



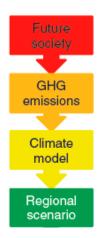


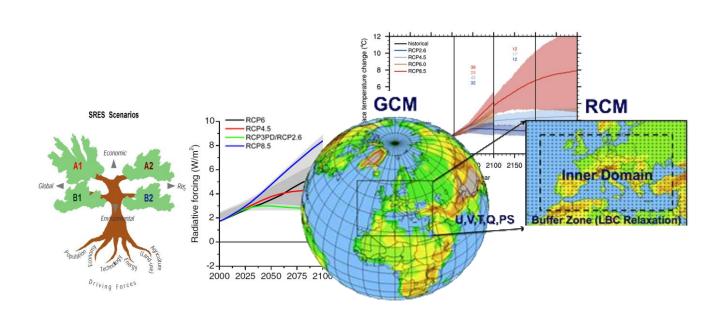
### Soulivanh THAO (LSCE), Yoann ROBIN (IPSL), Mathieu Vrac (LSCE)

# Introduction au downscaling statistique et à la correction de biais des simulations climatiques

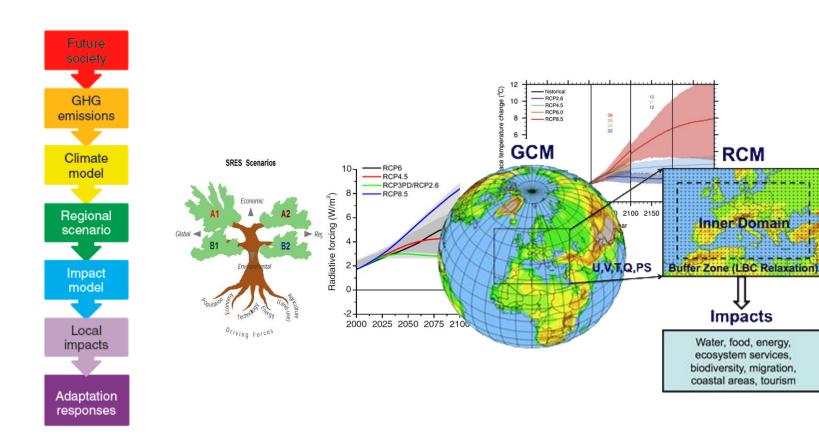
Ecole de Printemps « Impacts hydrologiques des changements climatiques » Banyuls-sur-mer, 09-13 mai 2022

# From scenarios to adaptations/mitigations

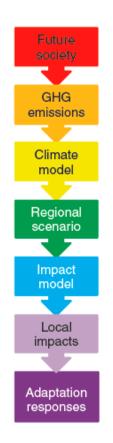




# From scenarios to adaptations/mitigations

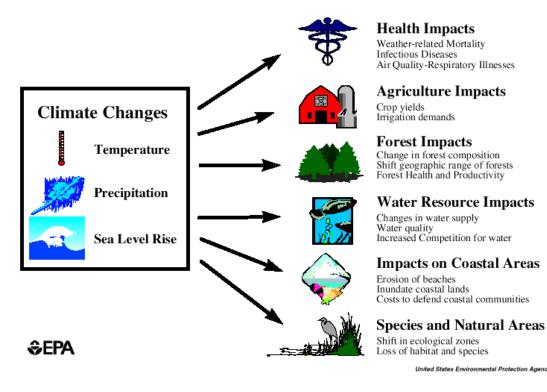


# From climate simulations to impact models

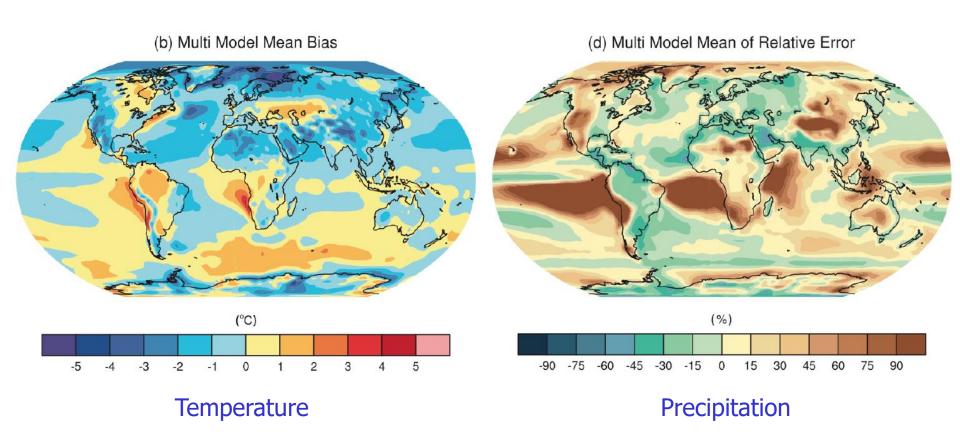


#### Many impact models

United States Environmental Protection Agency

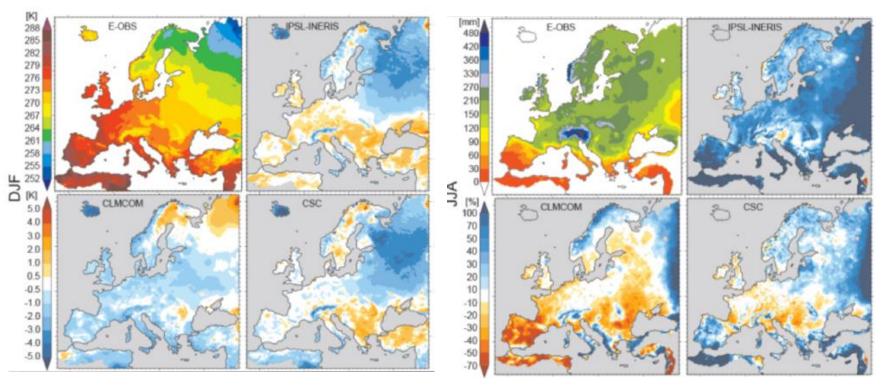


### But GCMs have biases...



Source: Flato et al., IPCC AR5, 2013

# ... RCMs as well ...



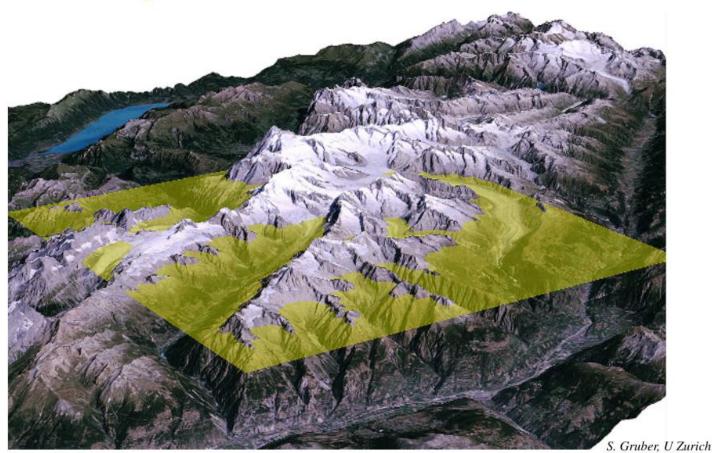
**Temperature** 

Precipitation

Source: Kotlarski et al., Geosci. Model Dev, 2014

# Some are due to differences in resolution

18 km x 18 km RCM grid box



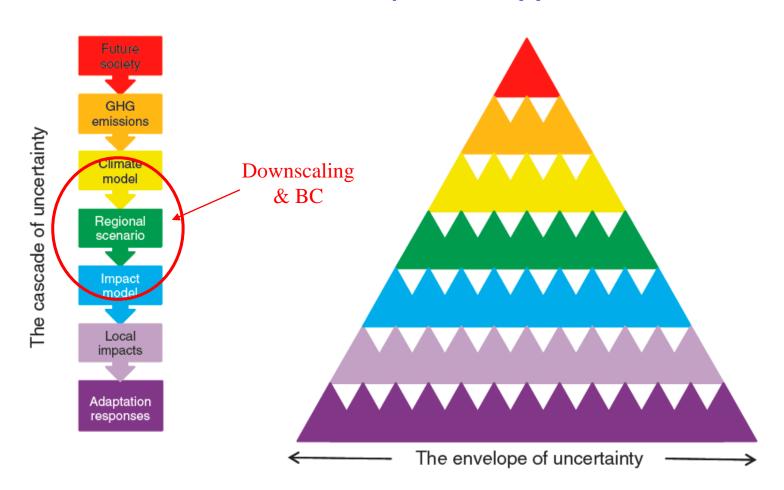
# Two types of post-processing methods

1. Statistical downscaling (Perfect prog)

2. Bias correction (Model Output Statistics)

Philosophy and principles, not all variants and methods !!

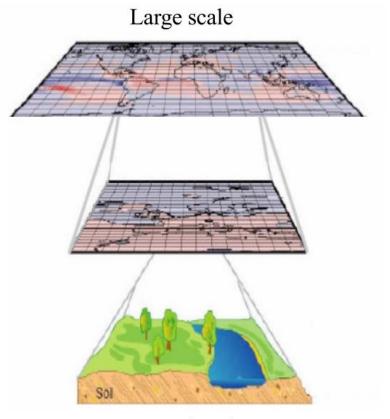
# Not that easy: some (!) uncertainties



# What is downscaling ???

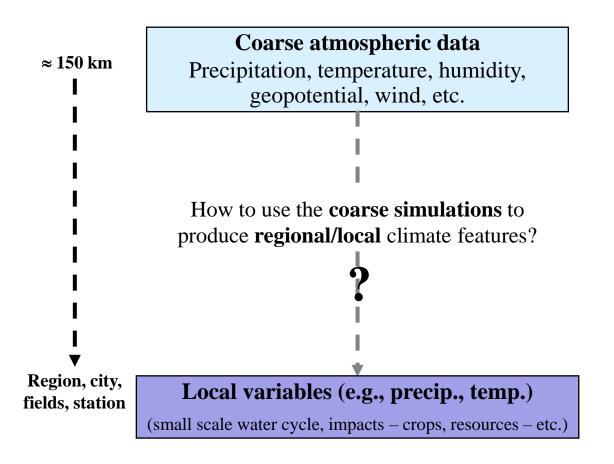
#### Definition:

Downscaling is the action of generating climatic or meteorological values and/or characteristics at a local scale, based on information (e.g., from GCM or reanalyses) given at a large scale.

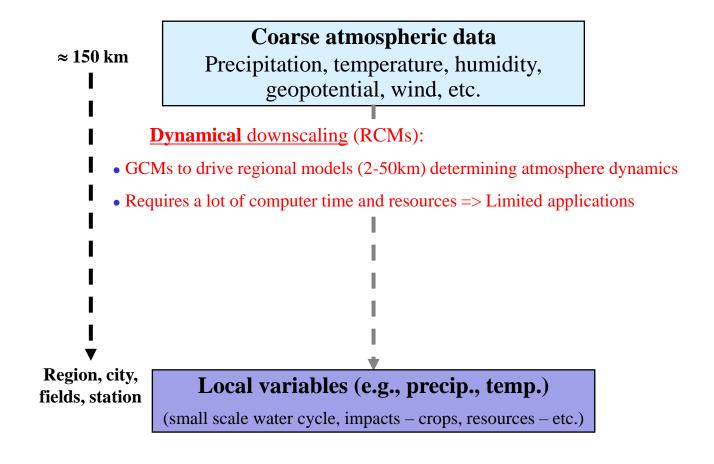


Local scale

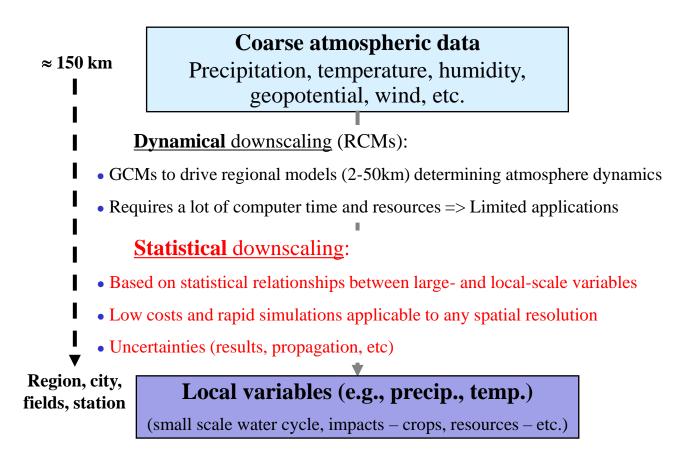
#### **How** to downscale?: The basics



#### **How** to downscale?: The basics



#### **How** to downscale?: The basics



# Perfect Prognosis

Trouver comment transformer une réalisation  $x_i$  en une réalisation  $y_i$ .

Un exemple de modèle classique

$$Y = f(X) + \epsilon$$

• approche par fonction de transfert / régression, on cherche à modéliser

$$\mathbb{E}[Y|X=x] = f(x)$$

 approche par générateur de temps stochastique, on cherche à modéliser la variable aléatoire

$$(Y|X=x)=f(x)+\epsilon$$

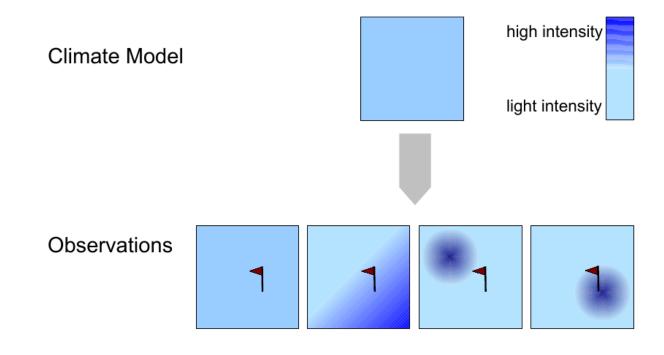
en essayant d'estimer la fontion f et de modéliser la distribution du bruit  $\epsilon$  .

Les predicteurs sont supposés parfaits.

On a besoin de l'appariement temporelles entre les variables  $X_i$  et  $Y_i$ .



#### One Grid Box State Several Local States



Grid box variability does not explain all local variability

Douglas Maraun Bias adjustment October 14, 2021 24 / 47

# Large-scale & Local-scale data

Large-scale data = Data to

downscale or BC

• Global Climate Model = General Circulation Model (GCM)

Computer code(s) simulating the main characteristics (pressure, temp., prec., etc.) of the Earth system

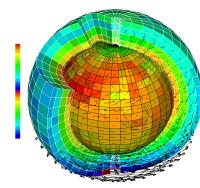
Based on physical equations

Local-scale data

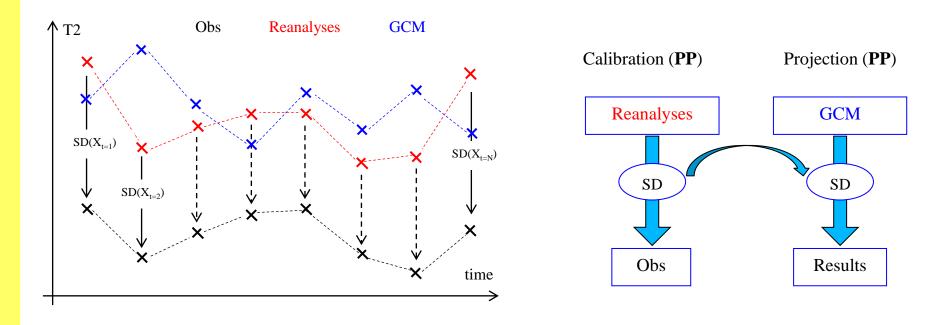
=
 « reference »
data to calibrate
 SD or BC
 methods

- Reanalyses
  - > Based on "data assimilation" approach
- Observations
  - weather stations / satellites / radar





# Perfect prognosis



Main PP hypotheses: GCM predictors have no bias & SD valid in proj. period

# Main (implicit) assumptions of SDM

#### For calibration under (near-) **present** climate:

- ightharpoonup A1: local scale = f(large scale, regional characteristics)
- > F1: We need local-scale data!!!

#### Using SDMs under **climate change**:

- > A2: Predictors are relevant and realistically modeled by GCM
- > A3: The predictors fully represent the climate change signal
- > A4: The SDM is valid also under altered climatic conditions

# TP Downscaling with linear regression and analogues

There are more complex methods, e.g.:

Deep learning for downscaling: https://github.com/paulaharder/deepdownscaling-overview