Jianzhui County Plant Protection Station of Yunnan Province, Jianzhui 654300, China

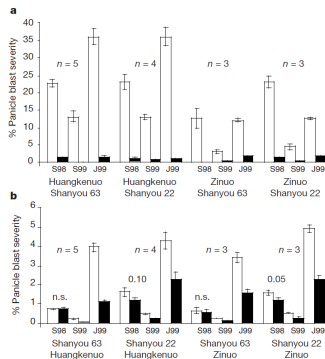
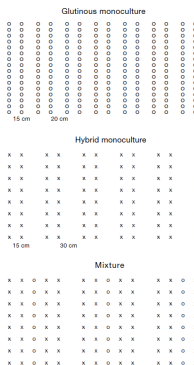
<sup>4</sup> Shipping County Plant Protection Station of Yunnan Province, Shipping 662200, China

<sup>5</sup> Division of Entomology and Plant Pathology, International Rice Research Institute, MCPO Box 3127, 1271 Makati City, The Philippines

<sup>6</sup> Department of Botany and Plant Pathology, 2082 Cordley Hall, Oregon State University, Corvallis, Oregon 97331-2902, USA

Crop heterogeneity is a possible solution to the vulnerability of monocultured crops to disease<sup>1-3</sup>. Both theory<sup>4</sup> and observation<sup>2,5</sup> indicate that genetic heterogeneity provides greater disease suppression when used over large areas, though experimental data are lacking. Here we report a unique cooperation among farmers,

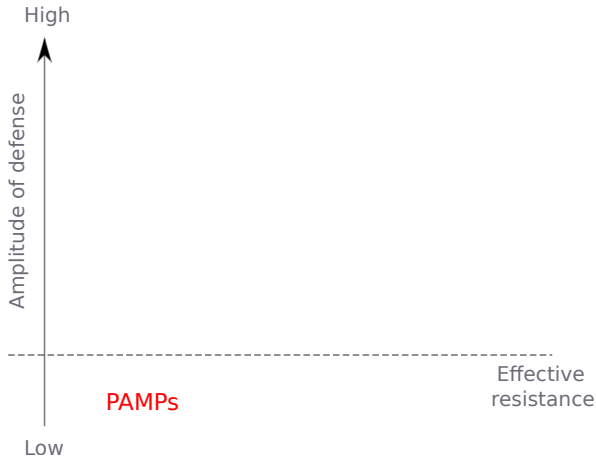
NATURE | VOL 406 | 17 AUGUST 2000 | www.nature.com



On resistant varieties compared to resistant monocultures, the prevalence decreased from 2% to 1%. Why is that?

# Plant immune system

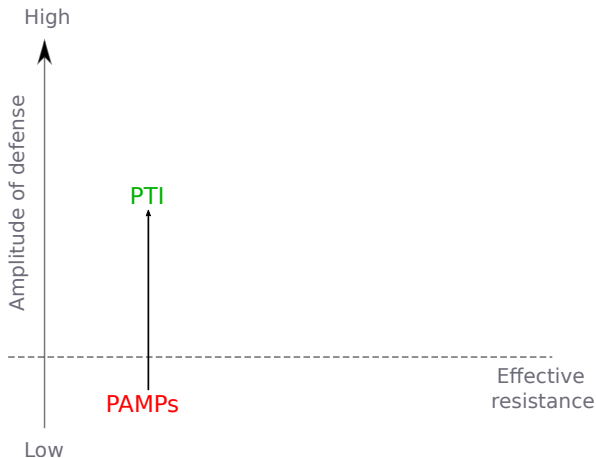
2 levels of immunity (Jones and Dangl, 2006; Milgroom, 2015)



**PAMPs** = Pathogen molecules, **PTI** = PAMP triggered immunity,  
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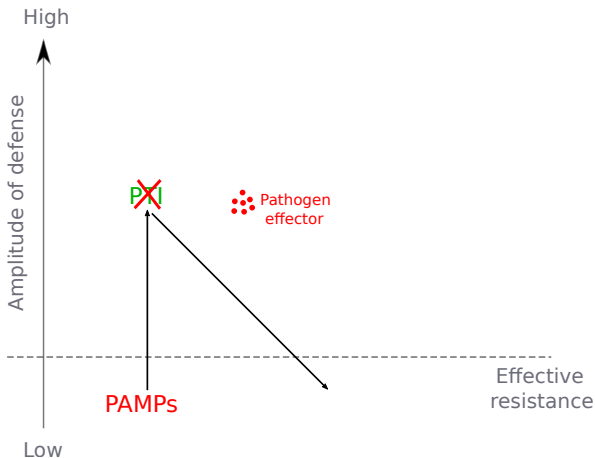
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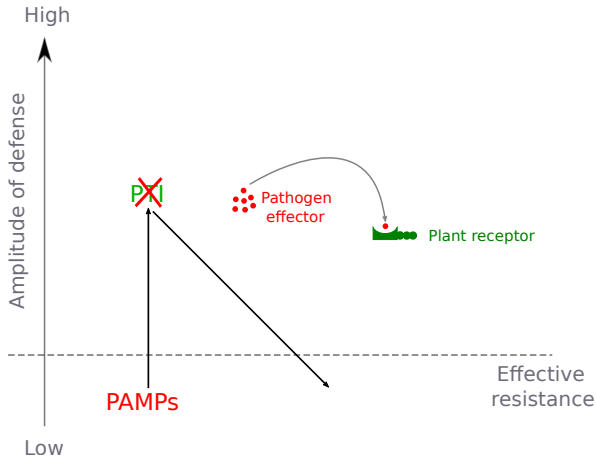


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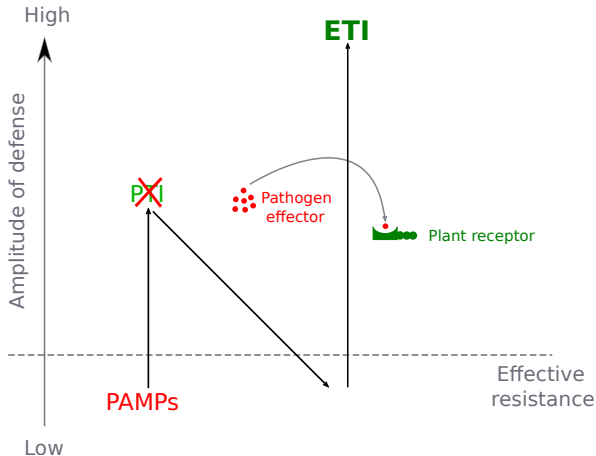
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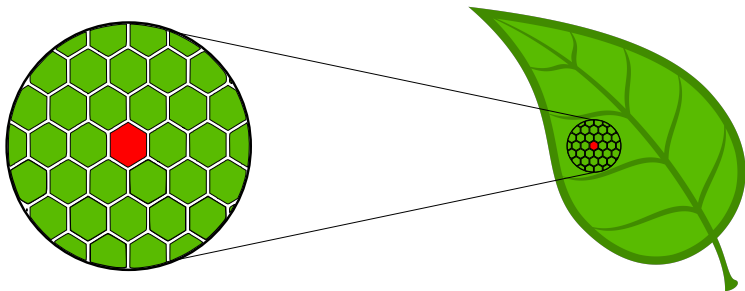
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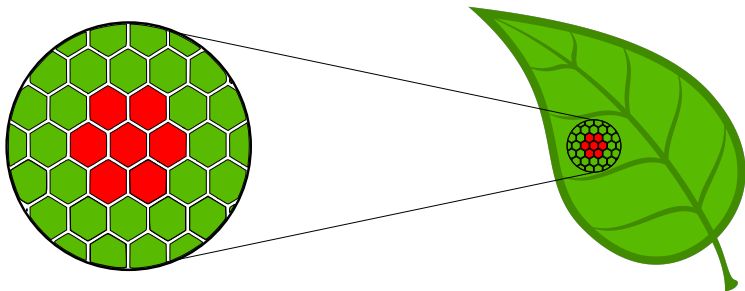
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Infection by **an avirulent pathogen** on a **resistant plant**:



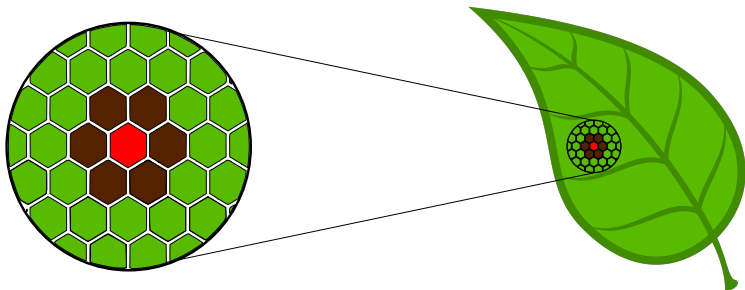
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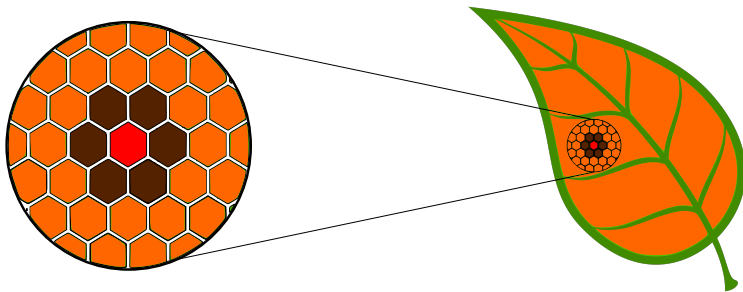


## Hypersensitive response

Programmed death of cells located where the infection occurred

# HR induces systemic acquired resistance

Infection by **an avirulent pathogen on a resistant plant**:



## Systemic acquired resistance SAR

Resistance response that applies to the entire plant  
Derives from ETI

# PRIMING

# SAR and priming in host mixtures

## Infection probability

Pathogen \ Host	Susceptible	Resistant
Wild-type	1	0
Resistance-breaking	1	1

→ What is the impact of the priming on epidemiological dynamics?

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# SAR and priming in host mixtures

## Infection probability

Pathogen \ Host	Susceptible	Resistant	Primed resistant
Wild-type	1	<b>Priming</b>	0
Resistance-breaking	1	1	$1-\rho$

→ What is the impact of the priming on epidemiological dynamics?

# Epidemic model: 3 important parameters

## Proportion of resistant hosts $p$

## Resistance-breaking cost $c$

Decreases the fitness of the resistance-breaking variant on both susceptible and resistant plants

*Xanthomonas axonopodis*, bacteria (Wichmann and Bergelson, 2004 - Genetics)

*Meloidogyne incognita*, nematode (Castagnone-Sereni et al, 2007 - Evo. Eco.)

*Potato virus Y* (Janzac et al, 2010 - MPMI)

*Phytophthora infestans*, oomycete (Montarry et al, 2010 - Evolutionary Biology)

*Soybean mosaic virus* (Khatabi et al, 2013 - MPP)

*Leptosphaeria maculans*, fungi (Bousset et al, 2018 - Evolutionary applications)

## Priming efficiency $\rho$

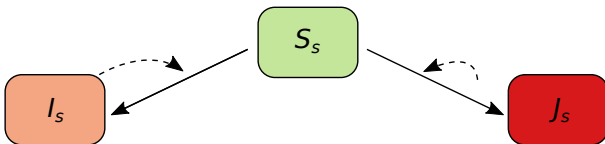
Reduces the infection success of the resistance-breaking variant on resistant plants

*Tobacco mosaic virus* (Ross, 1961 - Virology)

Full priming efficiency (Kuc, 1982 - BioSciences)

*A. thaliana* (Maleck et al., 2000 - Nature genetics)

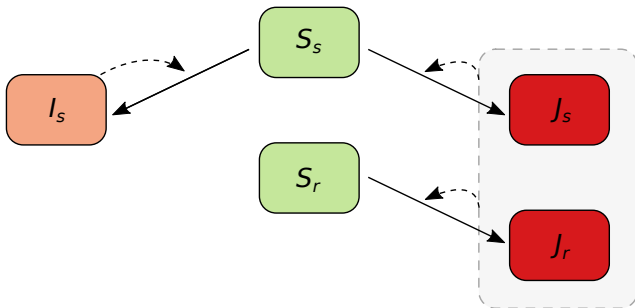
# Epidemic model with wild and resistance-breaking variants



$S_s = N_s - I_s - J_s$  and  $S_r = N_r - S_r^* - J_r$ , where  $N = \text{constant}$   
 $\beta = \text{pathogen transmission rate}$  and  $\alpha = \text{harvest and replanting rate}$

$$\begin{cases} \dot{I}_s &= \beta I_s S_s - \alpha I_s \\ \dot{J}_s &= (1 - c)\beta(J_s \quad )S_s - \alpha J_s \end{cases}$$

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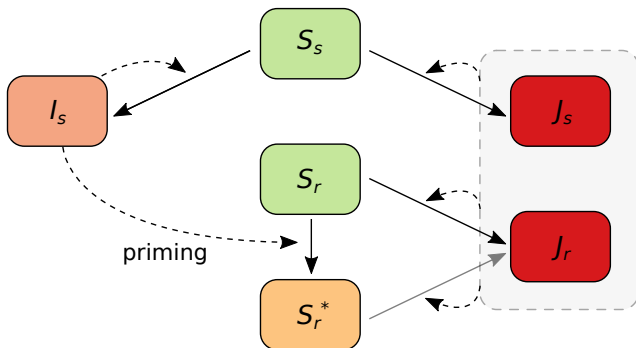


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$-\alpha J_r$ .

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$$\begin{cases} \dot{I}_s &= \beta I_s S_s - \alpha I_s \\ \dot{S}_r^* &= \beta I_s S_r - (1 - \rho)(1 - c)\beta(J_s + J_r)S_r^* - (\gamma + \alpha)S_r^* \\ \dot{J}_s &= (1 - c)\beta(J_s + J_r)S_s - \alpha J_s \\ \dot{J}_r &= (1 - c)\beta(J_s + J_r)S_r + (1 - \rho)(1 - c)\beta(J_s + J_r)S_r^* - \alpha J_r. \end{cases}$$

# Model

Let be  $x = \frac{I_s}{N}$ ,  $m = \frac{S_r^*}{N}$ ,  $y = \frac{J_s}{N}$ ,  $z = \frac{J_r}{N}$ , and  $t^* = \alpha t$  with

$N$  the number of plants

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$$\nu = \frac{\gamma + \alpha}{\alpha} \geq 1 \quad \text{where } \gamma \geq 0 \text{ corresponds to the loss of priming,}$$

and  $\alpha \geq 0$ , the harvest and replanting rate.

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Always exists

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**Equilibria stability ?**

# Cooperative systems

*Hal Smith, 2008 ; Hirsch, 1989*

## Conditions

- Positive interactions between variables,
- Irreducible jacobian matrix.

# Cooperative systems

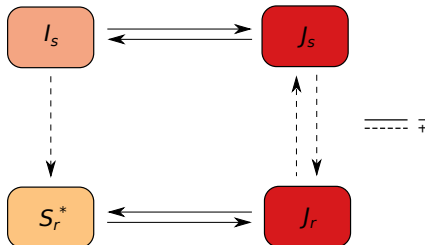
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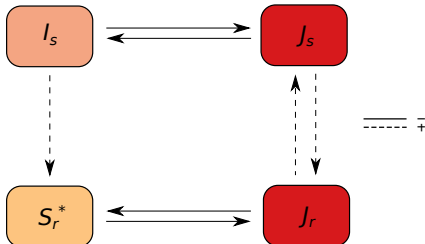
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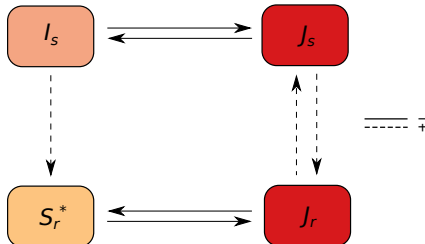
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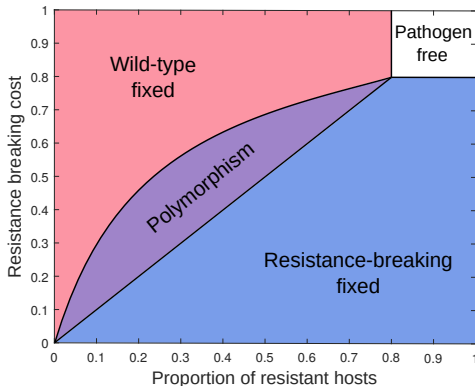
The system converges towards an equilibrium **which can only be the coexistence equilibrium!**

# Epidemiological dynamics stabilize genetic polymorphism

Two conditions:

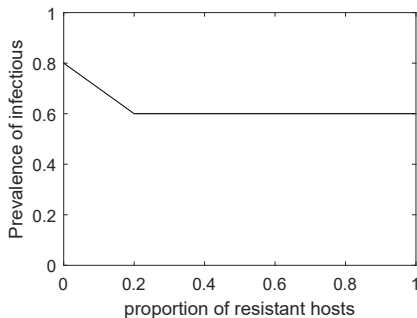
$$c > p \text{ and } \rho < \frac{(R(1-p)+\nu-1)(R(1-c)p-c)}{Rp(1-c)(R(1-p)-1)} = \frac{[pR_v-c][R_a+\nu-1]}{[R_a-1]R_v p}$$

For given  $R = 5$ ,  $\nu = 1$ , and  $\rho = 0.8$ :



# Priming increases the effectiveness of host mixtures

Prevalence of the disease,  $P = I_s + J_s + J_r$ , when  $\rho$  (Priming effectiveness) increases:



$\rho = 0$  for black line

$\rho = 0.2$  for blue line

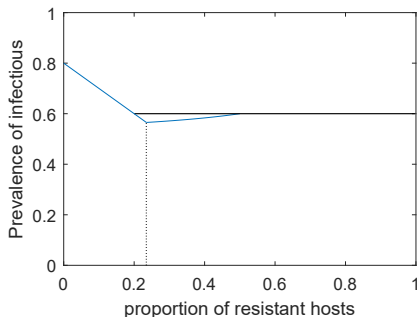
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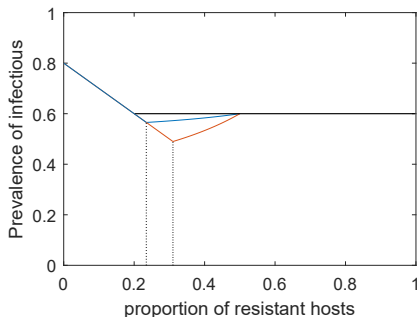
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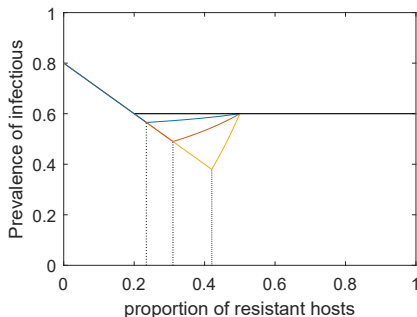
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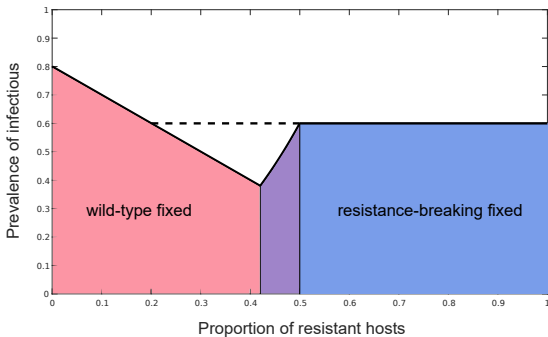
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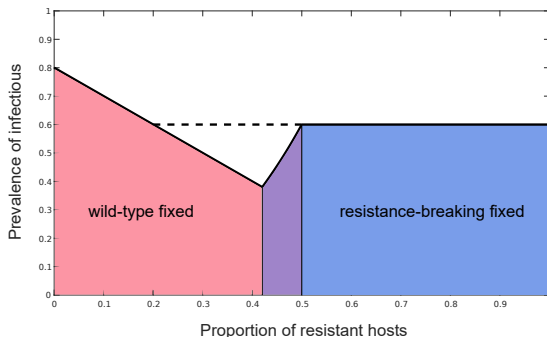
# Take-Home Messages



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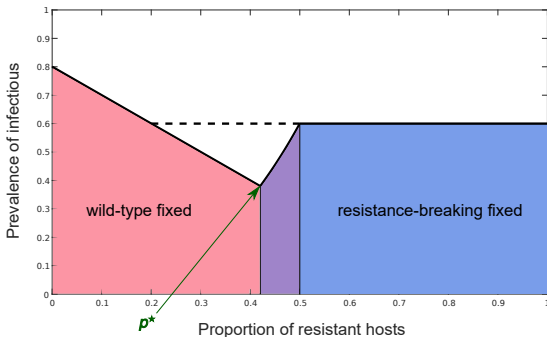


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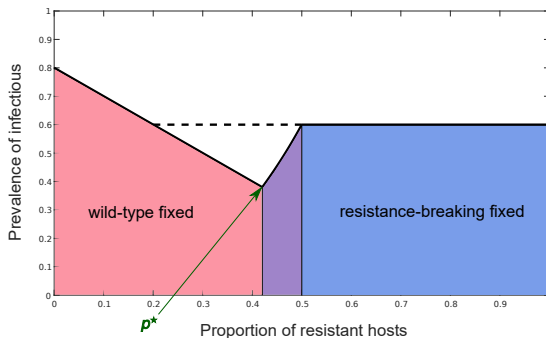
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- Priming induced direct **Cross-protection** between hosts

# Discussion

- Evidence of the interest of priming in host mixtures shown analytically for the first time on an epidemiological model.
  - Good agreement with experimental studies, which indicate that:
    - Priming may account for 20% to 40% of the disease reduction in mixtures
- (Lannou & Pope 1997, Calonnec et al 1996)
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- The use of an optimal proportion of resistant,  $p^*$ , prevents the emergence of virulent pathogens.

In addition to reducing the prevalence of the disease :

**Host mixtures and priming increase durability of resistances.**

**Thanks for listening!**  
**Questions ?**

## **Behavioural epidemiology:**

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- Conformism vs Stubbornness vs Responsiveness behavior, Mcquaid et al, 2017.



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Behavioural model  
Epidemiological model } linked by a decision variable,  $p$ .

# Transient phase

Prevalence is minimized at equilibrium, i.e. after a long periode of time.

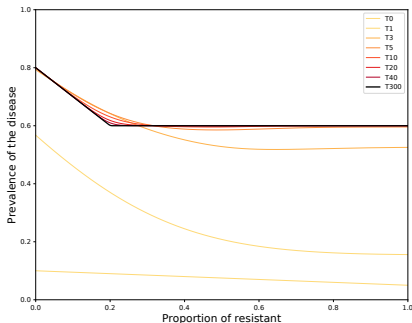
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$R = 5, c = 0.5, \rho = 0, \nu = 1$



$R = 5, c = 0.5, \rho = 0.8, \nu = 1$

