

INRAE Occitanie-Montpellier - MISTEA

# Weather generators and phenological models to study climate change impacts on grapevines and apples

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## 1 Stochastic weather generators

- Principle
- Model
- Estimation
- Simulation

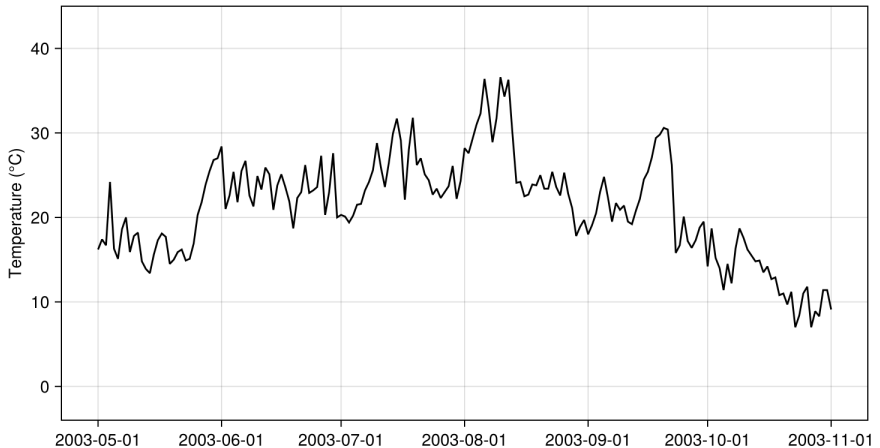
## 2 Phenological models

- Introduction
- Grapevine Model
- Apple Model
- With simulated TG (Apple)

## 3 Conclusion

# I. Stochastic weather generators

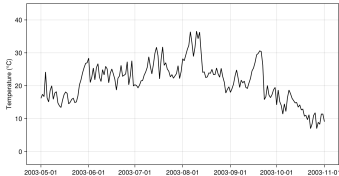
# I. Stochastic weather generators | Principle



Recorded daily maximum temperature (TX) at Lille from 1<sup>st</sup> May 2003 to 1<sup>st</sup> November 2003 (source : ECA&D)

# I. Stochastic weather generators | Principle

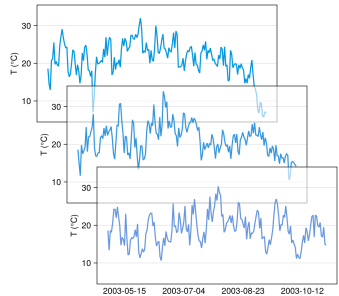
**Weather data**



Stochastic  
Model

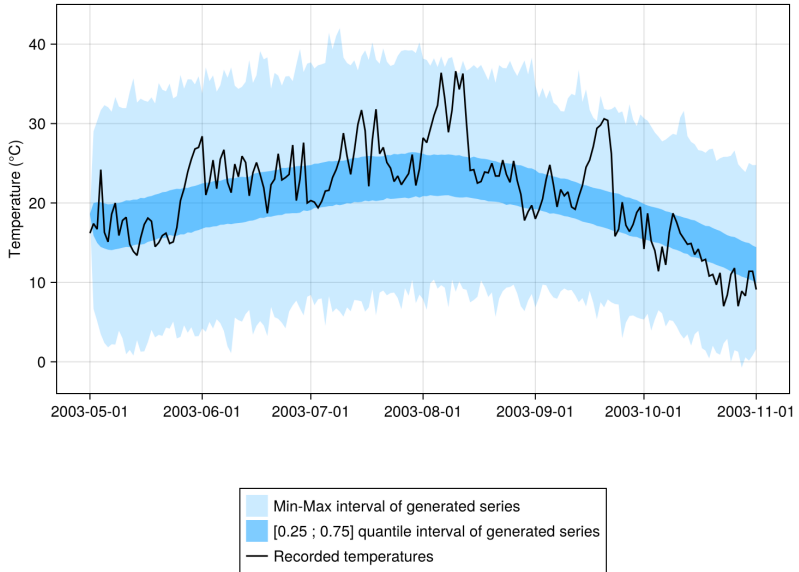


**Simulations**



# I. Stochastic weather generators | Principle

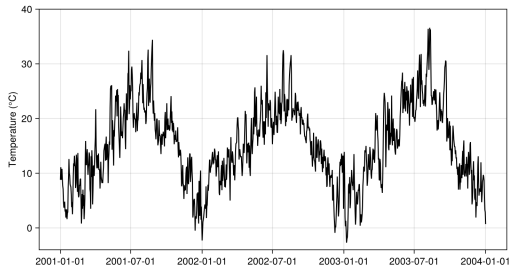
With 5000 simulations :



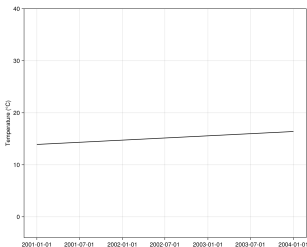
# I. Stochastic weather generators | Model

$Y_t = T_t + S_t + X_t$  where

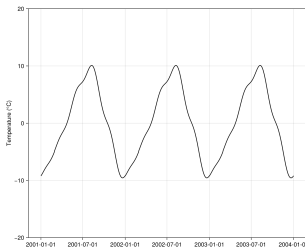
- $T_t$  is the trend
- $S_t$  is the seasonality
- $X_t$  is the stochastic part



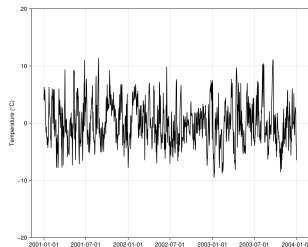
$Y_t$



$T_t$



$S_t$



$X_t$

# I. Stochastic weather generators | Model

- Trend :  $T_t = \mu$  (for now)
- Seasonality :

$$S_t = \sum_{k=1}^K \alpha_k \cos(\omega kt) + \beta_k \sin(\omega kt)$$

With  $\omega = 2\pi/365.25$ ,  $\alpha_k$  and  $\beta_k$  coefficients to estimate and  $K$  the order ( $K = 5$  in our work).



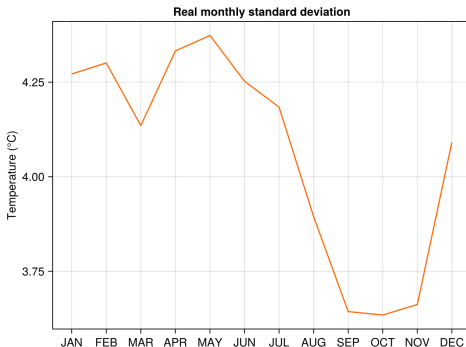
# I. Stochastic weather generators | Model

- Stochastic part  $\sim \text{AR}(p)$  model :

$$X_t = \sum_{i=1}^p \phi_i X_{t-i} + \sigma \varepsilon_t$$

With  $\varepsilon_t \stackrel{\text{iid}}{\sim} \mathcal{N}(0, 1)$ ,  $\phi_i$  and  $\sigma$  coefficients to estimate and  $p$  the order (we have chosen  $p = 1$ ).

Problem : Some features change during a year :



Model with different parameters for each month :

$$X_t = \sum_{i=1}^p \phi_{i,m(t)} X_{t-i} + \sigma_{m(t)} \varepsilon_t$$

With  $\varepsilon_t \stackrel{\text{iid}}{\sim} \mathcal{N}(0, 1)$  and  $m(t)$  the month of the date  $t$ .

→  $(p + 1) \times 12$  parameters

- 1 **Ordinary Linear Square (OLS) regression** to estimate  $\mu$ ,  $\alpha_k$  and  $\beta_k$  for  $k = 1, \dots, K$  :

$$Y_t = \mu + \sum_{k=1}^K (\alpha_k \cos(\omega kt) + \beta_k \sin(\omega kt)) + X_t$$

With  $X_t$  the residuals.

- 2 **Maximum likelihood estimation** on these residuals to estimate  $\phi_{i,j}$  and  $\sigma_j^2$ , for  $i = 1, 2, \dots, p$  and  $j = 1, 2, \dots, 12$  :

$$X_t = \sum_{i=1}^p \phi_{i,m(t)} X_{t-i} + \sigma_{m(t)} \varepsilon_t$$

# I. Stochastic weather generators | Simulation

We have estimated  $\hat{\mu}$ ,  $\hat{\alpha}_k$  and  $\hat{\beta}_k$  for  $k = 1, \dots, K$  and  $\hat{\phi}_{i,j}$  and  $\hat{\sigma}_j^2$ , for  $i = 1, 2, \dots, p$ ,  $j = 1, 2, \dots, 12$ . To simulate new temperatures series :

- 1 Initial conditions :  $\hat{X}_1 = X_1, \hat{X}_2 = X_2, \dots, \hat{X}_p = X_p$
- 2 For  $t > p$  :

$$\hat{X}_t = \sum_{i=1}^p \hat{\phi}_{i,m(t)} \hat{X}_{t-i} + \hat{\sigma}_{m(t)} \varepsilon_t$$

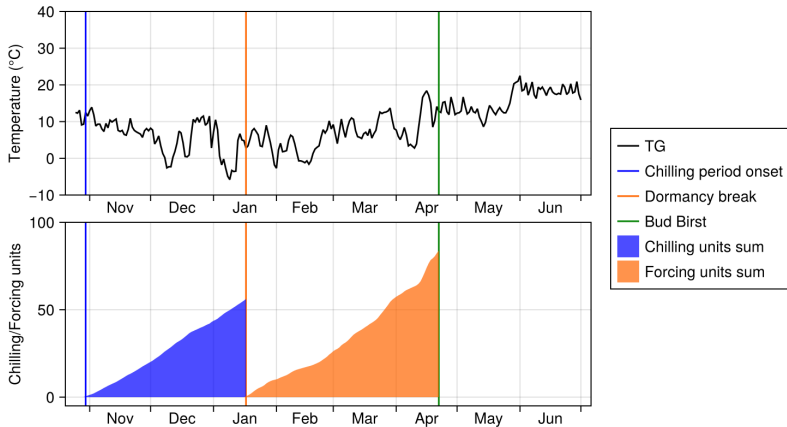
With  $\varepsilon_t \sim \mathcal{N}(0, 1)$ .

3

$$\hat{Y}_t = \hat{\mu} + \sum_{k=1}^K (\hat{\alpha}_k \cos(\omega kt) + \hat{\beta}_k \sin(\omega kt)) + \hat{X}_t$$

## II. Phenological models

## II. Phenological models | Introduction



Apple phenology dates simulated for the 2002-2003 period in Lille, according to the parameters estimated in Legave and al. - 2013

## II. Phenological models | Grapevine

Dormancy break date :

$$n_{db} = \min\left\{N, \sum_{n=CPO}^N C_U(n) > C_c\right\}$$

Budburst date :

$$n_{bb} = \min\left\{N, \sum_{n=n_{db}}^N A_c(n) > G_{hc}\right\}$$

$CPO$  : Chilling period onset

$C_c, G_{hc}$  : Chilling and heating quantity required

$C_U, A_c$  : Chilling and heating function

## II. Phenological models | Grapevine

Chilling function :

$$C_U(n) = Q_{10c}^{-\frac{TX(n)}{10}} + Q_{10c}^{-\frac{TN(n)}{10}}$$

Heating function :

$$T^*(h, n) = \begin{cases} TN(n) + h \left( \frac{TX(n) - TN(n)}{12} \right) & \text{if } h \leq 12 \\ TX(n) - (h - 12) \left( \frac{TX(n) - TN(n+1)}{12} \right) & \text{if } h > 12 \end{cases}$$

$$T(h, n) = \begin{cases} 0 & \text{if } T^*(h, n) < T_{0Bc} \\ T^*(h, n) - T_{0Bc} & \text{if } T_{0Bc} \leq T^*(h, n) \leq T_{MBc} \\ T_{MBc} - T_{0Bc} & \text{if } T_{MBc} < T^*(h, n) \end{cases}$$

$$A_c(n) = \sum_{h=1}^{24} T(h, n)$$

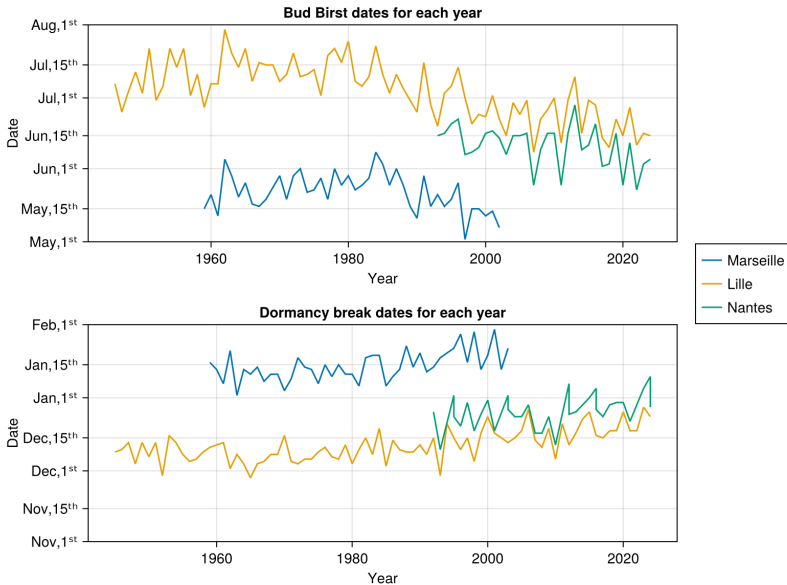
$Q_{10c}$  and  $T_{0Bc}$  are parameters and  $T_{MBc}$  is fixed at 25°C.



Parameters considered for our simulation :

- $CPO$  : 1<sup>st</sup> of August
- $C_c$  : 119.0 (chilling units)
- $G_{hc}$  : 13236°C
- $Q_{10c}$  : 2.17
- $T_{0Bc}$  : 8.19°C

## II. Phenological models | Grapevine



Dormancy break date :

$$n_{db} = \min\left\{N, \sum_{n=CPO}^N F_c(TG(n)) > C\right\}$$

Budburst date :

$$n_{bb} = \min\left\{N, \sum_{n=n_{db}}^N F_h(TG(n)) > H\right\}$$

$CPO$  : Chilling period onset

$C, H$  : Chilling and heating quantity required

$F_c, F_h$  : Chilling and heating function

## II. Phenological models | Apple

Parameters considered for our simulation :

- $CPO$  : 30<sup>th</sup> of October
- $C$  : 56.0 (chilling units)
- $H$  : 83.58 (forcing units)
- $F_c$  : Triangular chilling

$$F_c(T) = \begin{cases} 1 - (|T - T_c|/l_c) & \text{if } T \in (T_c - l_c, T_c + l_c) \\ 0 & \text{else} \end{cases}$$

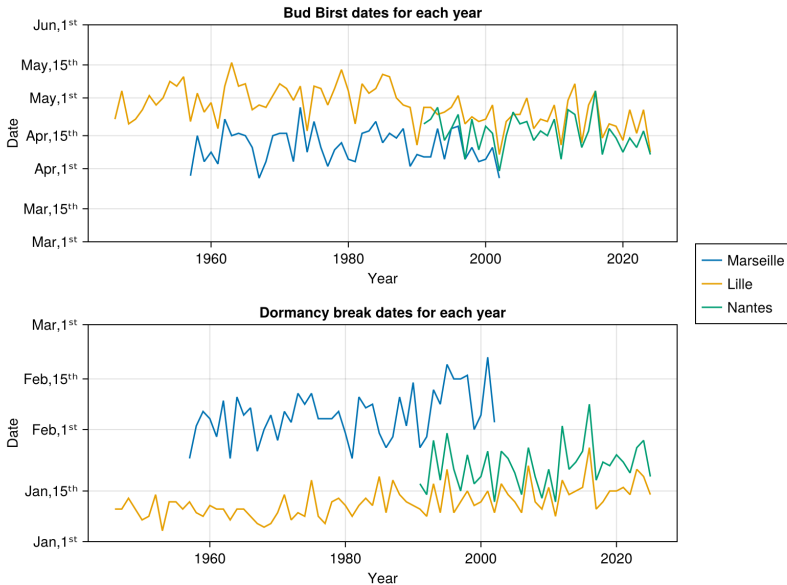
With  $T_c = 1.1^\circ\text{C}$  and  $l_c = 20.^\circ\text{C}$

- $F_h$  : Exponential forcing

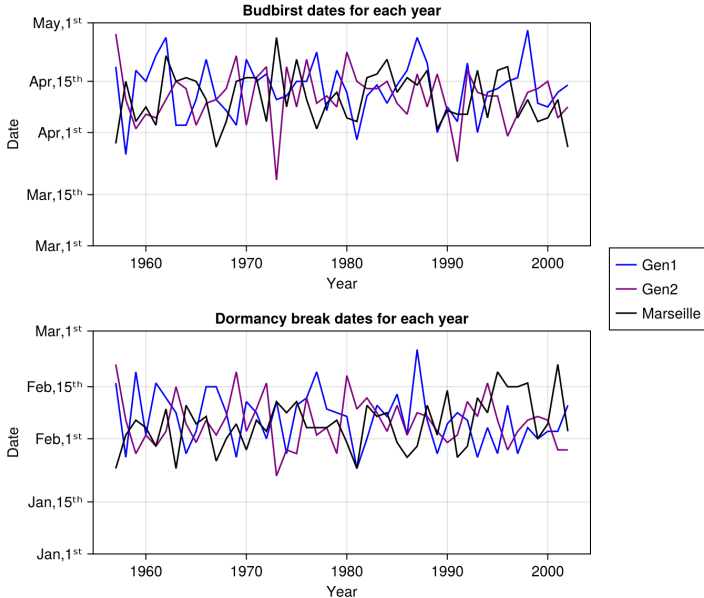
$$F_h(T) = \exp(T/T_h)$$

With  $T_h = 9.0^\circ\text{C}$

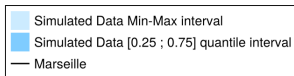
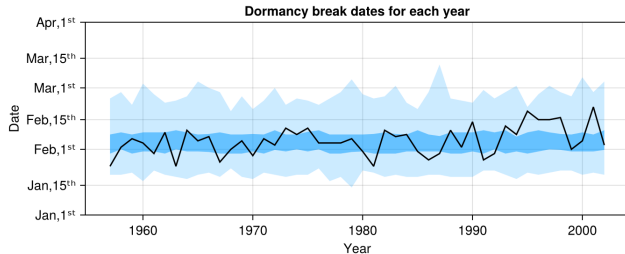
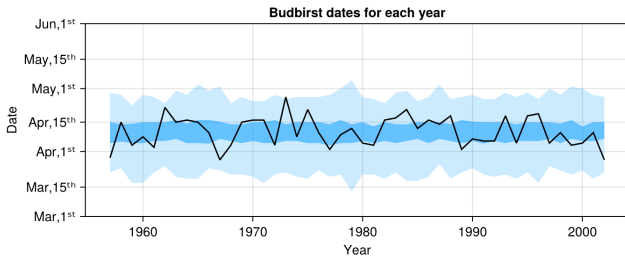
## II. Phenological models | Apple



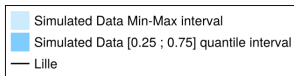
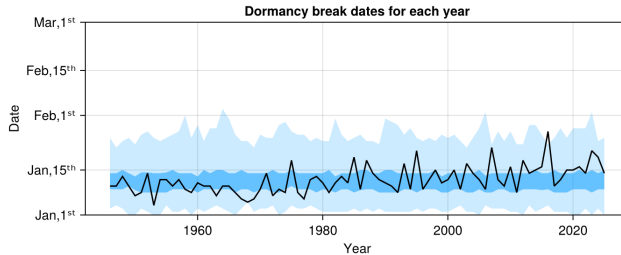
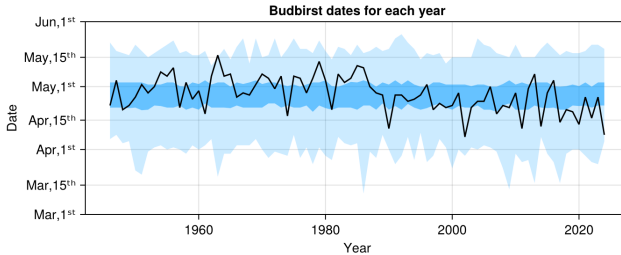
## II. Phenological models | With simulated TG (Apple)



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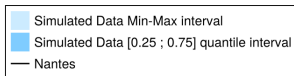
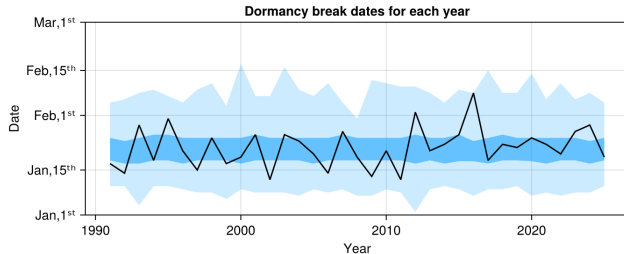
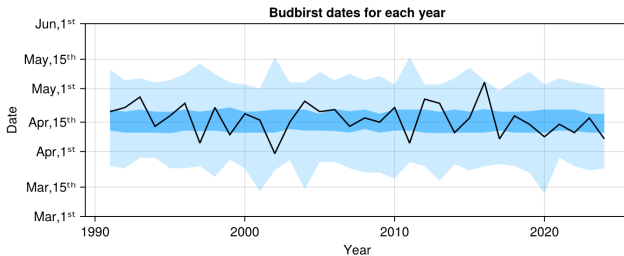


## II. Phenological models | With simulated TG (Apple)





## II. Phenological models | With simulated TG (Apple)



- We have studied and implemented a model that is able to generate temperature series but with a too simple modeling of the trend.
- We have implemented phenological models and applied them on stations from different climates in France, and we noticed that there is an evolution across the years since  $\approx 1980$ .
- Finally, we have applied phenological on simulated series, and the results show that it is required to consider a better model for the trend if we want more realistic temperature series.

Thanks you !