

# **Impact of fungi on control of bollworms**

## *Chrysodeixis includens and Helicoverpa armigera*

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# Outline

- 1. Introduction**
- 2. Models and methods**
- 3. Results and discussion**
- 4. Final remarks**

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# Introduction

# Entomological motivation

## Species of interest:

- ▶ *Chrysodeixis includens*: the “soybean looper”;
- ▶ *Helicoverpa armigera*: the “cotton bollworm”;
- ▶ They feed on a wide range of plants, including many important cultivated crops (agronomic crops: soybean, cotton, maize, etc. and also vegetable and floricultural crops).

## Biological pest control:

- ▶ Controlling pest population using other organisms;
- ▶ An important biological control agents are pathogenic fungi.

## Research question:

- ▶ The inoculation of fungi in soybean plants may inhibit the development of bollworms *Chrysodeixis includens* and *Helicoverpa armigera*?

# Design of the experiment

## Treatments:

- ▶ Three species of fungi:  
*Metarhizium anisopliae* ESALQ-1638 (Met 1638);  
*Beauveria bassiana* ESALQ-3399 (Bb 3399); and  
*Isaria fumosorosea* ESALQ-3422 (If 3422);
- ▶ Control (Tween 80).

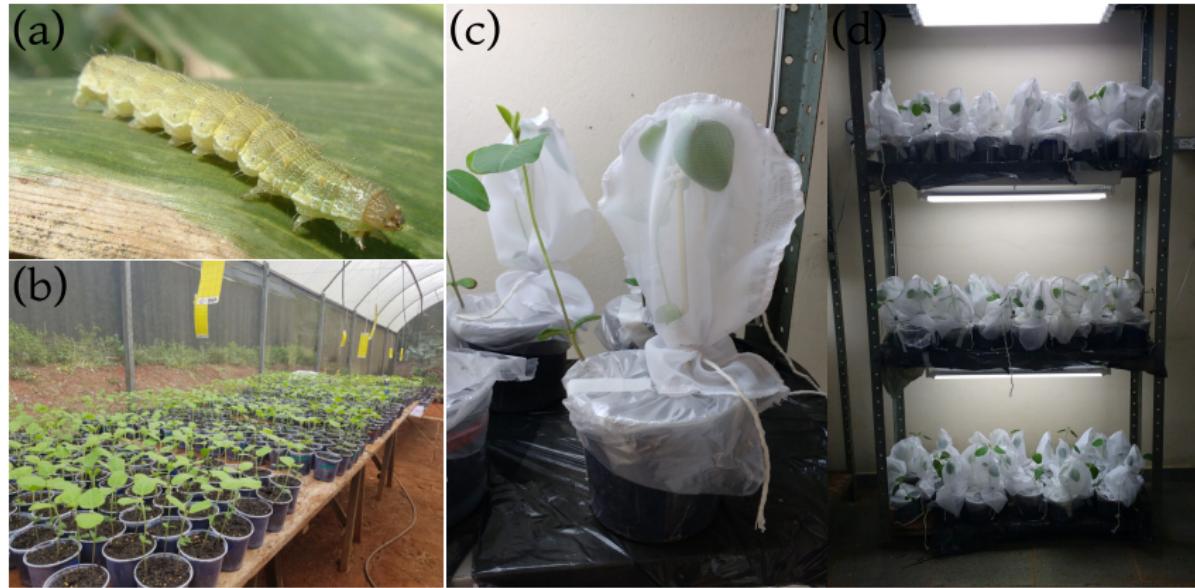
## Experiment with whole plants:

- ▶ The fungi were inoculated on the commercial substrates to cultivate the soybean plants.
- ▶ 30 bollworms (for each treatment) were confined in a pot with a plant, where the substrate was isolated.
- ▶ The plots were evaluated every three days during 18 days for *Chrysodeixis includens* and 21 days for *Helicoverpa armigera*.
- ▶ This design were repeated two times in different periods.

## Outcomes:

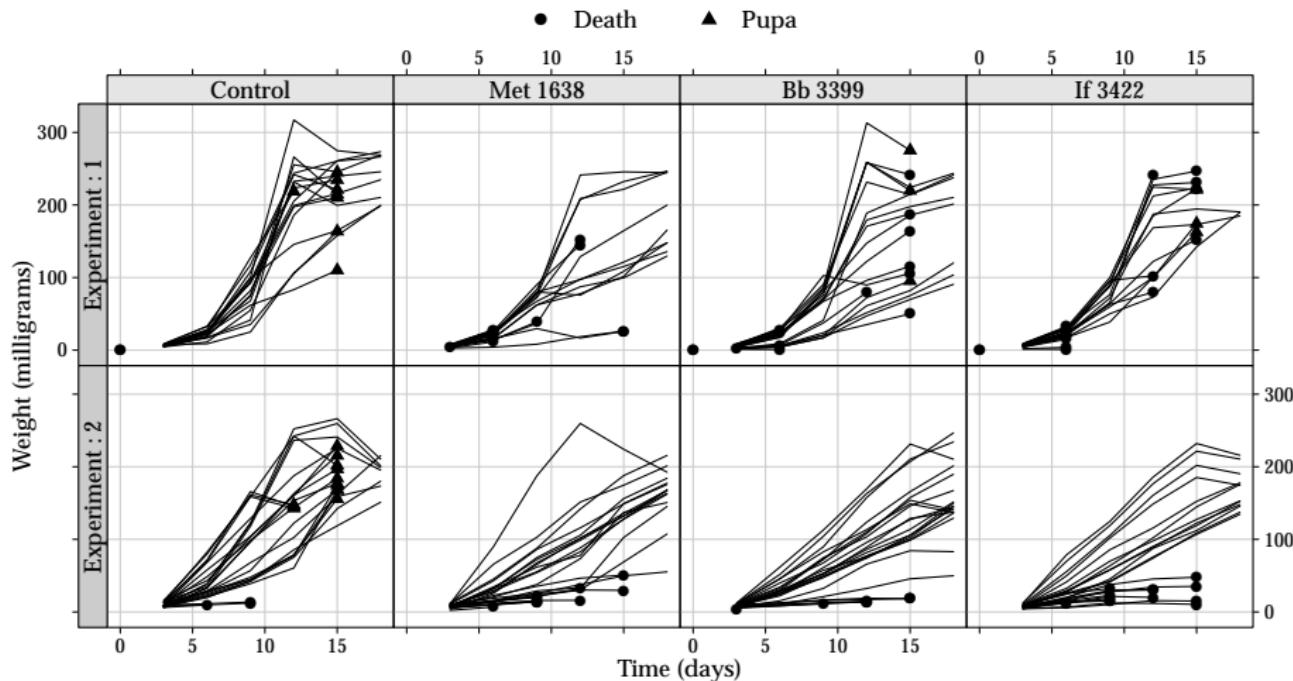
- ▶ Weight of bollworms over time (longitudinal data) and
- ▶ Time to death of the bollworms (time-to-event data).

# Design of the experiment



**Figure:** Pictures of the experiment: (a) *Helicoverpa armigera*, (b) soybean plants, (c)-(b) bollworms confined in the pots.

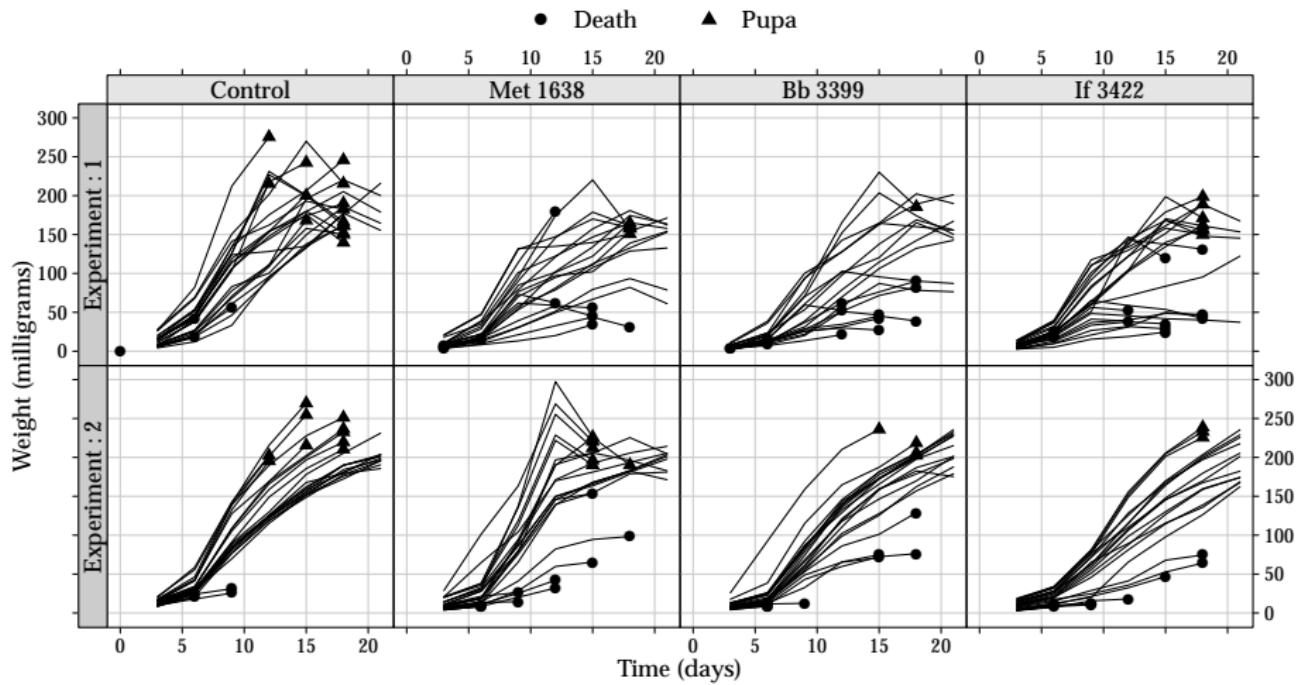
# Descriptive analysis (longitudinal data)



(a) Specie: ***Chrysodeixis includens***

**Figure:** Data on weights of the *Chrysodeixis includens* bollworms over time. The symbols ● and ▲ indicate death and pupa stage, respectively.

# Descriptive analysis (longitudinal data)



(b) Specie: ***Helicoverpa armigera***

**Figure:** Data on weights of the *Helicoverpa armigera* bollworms over time. The symbols • and ▲ indicate death and pupa stage, respectively.

# Descriptive analysis (time-to-event data)

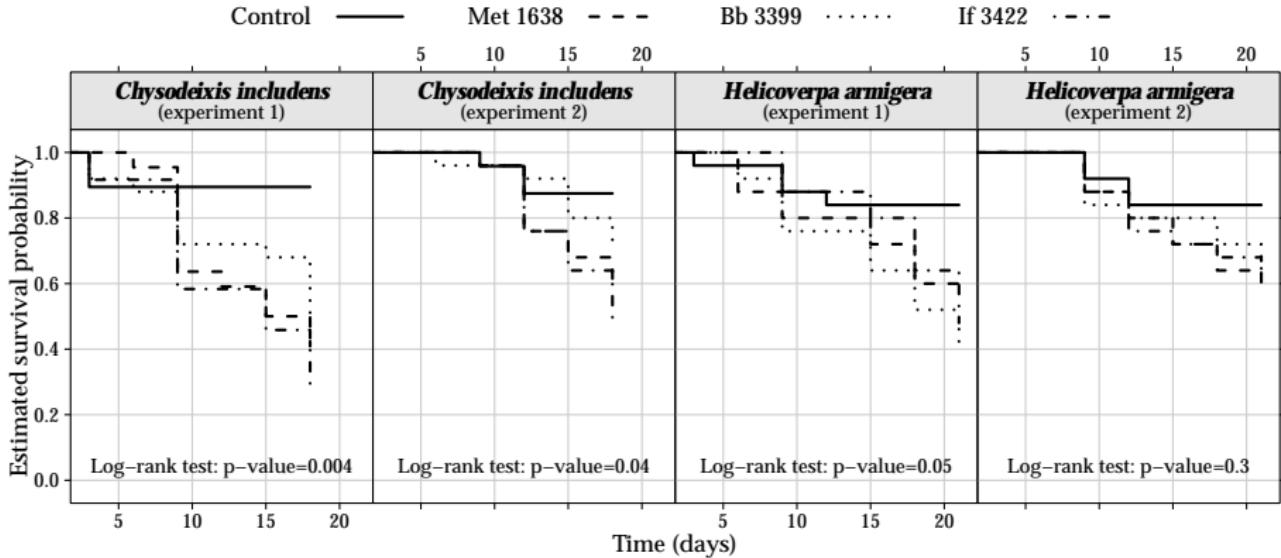


Figure: Kaplan-meier survivor function estimates for the times to death of the bollworms.

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## Models and methods

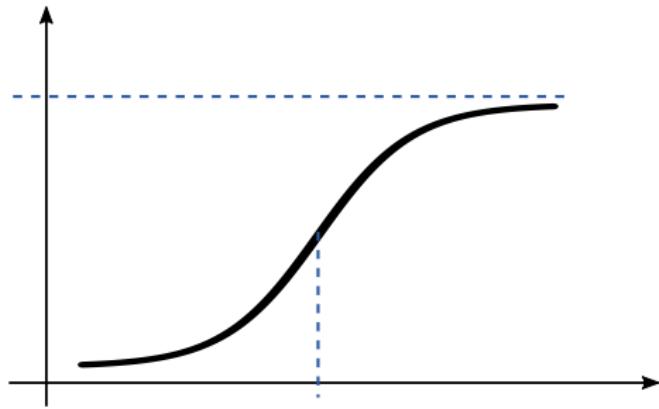
# Non-linear models (logistic growth)

To model the weight growth of bollworms ( $Y$ ) over time ( $t$ ), we consider the logistic growth,

$$E(Y) = f(t) = \frac{\theta_A}{1 + \exp[(\theta_M - t)/\theta_S]},$$

where

- ▶  $\theta_A$  is the horizontal asymptote ( $f(t)$  when  $t \rightarrow \infty$ , if  $\theta_S > 0$ ),
- ▶  $\theta_M$  is the inflection point of the curve and
- ▶  $\theta_S$  is the scale parameter.



# Heteroscedastic non-linear mixed models

## Statistical challenges

- ▶ Model the correlation between measures of the same bollworm and
- ▶ Model the heteroscedastic within-error.

## Fitted model

- ▶ Let  $(y_{ijk}, t_{ijk})$  denote the weight and time (days) of the  $i$ -th bollworm on the  $j$ -th treatment at the  $k$ -th time, the fitted model can be expressed as

$$y_{ijk} = \frac{\theta_{Aj} + b_{Ai}}{1 + \exp[(\theta_{Mj} + b_{Mi} - t_{ijk})/\theta_{Sj}]} + \varepsilon_{ijk},$$

## Variance components

- ▶  $\begin{bmatrix} b_{Ai} \\ b_{Mi} \end{bmatrix} \sim \mathcal{N} \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \Sigma = \begin{bmatrix} \sigma_A^2 & \sigma_{AM} \\ \sigma_{AM} & \sigma_M^2 \end{bmatrix} \right),$
- ▶  $\text{Var}(\varepsilon_{ijk}) = \sigma^2 \delta_{1j}^2 \delta_{2K}^2.$

# Accelerated failure time models

Let  $T_{ij}$  be the time-to-death of the  $i$ -th treatment and  $j$ -th bollworm, the AFT model can be expressed (where  $\omega = \log(\varepsilon)$ ) as

$$T_{ij} = \exp(\mathbf{x}_{ij}^\top \boldsymbol{\beta}) \varepsilon^\sigma$$

$$\log(T_{ij}) = \mathbf{x}_{ij}^\top \boldsymbol{\beta} + \sigma\omega.$$

The distributional assumption of  $T_{ij}$  implies distribution of  $\omega$  (e.g.  $T_{ij} \sim \text{Weibull}(\alpha, \lambda) \implies \omega \sim \text{E.V.}(\lambda, \alpha)$ ).

## Censoring times

- ▶ The time of event endpoint is not observed exactly;
- ▶ Right-censoring: the censored time-to-death will be a time beyond the observed time.

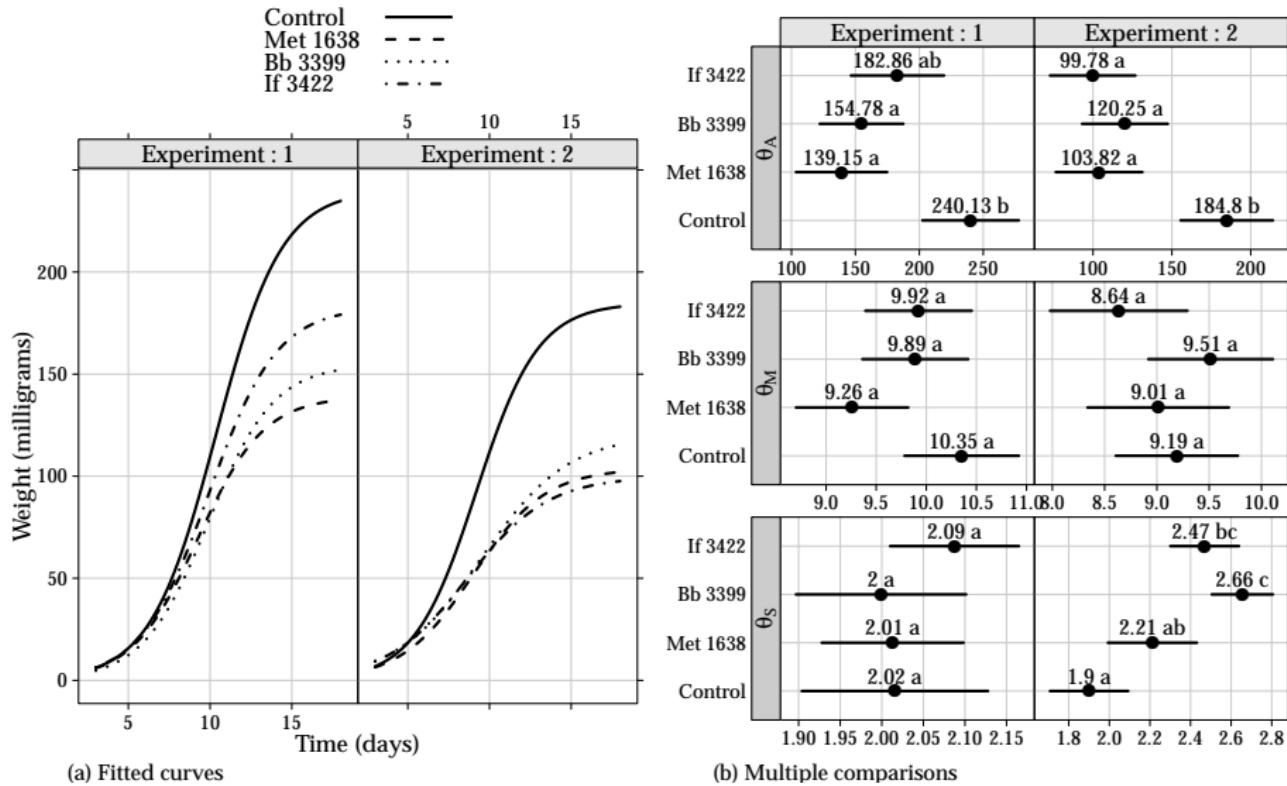
$$\mathcal{L}(\boldsymbol{\theta} \mid \mathbf{t}) = \prod_{i=1}^n [f(t_i; \boldsymbol{\theta})]^{cens_i} [S(t_i; \boldsymbol{\theta})]^{(1 - cens_i)},$$

- ▶  $t_i$  is the  $i$ -th recorded time,  $i = 1, \dots, n$ ,
- ▶  $cens_i = 0$ , if  $t_i$  is a censored time and 1 otherwise.

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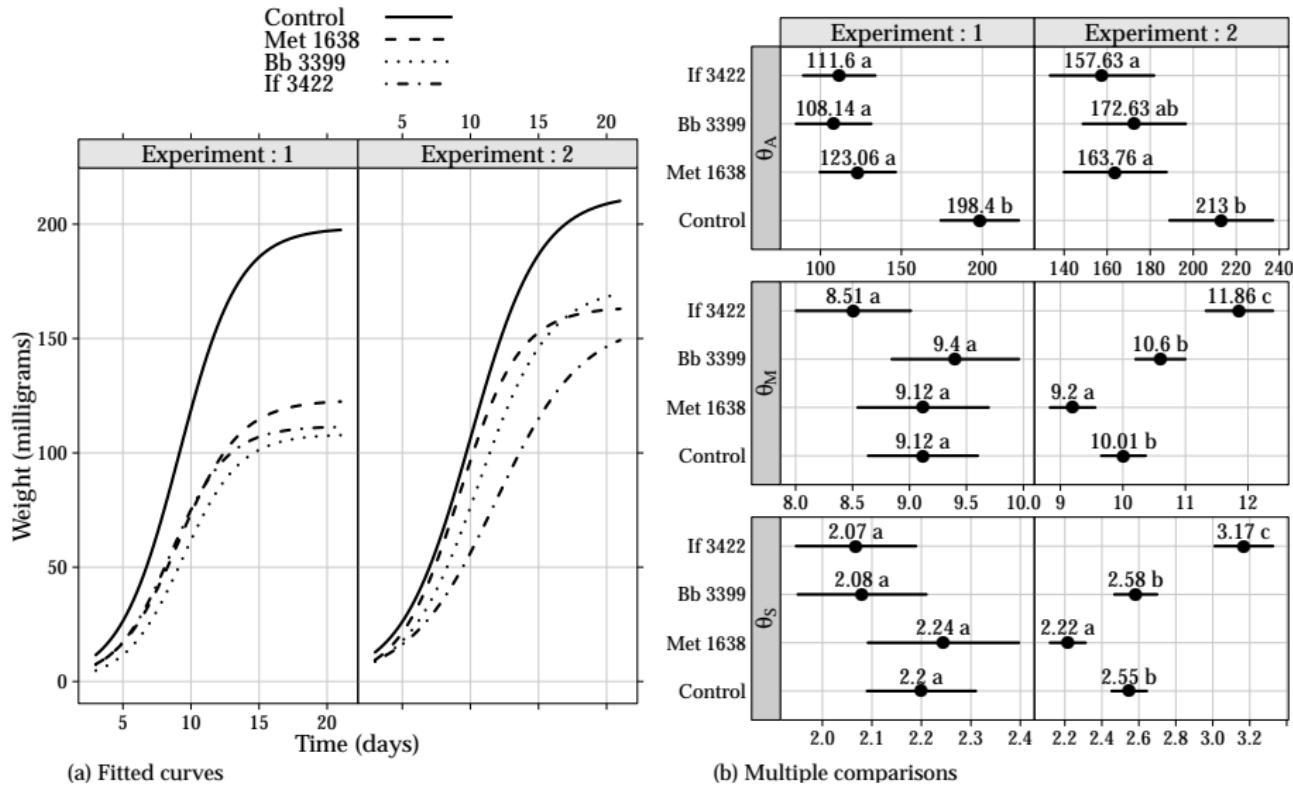
## Results and discussion

# Non-linear models (*Chrysodeixis includens*)



**Figure:** Results for *Helicoverpa armigera*. (a) fitted logistic curves and (b) parameter estimates and multiple comparisons (5% significance level).

# Non-linear models (*Helicoverpa armigera*)



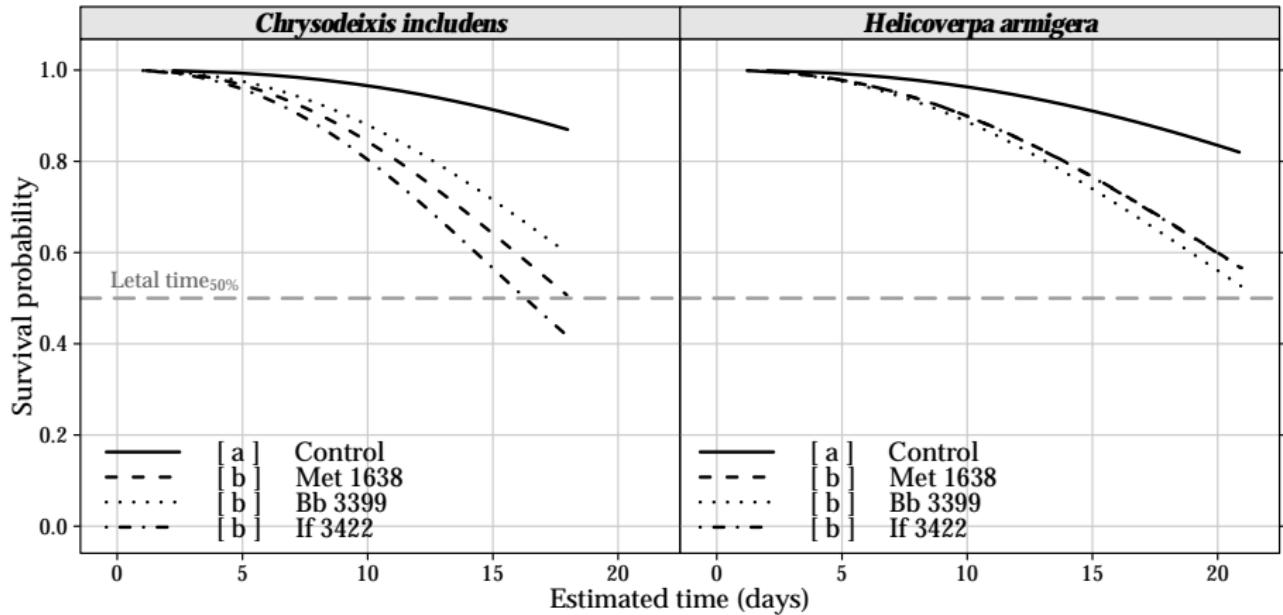
**Figure:** Results for *Helicoverpa armigera*. (a) fitted logistic curves and (b) parameter estimates and multiple comparisons (5% significance level).

# AFT models

**Table:** Analysis of deviance for the models fitted to the time-to-death of the bollworms *Chrysodeixis includens* and *Helicoverpa armigera*.

Specie	Effect	df	Deviance	Diff df	2 logLik	p-value
Chrysodeixis	Null			187	679.9222	
	Treatment	3	23.4077	184	656.5144	0.00003
	Experiment	1	7.7327	183	648.7817	0.00542
	Interaction	3	1.1316	180	647.6501	0.76946
Helicoverpa	Null			198	693.8224	
	Treatment	3	11.3926	195	682.4298	0.00978
	Experiment	1	1.6096	194	680.8202	0.20455
	Interaction	3	1.2429	191	679.5772	0.74272

# AFT models



**Figure:** Survival curves for the estimated times to death.

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## Final remarks

## Data analysis results

- ▶ In this work, we evaluated the effect of fungus inoculation in soybean plants on delayed development of bollworms;
- ▶ The data analysis has shown that inoculation of fungi may delay the bollworm development.
  - The weight gain and time-to-death were lower for bollworms fed with fungus inoculated plants.

## Methodological contributions

- ▶ Use of non-linear models with interpretive parameters and multiple comparison tests for each parameter;
- ▶ Use of parametric models for time-to-event data with comparison of the survival curves by using the likelihood ratio test.

## Future research

- ▶ Joint modelling longitudinal outcomes and time-to-event data (Rizopoulos 2012).

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