

# Mathematical modelling for sustainable crop protection

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**MOVI 2024** – Rennes, 31 May 2024

- Population & food demand are increasing  
*“By 2050, global agricultural production must increase by 70% [...] to meet the demand from a population of 9 billion” [FAO]*
  - Crop pests, diseases and weeds threaten food security  
*20–40% of crop yields destroyed every year*
  - Agriculture is a major sector for employment and revenues in many (developing) countries  
*nearly 80% of working poor live in rural areas [FAO]*
- ➔ **Controlling crop pests is a major issue**
- Chemical pesticides:
    - negative impact on human health & the environment
    - variable effectiveness, induce pest resistance
    - high financial and labour costs
- ➔ **Need for sustainable control methods**

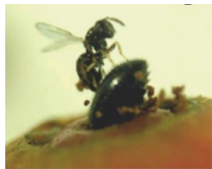


## Alternatives to chemical pesticides

- **Physical methods:** traps, soil solarisation...
- **Cultural practices:** rotation, strip-cropping, destruction of residues, fallow, stump pruning...
- **Biocontrol agent releases:**
  - biopesticides: micro-organisms (bacteria, fungi, viruses), bio-derived chemicals (pheromones...)
  - macro-organisms: predators, parasites
  - sterile insect technique
- **Plant resistance deployment:**
  - qualitative (gene-for-gene / complete)
  - quantitative (polygenic / partial) resistance
  - plant tolerance



©Brocap



Castillo & Infante

*Compartmental models to represent the dynamics (progression over time) of an infectious disease in a population, where individuals are:*

**Susceptible** = healthy, naive

**Exposed** = infected, latent, non infectious

**Infected** = infectious

**Recovered** = immune, resistant / removed

E.g. SEIRS model

$S(t)$

$E(t)$

$I(t)$

$R(t)$

$$\begin{cases} \dot{S} = \\ \dot{E} = \\ \dot{I} = \\ \dot{R} = \end{cases}$$

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$$\left\{ \begin{array}{l} \dot{S} = -\beta I S \\ \dot{E} = \beta I S \\ \dot{I} = \\ \dot{R} = \end{array} \right. \quad \text{force of infection } \beta I$$

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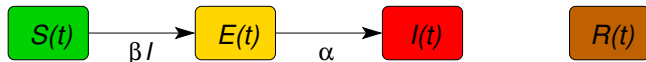
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$$\begin{cases} \dot{S} = -\beta I S \\ \dot{E} = \beta I S - \alpha E & \text{latency period } 1/\alpha \\ \dot{I} = \alpha E \\ \dot{R} = \end{cases}$$

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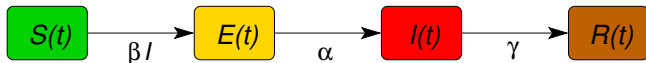
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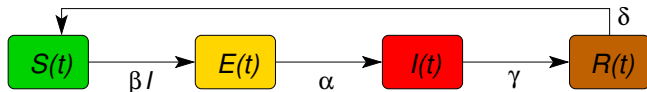
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$$\begin{cases} \dot{S} = -\beta I S + \delta R \\ \dot{E} = \beta I S - \alpha E \\ \dot{I} = \alpha E - \gamma I \\ \dot{R} = \gamma I - \delta R \end{cases} \quad \text{duration of immunity } 1/\delta$$



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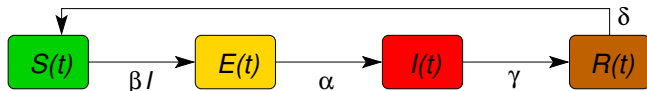
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Constant population:  $P = S(t) + E(t) + I(t) + R(t)$

Equilibria ( $\dot{S} = \dot{E} = \dot{I} = \dot{R} = 0$ ):

- disease-free (DFE):  $S^* = P, E^* = I^* = R^* = 0$
- endemic (with disease) if  $\gamma < \beta P$

# Basic reproduction number $\mathcal{R}_0$

Number of secondary cases generated by an average index case during its entire infectious period, when introduced in a fully susceptible population

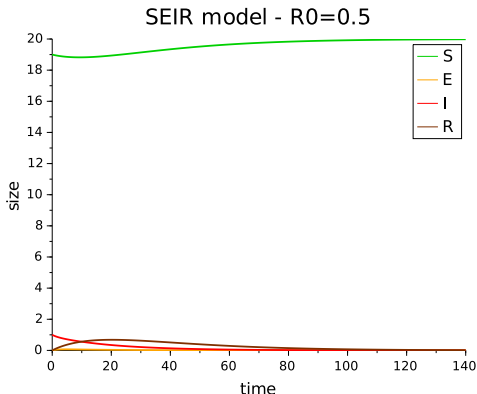
$\mathcal{R}_0$  is a **threshold** (DFE local stability)

- $\mathcal{R}_0 < 1 \rightarrow$  no epidemic, infection cannot settle in
- $\mathcal{R}_0 > 1 \rightarrow$  epidemic

E.g. SEIRS model

$$\mathcal{R}_0 = \frac{\beta P}{\gamma}$$

If  $\mathcal{R}_0 < 1$  : stable DFE



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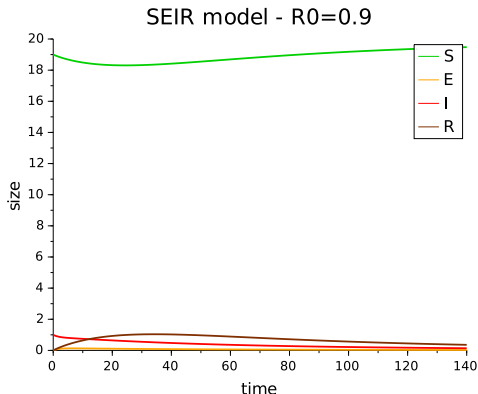
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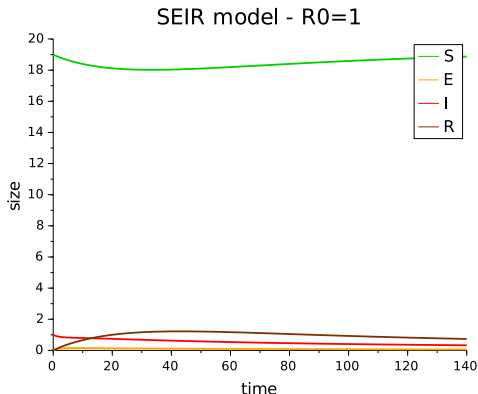
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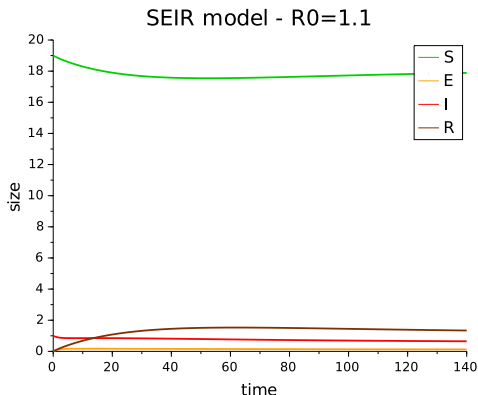
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If  $\mathcal{R}_0 < 1$  : stable DFE

If  $\mathcal{R}_0 > 1$  :

- unstable DFE
- endemic equilibrium



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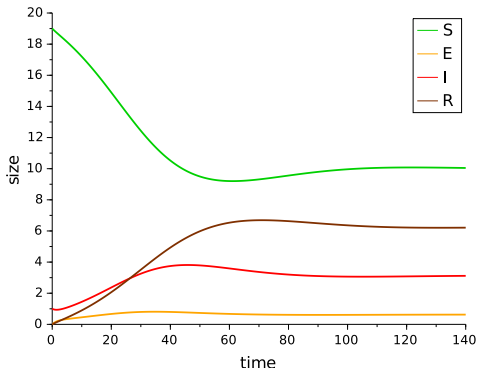
$$\mathcal{R}_0 = \frac{\beta P}{\gamma}$$

If  $\mathcal{R}_0 < 1$  : stable DFE

If  $\mathcal{R}_0 > 1$  :

- unstable DFE
- endemic equilibrium

SEIR model -  $R_0=2$



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E.g. SEIRS model

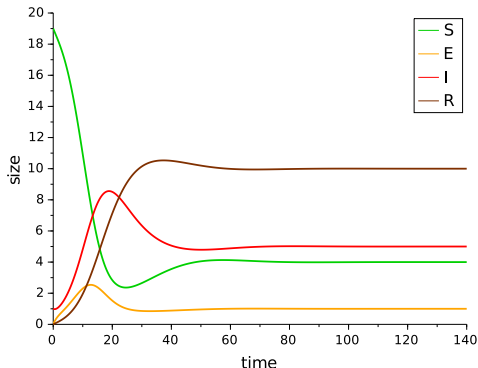
$$\mathcal{R}_0 = \frac{\beta P}{\gamma}$$

If  $\mathcal{R}_0 < 1$  : stable DFE

If  $\mathcal{R}_0 > 1$  :

- unstable DFE
- endemic equilibrium

SEIR model -  $R_0=5$



Epidemiological models for human populations, but also animal and **plant** populations

## Plant & crop specificities

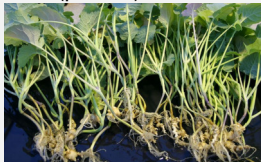
- Definition of a (healthy/infected) **individual**: plant/tree, (part of) leaf, root, fruit...?



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- Plants affected by diseases and **pests**: grazers, phytophagous insects...
- Plants **don't move**: "contacts" via vectors, wind, water, free-living pathogen stages...
- Plants usually **don't recover**, but variable susceptibility
- Crops **managed by humans**: planting, harvest, partial environmental control...
- **Seasonality** plays an important role in annual & perennial crops



Design and analyse **epidemiological models** to:

- better understand plant–parasite interactions
- predict the evolution of damages
- provide **efficient and sustainable control strategies** to limit damages and crop losses

Tools: **optimisation** and **control** theory

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Tools: **optimisation** and **control** theory

Different pathosystems

- o **single** or **multiple** cropping seasons
- o spatial scale:



A. Canaud

plant



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greenhouse



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under cover



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field



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landscape

1 Optimising cultural practices – Banana burrowing nematodes

PhD (2021): **Israël TANKAM CHEDJOU**



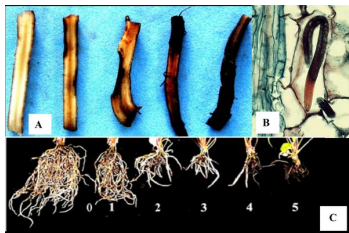
*Inria*

Frédéric GROGNARD, Jean Jules TEWA + Ludovic MAILLERET

2 Optimal biopesticide-based control – Coffee berry borer

3 Self-financing model for cabbage crops with pest management

# Banana burrowing nematodes (*Radopholus similis*)



A: [Jesus, Agron Sustain Dev 2014]; B: M. MacClure, Univ. Arizona; C: [Zhang, EJPP 2012]

- Banana, including plantain: major staple food – *Cameroon: 2% GDP*
- Burrowing nematodes develop, feed and reproduce in roots
- Severe crop losses (up to plant toppling)
- Control
  - chemical nematicides: harmful to environment and human health
  - cropping practices (soil sanitation)
  - biological control: limited options
  - tolerant or resistant banana cultivars

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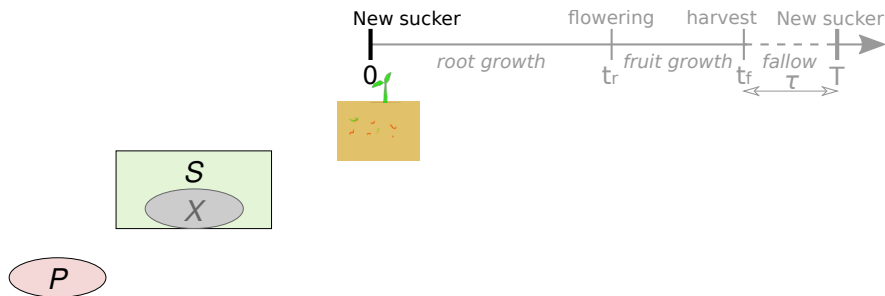


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- Burrowing nematodes develop, feed and reproduce in roots
- Severe crop losses (up to plant toppling)
- Control
  - chemical nematicides: harmful to environment and human health
  - cropping practices (soil sanitation): **fallow**
  - biological control: limited options
  - tolerant or resistant banana cultivars

**How best to implement fallows to limit pest damages and preserve yield?**

# Model: initialisation

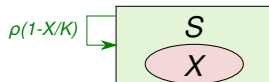
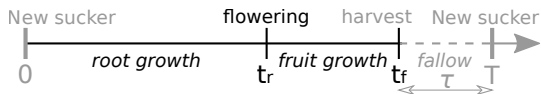


$$\begin{cases} S(0) = S_0 & \text{plant root} \\ X(0) = 0 & \text{nematodes in root} \\ P(0) = P_0 & \text{nematodes in soil} \end{cases}$$

## Hypotheses:

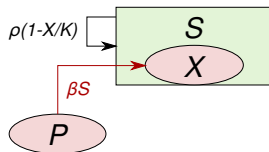
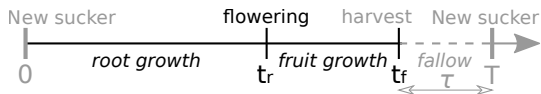
- nursery-grown pest-free sucker (no asexual reproduction by offshoots)
- no male nematodes (not infective & not necessary for reproduction)

# Model: cropping season



$$\left\{ \begin{array}{l} \dot{S} = \overset{\text{root growth}}{\rho(t) S \left(1 - \frac{S}{K}\right)} \\ \dot{X} = \\ \dot{P} = \end{array} \right.$$

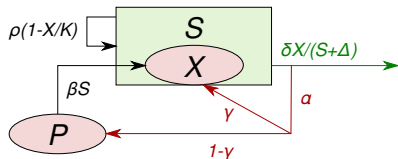
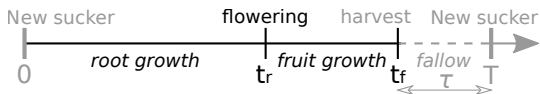
# Model: cropping season



$$\left\{ \begin{array}{l} \text{root growth} \\ \dot{S} = \rho(t) S \left(1 - \frac{S}{K}\right) \\ \dot{X} = \phantom{\rho(t) S \left(1 - \frac{S}{K}\right)} + \beta P S \\ \dot{P} = \phantom{\rho(t) S \left(1 - \frac{S}{K}\right)} - \beta P S \\ \phantom{\dot{P} =} \text{root entering} \end{array} \right.$$



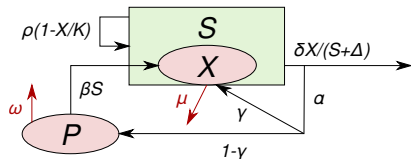
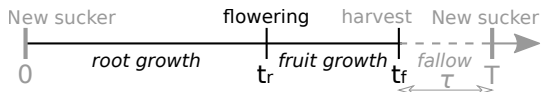
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root growth	-	root consumption
	+	
	+	
	-	
	+	
root entering		feeding & reproduction

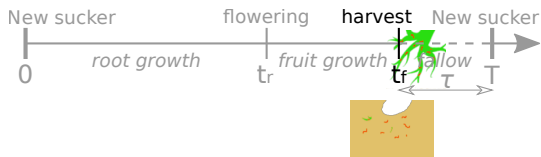
# Model: cropping season



$$\left\{ \begin{array}{l}
 \dot{S} = \rho(t) S \left(1 - \frac{S}{K}\right) - \delta \frac{S X}{S + \Delta} \\
 \dot{X} = + \beta P S + \delta \frac{S X}{S + \Delta} \alpha \gamma - \mu X \\
 \dot{P} = - \beta P S + \delta \frac{S X}{S + \Delta} \alpha (1 - \gamma) - \omega P
 \end{array} \right.$$

root growth
root consumption

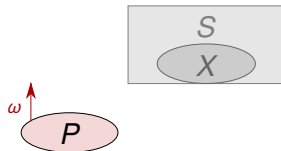
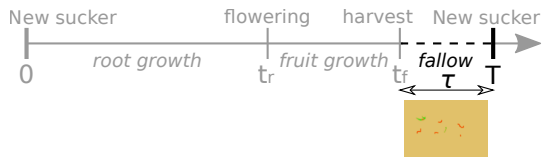
root entering
feeding & reproduction
mortality



$$\begin{cases} S(t_f^+) = 0 \\ X(t_f^+) = 0 \\ P(t_f^+) = P(t_f) + q X(t_f) \end{cases}$$

**Hypothesis:** some infested roots remain in soil at uprooting

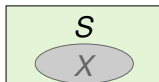
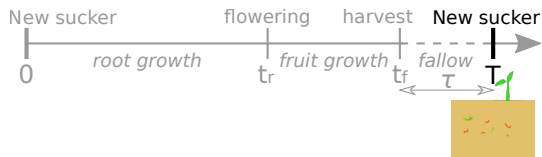
# Model: fallow



$$\begin{cases} \dot{S} = 0 \\ \dot{X} = 0 \\ \dot{P} = -\omega P \end{cases}$$

**Hypothesis:** no alternative hosts for nematodes during fallow

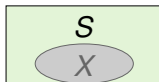
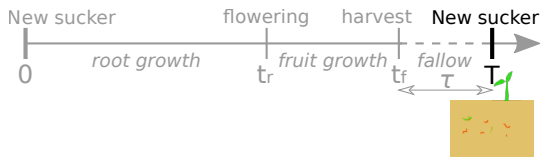
## Model: new sucker



$$\begin{cases} S(T^+) = S_0 \\ X(T^+) = 0 \\ P(T^+) = P(T) = (P(t_f) + q X(t_f)) e^{-\omega \tau} \end{cases}$$

**Hypothesis:** new nursery-grown pest-free sucker

## Model: new sucker

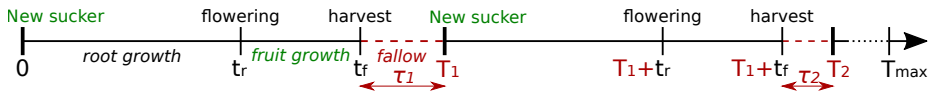


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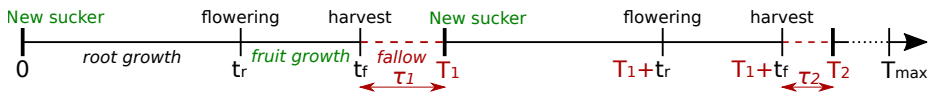
**Etc. for the next seasons**

# Optimal fallow deployment



- Seasonal yield proxy:  $Y_1 = \int_{t_r}^{t_f} w S(t) dt$
- Cost of a pest-free sucker:  $c$
- Seasonal profit:  $R_1 = Y_1 - c$

# Optimal fallow deployment



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## Optimisation problem

Determine the number and duration of fallow periods ( $\tau_i$ ) which **maximise the cumulated profit** on a fixed multi-seasonal time horizon ( $T_{\max}$ ):

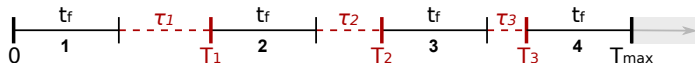
$$\max_{N, \tau_j} \sum_{i=1}^N R_i(\tau_j, j < i)$$

Numerical method: Adaptive Random Search algorithm



# Optimal fallow deployment

Admissible fallows ( $\tau_i$ ) such that last cropping season ends at  $T_{max}$ , e.g.:



Maximum number of fallows:  $n_{max} = \left\lfloor \frac{T_{max}}{t_f} \right\rfloor - 1$

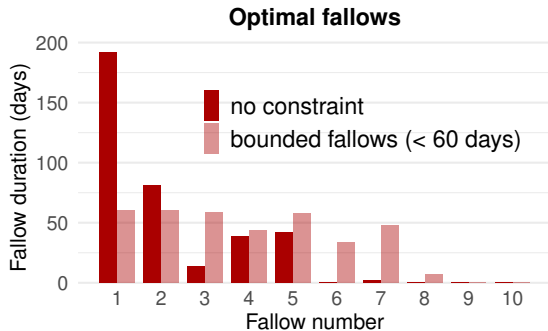
1. For  $n = 1, \dots, n_{max}$

optimisation over  $n$ -simplex:

$$\sum_{i=1}^n \tau_i = T_{max} - (n+1)t_f$$

$\rightarrow$  optimal profit  $R^{n*}$

2. Select highest  $R^{n*}$



Example for  $T_{max} = 4000$  days ( $n_{max} = 11$ )

## How best to implement fallows to limit pest damages and preserve yield?

Fallows can limit nematode infestation and maintain profit

- especially with **long fallows early on to sanitise the soil**
- but expensive pest-free suckers → follow-up with fallows and natural reproduction

I. Tankam Chedjou et al., 2021. *Applied Mathematics and Computation* 397:125883.  
doi: 10.1016/j.amc.2020.125883

I. Tankam Chedjou et al., 2021. *Journal of Interdisciplinary Methodologies and Issues in Science* 8 – Digital Agriculture in Africa. doi: 10.18713/JIMIS-120221-8-4



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### In terms of behavioural epidemiology?

- basic economic criterion
- long term optimisation of cultural practices, but open-loop

*One step further: infestation feedback to represent the grower's decision (fallow or not fallow, etc.) each season?*

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2 Optimal biopesticide-based control – Coffee berry borer

PhD (2022): **Yves FOTSO FOTSO**



*Inria*

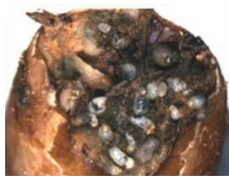
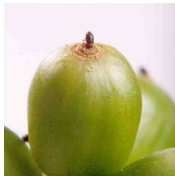
Samuel BOWONG, Frédéric GROGNARD, Berge TSANOU

3 Self-financing model for cabbage crops with pest management

# Coffee berry borers (*Hypothenemus hampei*)



Uccao Cameroun



[Burbano, JIS 2011]

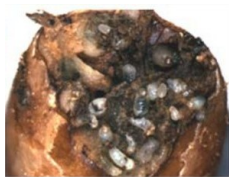
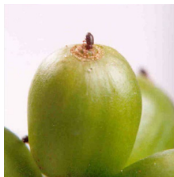
- Coffee: cash crop for tropical developing countries – 25 million households [FAO]
- CBB: mostly develop and feed in coffee berries
- In all production countries, economic losses > 500 million \$/year
- Control
  - chemical insecticides: poorly efficient (cryptic pest)
  - trapping
  - cropping practices: strip-picking, stump pruning
  - biological control: parasitoid or predator insects, entomopathogenic fungi



# Coffee berry borers (*Hypothenemus hampei*)

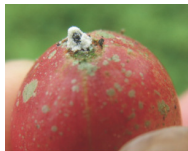


Uccao Cameroun



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A. Ramirez

**How best to apply a biopesticide to control CBB during a growing season?**

# Single season model



$\dot{s} =$

healthy berries

$\dot{i} =$

infested berries

$\dot{y} =$

colonising ♀ ( $\sigma$  not limiting)

$\dot{z} =$

infesting ♀

# Single season model



$$\dot{s} = \text{new berries } \Lambda$$

healthy berries

$$\dot{i} =$$

infested berries

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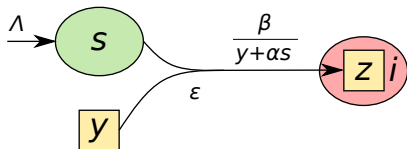
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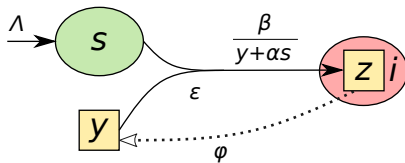


# Single season model



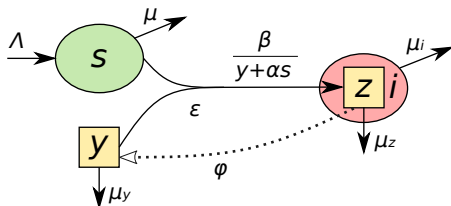
$\dot{s} =$	$\Lambda$	$-$	$\beta \frac{sy}{y + \alpha s}$	healthy berries
$\dot{i} =$		$+$	$\beta \frac{sy}{y + \alpha s}$	infested berries
$\dot{y} =$		$-$	$\epsilon \beta \frac{sy}{y + \alpha s}$	colonising ♀ ( $\sigma$ not limiting)
$\dot{z} =$		$+$	$\epsilon \beta \frac{sy}{y + \alpha s}$	infesting ♀

# Single season model



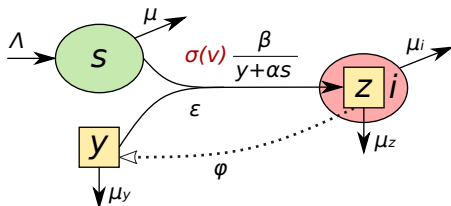
$\dot{s}$ =	new berries	$\Lambda$	-	$\beta \frac{sy}{y + \alpha s}$	infestation	healthy berries
$\dot{i}$ =			+	$\beta \frac{sy}{y + \alpha s}$		infested berries
$\dot{y}$ =	emergence	$\varphi z$	-	$\varepsilon \beta \frac{sy}{y + \alpha s}$		colonising ♀ ( $\sigma$ not limiting)
$\dot{z}$ =			+	$\varepsilon \beta \frac{sy}{y + \alpha s}$		infesting ♀

# Single season model



$\dot{s}$	=	$\Lambda$	-	$\beta \frac{sy}{y + \alpha s}$	-	$\mu s$	healthy berries
$\dot{i}$	=		+	$\beta \frac{sy}{y + \alpha s}$	-	$\mu_i i$	infested berries
$\dot{y}$	=	$\epsilon s$	-	$\epsilon \beta \frac{sy}{y + \alpha s}$	-	$\mu_y y$	colonising ♀ ( $\sigma$ not limiting)
$\dot{z}$	=		+	$\epsilon \beta \frac{sy}{y + \alpha s}$	-	$\mu_z z$	infesting ♀

# Single season model



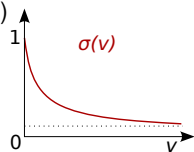
$$\dot{s} = \begin{array}{l} \text{new berries} \\ \Lambda \end{array} - \begin{array}{l} \text{infestation} \\ \sigma(v) \beta \frac{sy}{y + \alpha s} \end{array} - \begin{array}{l} \text{mortality} \\ \mu s \end{array} \quad \text{healthy berries}$$

$$\dot{i} = \begin{array}{l} \text{infestation} \\ + \sigma(v) \beta \frac{sy}{y + \alpha s} \end{array} - \begin{array}{l} \text{mortality} \\ \mu_i i \end{array} \quad \text{infested berries}$$

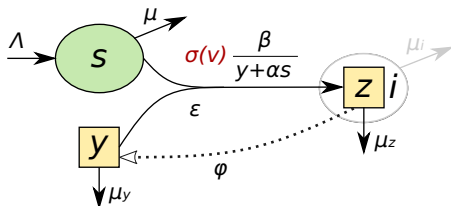
$$\dot{y} = \begin{array}{l} \text{emergence} \\ \varphi z \end{array} - \begin{array}{l} \text{infestation} \\ \sigma(v) \varepsilon \beta \frac{sy}{y + \alpha s} \end{array} - \begin{array}{l} \text{mortality} \\ \mu_y y \end{array} \quad \text{colonising } \varphi \text{ (}\sigma \text{ not limiting)}$$

$$\dot{z} = \begin{array}{l} \text{emergence} \\ + \sigma(v) \varepsilon \beta \frac{sy}{y + \alpha s} \end{array} - \begin{array}{l} \text{mortality} \\ \mu_z z \end{array} \quad \text{infesting } \varphi$$

$$\dot{v} = \begin{array}{l} \text{control} \\ h(t) \end{array} - \begin{array}{l} \text{decay} \\ \gamma v \end{array} \quad \text{fungus load}$$



# Single season model



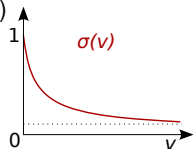
$$\dot{s} = \begin{array}{l} \text{new berries} \\ \Lambda \end{array} - \begin{array}{l} \text{infestation} \\ \sigma(v) \beta \frac{sy}{y + \alpha s} \end{array} - \begin{array}{l} \text{mortality} \\ \mu s \end{array} \quad \text{healthy berries}$$

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## Optimal control problem

Determine the entomopathogenic fungus application  $h(t)$

**maximising the yield** at the end of the cropping season  $s(t_f)$ ,

:

$$\mathcal{J}(h) = \zeta_s s(t_f)$$

yield

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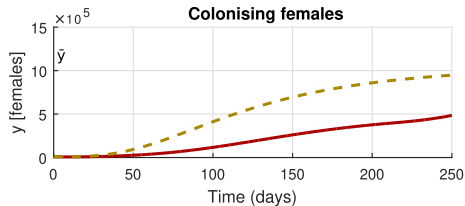
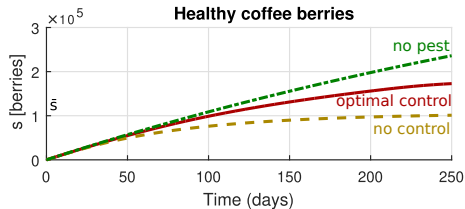
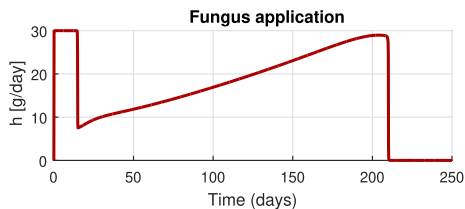
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Pontryagin's Maximum Principle: **bang-singular-bang** solution

Numerical method: **BOCOP**



➔ **Efficient biopesticide control:**

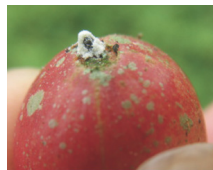
- CBB population halved
- penalised profit  $\mathcal{J}$  doubled

## How to best apply a biopesticide to control CBB during a growing season?

- Optimal control gives a rough idea of how to apply pest control: **start high**
- Extension with 2 controls: biopesticide + traps

Y. Fotso Fotso et al., 2021. *Mathematical Methods in the Applied Sciences* 44(18):14569–14592.  
doi: 10.1002/mma.7726

Y. Fotso Fotso et al., 2023. *Journal of Optimization Theory and Applications* 196(3):882–899.  
doi: 10.1007/s10957-022-02151-7



A. Ramírez

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A. Ramirez

### In terms of behavioural epidemiology?

- basic economic criterion
- optimal control(s), but open-loop and short term

#### *Some steps further:*

- *feedback for grower's decision*
- *information on local/regional prevalence*
- *risk perception to determine between-season controls (strip-picking, etc.)*

- 1 Optimising cultural practices – Banana burrowing nematodes
- 2 Optimal biopesticide-based control – Coffee berry borer
- 3 Self-financing model for cabbage crops with pest management

Ongoing PhD: **Aurelien KAMBEU YOUNBI**



**INRAO**

Frédéric GROGNARD, Berge TSANOU

# Diamondback moth (*Plutella xylostella*)



Andrew Weeks



Alton N. Sparks, Jr

- Cabbages (*Brassica oleracea*): important staple food and source of income for smallholder farmers
- DBM: cosmopolitan insect, whose larvae graze mostly on cabbage plants
- Major pest, especially in regions with mild winters
- Control
  - chemical pesticides ⇒ moth resistance – botanical pesticides
  - cultural practices: inter-cropping, rotation...
  - biological control: parasitoid wasps

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**How should a smallholder farmer best use the cabbage crop revenues?**

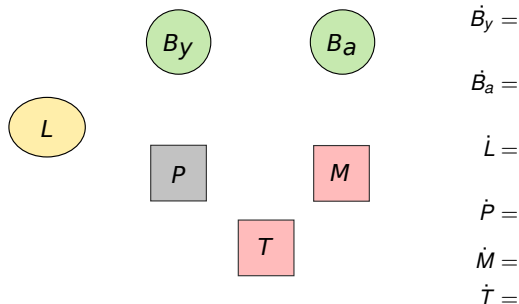
# Self-financing model

**Crop:**  $B_y$  young biomass (susceptible) &  $B_a$  adult biomass (resistant)

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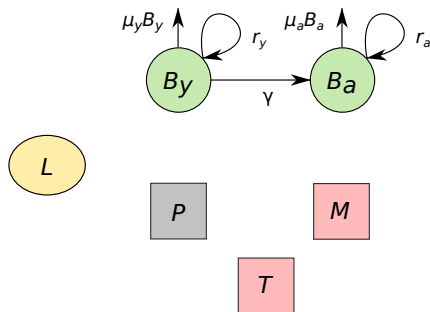
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cabbage growth & ageing

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$$\dot{L} =$$

$$\dot{P} =$$

$$\dot{M} =$$

$$\dot{T} =$$

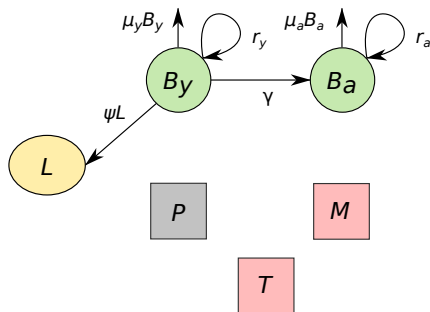
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cabbage growth & ageing

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$$\dot{L} =$$

$$+ c'_B \frac{\text{recruitment } \psi L B_y}{1}$$

$$\dot{P} =$$

$$\dot{M} =$$

$$\dot{T} =$$

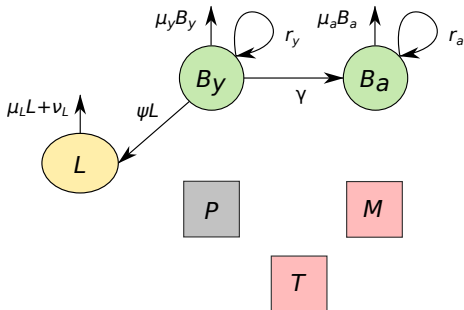
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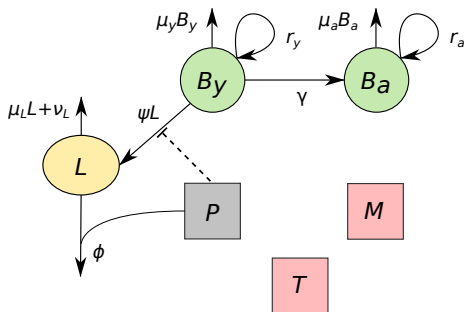
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$$\dot{B}_y = r_y B_y - \mu_y B_y^2 - \gamma B_y - \frac{\text{grazing } \psi L B_y}{bP + 1}$$

cabbage growth & ageing

$$\dot{B}_a = r_a B_a - \mu_a B_a^2 + \gamma B_y$$

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$$\dot{P} = \text{larva uptake } -\phi P L$$

$$\dot{M} =$$

$$\dot{T} =$$

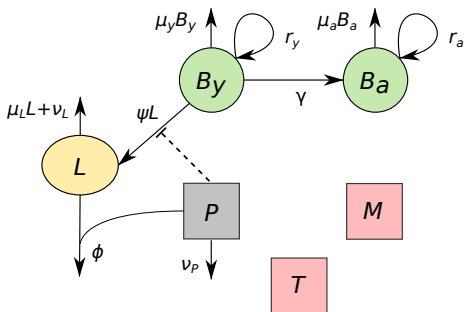
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$$\dot{T} =$$

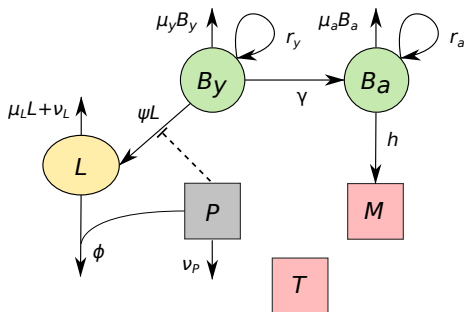
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cabbage growth & ageing

$$\dot{B}_a = r_a B_a - \mu_a B_a^2 + \gamma B_y - \frac{h B_a}{bP + 1}$$

harvest

$$\dot{L} = -\nu_L L - \mu_L L^2 - c_L \phi P L + c'_B \frac{\psi L B_y}{bP + 1}$$

larva mortality

recruitment

$$\dot{P} = -\phi P L - \nu_P P$$

larva uptake

decay

$$\dot{M} = c_B h B_a$$

revenue

$$\dot{T} =$$

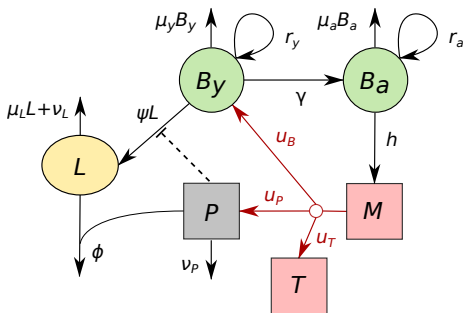
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$$\dot{B}_a = r_a B_a - \mu_a B_a^2 + \gamma B_y - \frac{\text{harvest}}{h} B_a$$

$$\dot{L} = -\nu_L L - \mu_L L^2 - c_L \phi P L + c'_B \frac{\text{recruitment}}{\psi L B_y} \frac{\psi L B_y}{bP + 1}$$

$$\dot{P} = \frac{\text{larva uptake}}{-\phi P L} - \frac{\text{decay}}{\nu_P P} + c'_M u_P$$

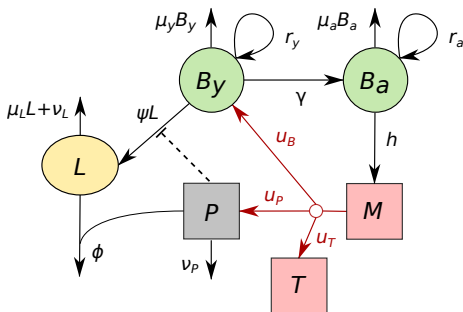
$$\dot{M} = \frac{\text{revenue}}{c_B h B_a} - (u_B + u_P + u_T)$$

$$\dot{T} = u_T$$

Automatic controls:  $u_B$  new seedlings,  $u_P$  protection costs, and  $u_T$  net income

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$$\begin{aligned} \dot{B}_y &= r_y B_y - \mu_y B_y^2 - \gamma B_y - \frac{\text{grazing } \psi L B_y}{bP + 1} + c_M u_B \\ &\text{cabbage growth \& ageing} \\ \dot{B}_a &= r_a B_a - \mu_a B_a^2 + \gamma B_y - \frac{\text{harvest } h B_a}{bP + 1} \\ &\text{larva mortality} \\ \dot{L} &= -\nu_L L - \mu_L L^2 - c_L \phi P L + c'_B \frac{\text{recruitment } \psi L B_y}{bP + 1} \\ &\text{larva uptake} \quad \text{decay} \\ \dot{P} &= -\phi P L - \nu_P P + c'_M u_P \\ &\text{revenue} \\ \dot{M} &= c_B h B_a - (u_B + u_P + u_T) \\ \dot{T} &= u_T \end{aligned}$$

Automatic controls:  $u_B$  new seedlings,  $u_P$  protection costs, and  $u_T$  net income



## Optimisation problem

Determine controls  $u_B$  new seedlings,  $u_P$  protection costs, and  $u_T$  net income to **maximise the total earnings**, i.e. the final  $T$ .



- Static optimisation
- Open-loop or feedback controls
- Discrete controls

- **Modelling crop pests and diseases + ecofriendly control strategies**  
→ **insights in sustainable control deployment**
- Some (necessary) **simplifications**: no abiotic factors, single pest, open-loop control...
- A few **challenges** (still):
  - “small data” → link with remote sensing?
  - growers' decisions → behavioural epidemiology?

# Conclusion

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→ **insights in sustainable control deployment**
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Work ∈ **EPITAG** = EPIdemiological modelling and control for Tropical AGriculture

*French & Cameroonian researchers and students, with a background in applied mathematics, and an interest in crop diseases*

**INRAE** **AGROBIOTECH** **Sophia Antipolis** **cirad** **IRD**

**Inria** **Cameroon** **Univ. Dschang** **Univ. Yaoundé I** **Univ. Douala**

**Joint PhD supervision**

More on EPITAG: <https://team.inria.fr/epitag/>