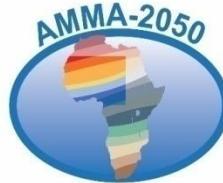




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Reduction of CMIP5 models bias using Cumulative Distribution Function transform and impact on crops yields simulations across West Africa

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Objectives

- We make this work in the framework of AMMA 2050 which is part of Future Climate for Africa (FCFA) program.
- One of the aims is to analyse climate projection of Africa around 2050 and impacts on agriculture and hydrology.
This is based on GCMs simulations that still have large biases.
- So bias correction is necessary for Agronomical and hydrological use.
This has been done over Africa with:

GCMs: 29 CMIP5 GCMs

- ❖ Interpolate to WFDEI grid;
- ❖ Period of covering:
 - Present period: 1950-2005
 - Future period: 2006-2099
- ❖ Historical, RCP2.6, RCP4.5 and RCP8.5

Reference data: WFDEI (new version of WFD)

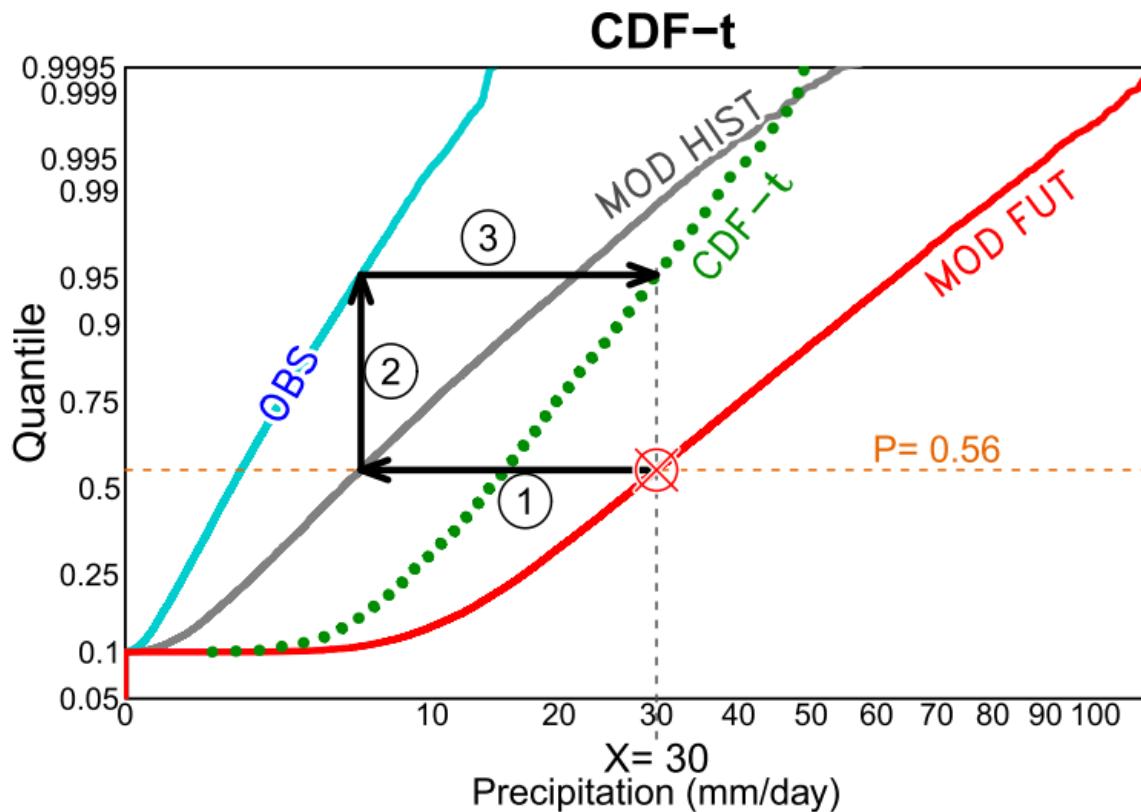
- ❖ Grid: $0.5^\circ \times 0.5^\circ$
- ❖ Period of covering: 1979-2013
- ❖ Variables: tas, tasmin, tasmax, pr, rsds and wind

- Use of this bias correction on one GCM to predict crops yields over West Africa.

- ✓ Validation of bias correction method using IPSL-CM5A-LR;
- ✓ Impact on crops yields using Sarra-H.

Cumulative Distribution Function-transform (CDF-t)

- CDFt is developed by Michelangeli et al. (2009)

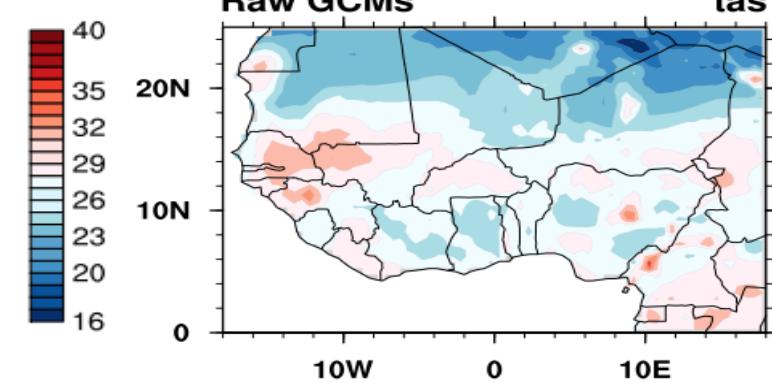
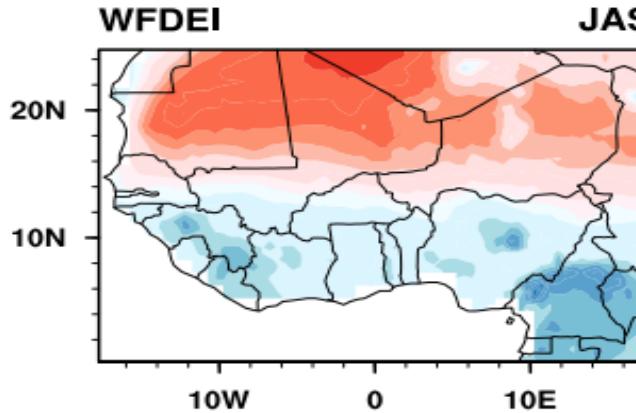


- CDFt aims to relate the CDF of climate variable at large scale to the CDF of this variable at locale scale.

- CDFt has been widely used in many studies over differents regions (Vrac et al., 2012, Vrac et al., 2016; Oettli et al., 2011; Kallache et al., 2011).

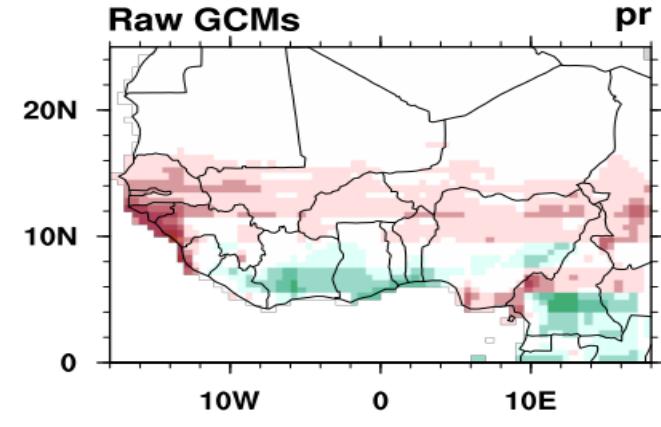
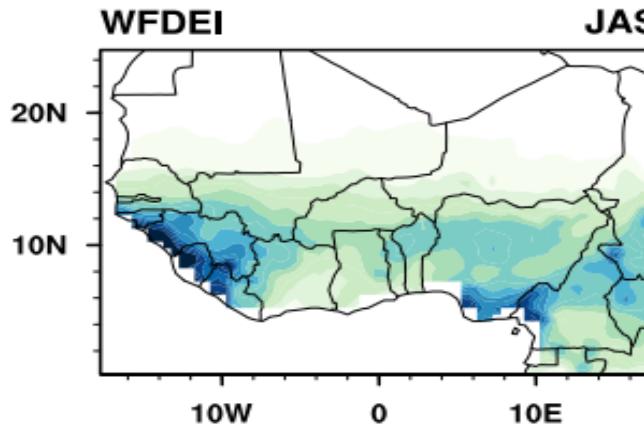
Biases GCM-IPSL / WFDEI JAS

tas



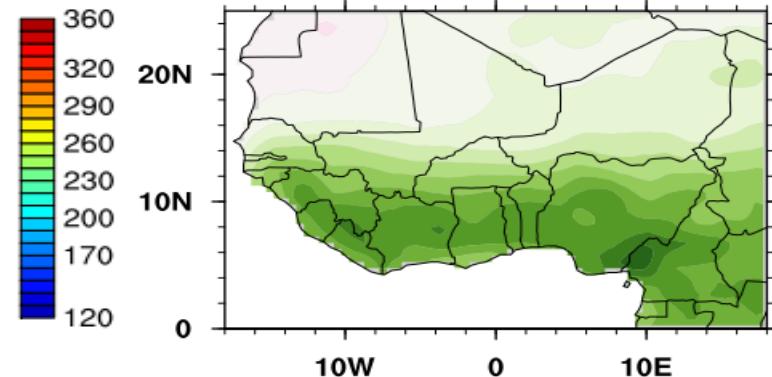
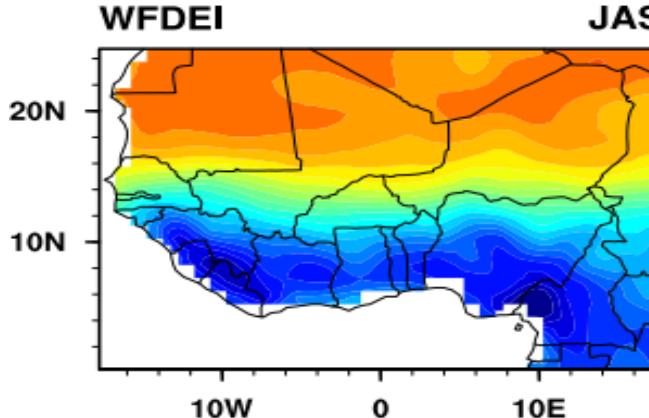
6
5
4
3
2
1
0
-1
-2
-3
-4
-5
-6

pr



20
16
12
8
4
0
-4
-8
-12
-16
-20

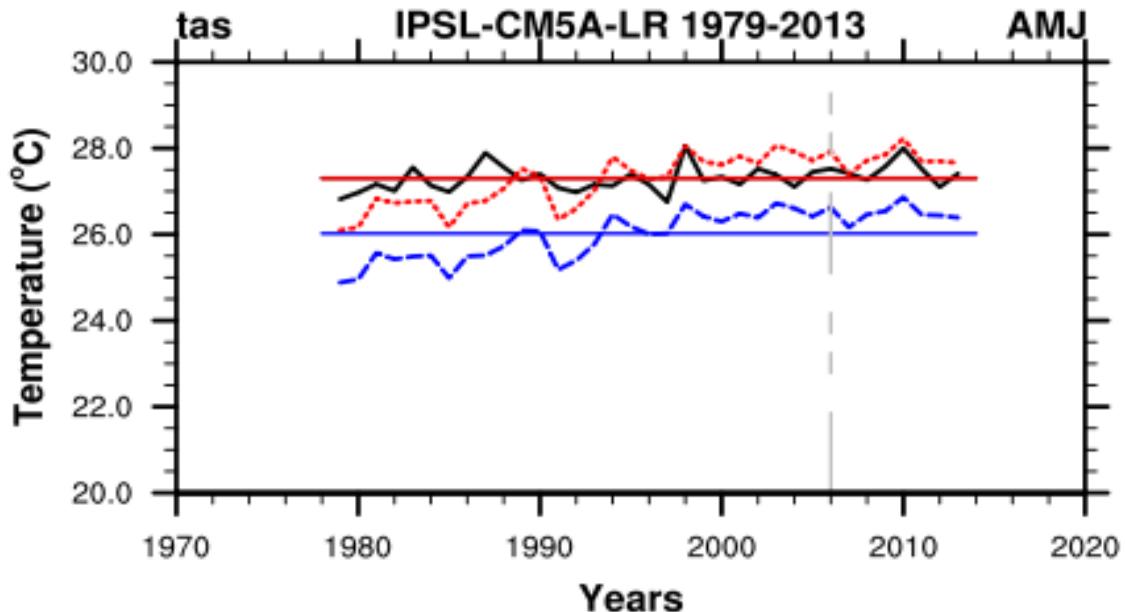
rsds



160
120
80
40
0
-40
-80
-120
-160

Raw vs BC IPSL data AMJ: Temperature

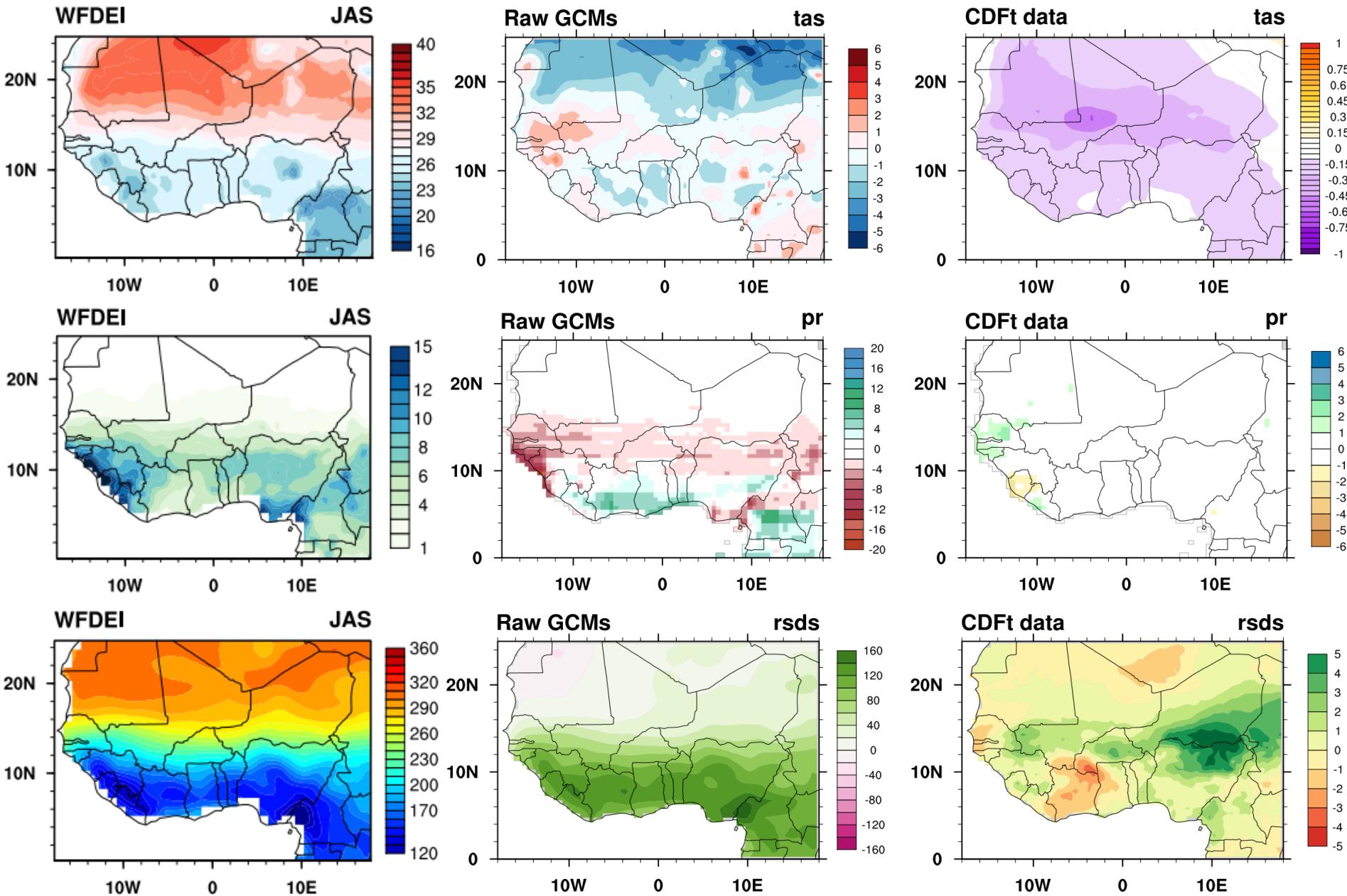
Guinean coast AMJ



Mean bias

	Calibration
IPSL (Raw)	- 1.29
IPSL (CDFt)	-0.01

Raw vs BC IPSL data JAS



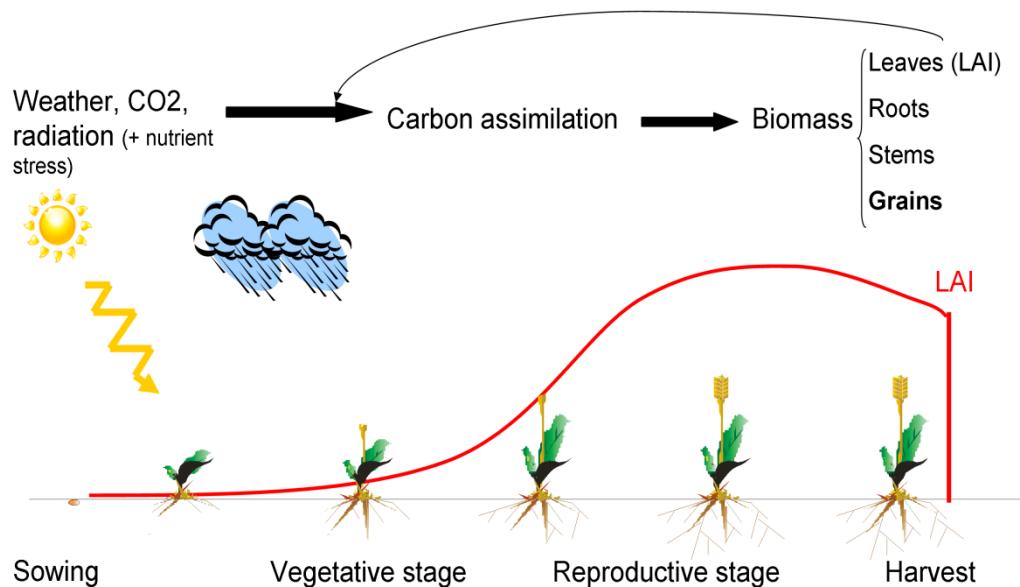
The Crop model Sarra-H

SarraH

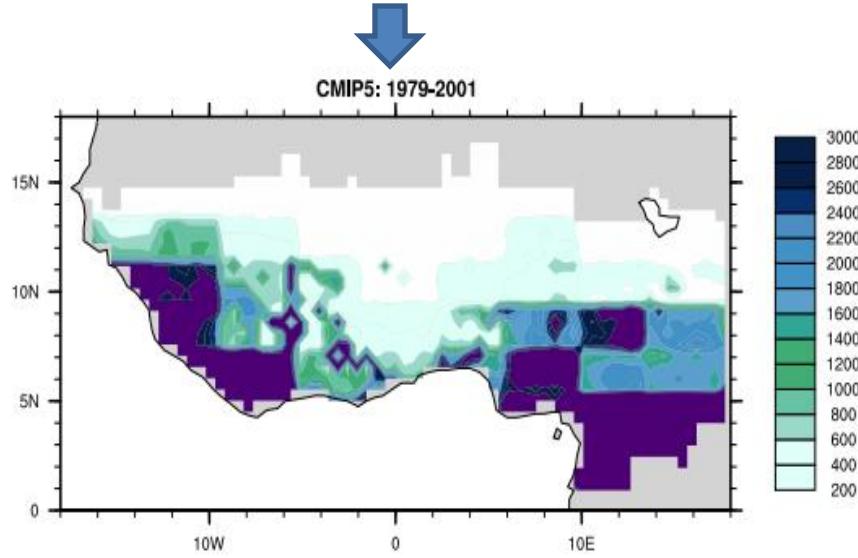
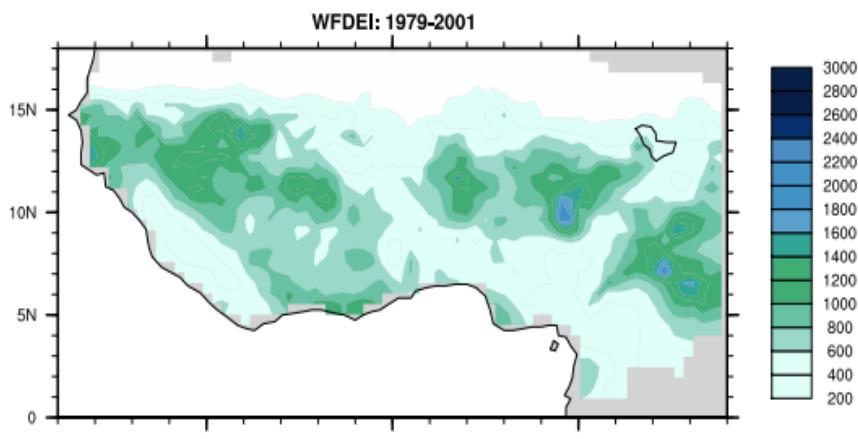


Dingkuhn et al. (2003)

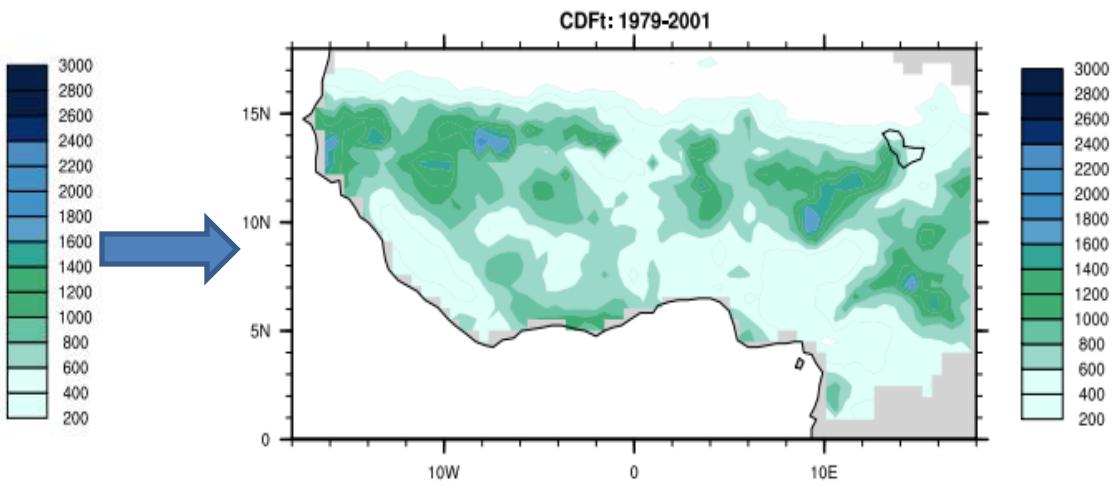
- **Water balance model** (water demand, soil water availability, stress index, critical phases)
- **Carbon assimilation and partitioning** (radiation use efficiency, phenology, evolution of biomass and LAI)



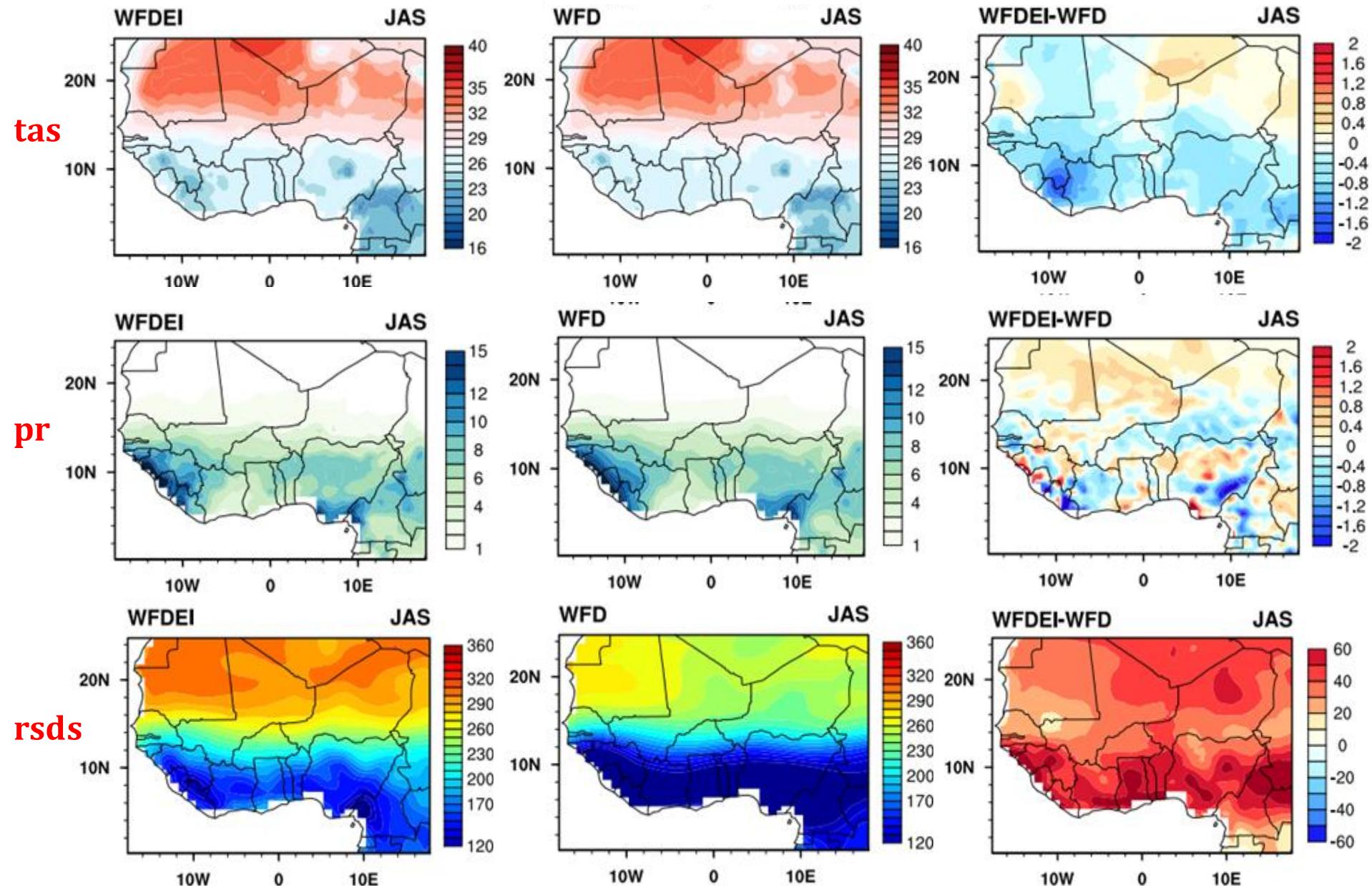
Crop yield mean: 1979-2001



□ Simulation driven by CDFt data produce better crop yield than those driven by raw GCMs data.

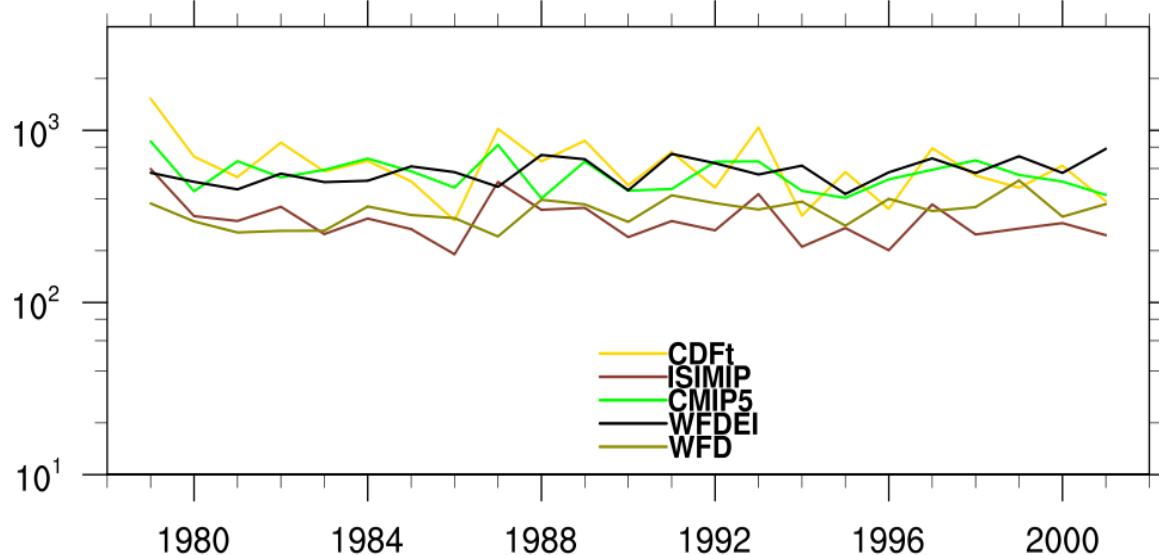


JAS Climatology WFDEI vs WFD: 1979-2001



Crop yield Maize: Present day 1979-2001

SAHEL

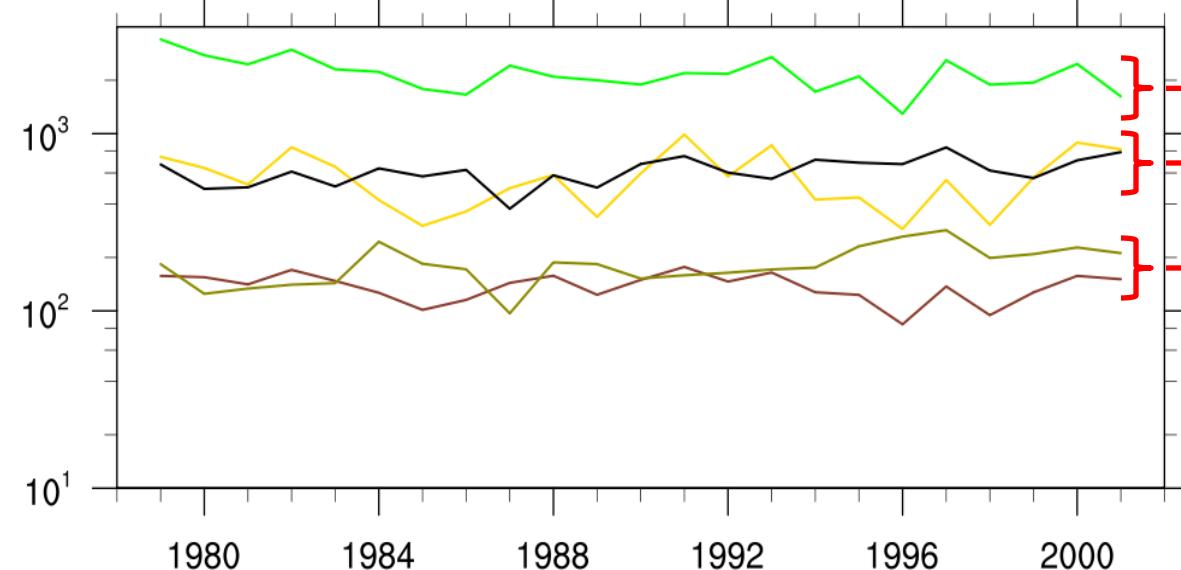


Bias sensitivity experiments

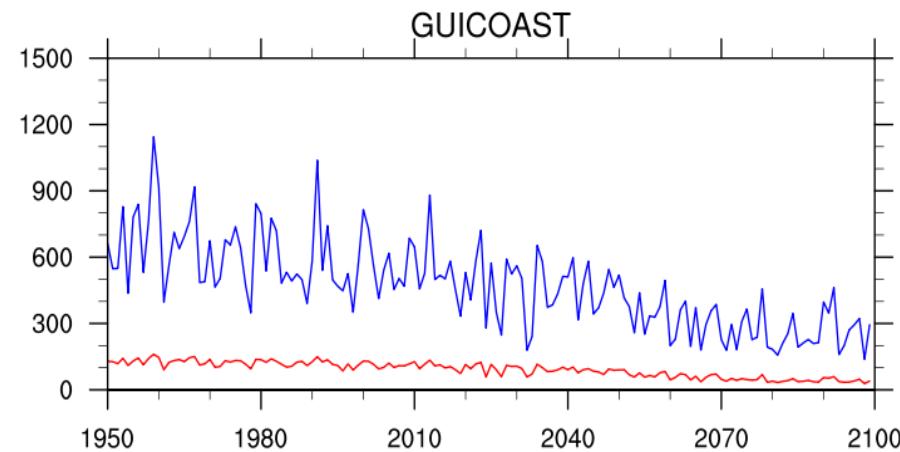
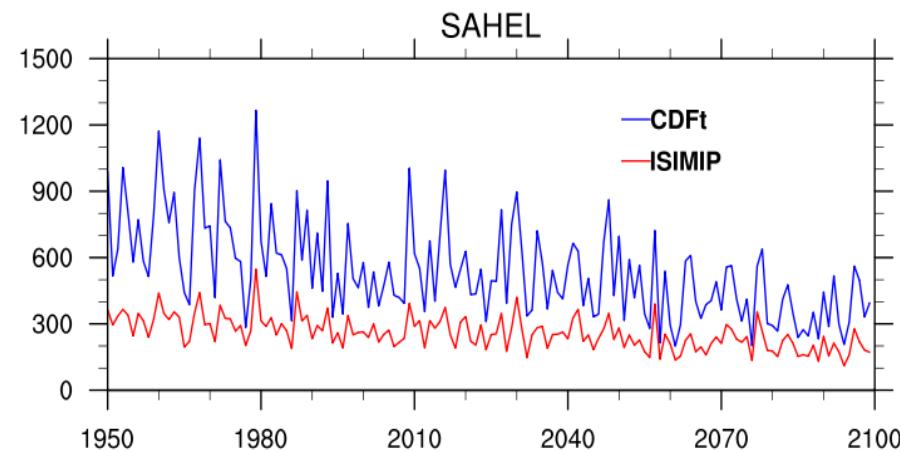
	SAHEL	GUICOAST
WFDEIetp	21.06	23.75
WFDEIpr	29.23	-29.09
WFDEIrsds	48.61	40.74
WFDEItminmax	1.94	9.04
WFDEIWFDrsds	-236.84	-462.17
WFD	-187.23	-425.42



GUICOAST



Crop yield Maize: Projections



Variance

	SAHEL		
	tas	pr	rsds
WFDEI	0.09	0.03	2.46
WFD	0.09	0.03	3.30
CMIP5	0.27	0.01	3.08
CDFt	0.29	0.06	4.23
ISIMIP	0.28	0.04	2.05

Variance

	GUINEAN COAST		
	tas	pr	rsds
WFDEI	0.05	0.06	5.35
WFD	0.05	0.08	10.68
CMIP5	0.15	0.03	3.15
CDFt	0.21	0.17	12.02
ISIMIP	0.18	0.05	0.77

CONCLUSION

Results

- ❑ Bias correction performed over Africa on 29 GCMs (1950-2099).
 - ✓ CDF-t corrects well mean state of variables ;
 - ✓ CDF-t preserves models trend;
 - ✓ CDF-t improves crop yield simulation;
- ❑ Crop model is sensitve to input data mainly to surface radiation;
- ❑ Crop model shows decreasing crop yields over West Africa (RCP8.5)

On going works

- ❑ Crop yield simulation with Sarra-H driven by 29 GCMs (historical,RCP2.6, RCP4.5 and RCP8.5)
- ❑ Explore the question of time of emergence with BC Data;
- ❑ ISIMIP2 project which use new version of data: EWEMBI (Lange et al., 2016)

Thank you!