

# DEVELOPMENT OF A SNOW REANALYSIS PIPELINE USING DOWNSCALED ERA5 DATA

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Sharing is  
encouraged

# INTRODUCTION

- **Reanalyses products** are too coarse to be used as direct input to snow model
- MicroMet-SnowModel (Liston et al. 2006) can be useful to **downscale** reanalysis data and simulate SWE distribution (e.g. Andes: Mernild et al. 2017; Atlas: Baba et al. 2018)
- **Assimilation of remote sensing snow cover area** can correct bias (Andreadis & Lettenmeier 2006)

# 1 GENERAL PRESENTATION OF THE ESP

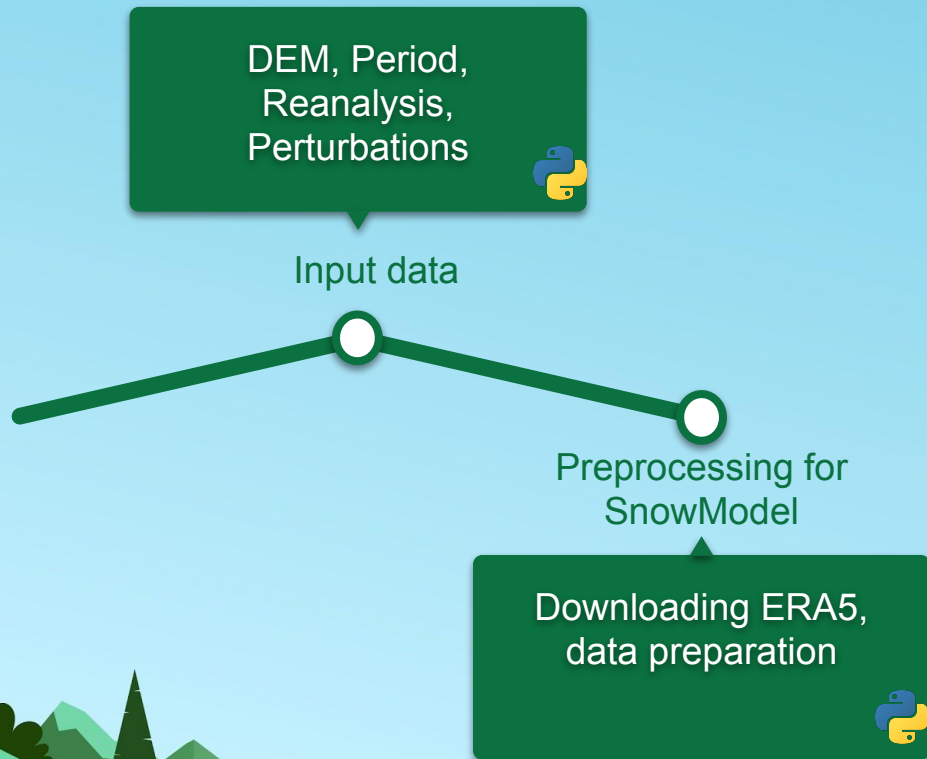
DEM, Period,  
Reanalysis,  
Perturbations



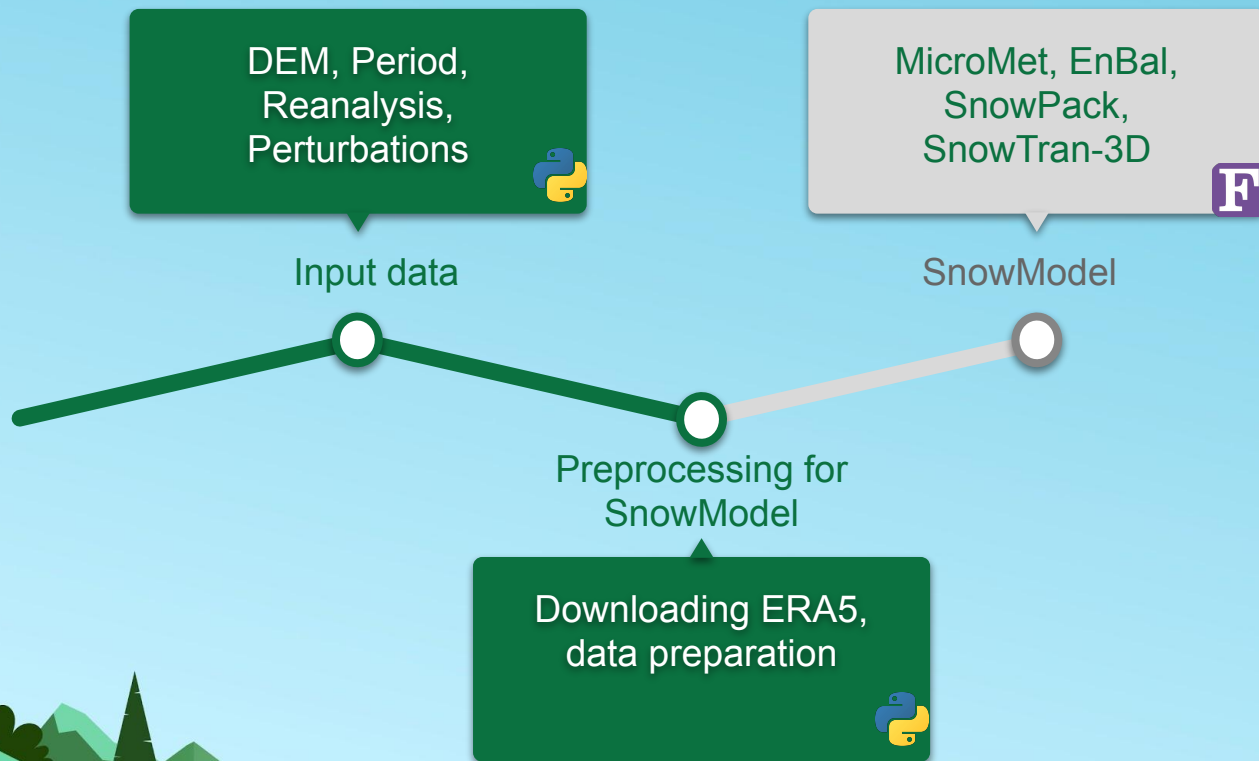
Input data



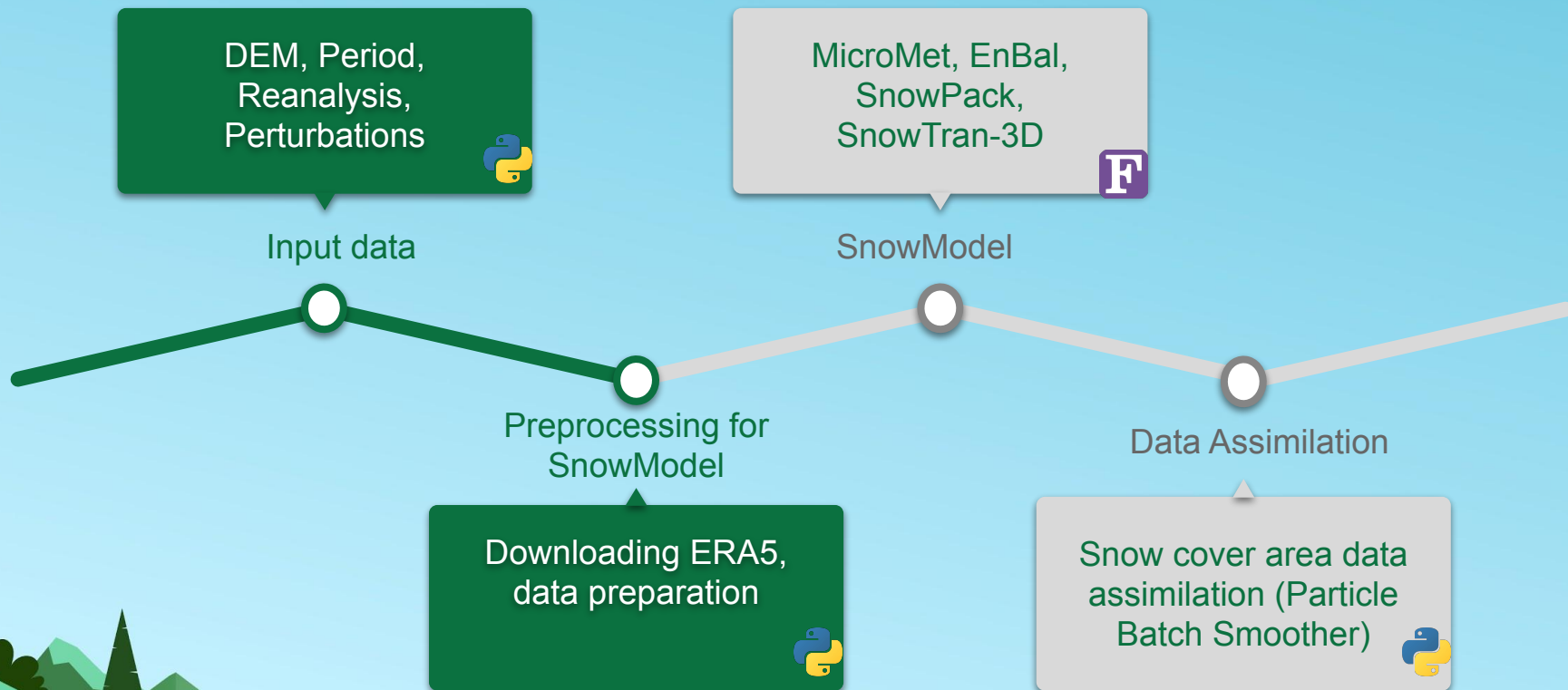
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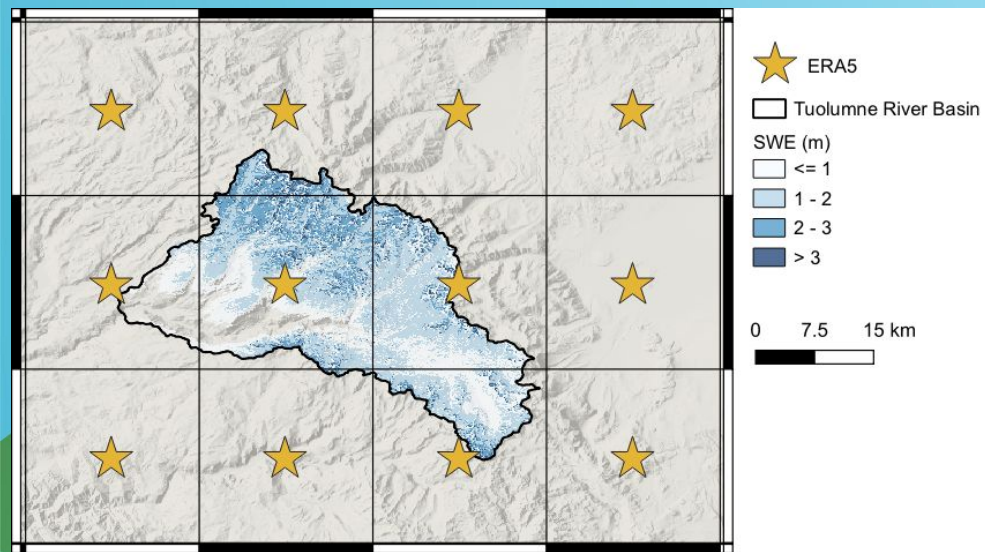


## 2

## EVALUATION

A

Tuolumne (USA) ASO HS  
(2012-2019)



# 2

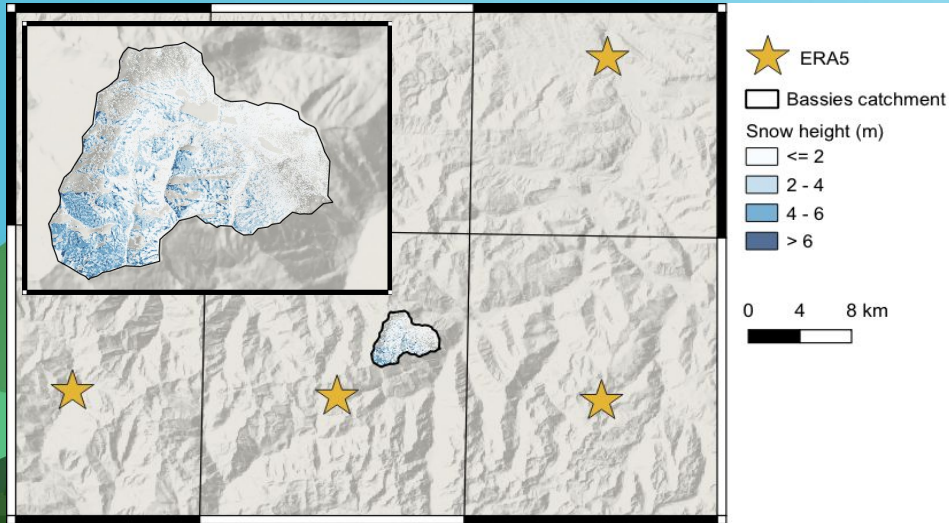
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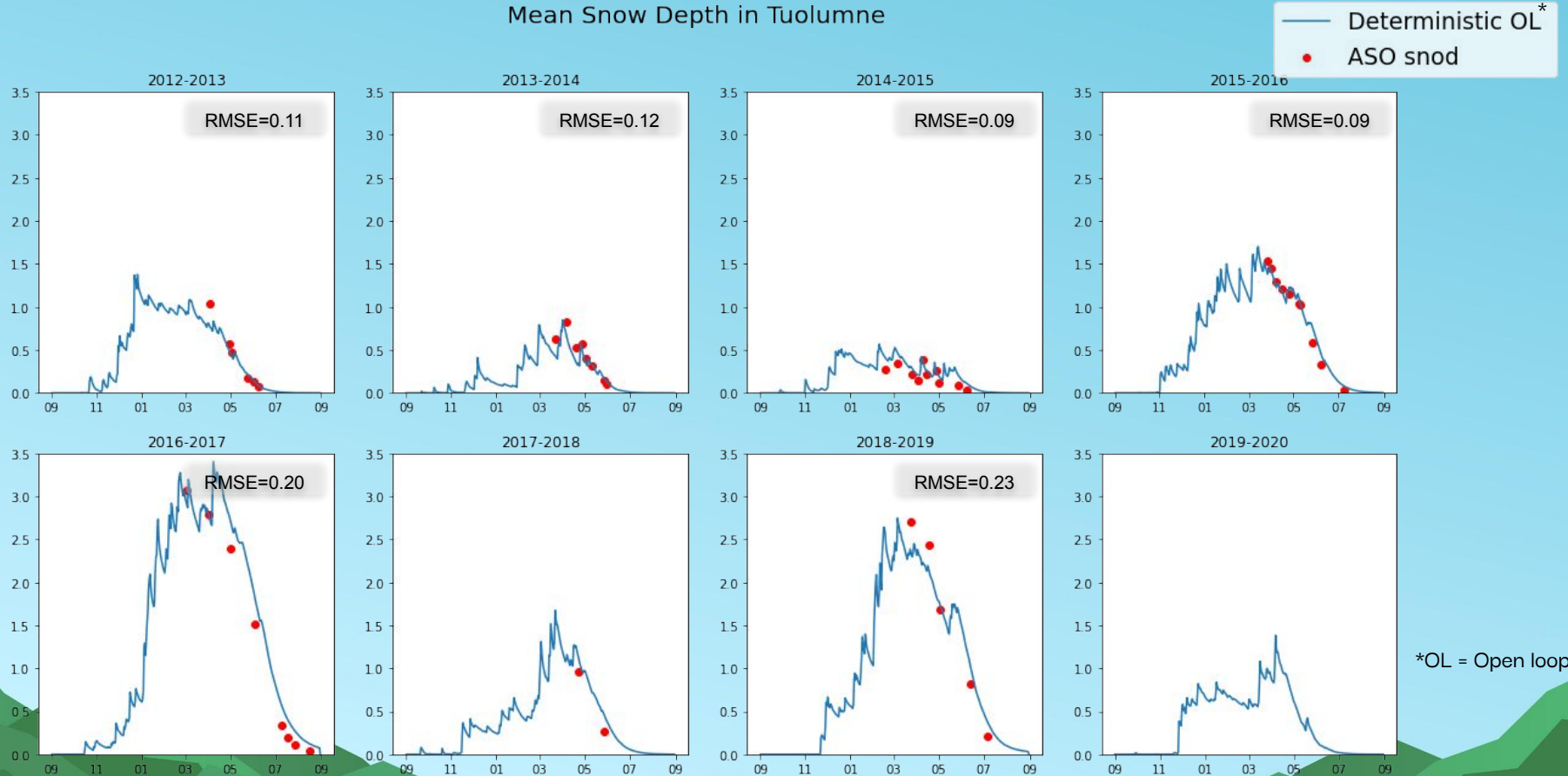
B

Bassières (France) Pléiades  
HS (2015-2020)



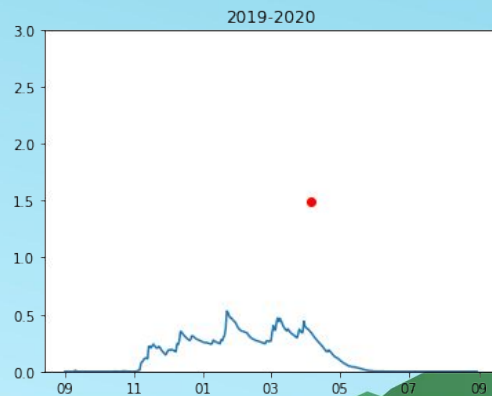
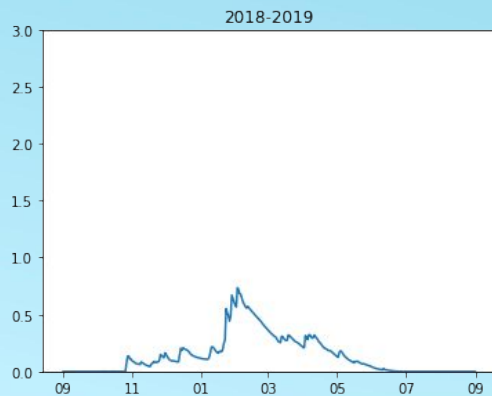
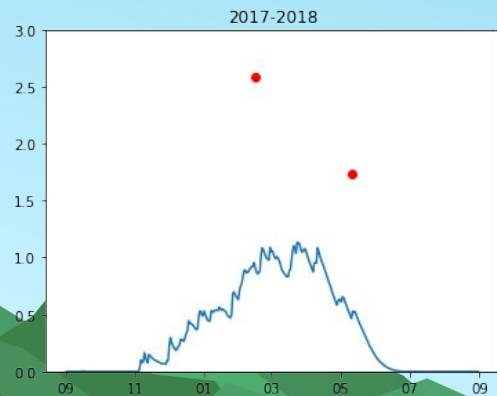
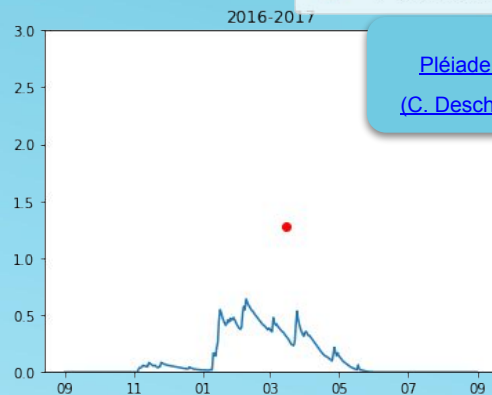
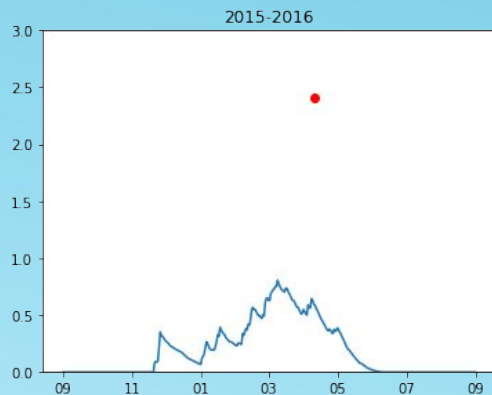
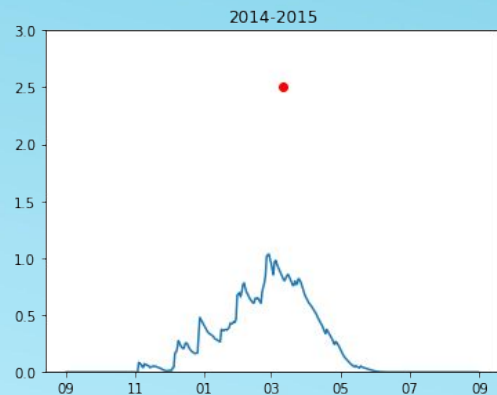
# 2A EVALUATION IN TUOLUMNE (1200 KM<sup>2</sup>)

Mean Snow Depth in Tuolumne



# 2B EVALUATION IN BASSIES (14 KM<sup>2</sup>)

Mean Snow Depth in Bassies

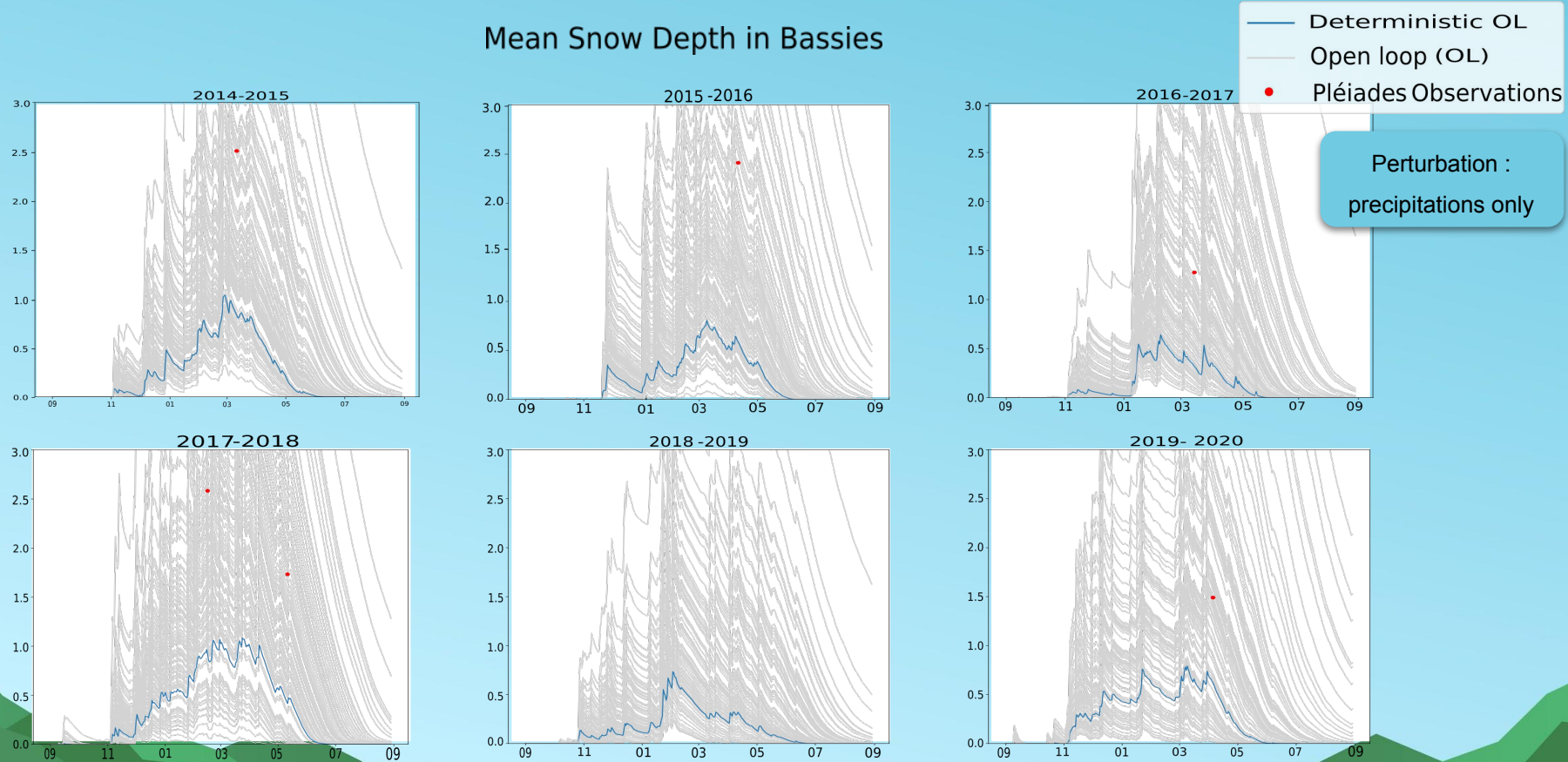


— Deterministic OL  
• Pléiades Observations

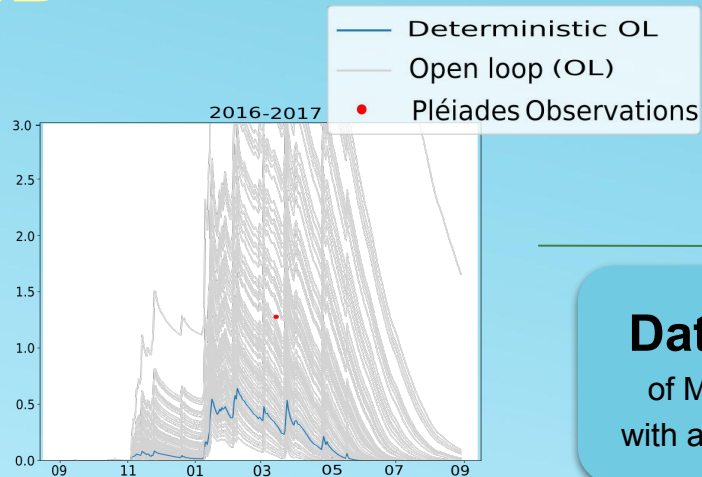
[Pléiade Observations HS](#)  
(C. Deschamps-Berger et al.)

# 2B ENSEMBLE RUN IN BASSIES (14 KM<sup>2</sup>)

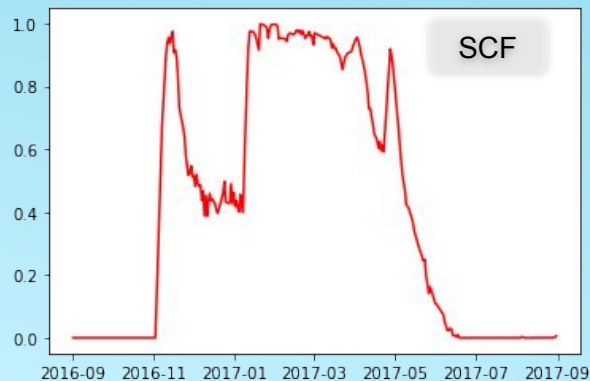
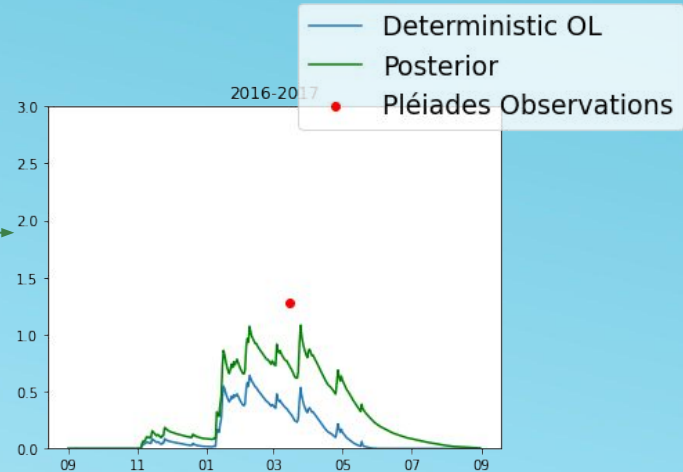
Mean Snow Depth in Bassies



# 2B DATA ASSIMILATION IN BASSIES (14 KM<sup>2</sup>)

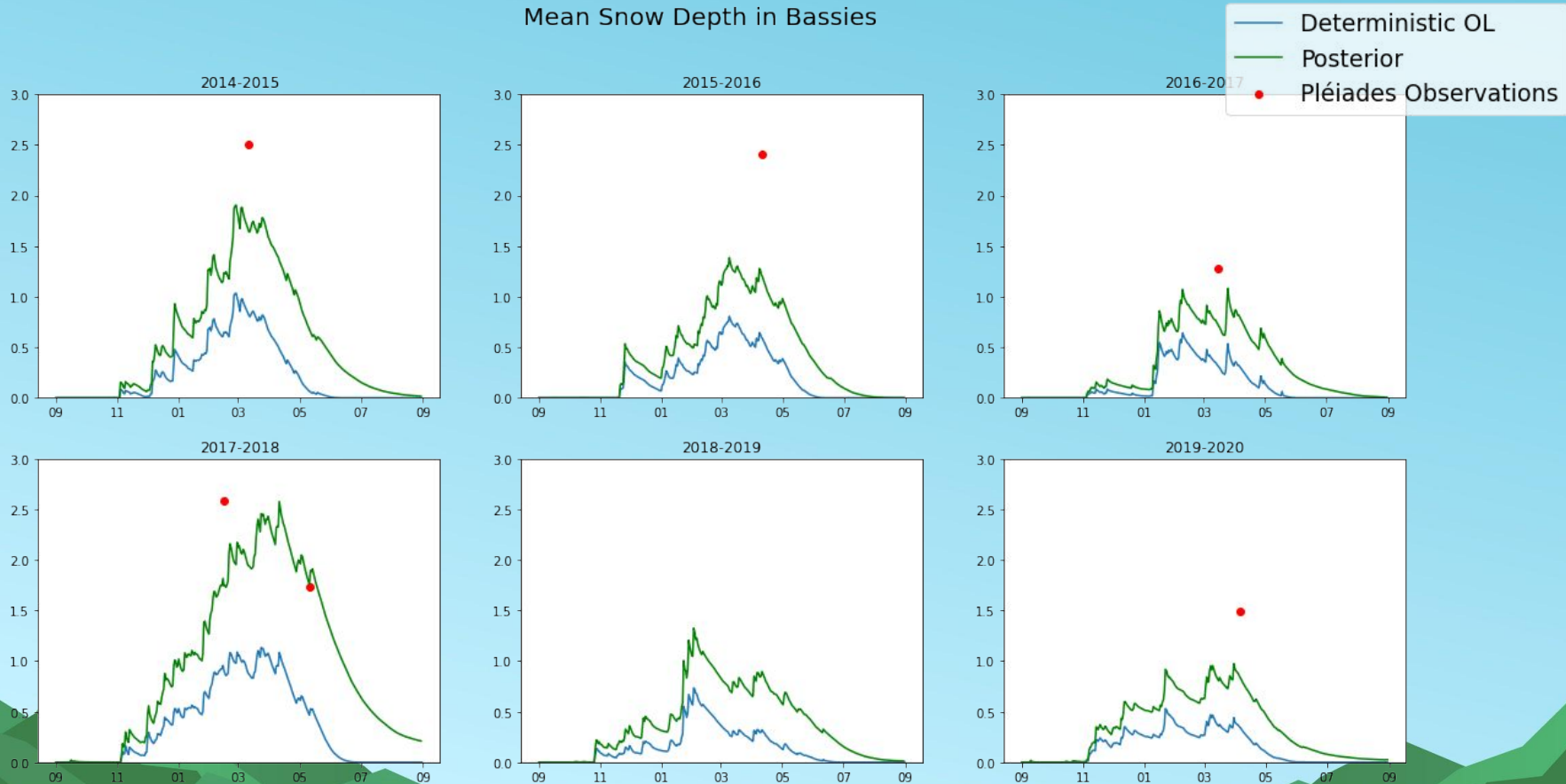


**Data Assimilation**  
of MODIS snow cover area  
with a Particle Batch Smoother



# 2B DATA ASSIMILATION IN BASSIES (14 KM<sup>2</sup>)

Mean Snow Depth in Bassies



# CONCLUSION

- THE ADVENT OF REANALYSES IS A GAME CHANGER IN ENVIRONMENTAL SCIENCE
- THE ESP PROVIDES AN EASY WAY TO COMPUTE SNOW COVER WITH REANALYSES
- POSSIBILITY TO RUN ENSEMBLE SIMULATIONS BY PERTURBING PRECIP & T°C
- POSSIBILITY TO ASSIMILATE OTHER SNOW COVER AREA PRODUCTS (SENTINEL-2)
- DATA ASSIMILATION AVAILABLE, STILL UNDER EVALUATION



# THANK YOU

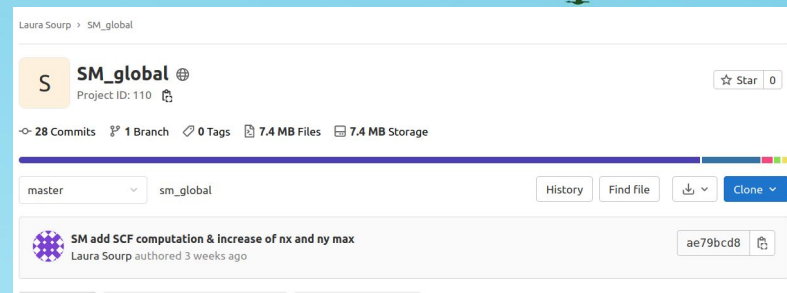
Any questions?

[laura.sourp@cesbio.cnes.fr](mailto:laura.sourp@cesbio.cnes.fr)

[https://gitlab.cesbio.omp.eu/lauras/sm\\_global.git](https://gitlab.cesbio.omp.eu/lauras/sm_global.git)



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# A. SNOWMODEL MODULES



## MICROMET

Distribution of the meteorological variables across given landscape by spatially interpolating data and physical sub-models improvements



## ENBAL

Computing energy fluxes between the snowpack and exterior environment



## SNOWPACK

Computing snow depths and snow density with the meteorological forcings, the energy fluxes and mechanical processes



## SNOWTRAN 3D

Redistribution of the snow according to the wind

# A. SNOWMODEL MODULES



## MICROMET

Distribution of the meteorological variables across given landscape by spatially interpolating data and physical sub-models improvements



Temperature

$$T_0 = T_{\text{stn}} - \Gamma(z_0 - z_{\text{stn}})$$

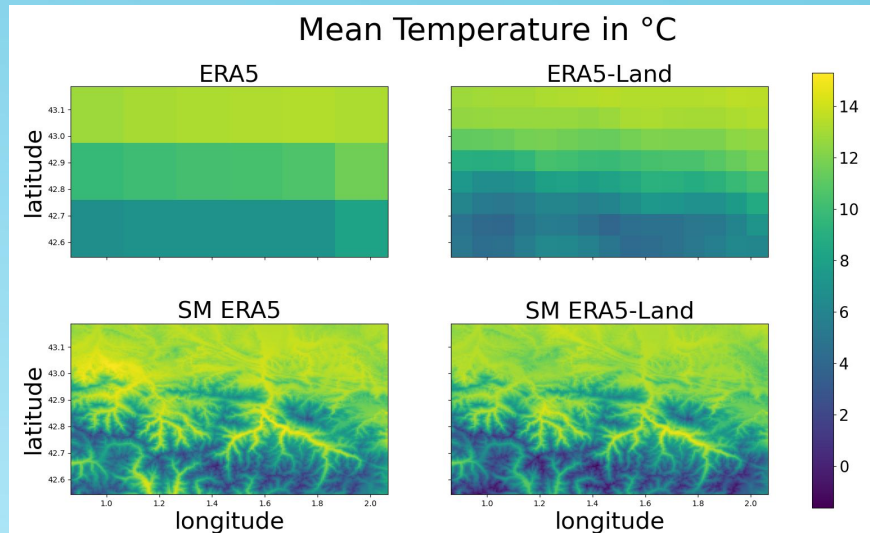
Spatial interpolation  
(Barnes 1964)

$$T = T_0 - \Gamma(z - z_0)$$

$T_{\text{stn}/0}$  : observed/sea level temperature

$z_{\text{stn}/0}$  : observed/sea level elevation

$\Gamma$  : lapse rate



# A. SNOWMODEL MODULES



## MICROMET

Distribution of the meteorological variables across given landscape by spatially interpolating data and physical sub-models improvements



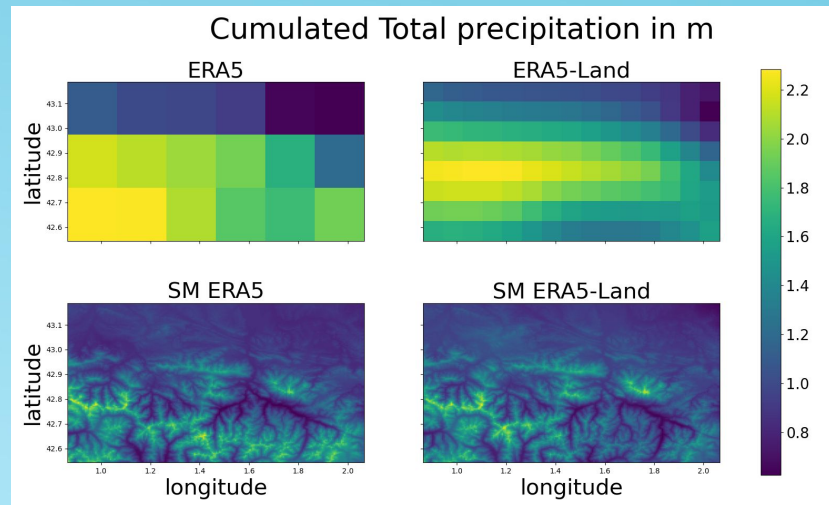
## Precipitations

$$P = P_0 \left[ \frac{1 + \chi(z - z_0)}{1 - \chi(z - z_0)} \right]$$

$P_0$  : sea level precipitations

$z_{.0}$  : DEM/sea level elevation

$\chi$ : precipitations adjustment factor



# A. SNOWMODEL MODULES



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Distribution of the meteorological variables across given landscape by spatially interpolating data and physical sub-models improvements

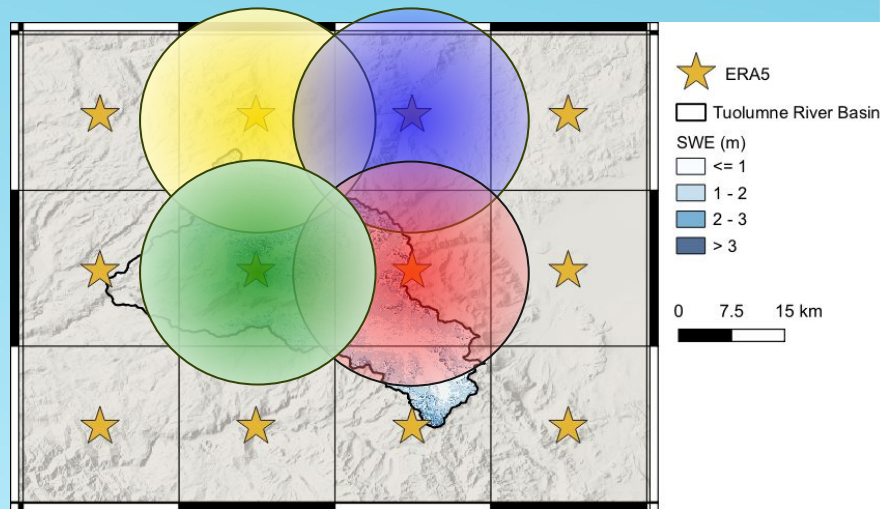


Spatial interpolation (Barnes 1994):

A Gaussian distance-dependent weighting function

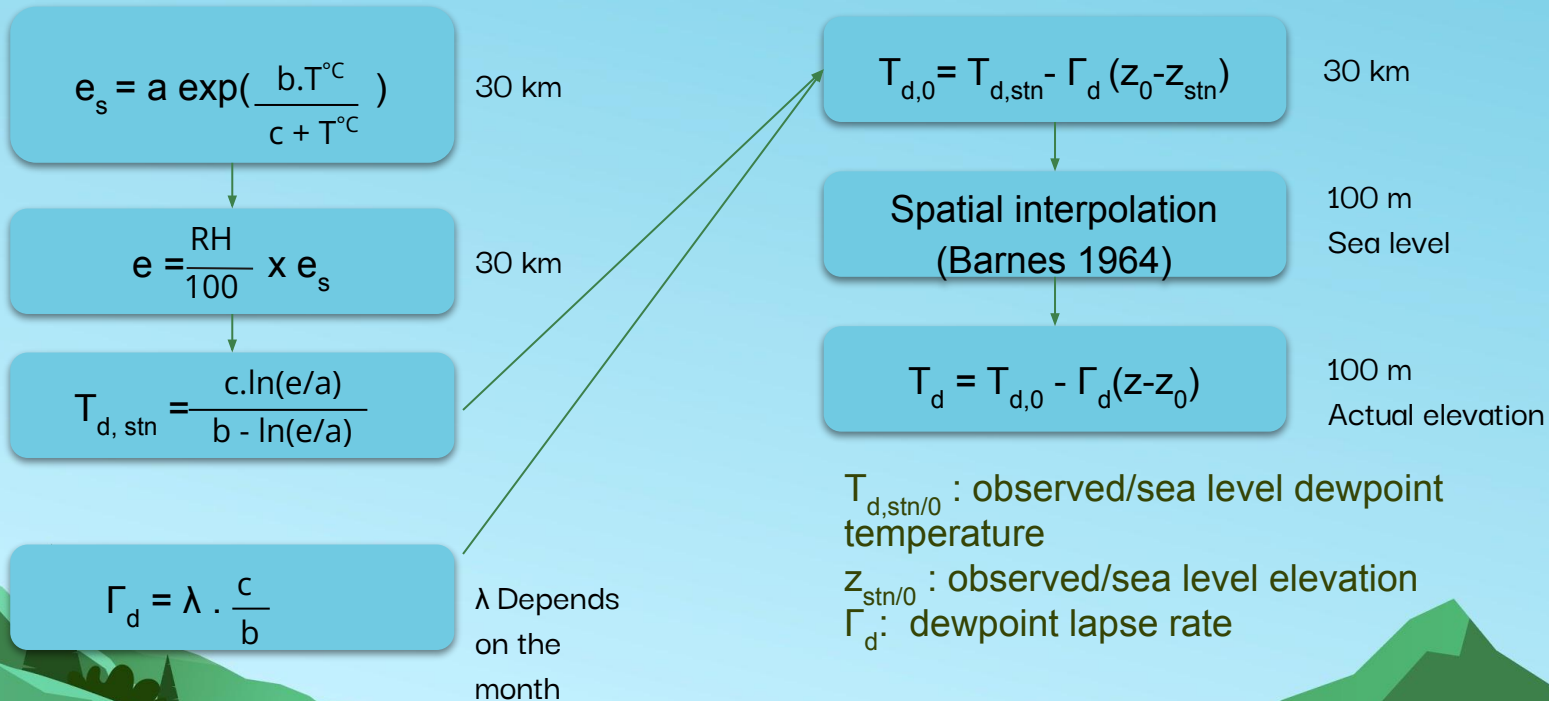
$$w = \exp\left(\frac{r^2}{f(dn)}\right)$$

2nd pass



# A. SNOWMODEL MODULES - MICROMET

Relative Humidity to dewpoint temperature



# A. SNOWMODEL MODULES - MICROMET

≈ Wind speed and direction

$$\begin{aligned} u &= -W_{\text{stn}} \sin(\Theta_{\text{stn}}) \\ v &= -W_{\text{stn}} \cos(\Theta_{\text{stn}}) \end{aligned} \quad 30 \text{ km}$$

Spatial interpolation  
(Barnes 1964) 100 m

$$\begin{aligned} W &= (u^2 + v^2)^{1/2} \\ \Theta &= \frac{3\pi}{2} - \tan^{-1}\left(\frac{v}{u}\right) \end{aligned} \quad 100 \text{ m}$$

Terrain slope :  $\beta$       Terrain slope azimuth :  $\xi$       Curvature :  $\Omega_c$

$$\text{Slope in wind direction: } \Omega_s = \beta \cos(\Theta - \xi)$$

$$\text{Wind weight factor : } W_w = 1 + \gamma_s \Omega_s + \gamma_c \Omega_c$$

$$\text{Terrain wind speed : } W_t = W_w \cdot W$$

$$\begin{aligned} \text{Diverting factor :} \\ \Theta_d &= -0.5 \Omega_s \sin[2(\xi - \Theta)] \end{aligned}$$

$$\text{Terrain wind dir : } \Theta_t = \Theta + \Theta_d$$

# A. SNOWMODEL MODULES - MICROMET



## Solar Radiation

$$Q_{si} = S^*(\Psi_{dir} \cos i + \Psi_{dif} \cos Z)$$

- $Z$  : day, latitude, hour
- $i$  : slope, slope azimuth,  $Z$
- $\Psi$  :  $Z$ , cloud fraction ( $T^\circ\text{C}$  &  $T_d$ )



## Longwave Radiation

$$Q_{li} = \varepsilon \sigma T^4$$

- $\varepsilon$  : elevation, cloud fraction,  $T^\circ\text{C}$
- $\sigma$  : constante



## Surface pressure

$$p = p_0 \exp\left(-\frac{z}{H_{atm}}\right)$$

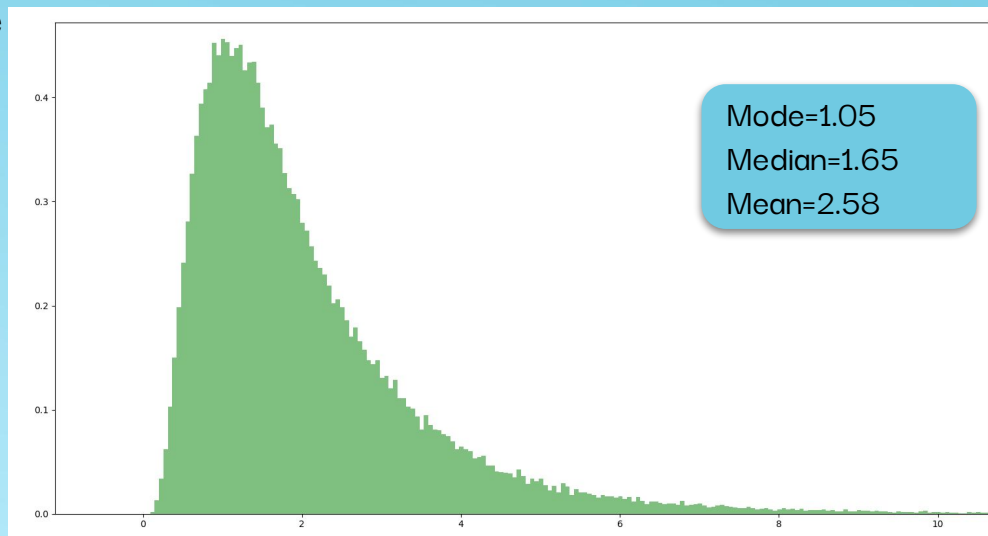
# A. PREPROCESSINGS : PERTURBATIONS

All the precipitations are multiplied with the same  $b$  coefficient for one year and one particle:

$$P_{\text{new}} = b * P_{\text{ERA}}$$

$$b \sim \text{LN}(\mu, \sigma)$$

When launching the pipeline, the user chooses  $\mu$  and  $\text{CV} = e^{\sigma} - 1$  (variation coefficient)



Density function for the log-normal perturbation with  $\mu=0.5$  and  $\text{CV}=0.75$

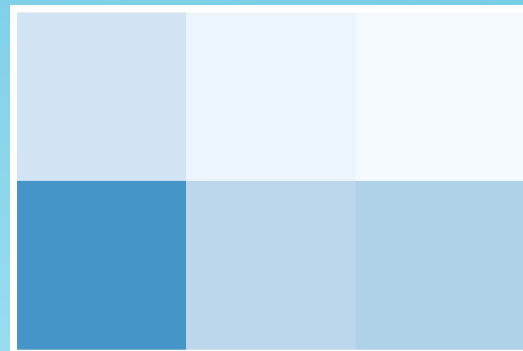
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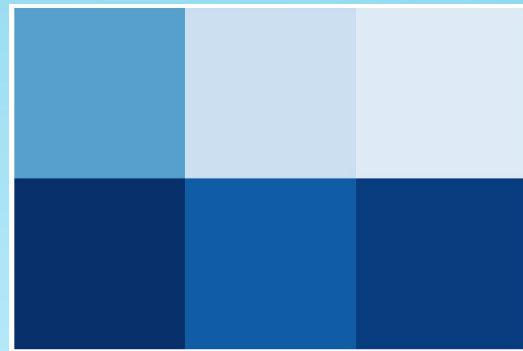
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ERA5 precipitations at one random time



Perturbed ERA5 precipitations at the same time ( $b=2.58$ )

# A. THE PBS FILTER IN THE ESP



## DATA ASSIMILATION

Introducing a particle filter  
(PBS) with MODIS surface  
cover

1 ensemble :  $j=1,\dots,N$

$Y$  : matrix of the  
measurements of the  
**whole season**

Anterior:  $Y_j^- = \text{SWE} E_j^-$  with  $w_j^- = 1/N$

Posterior :  $Y_j^+ = Y_j^- = \text{SWE} E_j^-$

with  $w_j^+ = \frac{c_0}{N} p_V(Z - M_j^-)$

# A. DATA ASSIMILATION



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(PBS) with MODIS surface  
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