

a model to highlight the  
relations between plant growth, powdery  
crop management and climate; a  
sensitivity analysis



Calonnec A.

Burie J-B.  
Langlais M.

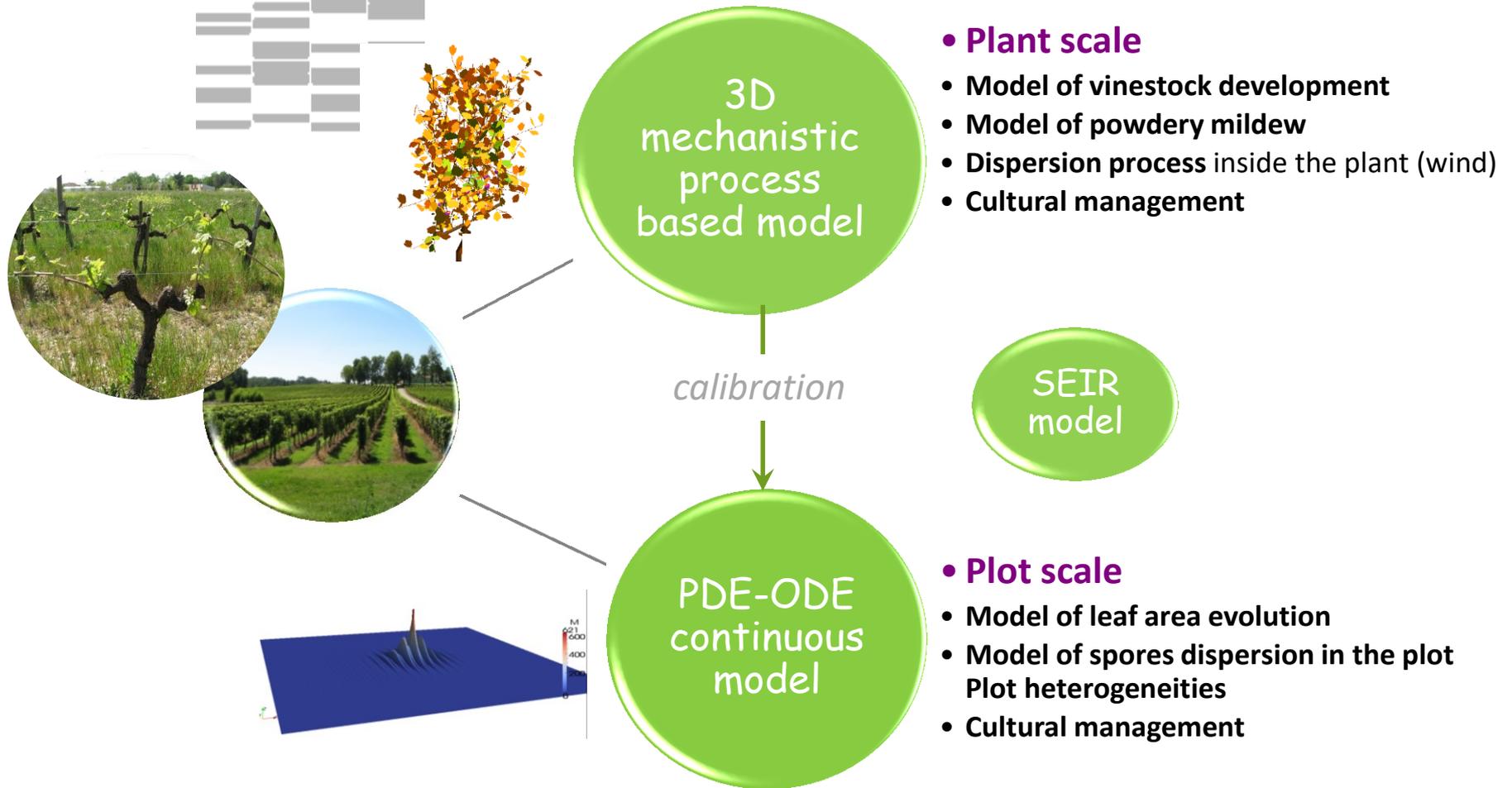
Mammeri Y.



Bruchou C.



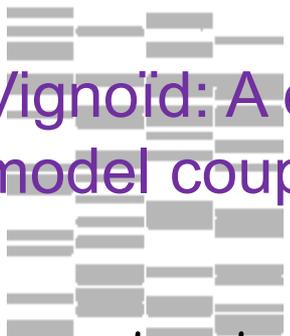
# Models for grapevine powdery mildew



Calonnec A, et al. (2008). *Plant Pathology*, 57, 493-508.

Burie JB, Langlais M, Calonnec A, (2011). *Annals of Botany* 107, 885-95.

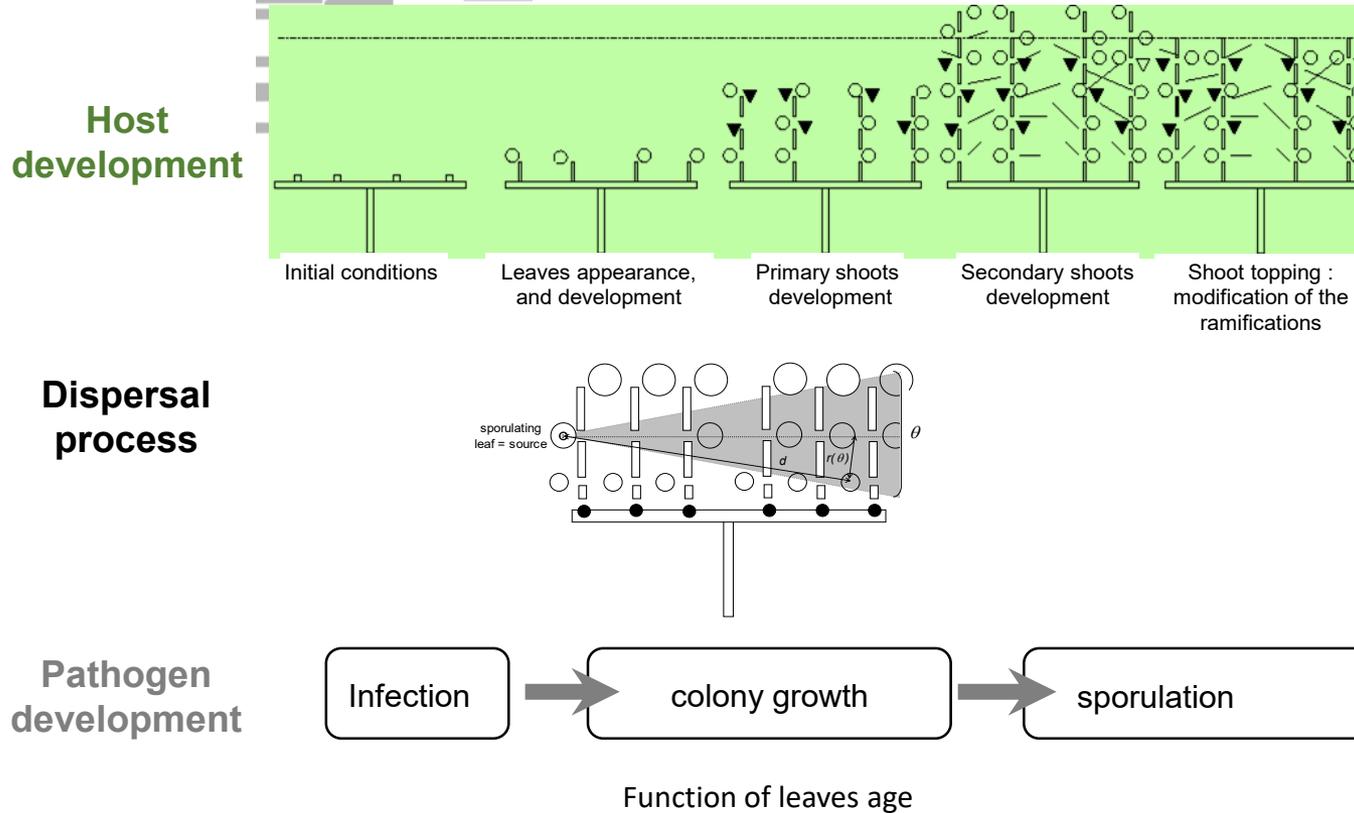
Mammeri Y, Burie JB, Langlais M, Calonnec A. (2014). *Ecological Modelling*, 290, 178-191



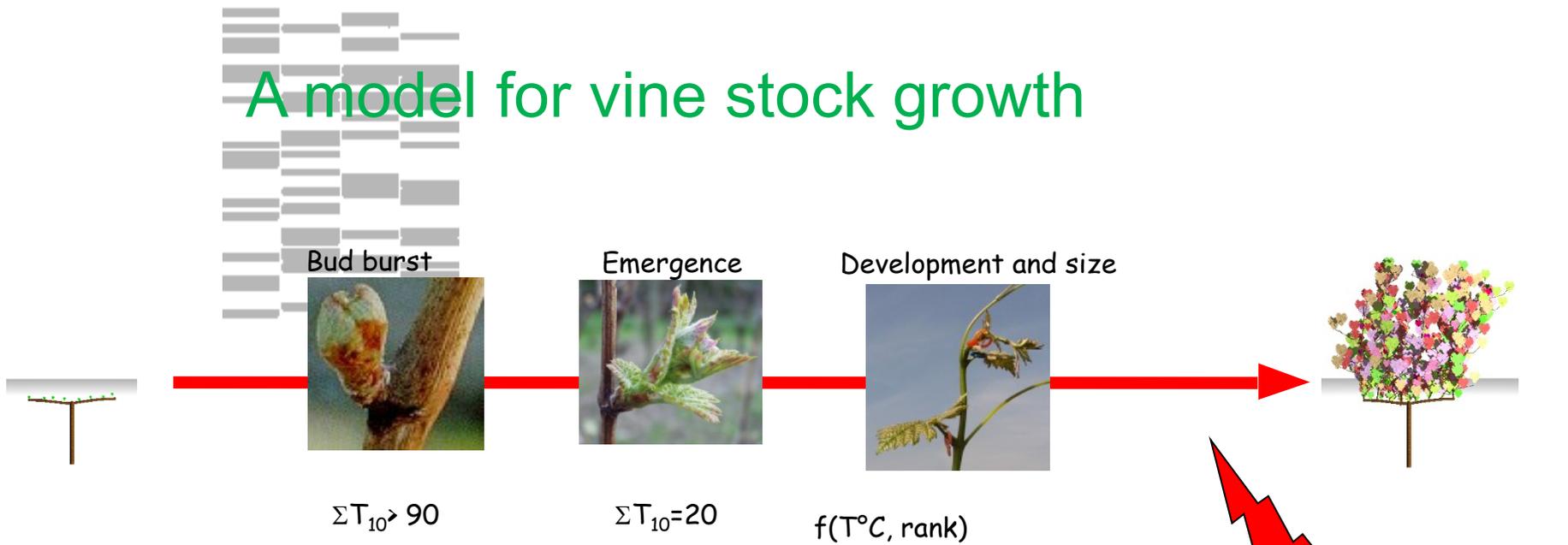
## Vignoid: A discrete mechanistic process based simulation model coupling grapevine and powdery mildew

- Understand and assess the dynamical interactions between the **Host** / **Pathogen** / **Environment** (plant growth, phenology, spore dispersion...)
- Simulate spatio-temporal dynamics from various scenarii (climate, training system, isolates, treatments...) and explore through the modelisation the capacity of host development to modify the disease progress
- Define plant conducts the most unfavorable to the pathogen development

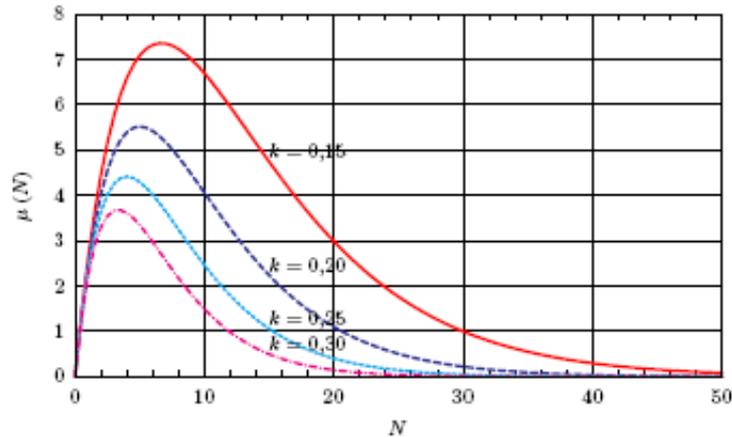
# An architectural model able to explore the host and pathogen processes involved



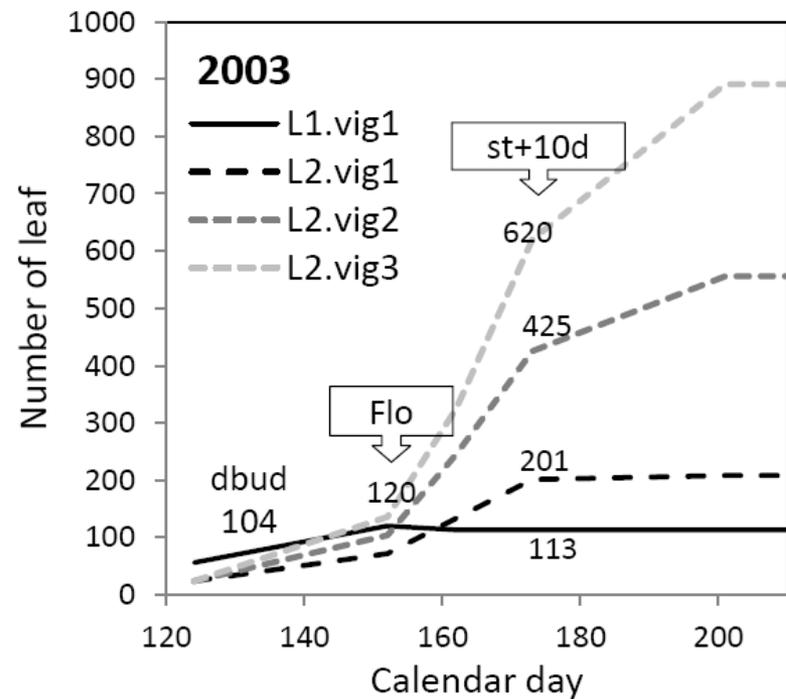
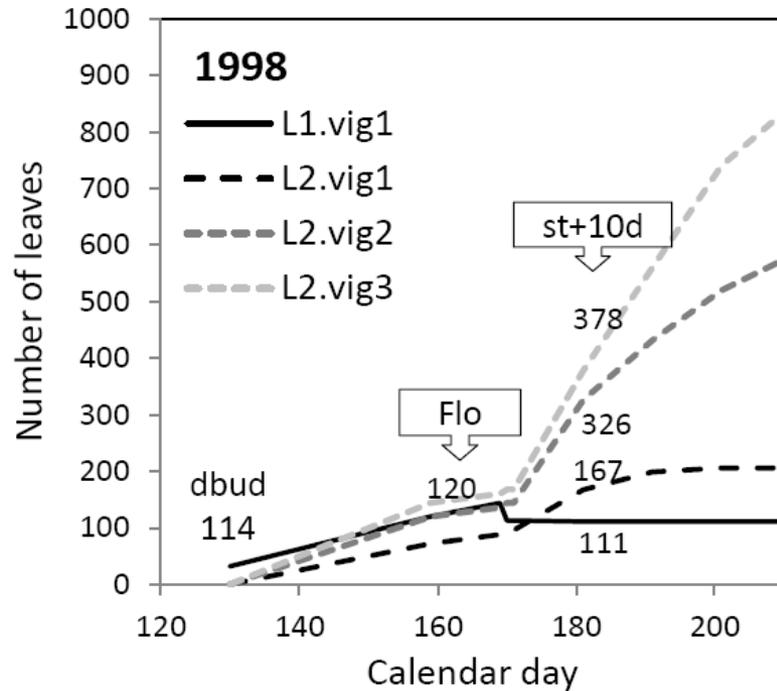
# A model for vine stock growth



- structure and dynamics of the canopy
- organs susceptibility
- the host variability



# Variation of Host development: vigour effect



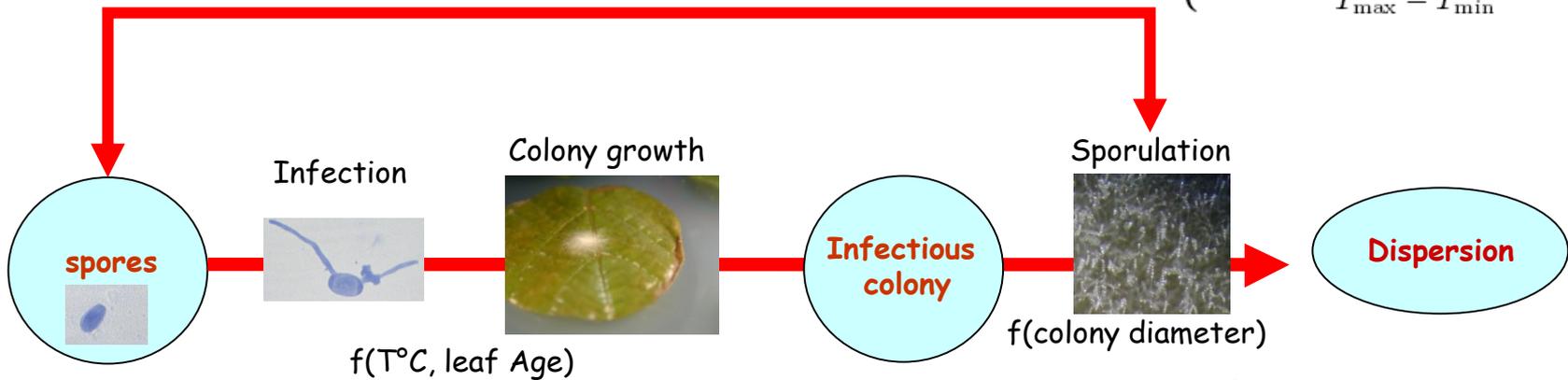
# A model for powdery mildew development



Latence period =  $f(T^{\circ}C)$

$$F(T) = RT_n(T)^n(1 - T_n(T))^m,$$

$$\left\{ \begin{array}{l} R = \frac{(m+n)^{m+n}}{m^m n^n} \\ T_n(T) = \frac{T - T_{\min}}{T_{\max} - T_{\min}} \end{array} \right.$$

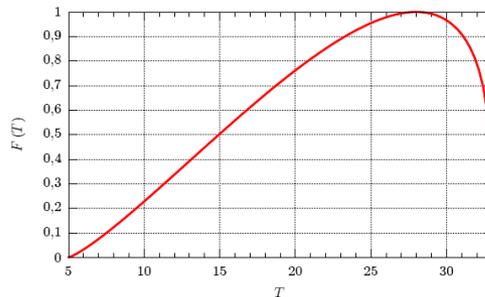


$f(T^{\circ}C, \text{leaf Age})$

$f(\text{colony diameter})$

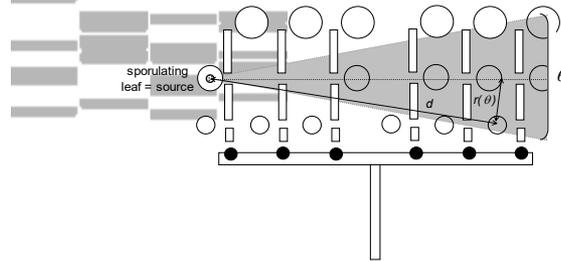
$$Q_r = Ae^{P_5 \text{Diam}}$$

$$I = I_{\max} F_g(T) e^{P_4 \text{Age} F}$$



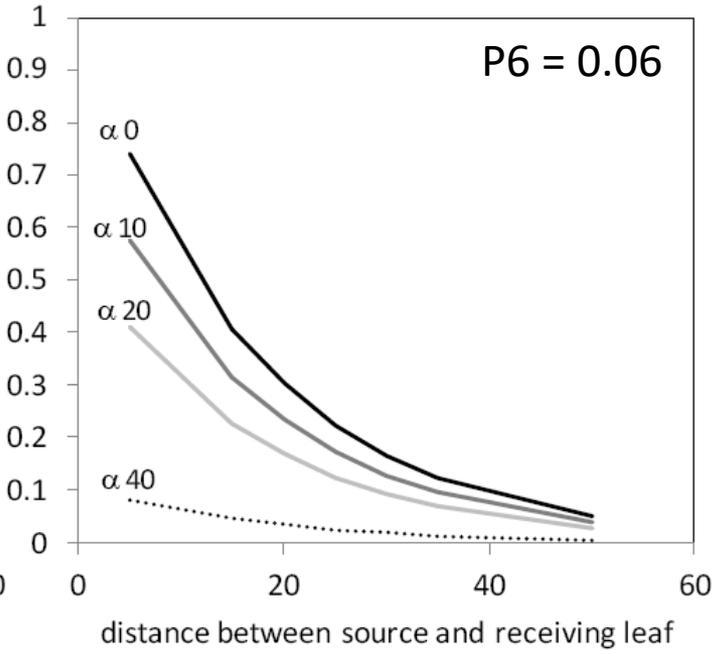
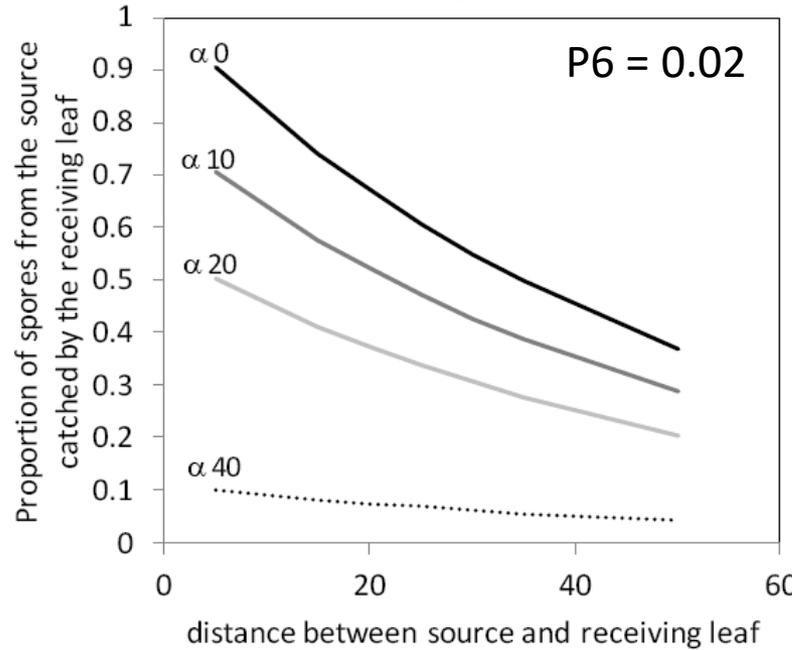
- The pathogen variability (isolate +/- aggressiveness,  $f(T^{\circ}C)$ , ontogenic resistance)

# A dispersion process

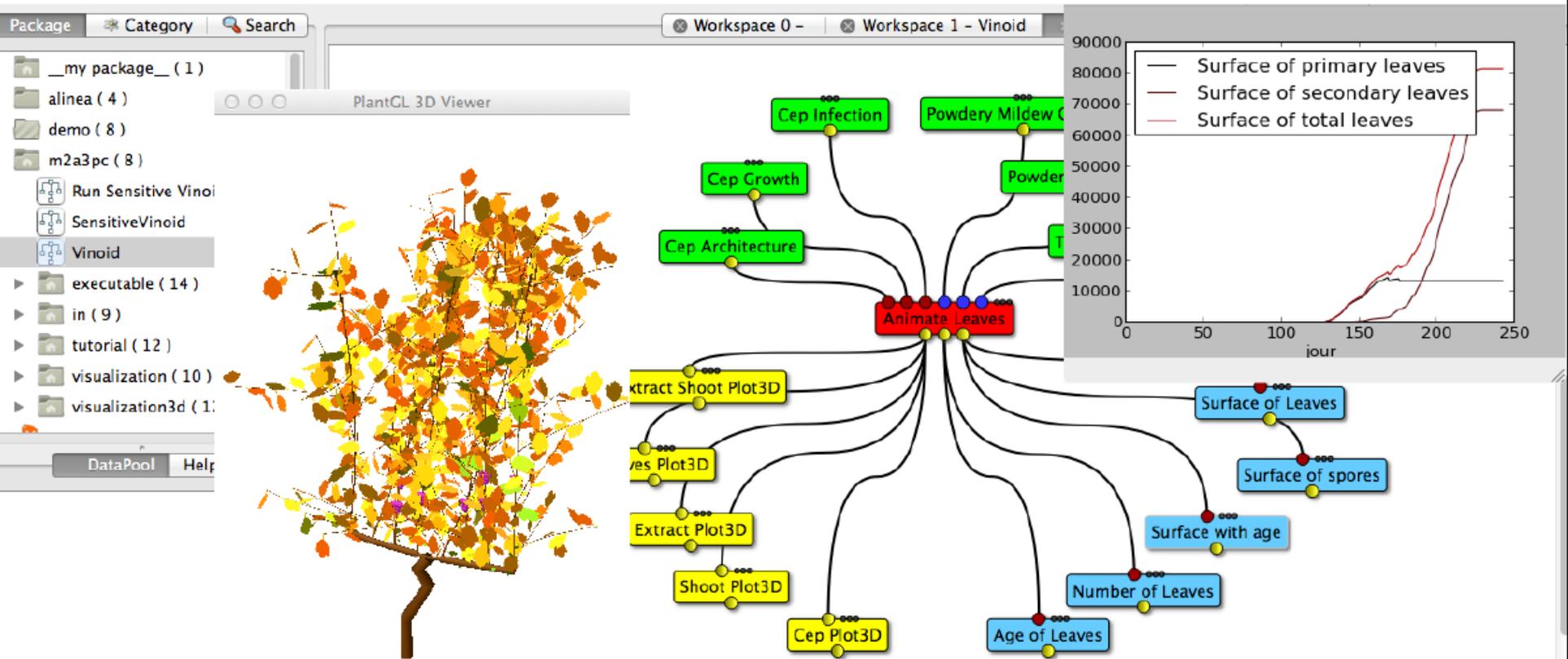


$$Q_c = Q_r \cdot S \cdot e^{-P6 \cdot d}$$

$Q_c$  = spores captured by the canopy  
 $Q_r$  = spores released by the source  
 $S$  = leaf surface  
 $d$  = distance to the source



# Platform Open Alea in Python language

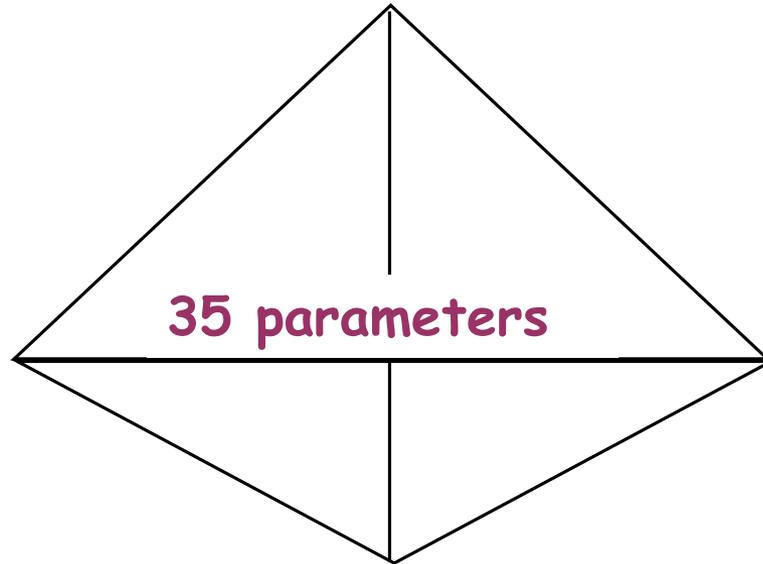


# Human

**Vine conduct** (shoot topping : date and **height**,  
pruning: number and **density of buds**)

## Host

(max leaves size, rate of shoot and leaves development, **leaves susceptibility**, structure and rate of development of secondary shoots, **vigor**)



## Pathogen

**Inoculation**  
(date, localization)

**Isolates**  
(infection, colony development, sporulation...)

## Climate

**Temperature**  
(development of host and pathogen)

**Wind**  
(spores release)



## Sensitivity analysis

Identify the plant-pathogen parameters that have the highest effect on the output (number and surface of diseased leaves)

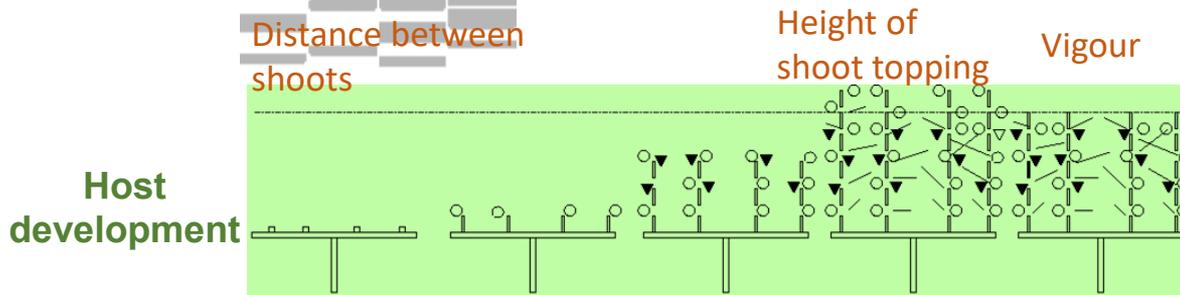
Can we define optimal conditions of plant management depending on the disease pressure or initiation or climatic conditions?

What will be the effect on disease development with a variety showing different pattern of ontogenic resistance?

Can we decrease the disease level by delaying the shoot topping or by increasing the distance between buds?

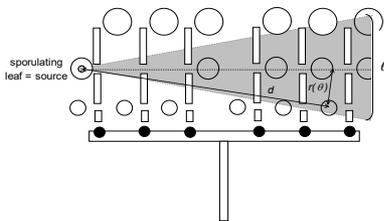
Are these effects dependant on initial and climatic conditions?

# Sensitivity analysis



4 Parameters from plant growth or plant management

## Dispersal process



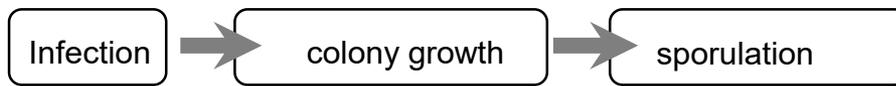
Dispersion

$$Q_c = Q_r \cdot S \cdot e^{-P_6 \cdot d}$$

2 Parameters from pathogen

Early inoculation (stage 1L) or Late inoculation (3L)

## Pathogen development



Leaves susceptibility/ontogenic resistance

$$I = I_{max} F_g(T) e^{P_4 Age F}$$

Sporulation

$$Q_r = A e^{P_5 Diam}$$

Five climatic scenarii

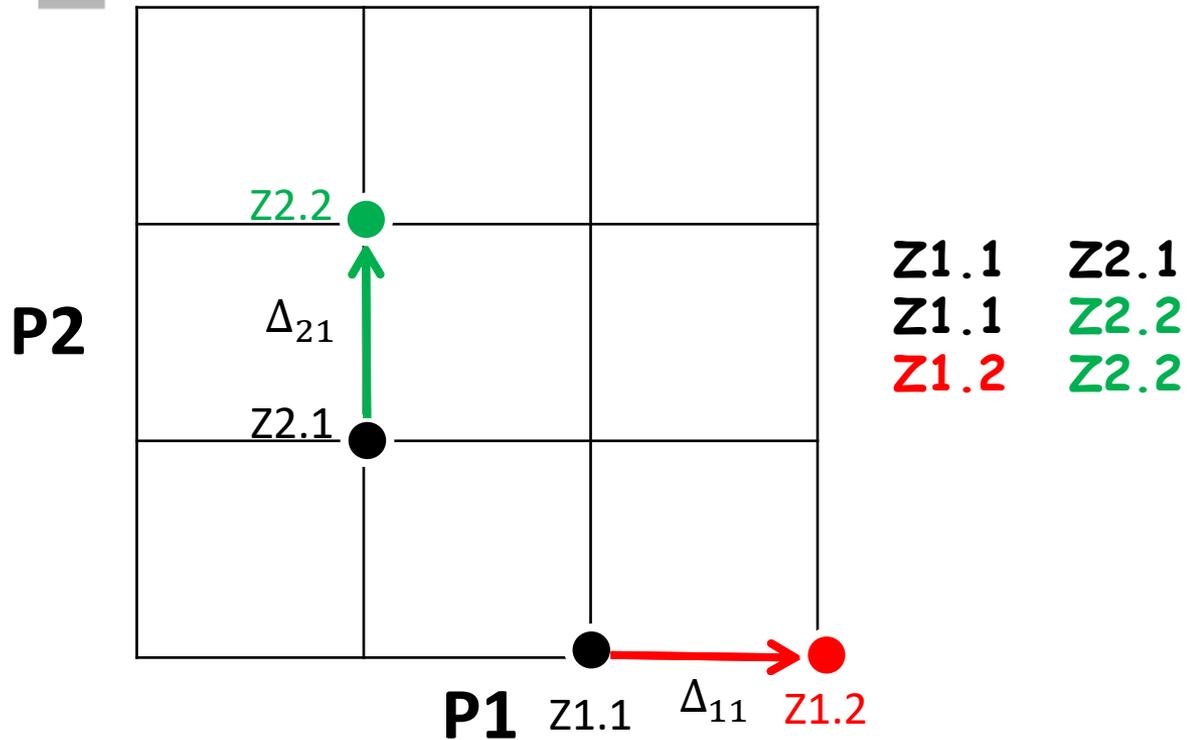
## Method : Morris 'one at a time'

- Define a design by combining  $k$  values (10) of the  $p$  parameters (6)
- Add a « jump »  $\Delta_{ij}$  to the  $i^{\text{th}}$  parameter



# Method : Morris 'one at a time'

Example of path ( $p=2$  parameters,  $k=2$  levels,  $r=1$  path)



## Method : Morris 'one at a time'

Example of path ( $p=6$  parameters,  $k=2$  levels,  $r=1$  path= 7 simulations)



## Method : Morris 'one at a time'

- Define a design by combining  $k$  values (10) of the  $p$  parameters (6)
- Add a « jump »  $\Delta_{ij}$  to the  $i^{\text{th}}$  parameter 
- Compute an « elementary effect »

$$d_{ij} = \frac{[y(z_1, \dots, z_i + \Delta_{ij}, \dots, z_p) - y(z_1, \dots, z_i, \dots, z_p)]}{\Delta_{ij}}$$

- Repeat the procedure for all parameters ( $i=1, \dots, p$ )
- Replicate  $r$  times ( $j=1, \dots, r$ ) (1000)
- Compute **absolute mean** and **variance of elementary effects** from  $r$  replicates

$$\mu^*_j = \sum_{i=1}^r |d_{ij}| / r \quad \sigma_i = \sqrt{\sum_{j=1}^r (d_{ij} - \mu_i)^2 / r}$$

7 000 simulations for 1 condition

70 000 simulations for 10 conditions (2 dates of contaminations, 5 climatic scenario)

# 'Number of diseased leaves': 1998, early inoculation

flowering

Height of shoot  
topping



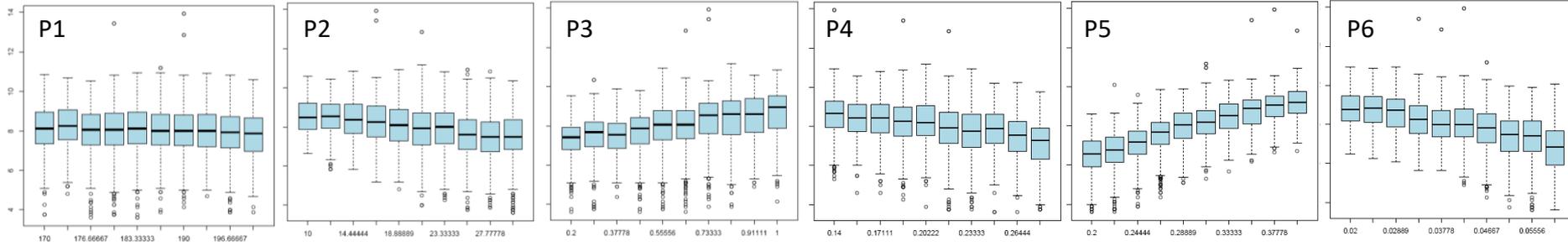
distance between  
shoots

Vigour

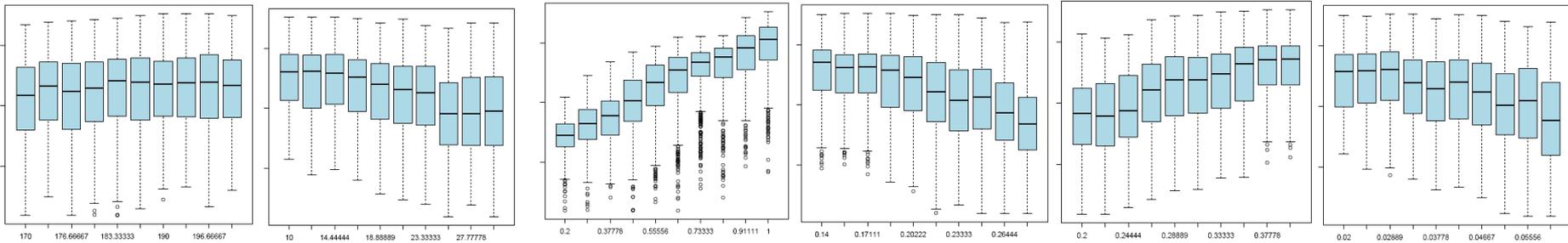
Decrease of  
ontogenic  
resistance

Sporulation  
level

Dispersion



20 days after shoot topping



# 'Number of diseased leaves': 1998, early inoculation

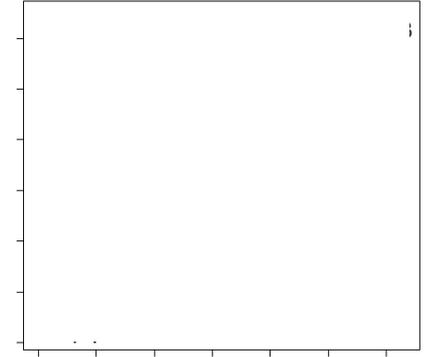
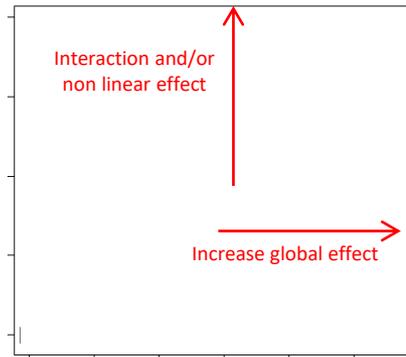
Flowering = d159

Shoot topping = d170

Shoot topping + 20 days

Day200

sigma = s.d. elementary effects



Negligible effect

mu\* = mean absolute effect of each parameter

# 'Number of diseased leaves': 1998, early inoculation

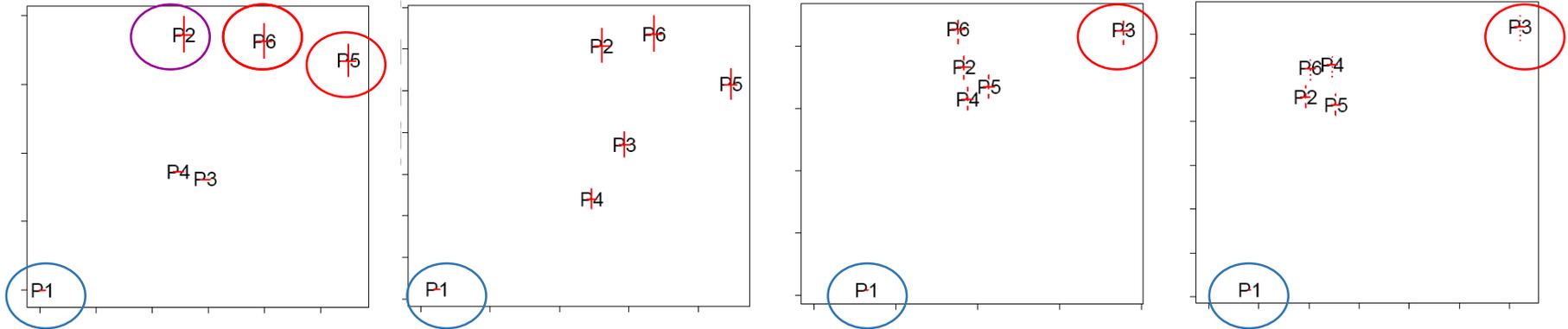
sigma = s.d. elementary effects

Flowering = d159

Shoot topping = d170

Shoot topping + 20 days

Day200



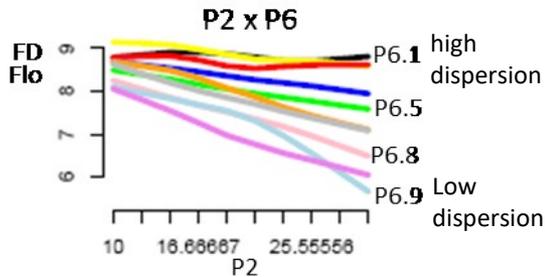
mu\* = mean absolute effect of each parameter

Early in the epidemic, P5 (sporulation) and P6 (dispersion) have strong effects

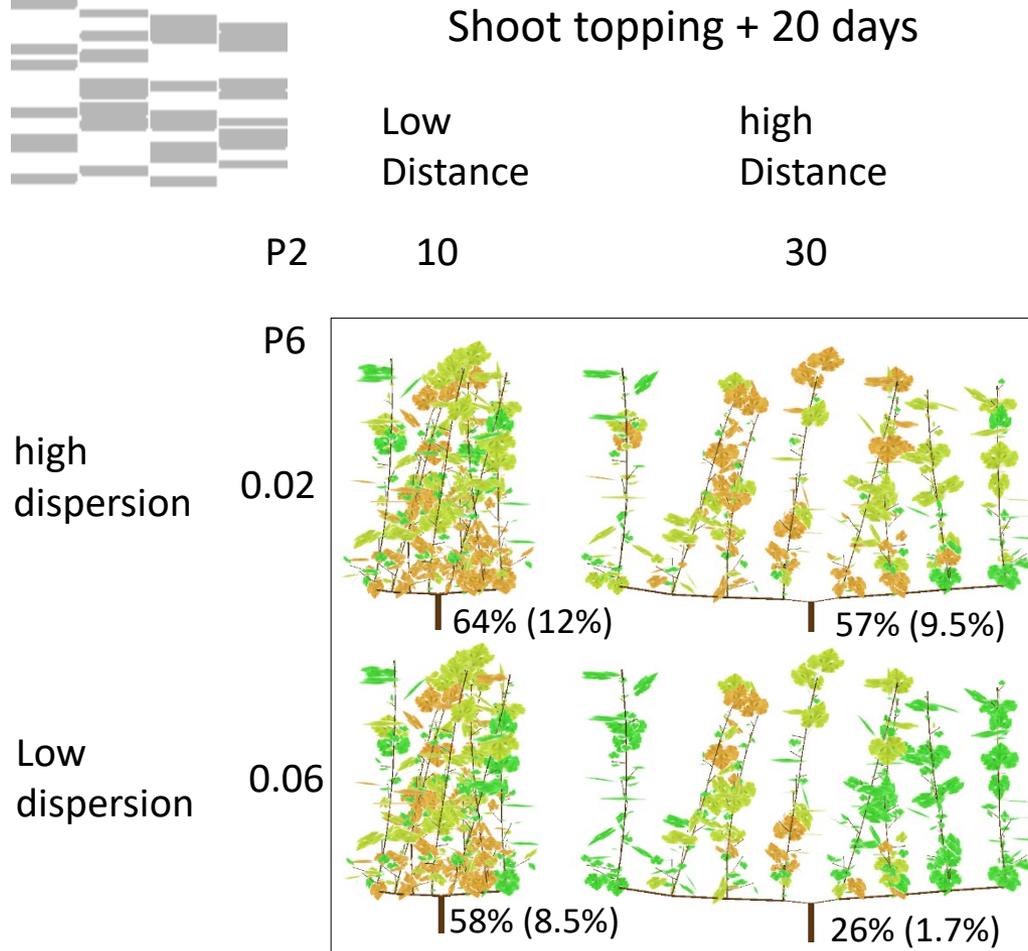
During plant growth, after shoot topping, P3 (vigour) become the parameter with the highest sensitivity

P1 (height of shoot topping) always has a negligible effect

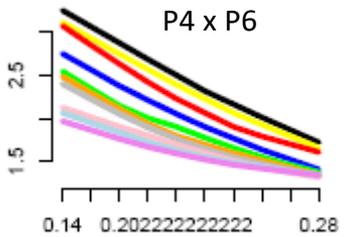
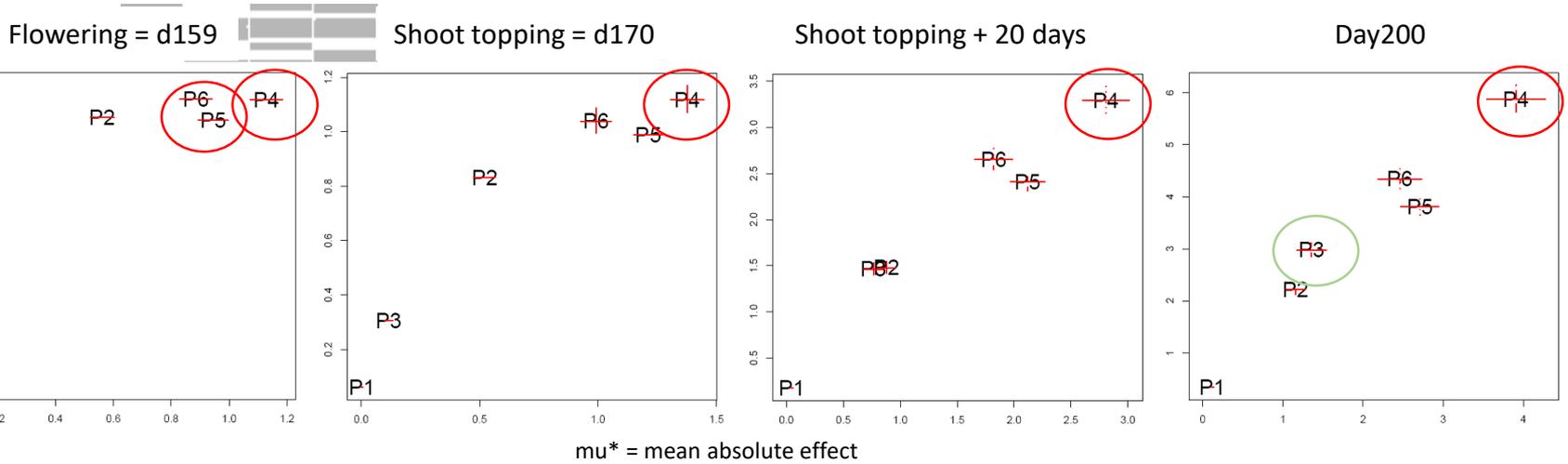
P2 (distance between shoot) is in interaction with P6 (dispersion)



# Interactions between parameters distance between shoots (P2) and dispersion (P6)

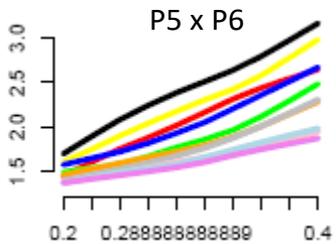


# 'Number of diseased leaves': 1998, late inoculation



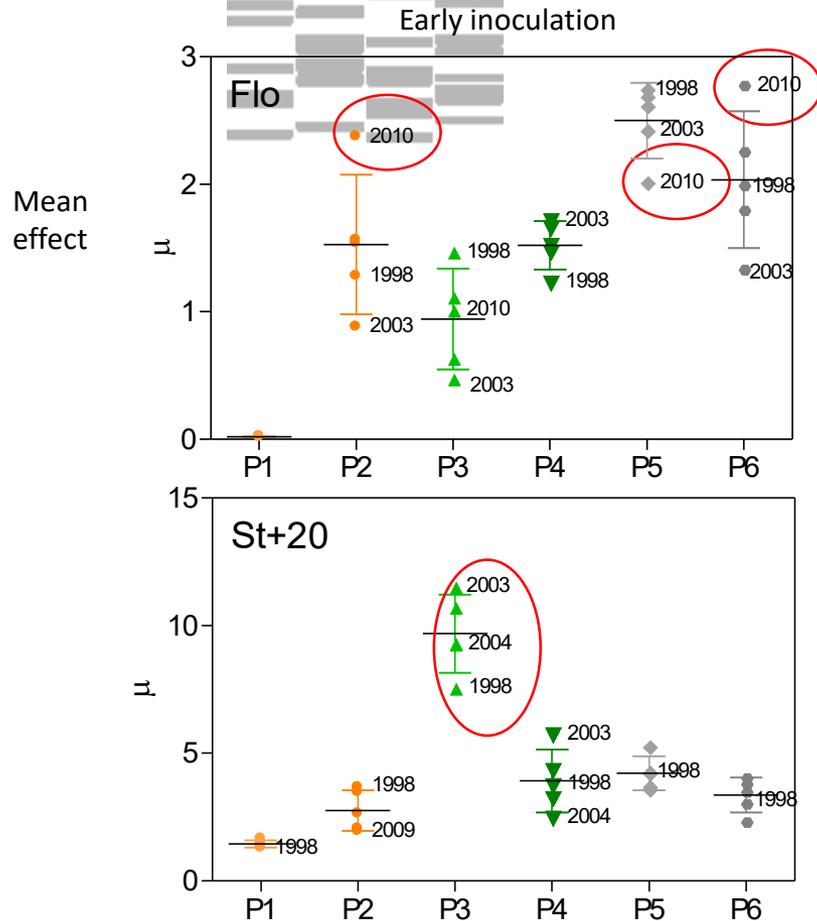
P4 (ontogenic resistance), P5, P6 become the most sensitive parameters

Vigour is a secondary except at the end of the growing season



There is a high level of interactions between parameters

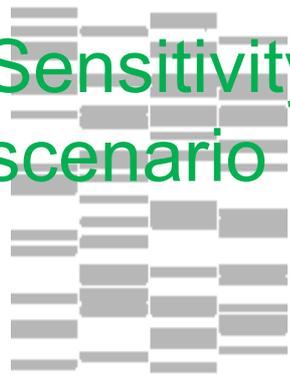
# Sensitivity of parameters function of climatic scenario



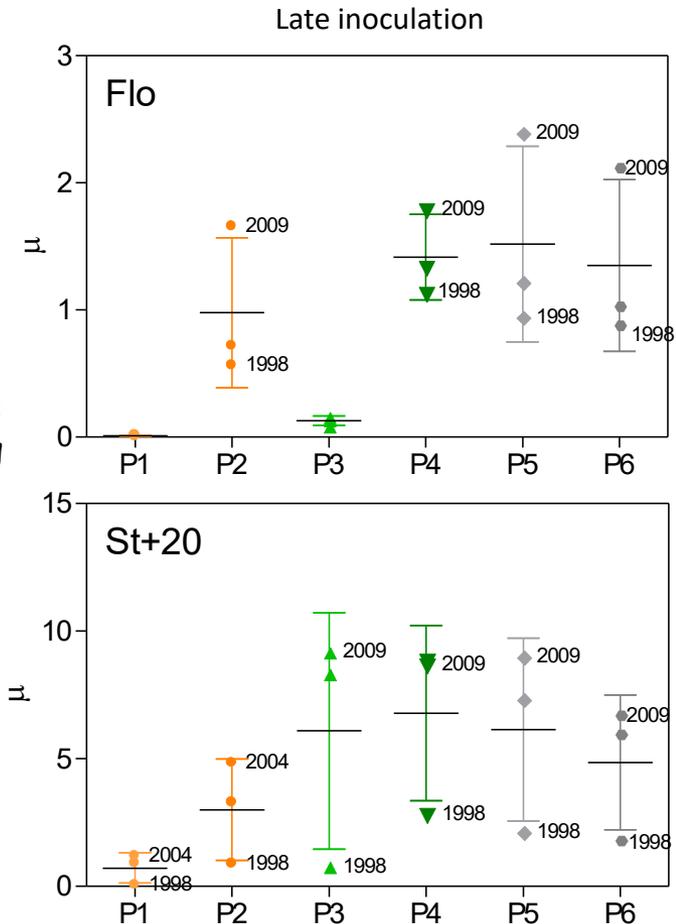
Some scénarii (like 2010) distinct at flowering : distance between shoot and dispersion are more important but sporulation less

Vigour is the most important factor as soon as shoot topping +20 days for all climatic scenario

# Sensitivity of parameters function of climatic scenario



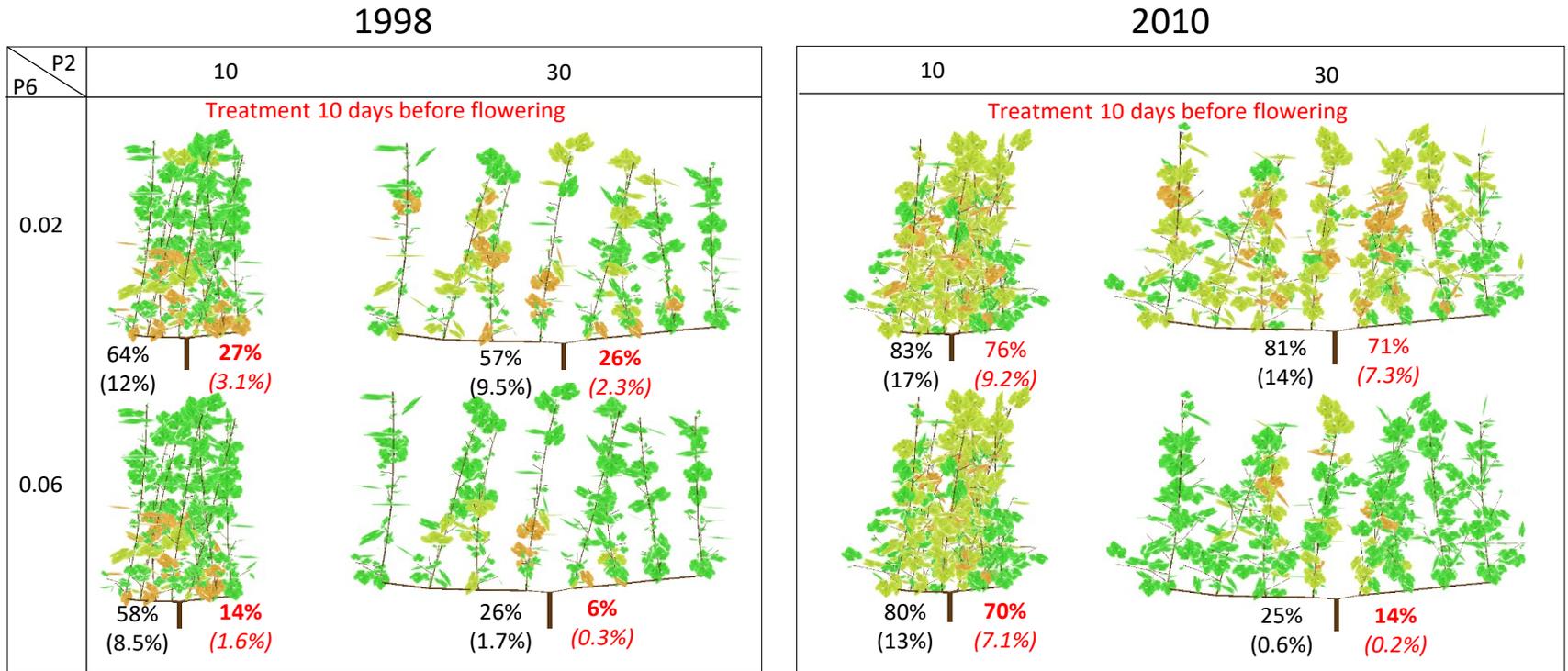
For late inoculation, we observe a lot of variation between the most influent parameters depending on climatic scenario



# Consequences for disease control

The control of the disease is crucial at early time

Depending on our understanding of the dispersion process, factors such as distance between shoots, fungicide treatments will have an effect on the disease or not!



High fungicide effect even at high dispersion  
Increased fungicide effect for low dispersion and high distance between buds

Fungicide effect only if the dispersion is low and the distance between shoot is high

## To conclude

The model is consistent with experimental results (*importance of the initial conditions and the vigour effect for early inoculation*)

The sensitivity analysis reveals that:

1. the initial condition of sporulation and dispersion are of main importance for disease severity
2. The plant vigour is of main importance for increasing disease severity after shoot topping
3. For late inoculation, Ontogenic resistance become the most sensitive parameter, whereas Vigour is a secondary factor
4. The potential effect of the distance between shoots (canopy density) is function of the spore dispersion