

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

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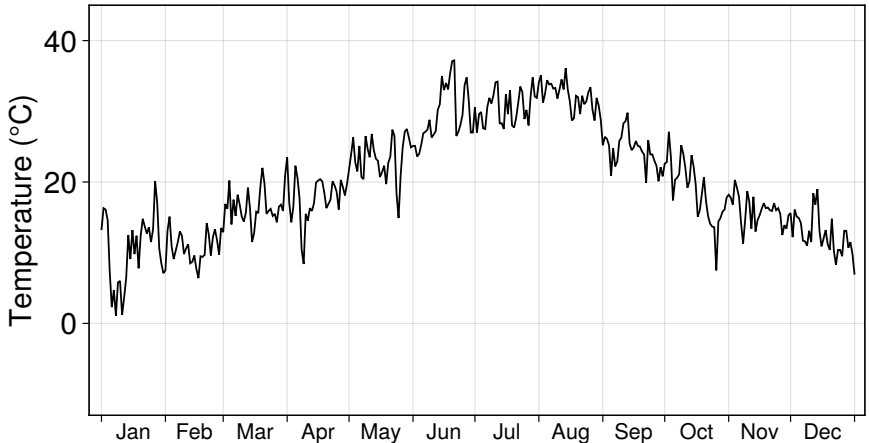
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I. Stochastic weather generators



Recorded daily maximum temperature (TX) at Montpellier from 1st May 2003 to 1st November 2003 (source : ECA&D)

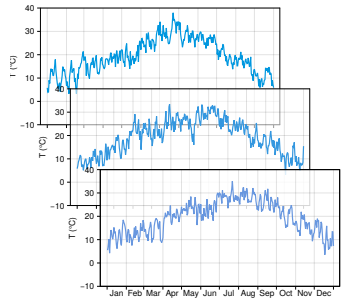
Weather data



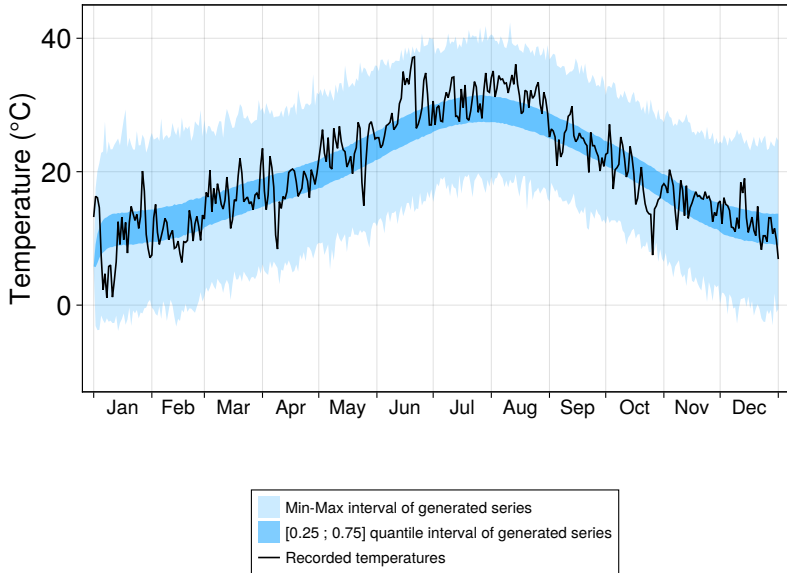
Stochastic Model



Simulations



With 5000 simulations :



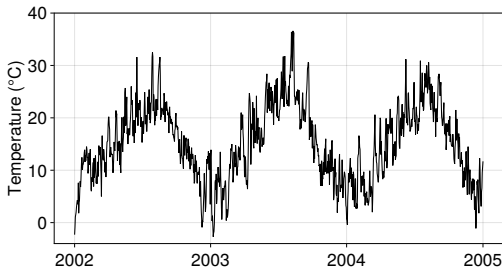
The SWG can simulate :

- Daily minimal temperature (TN) series.
- Daily average temperature (TG) series.
- Daily maximal temperature (TX) series.

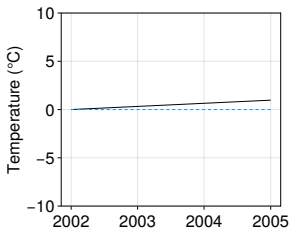
But for now it can't generate two of these series at the same time because we have to consider a correlation between them.

$T_t = M_t + S_t + X_t$ where

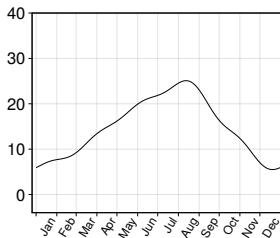
- T_t is the recorded temperature
- M_t is the trend
- S_t is the seasonality
- X_t is the stochastic part



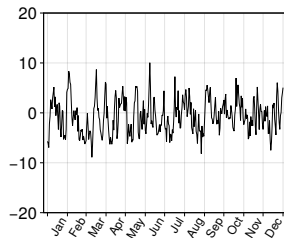
Recorded temperature T_t



Trend M_t



Seasonality S_t

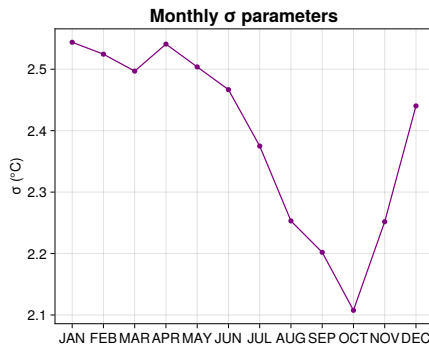
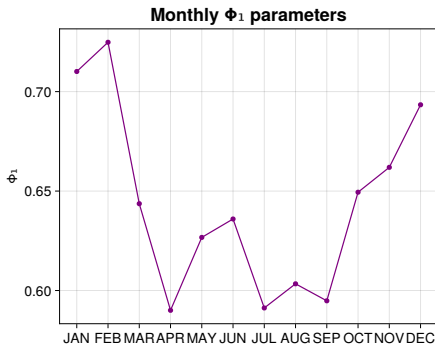


Stochastic part X_t

Model with different parameters for each month :

$$X_t = \phi_{1,m(t)}X_{t-1} + \phi_{2,m(t)}X_{t-2} + \dots + \phi_{p,m(t)}X_{t-p} + \sigma_{m(t)}\varepsilon_t$$

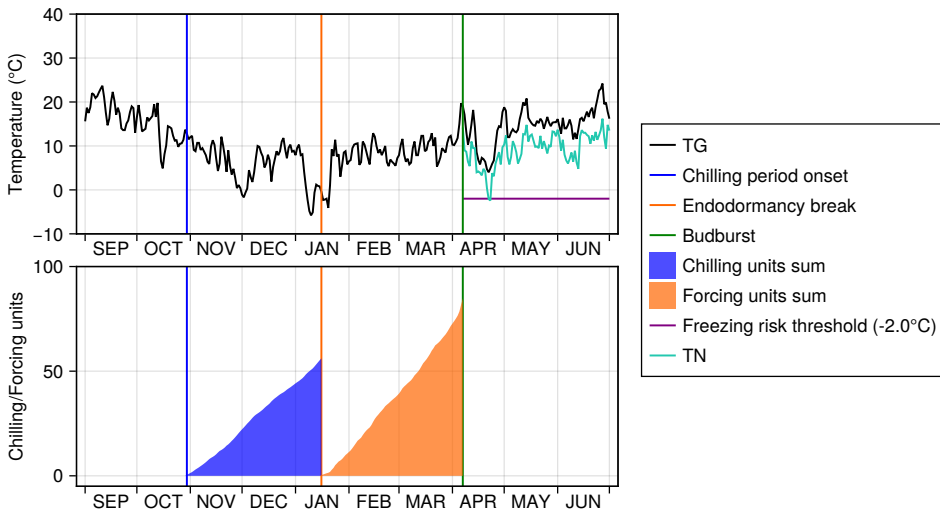
With $\varepsilon_t \stackrel{\text{iid}}{\sim} \mathcal{N}(0, 1)$ a noise and $m(t)$ the month of the date t . With $p = 1$, the model trained on the Montpellier TX series (1946-2025) gives :



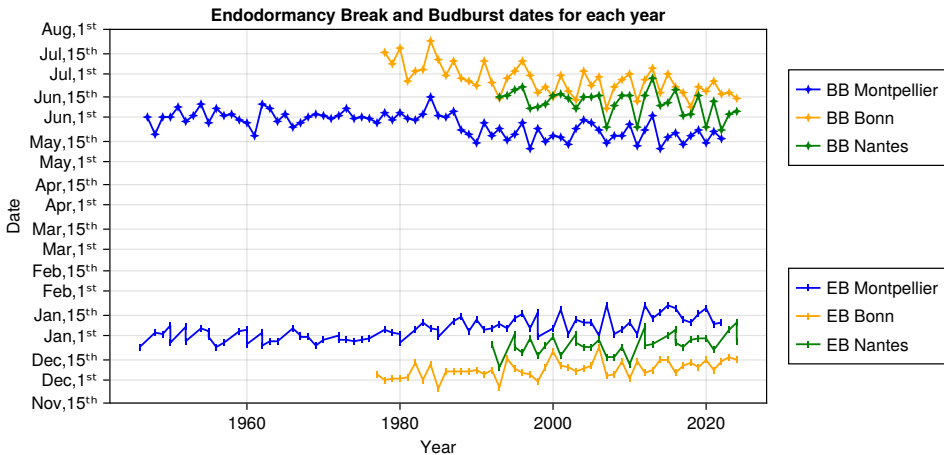
- Firstly, we estimate the parameters with a **maximum likelihood estimation**.
- To simulate a new temperature series (\hat{T}_t) :
 - ① Initial conditions : $\hat{X}_1 = X_1, \hat{X}_2 = X_2, \dots, \hat{X}_p = X_p$.
 - ② We simulate $\varepsilon_t \sim \mathcal{N}(0, 1)$ for $t > p$.
 - ③ \hat{X}_t calculated with the previous equation, for $t > p$.
 - ④ $\hat{T}_t = M_t + S_t + \hat{X}_t$.

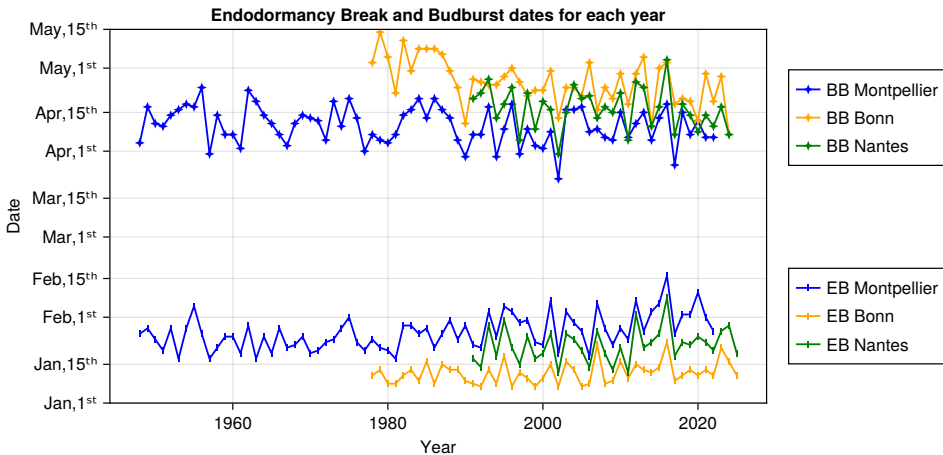
II. Phenological models applied to recorded temperatures

- For grapevine : **BRIN** (García de Cortázar-Atauri et al. -2009) :
 - Chilling period onset : 1st of August
 - Chilling quantity required C_c : 119.0 (chilling units)
 - Heating quantity required G_{hc} : 13236°C
 - Q_{10c} : 2.17
 - T_{0Bc} : 8.19°C
- For apple (Golden Delicious) : **F 1 Gold 1** (Legave et al. - 2013)
 - CPO : 30th of October
 - Chilling quantity required C : 56.0 (chilling units)
 - Heating quantity required H : 83.58 (forcing units)
 - Chilling function F_c : Triangular
 - T_c : 1.1°C
 - I_c : 20.°C
 - Forcing function F_h : Exponential
 - T_h : 9.0°C



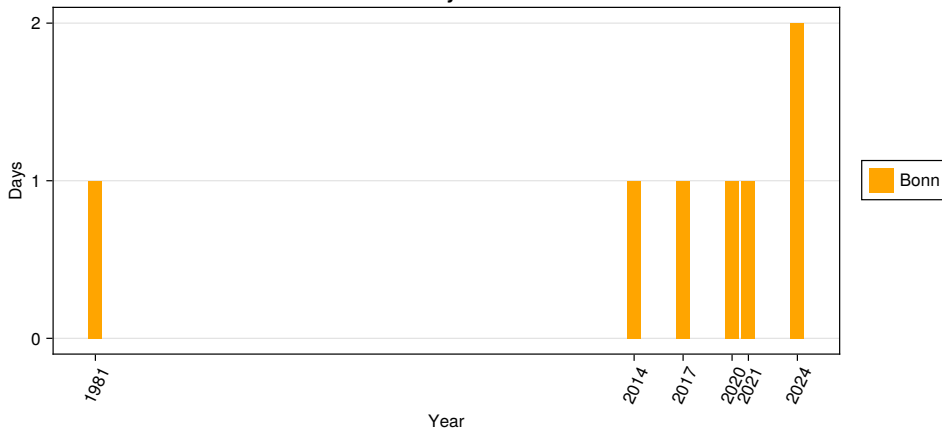
Apple phenology dates predicted for the 2023-2024 period in Bonn.





Too early budbursts make the plant vulnerable to a risk of freezing :

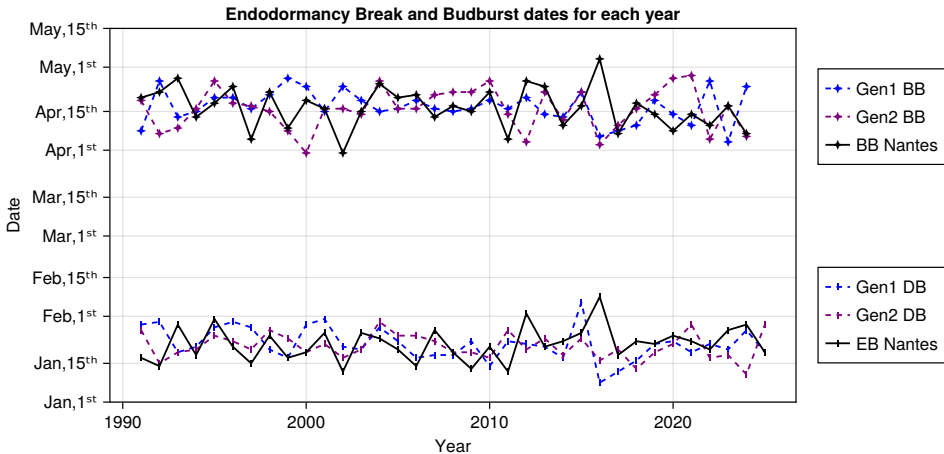
Max number of consecutives days with $TN \leq -2^{\circ}C$ after budburst



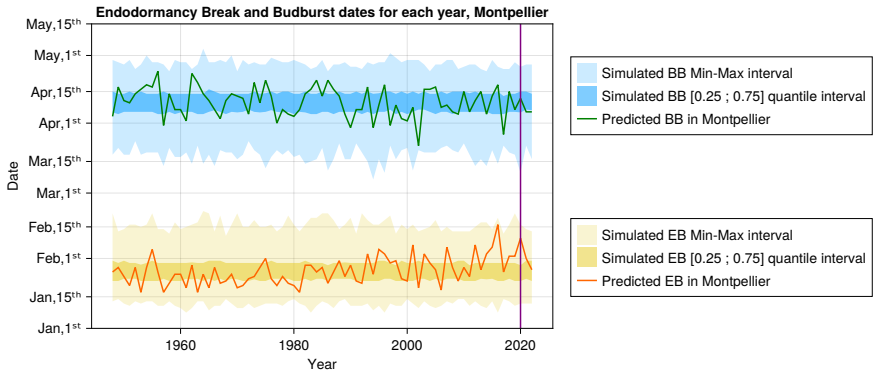
III. Apple phenological model applied to simulated temperatures

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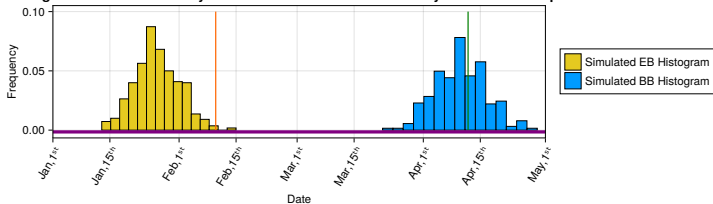
Nantes :



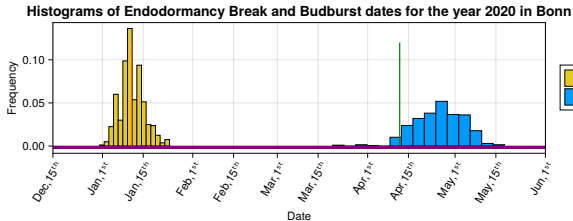
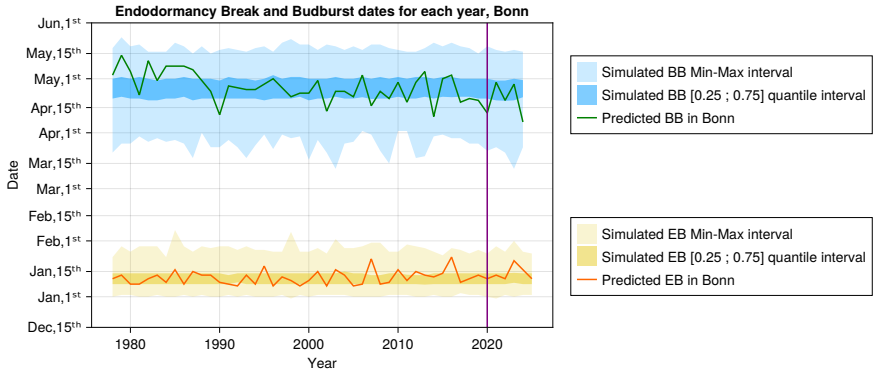
III. Apple phenological model applied to simulated temperatures



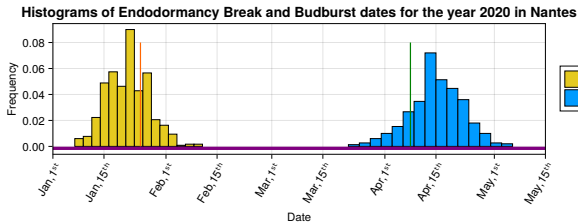
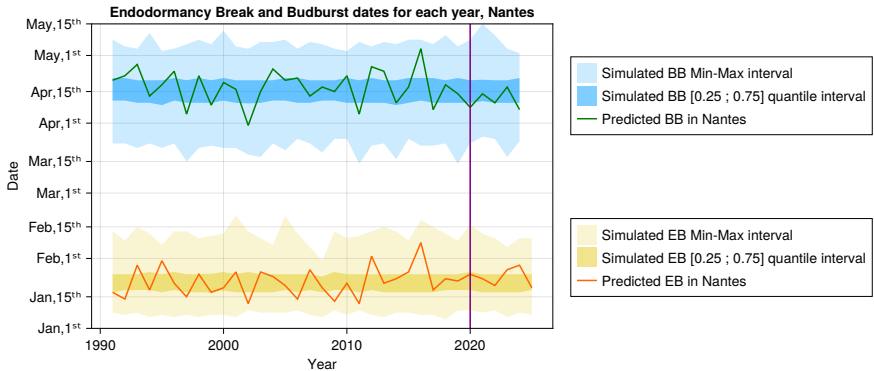
Histograms of Endodormancy Break and Budburst dates for the year 2020 in Montpellier



III. Apple phenological model applied to simulated temperatures

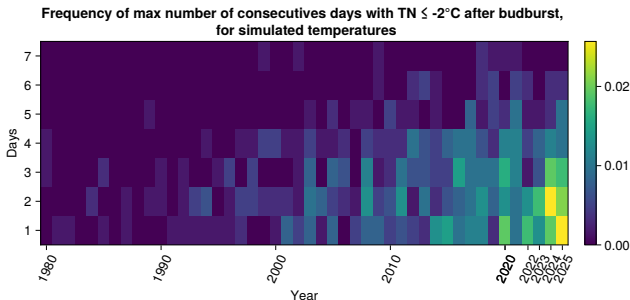


III. Apple phenological model applied to simulated temperatures



Prospects

- Modelling the trend with non-constant values → LOESS regression intended
- Making the SWG able to generate multiple type series correlated (e.g TN and TX or TN and TG) to :
 - Apply grapevine phenology models on generated TN and TX.
 - Model the risk of frost for phenology results from generated series. We can expect results like this :



Thanks you !

Appendix

- Seasonality : Parametric function with a periodicity of 365.25 days :

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

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

- "Monthly" AR(p) model :

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

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 μ , α_k and β_k for $k = 1, \dots, K$:

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

With X_t the residuals.

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

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

Endodormancy break date :

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

Budburst date :

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

CPO : Chilling period onset

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

C_U, A_c : Chilling and heating function

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.

Reminder : parameters considered for our simulation :

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

Endodormancy break date :

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

Budburst date :

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

CPO : Chilling period onset

C, *H* : Chilling and heating quantity required

F_c, *F_h* : Chilling and heating function

Reminder : parameters considered for our simulation :

- CPO : 30th of October
- C : 56.0 (chilling units)
- H : 83.58 (forcing units)
- F_c : Triangular chilling

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

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

- F_h : Exponential forcing

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

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