A large, faint watermark of the International Research Institute for Climate and Society (IRI) logo is visible on the left side of the slide. It consists of the letters 'IRI' in a serif font, enclosed within a circular emblem.

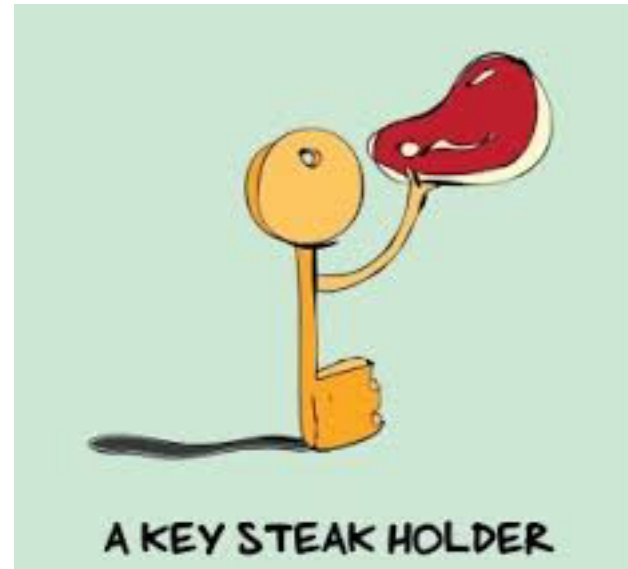
Connecting Prediction Information and Products to Applications

Lisa Goddard

International Research Institute
for Climate and Society
EARTH INSTITUTE | COLUMBIA UNIVERSITY

Thanks to Nachiketa Acharya, Angel Munoz,
Andy Robertson, and Nicolas Vigaud

Not all steakholders are the same

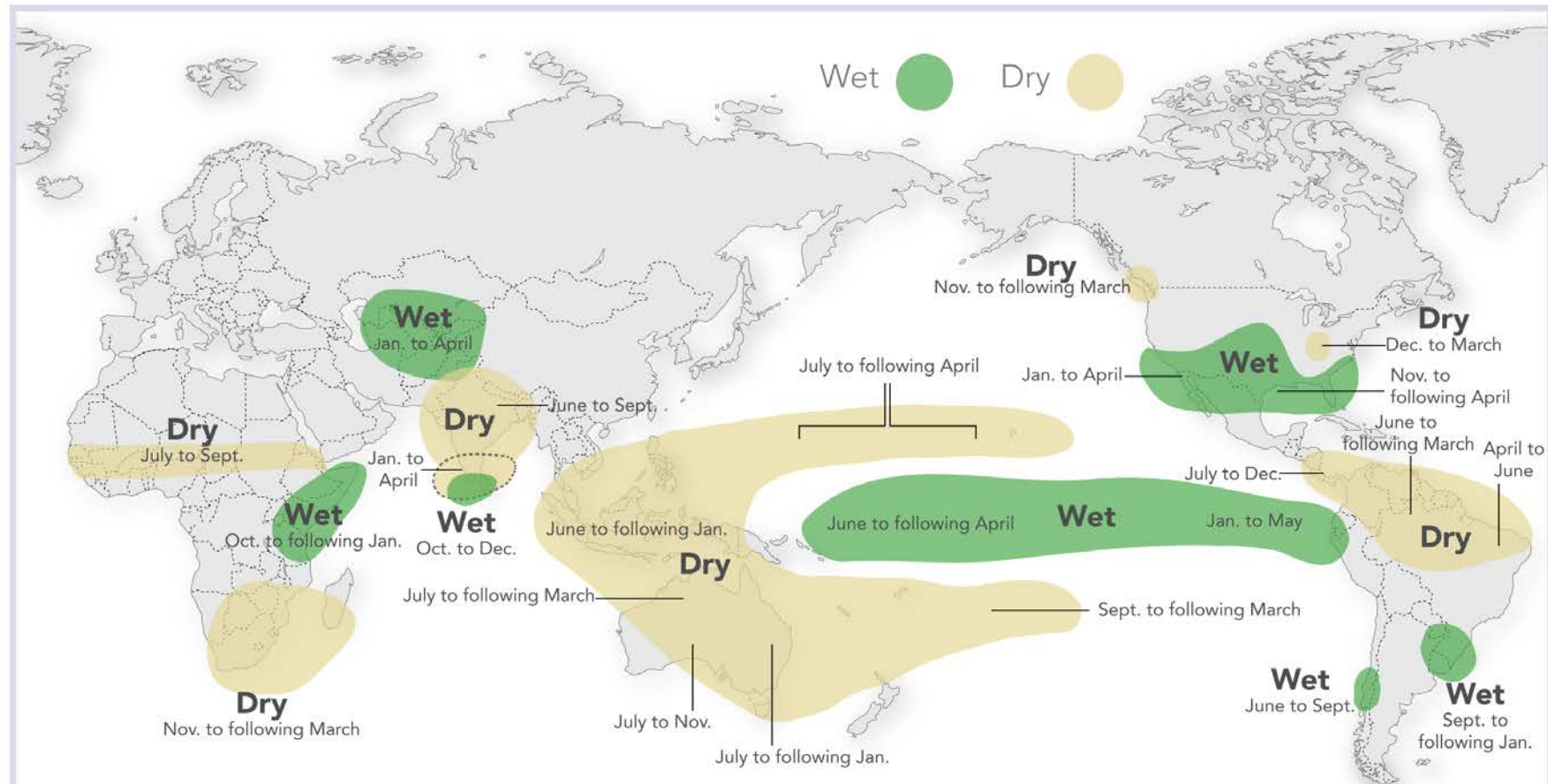


(Courtesy of Jim Buizer, UA)

“Expected” Climate Impacts During El Niño

El Niño and Rainfall

El Niño conditions in the tropical Pacific are known to shift rainfall patterns in many different parts of the world. Although they vary somewhat from one El Niño to the next, the strongest shifts remain fairly consistent in the regions and seasons shown on the map below.



For more information on El Niño and La Niña, go to: <http://iri.columbia.edu/enso>

Sources:

1. Ropelewski, C. F., and M. S. Halpert, 1987: Global and regional scale precipitation patterns associated with the El Niño Southern Oscillation. Mon. Wea. Rev., 115, 1606-1626;
2. Mason and Goddard, 2001. Probabilistic precipitation anomalies associated with ENSO. Bull. Am. Meteorol. Soc. 82, 619-638.

<http://iri.columbia.edu/ENSO/>



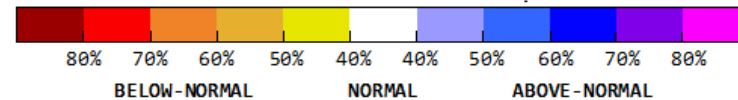
f /climatesociety @climatesociety

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Expected Flood Risk during an El Niño event

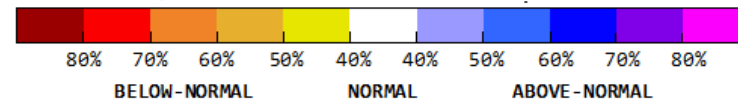
El Niño (SON)

SHADED AREA ARE PROBABILITY
BELOW AND ABOVE NORMAL
UNSHADED AREAS ARE NEAR-NORMAL
■ ARID AREAS (NO VALUE)



El Niño (DJF)

SHADED AREA ARE PROBABILITY
BELOW AND ABOVE NORMAL
UNSHADED AREAS ARE NEAR-NORMAL
■ ARID AREAS (NO VALUE)



FAO: Early Warning – Early Action

SOMALIA EXAMPLE: RESULTS

Situation Analysis



- El Niño 2015: Risk of floods affecting more than 90 000 people and 9100 ha of farmland along the Shebelle and Juba Rivers.

Example FAO Actions:

- strengthen riverbanks
- build flood barriers

Return on Investment

- Investment: 1.7m USD
- approx. 6.7m USD in maize production saved, 4 x the investment



Humanitarian Health Action

[Humanitarian Health Action](#)
[Crises](#)
[Technical guidelines](#)
[Key policy documents](#)
[Donor centre](#)
[Global network](#)
[Global Health Cluster](#)

Health preparedness for El Niño event 2015 - 2016

Global Situation Assessment of El Niño-related Health Risks

WMO has confirmed reports that a mature and strong El Niño is present in the tropical Pacific Ocean. International climate models suggest that the 2015/2016 El Niño will strengthen further before the end of 2015 and become one of the strongest El Niño events in the past twenty years.

The El Niño phenomenon is a major concern to global public health as it has the potential to exacerbate health risks associated with extreme weather in different parts of the world. The last major event of similar magnitude in 1997-1998 resulted in significant health impacts across multiple WHO Regions [WHO Factsheet : El Niño Fact sheet on the El Niño Southern Oscillation (ENSO) and health UPDATE OF NO.192 (2000)], for example:

- in Eastern Africa, extensive flooding led to food insecurity associated with widespread economic losses in agriculture and livestock, as well significant destruction of health infrastructure throughout the region;
- in Djibouti, Kenya, Mozambique, Somalia, Uganda and the United Republic of Tanzania unusually high rainfall triggered major outbreaks of cholera in, as well as unexpectedly large outbreaks of Rift Valley Fever (RVF) in Kenya, Somalia and Tanzania, for example, in Garissa District, Kenya (population 231,022) alone, there were 170 deaths from haemorrhagic fever and an estimated 27,500 infections;
- in Latin America, Peru and central Ecuador experienced rainfall more than 10 times the usual levels, with consequential flooding and extensive landslides that resulted in widespread destruction of homes and infrastructure;
- the Pacific Island region was hit by a drought so severe that 50% of Fiji's population required government water delivery, and the sugarcane harvest was slashed by 50% with widespread economic impacts.

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Related links

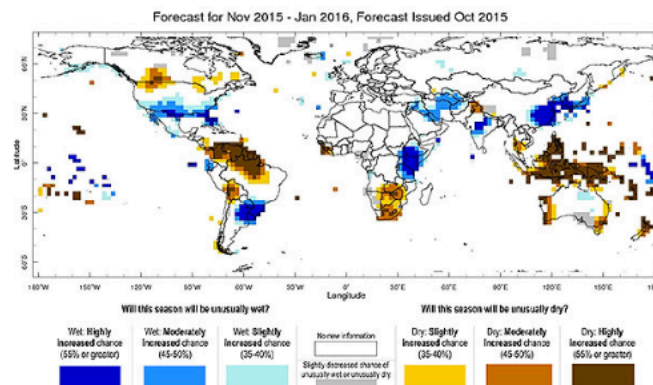
↓ Read this page in PDF format
pdf, 360kb

IGAD Climate Prediction and Applications Centre Monthly Bulletin, May 2015 October 2015 [↗](#)

Estado actual de las condiciones del Océano Pacífico Y su posible evolución durante el año 2015-2016. [↗](#)

Rift Valley Fever factsheet

WHO pre-positions emergency supplies in Somalia in preparation for El Niño [↗](#)



World Health Organization Reports
<http://www.who.int/hac/crises/el-nino/en/>

EL NIÑO EFFECTS AND HEALTH CONSEQUENCES



What Ministries of Health and WHO are doing for country and regional preparedness for El Niño

WHO is acutely aware of the high risk conditions of 2015 and providing support to WHO Member States and partners to enhance preparedness measures for the current El Niño event. The health sector should prepare for the specific risks associated with the El Niño event. This event also provides the opportunity to reinforce all-hazards preparedness of communities and countries, and the readiness of WHO, the Global Health Cluster and other national, regional and global actors for health emergencies.

EXAMPLES

Seasonal-to-Interannual (Seas):

1. [Philippines \(World Food Program\)](#): Food Security preparedness
2. [Ethiopia](#): Water resource management & hydro-power

Sub-Seasonal (SS):

1. [Colombia](#): Water resource management & hydro-power
2. [Bihar, India](#): Agriculture

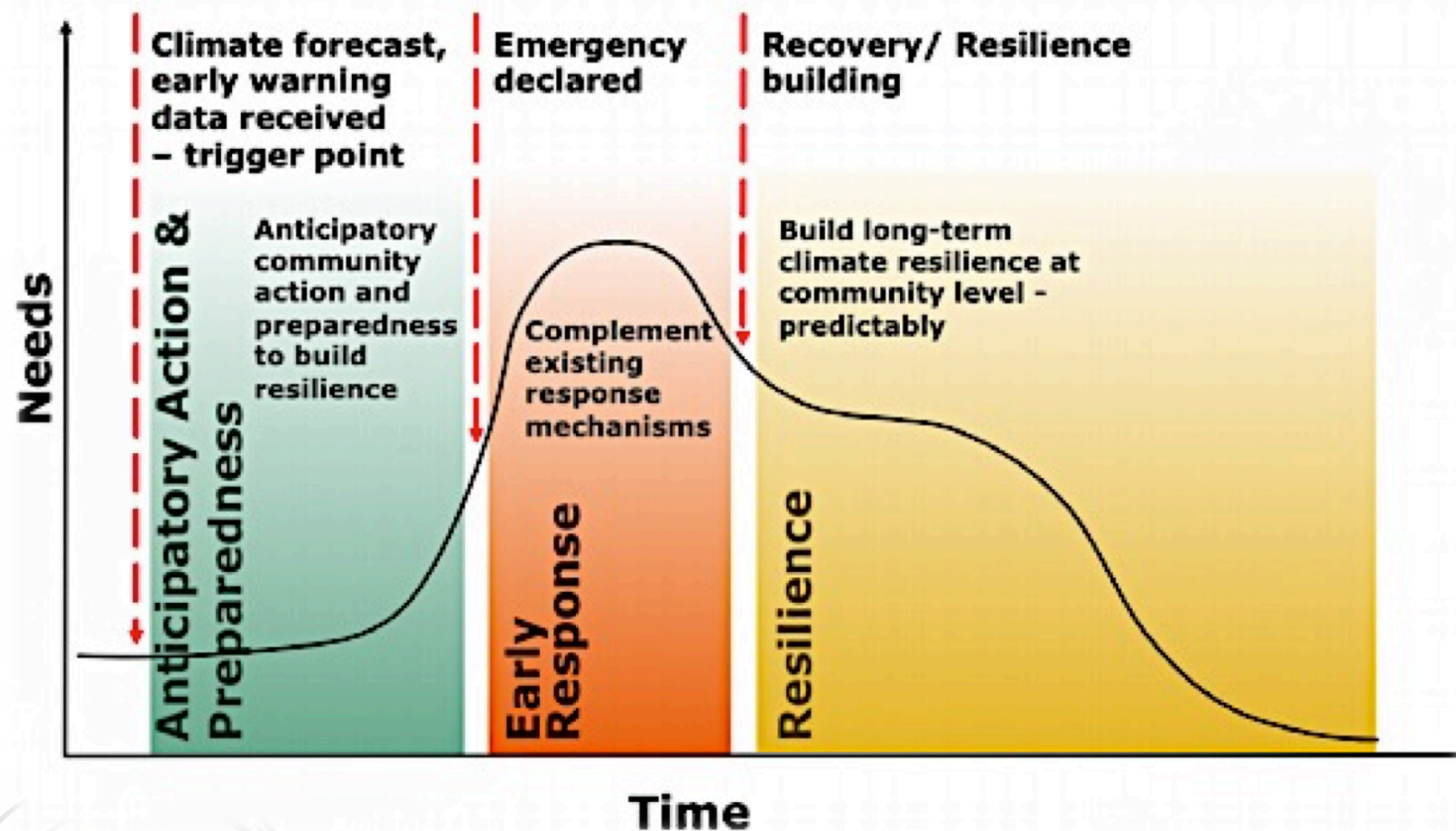
Seasonal-to-Decadal (Dec):

1. [South Africa \(western Cape\) Water Mgmt](#): System resilience



WFP's Food SECuRE

Pioneering Example of Forecast-based Financing

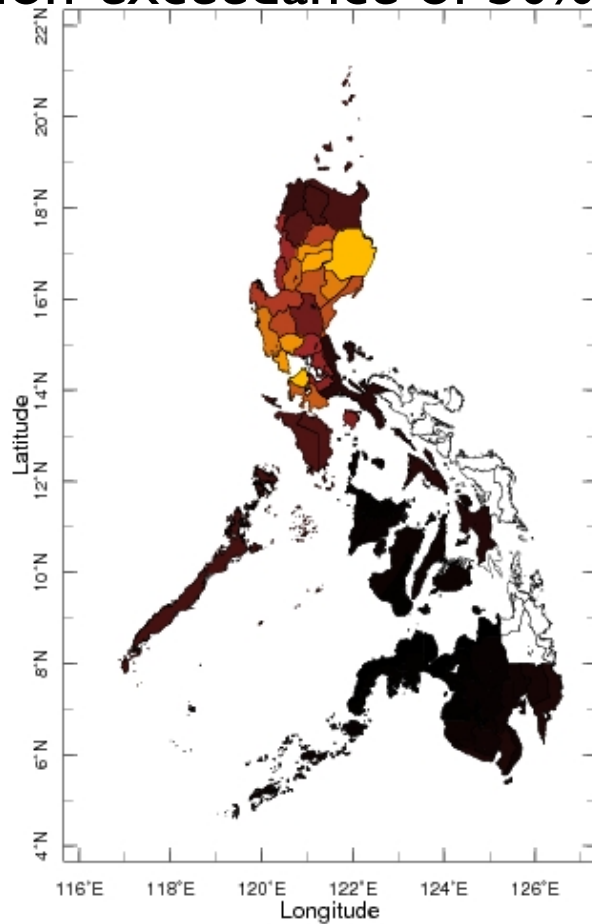


Seas-I

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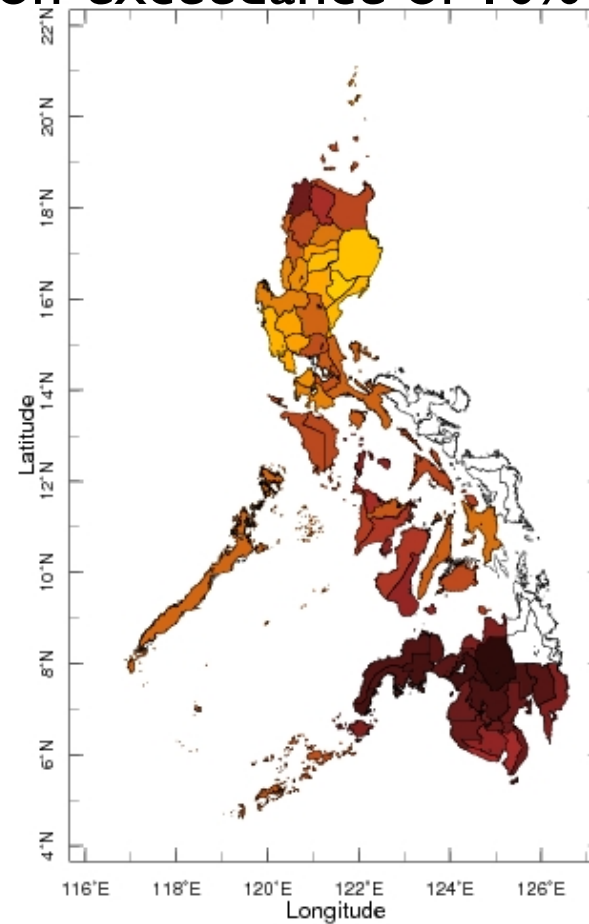
WFP's Food SECuRE – Drought Triggers

Non-exceedance of 50%-ile



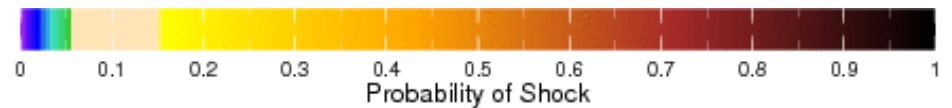
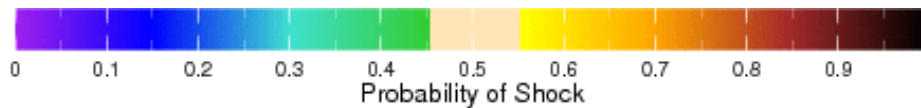
Jan-Apr 2016 Forecast issued September 2015

Non-exceedance of 10%-ile



Jan-Apr 2016 Forecast issued September 2015

Made
Sept.
2015

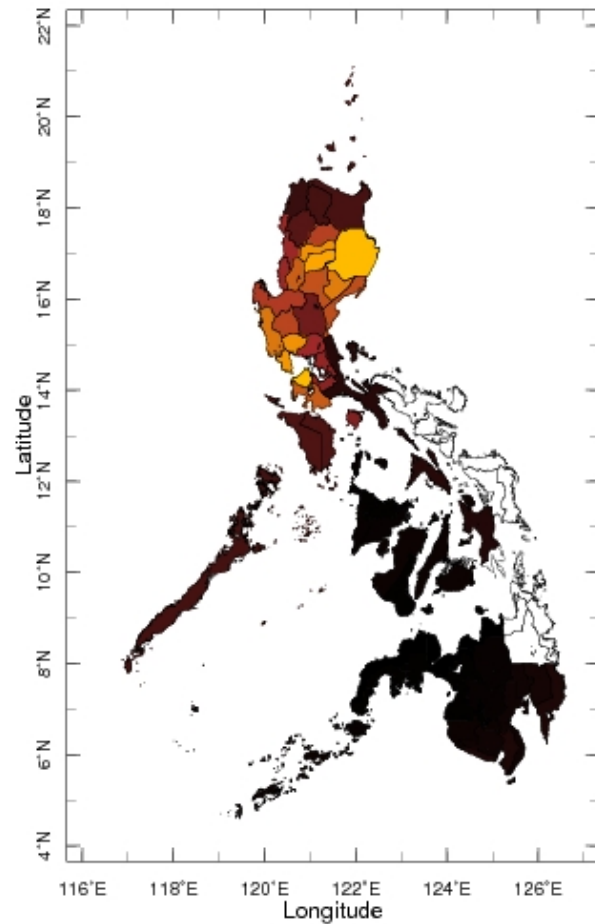


Seas-I

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WFP's Food SECuRE – Drought Triggers

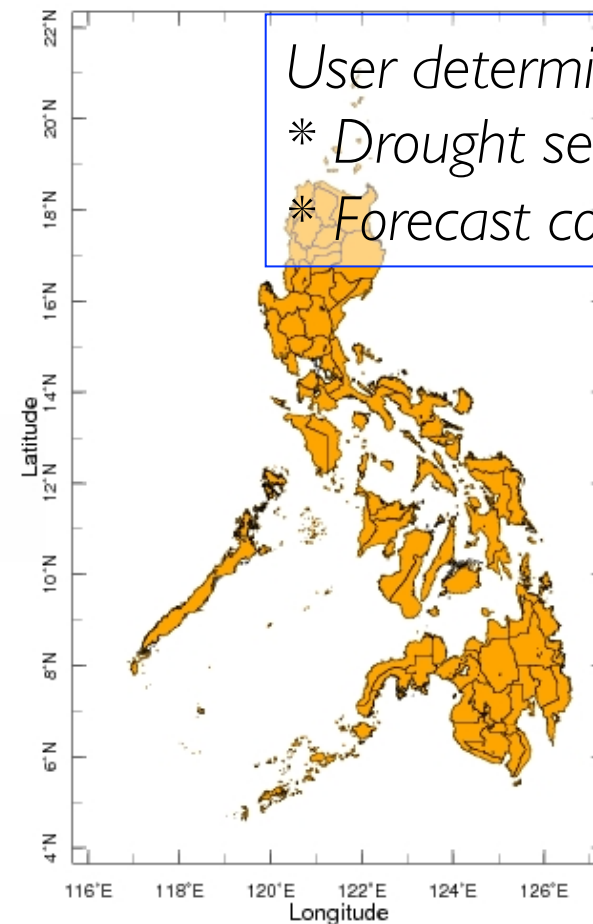
Non-exceedance of 50%-ile



Made
Sept.
2015

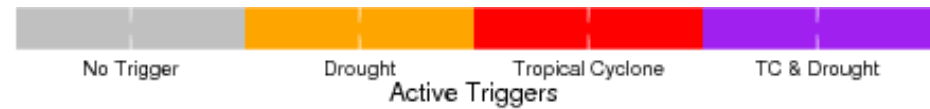


“ACTIVE TRIGGERS”



User determines:
* Drought severity
* Forecast confidence

Jan-Apr 2016 Forecast issued September 2015

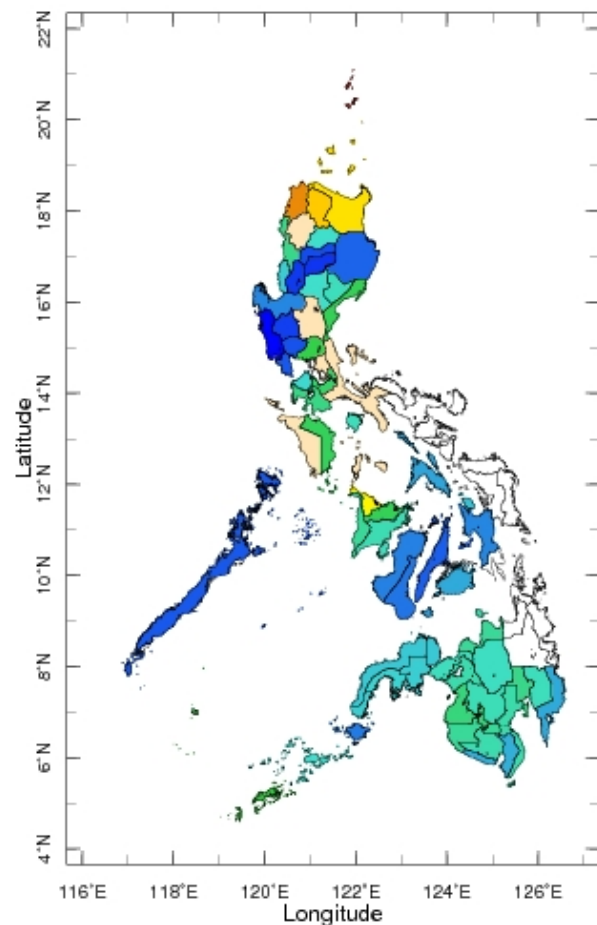


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WFP's Food SECuRE – Drought Triggers

Non-exceedance of 50%-ile



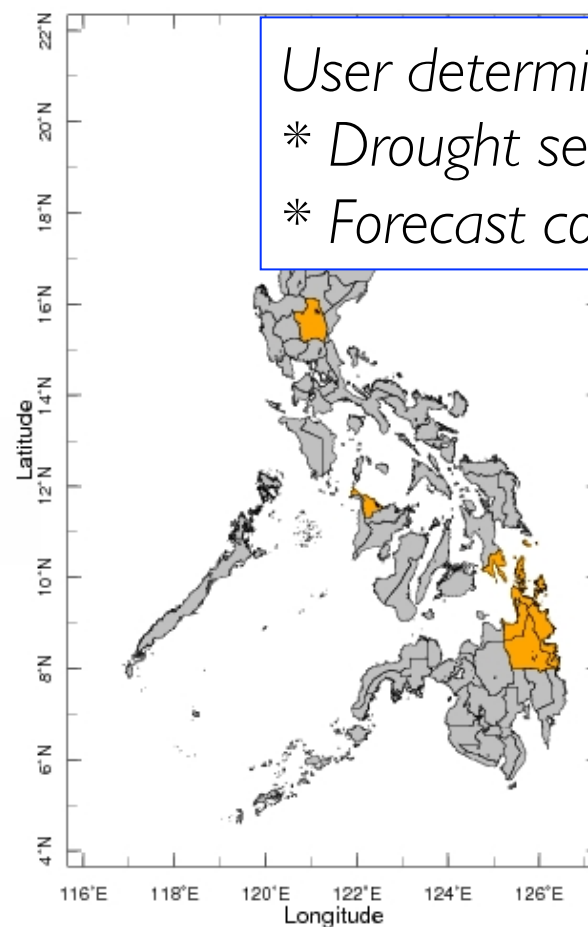
Made
Sept.
2016



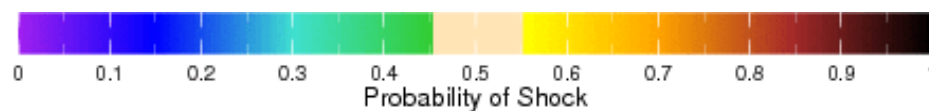
“ACTIVE TRIGGERS”

User determines:

- * Drought severity
- * Forecast confidence



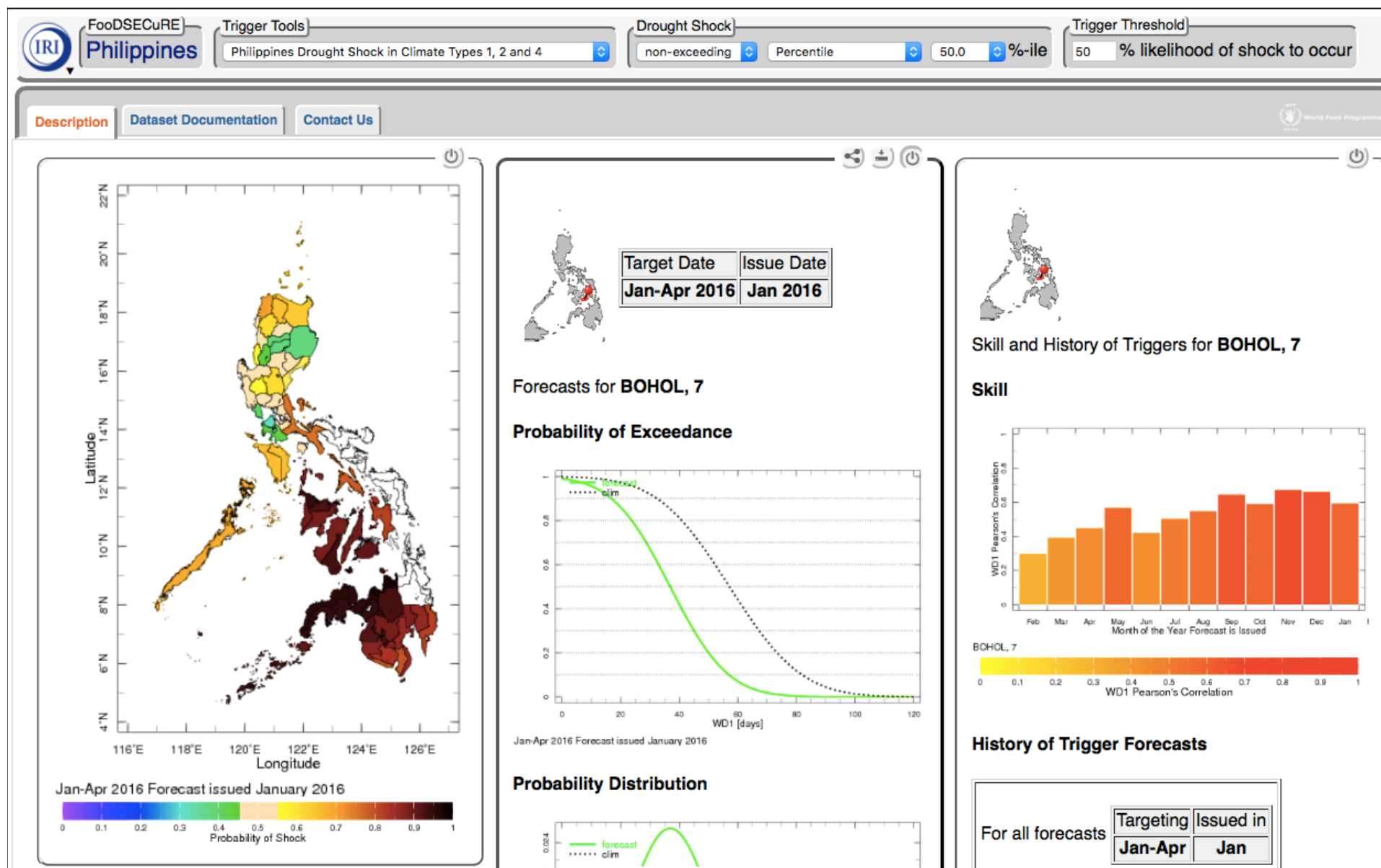
Jan-Apr 2017 Forecast issued September 2016



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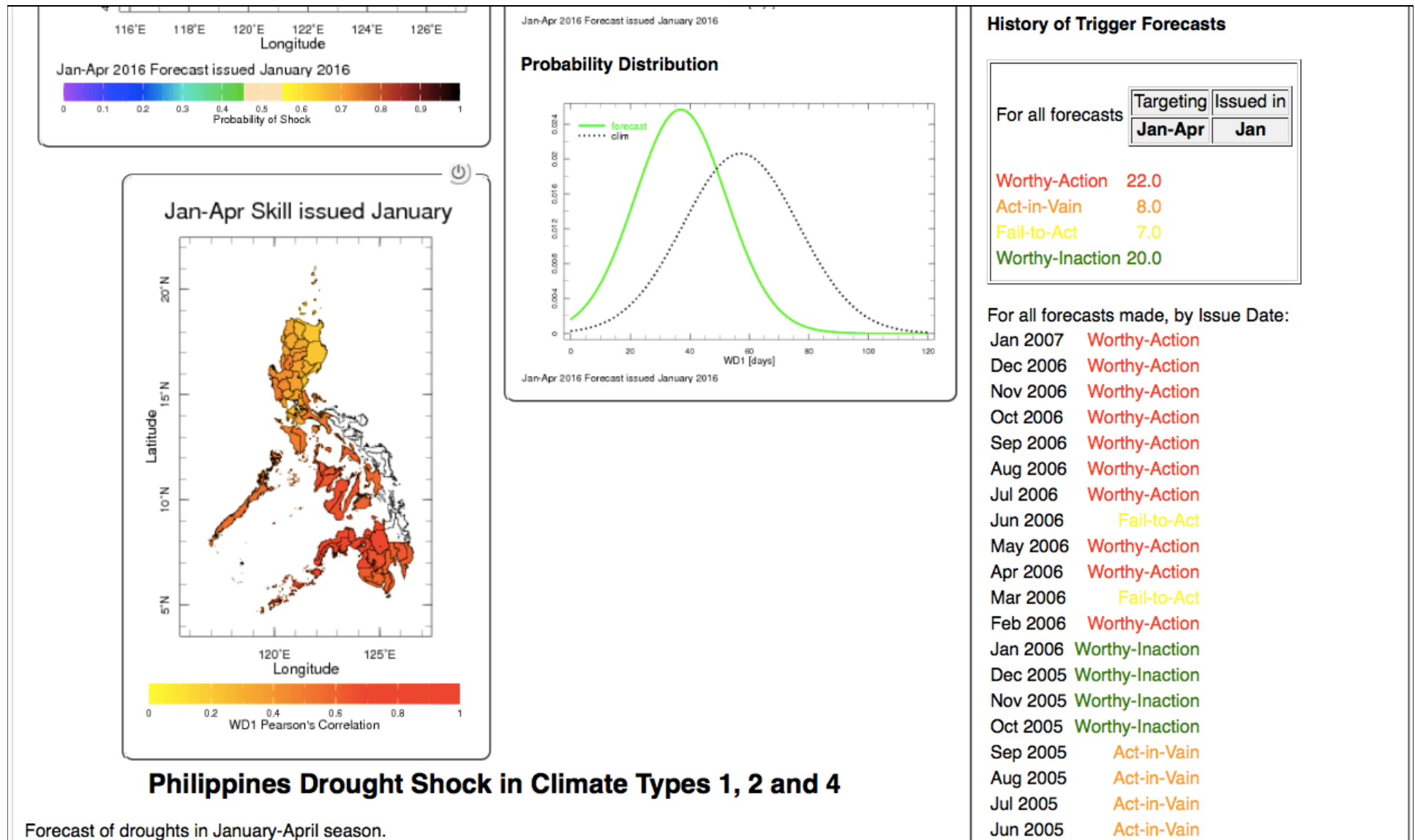
WFP's Food SECuRE – District Dashboard



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WFP's Food SECuRE – District Dashboard



ETHIOPIA — Water resource management & hydro-power

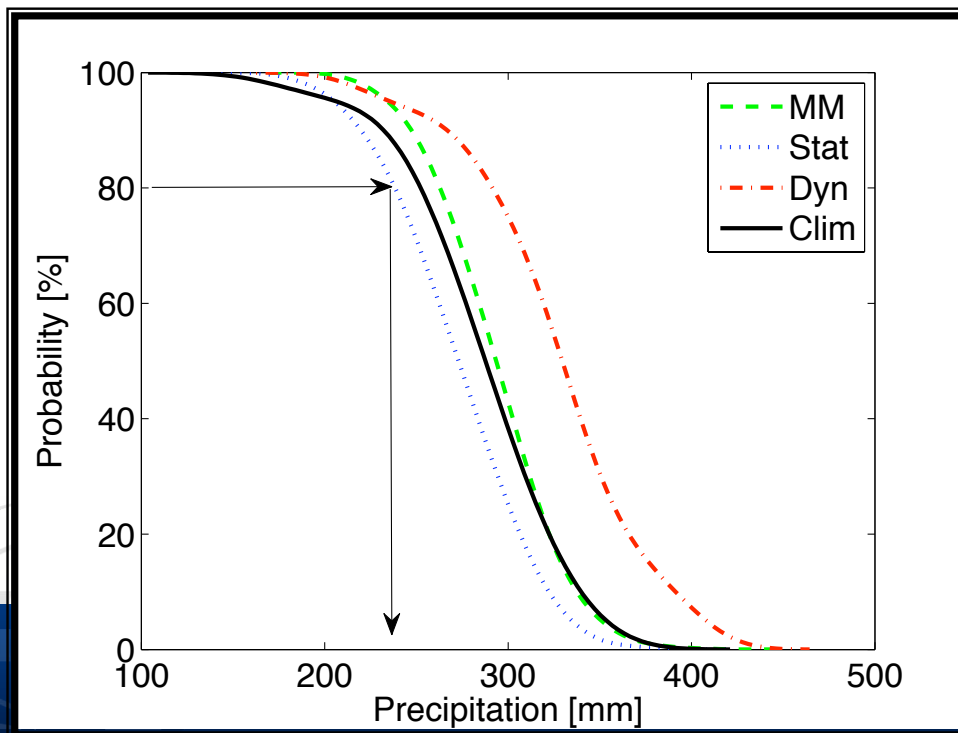
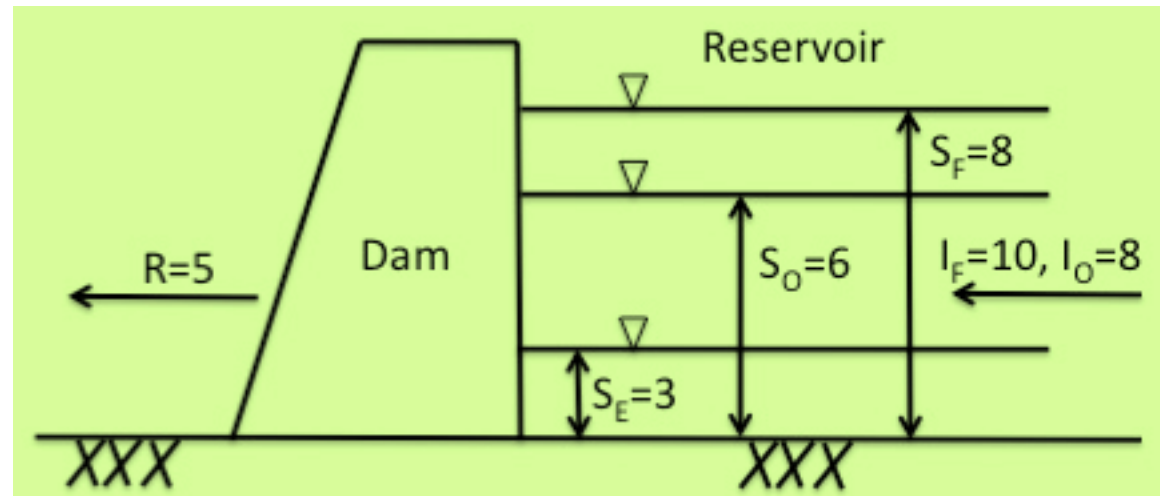


Seas-2

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Reservoir Management: Linked Model System

Monthly reservoir update
based on forecast and
observations



Select precipitation
exceedance probability,
apply to prediction
ensemble

(Block and Goddard, 2012, WRR)

Hydropower Benefits w/ different forecasts

Using probabilities of exceedance; threshold reliability = 200 GW hrs/mo

Forecast Technique	Mean Benefits [\$M/dec]	Mean Losses [\$M/dec (f)]	Dependability [%]	Superior Benefit Cases	
				Decadal [%]	Annual [%]
Perfect Forecast	3350	-	100%		
<i>prob = 80%</i>					
Statistical	2740	25 (10%)	100%	18	23
Dynamical	2610	100 (66%)	100%	0	15
Multi-model	2780	5 (2%)	100%	82	35
No Forecast	2610	-	>99%	0	27
<i>prob = 95%</i>					
Statistical	2500	35 (23%)	100%	8	20
Dynamical	2200	237 (100%)	100%	0	0
Multi-model	2550	48 (5%)	100%	88	44
No Forecast	2440	-	100%	4	36

For prob=50%: benefits and reliability drop

Hydropower Benefits w/ different forecasts

Using probabilities of exceedance; threshold reliability = 200 GW hrs/mo

Forecast Technique	Mean Benefits [\$M/dec]	Mean Losses [\$M/dec (f)]	Dependability [%]	Superior Benefit Cases Decadal [%] Annual [%]	
Perfect Forecast	3350	-	100%		
<i>prob = 80%</i>					
Statistical	2740	25 (10%)	100%	18	23
Dynamical	2610	100 (66%)	100%	0	15
Multi-model	2780	5 (2%)	100%	82	35
No Forecast	2610	-	>99%	0	27
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Statistical	2500	35 (23%)	100%	8	20
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Multi-model	2550	48 (5%)	100%	88	44
No Forecast	2440	-	100%	4	36

For prob=50%: benefits and reliability drop

Sub-Seasonal: weeks 1-4



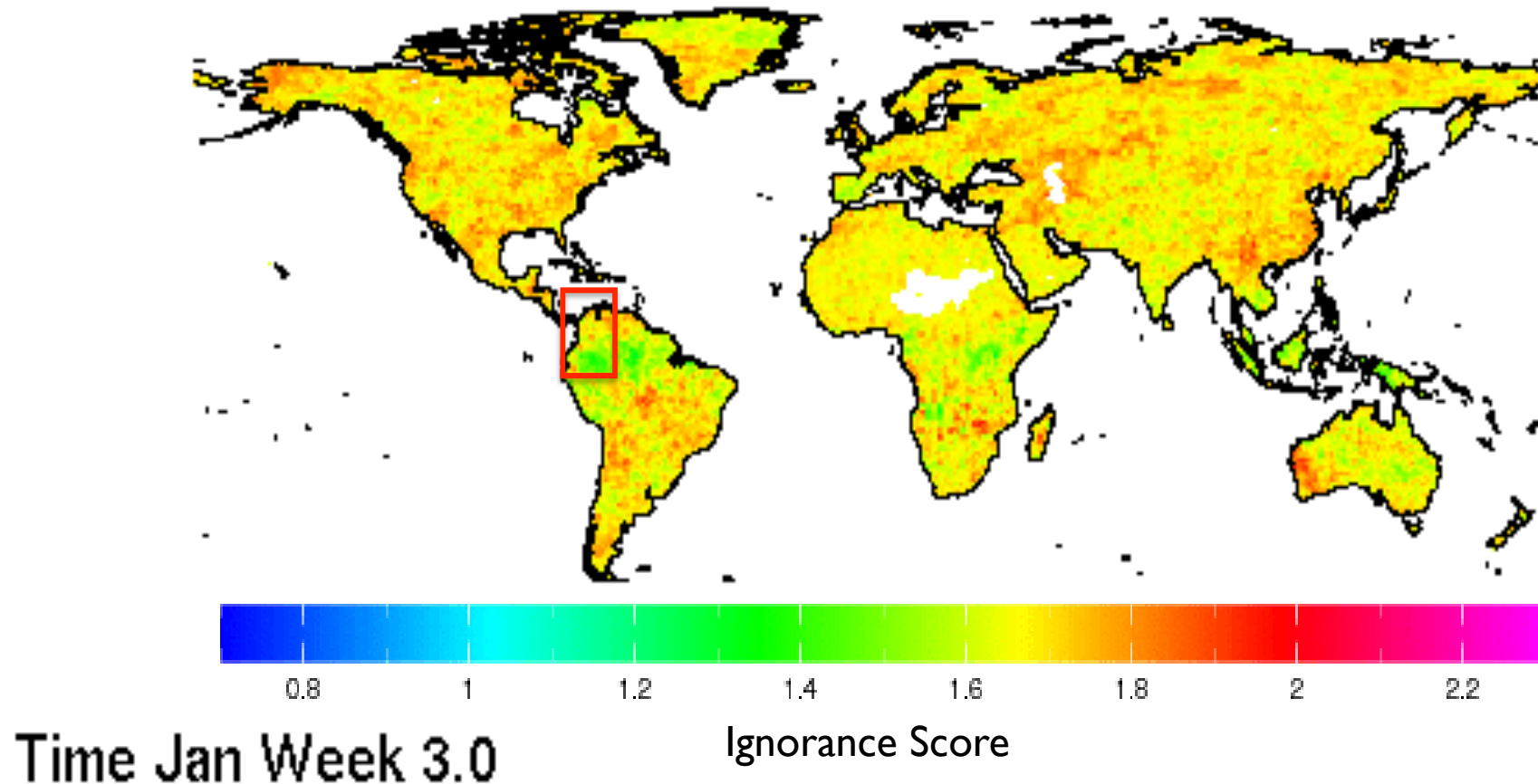
COLOMBIA – Rainfall forecasts for hydro-power



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Seasonality of Sub-seasonal Skill



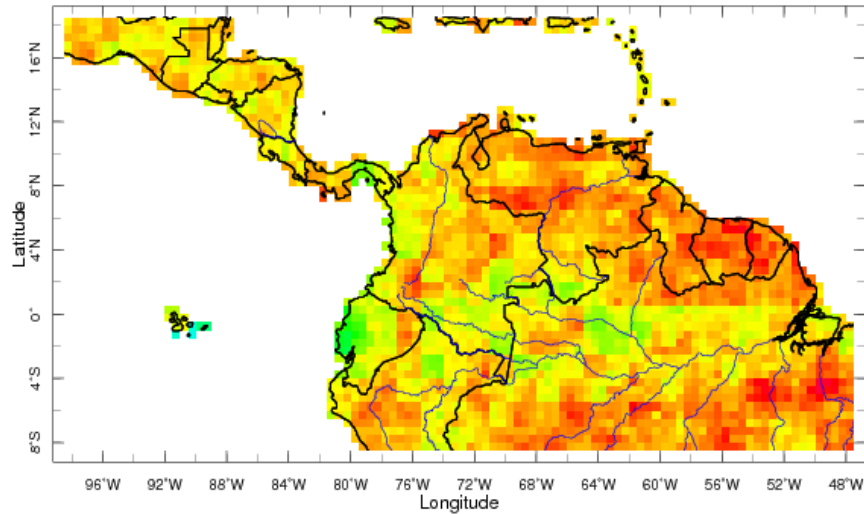
S2S, A3: Ángel Muñoz: “How Much Can M.O.S. Improve Sub-seasonal Prediction Skill”

SS-I

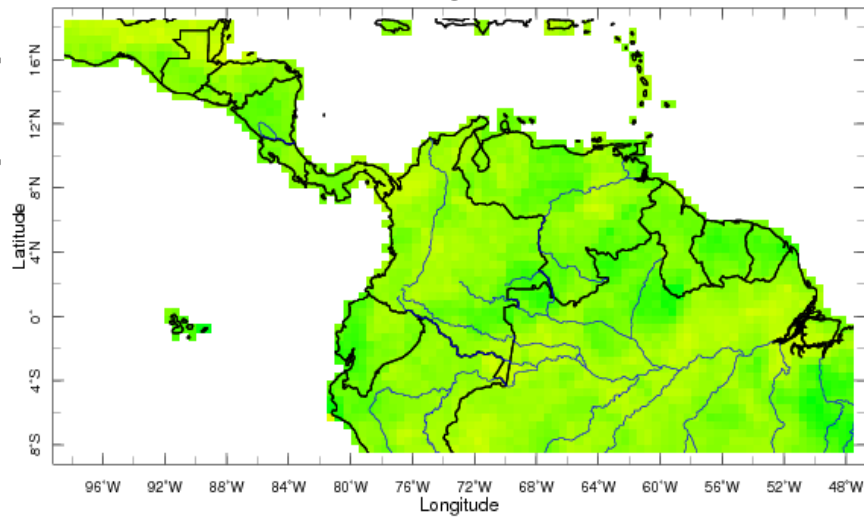
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Initializations: April Week 3

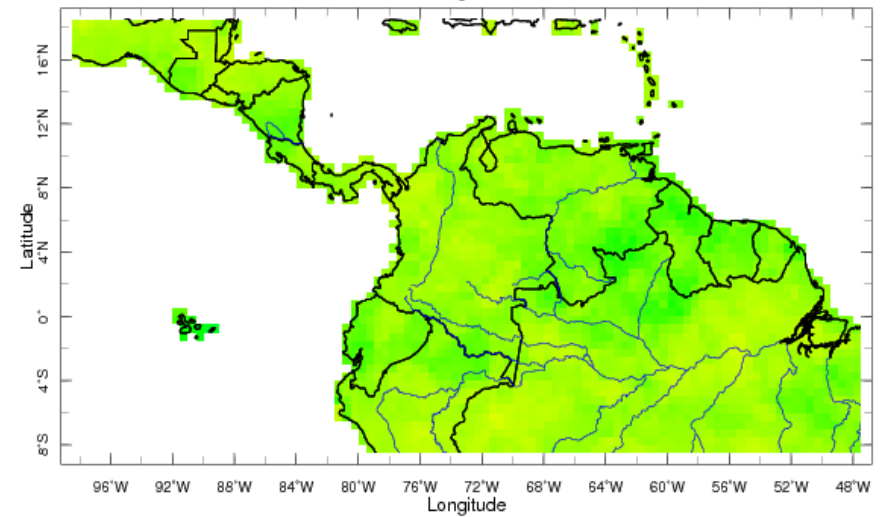
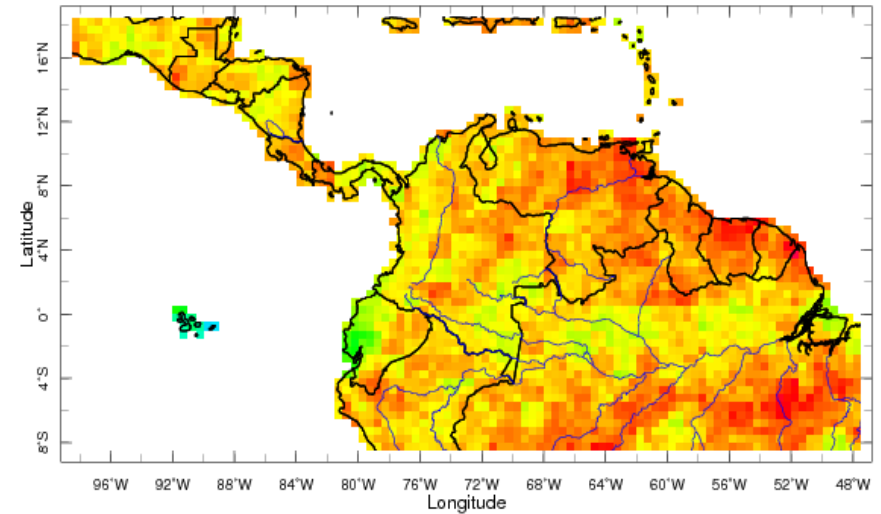
Uncalibrated



Calibrated (CCA)



Week 4



Better

Ignorance Score

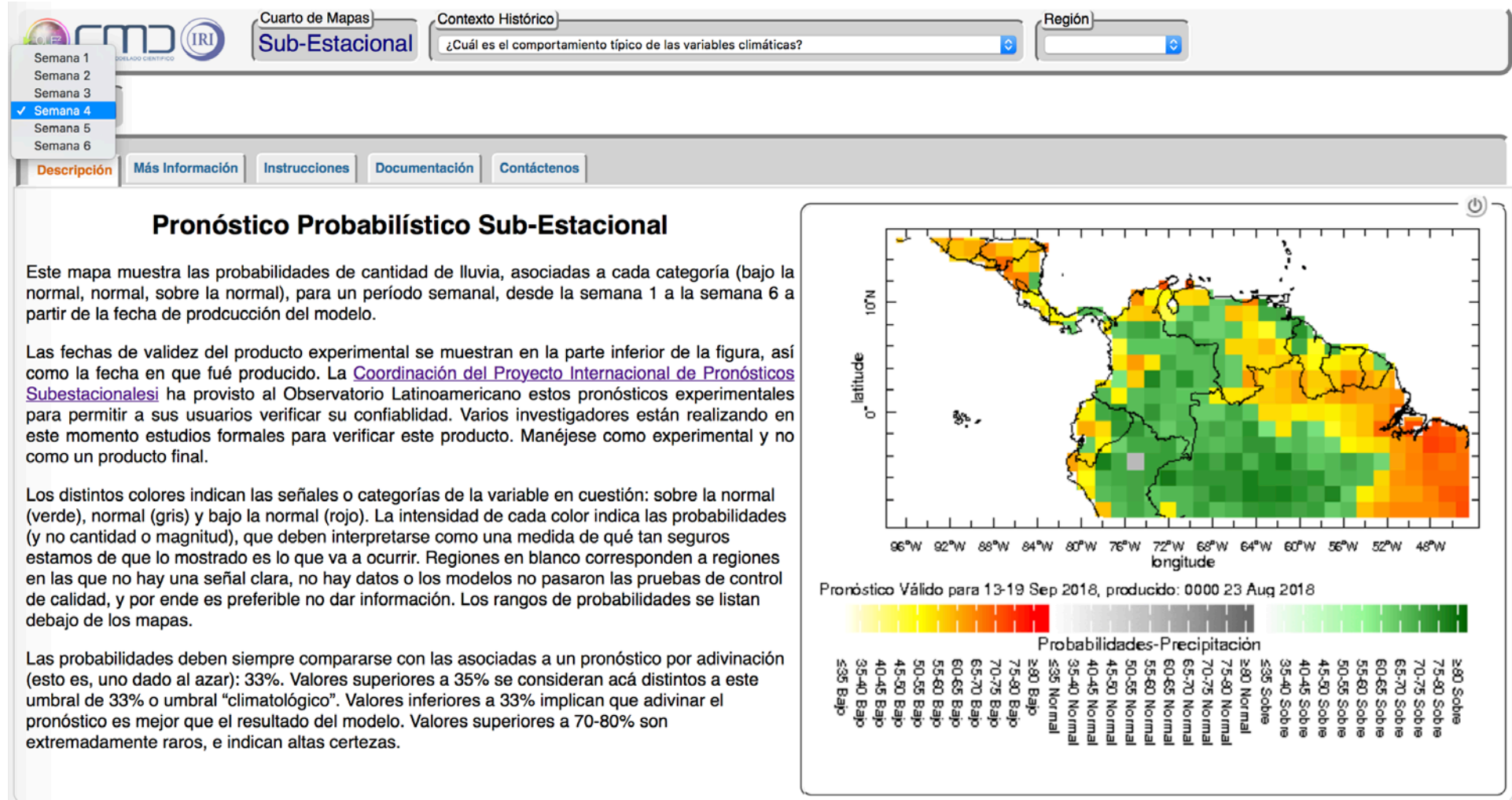
Worse

Muñoz et al (in prep)

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Proto-type Forecast Map Room: *in Spanish*



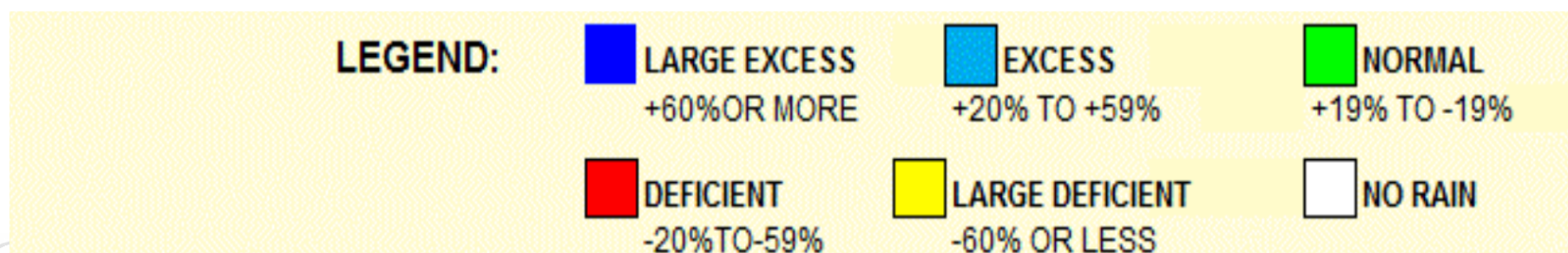
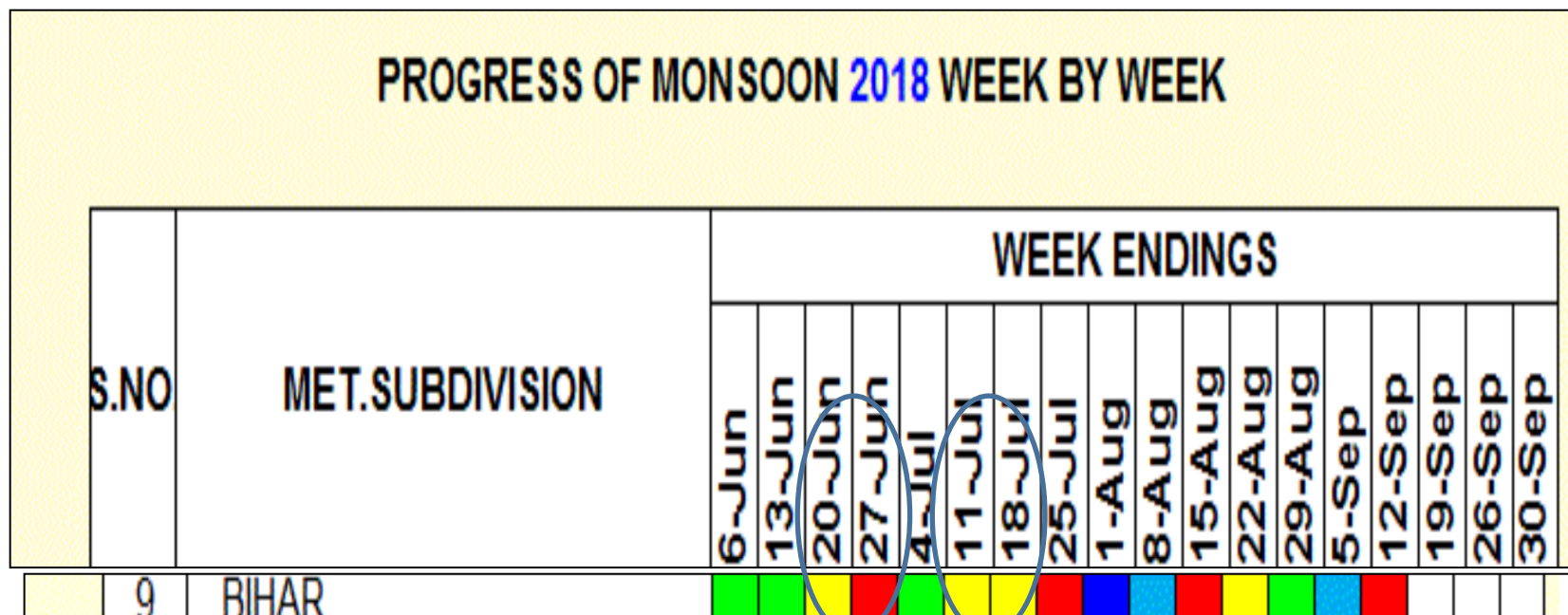
Bihar, INDIA – Rainfall forecasts for agriculture



SS-2

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Observed progress of monsoon, Bihar



(वास्तविक समय के आंकड़ों पर आधारित)

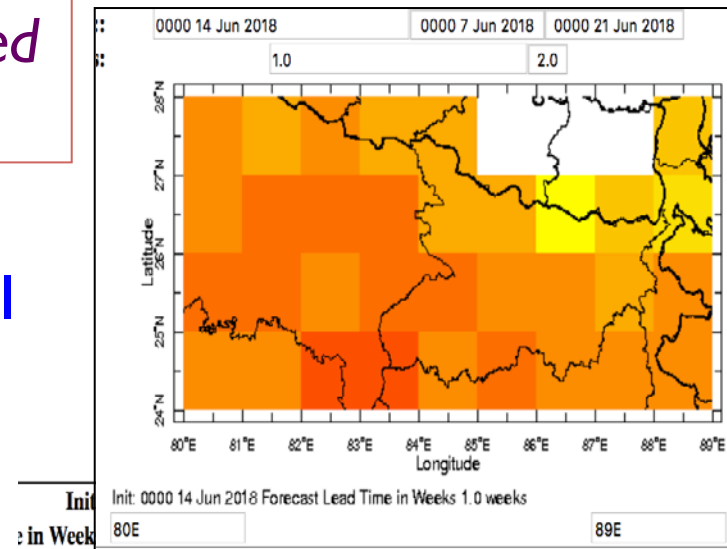
Source: <http://www.imd.gov.in>

Forecast progress of monsoon, Bihar

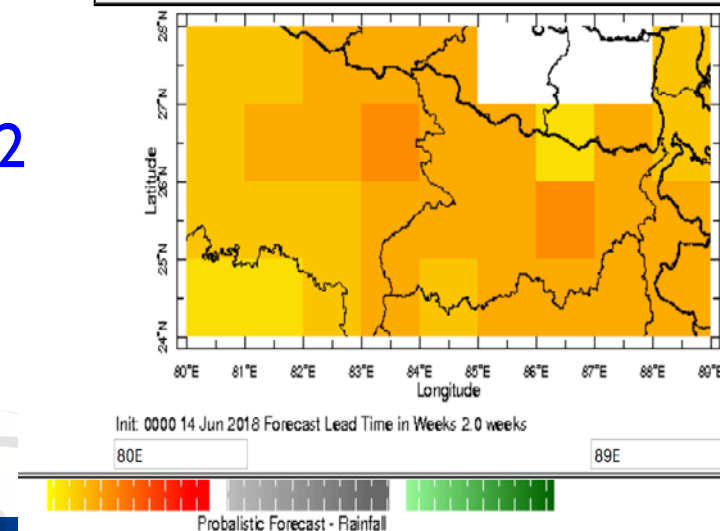
Forecast issued on June 14th

Calibrated
NMME

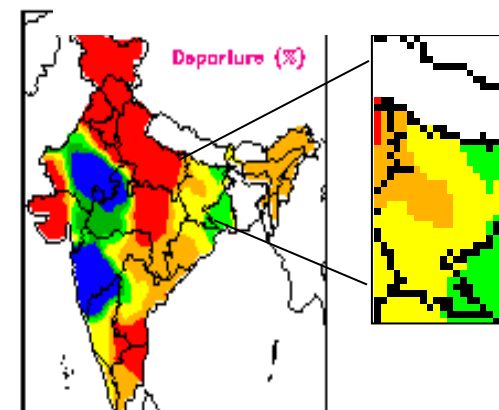
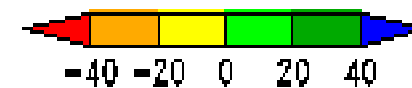
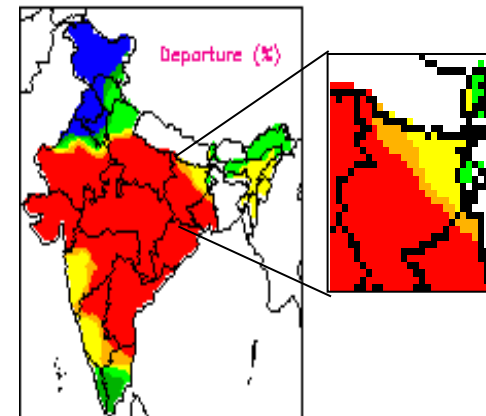
Week 1



Week 2



Observed weekly rainfall



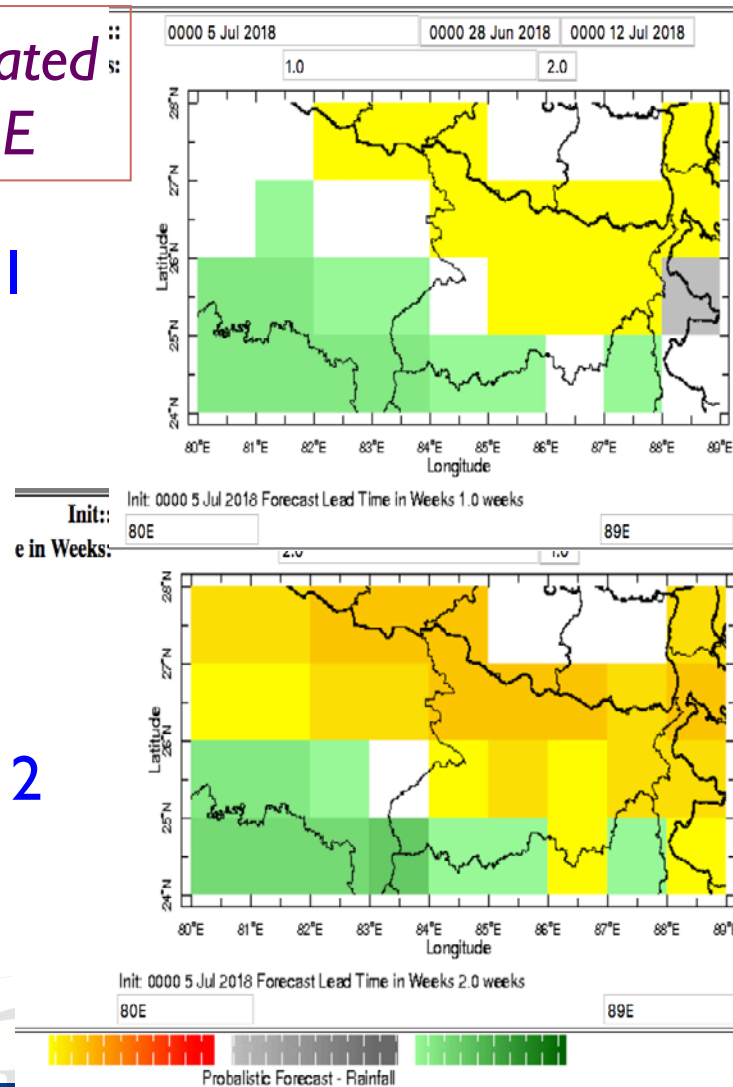
Source: <http://mol.tropmet.res.in>

Forecast progress of monsoon, Bihar

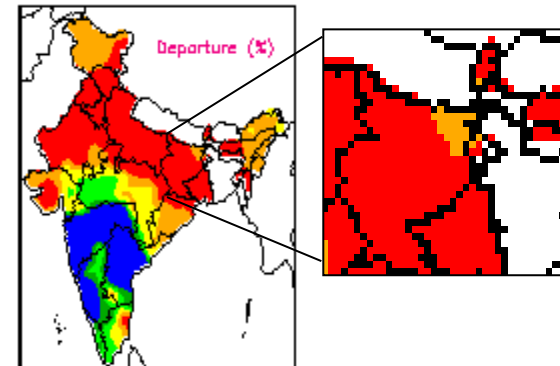
Forecast issued on July 5th

Calibrated
NMME

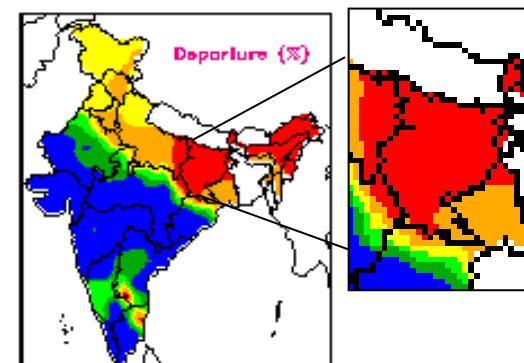
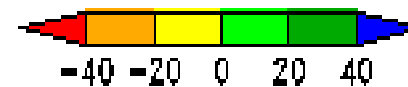
Week 1



Observed weekly rainfall



Week ending
July 11



Week ending
July 18


Source: <http://mol.tropmet.res.in>

Translation of forecasts into local language

Example: July 6, 2018

For East Champaran and Darbhanga by SAU-Pusha

डा० राजेन्द्र प्रसाद केन्द्रीय कृषि विश्वविद्यालय
पूसा, समस्तीपुर (बिहार)



सब-सीजनल मौसम पूर्वानुमान
(IRAP-IMD-RPCA Collaborative Effort)

दरभंगा एवं पूर्वी चम्पारण जिलों के लिए

उप-मौसमीय (सब-सीजनल) पूर्वानुमान


सप्ताह संख्या-१ (६-१२ जुलाई) में थोड़ा अधिक संभावना के साथ (३३ प्रतिशत की तुलना में ४० प्रतिशत) सामान्य से कम वर्षा हो सकती है।

सप्ताह संख्या-२ (१३-१९ जुलाई) में भी थोड़ा अधिक संभावना के साथ (३३ प्रतिशत की तुलना में ४० प्रतिशत) सामान्य से कम वर्षा हो सकती है।

(ए. सत्तार)
नोडल पदाधिकारी, कृषि मौसम
आर०पी०सी०ए०यू०, पूसा, समस्तीपुर

For Nawada and Jehanabad District, prepared by SAU-Sabour

बिहार कृषि विश्वविद्यालय, सबौर, भागलपुर



सीजनल मौसम पूर्वानुमान
(IRAP-IMD-BAU Collaborative Effort)

नवादा एवं जहानाबाद जिलों के लिए

मौसमीय (सीजनल) पूर्वानुमान

12 जुलाई से 30 सितम्बर 2018 तक (बचे हुए मॉनसून सीजन में) थोड़ा अधिक संभावना के साथ (33 प्रतिशत की तुलना में 40 प्रतिशत) सामान्य से अधिक वर्षा होने की संभावना है।

(बीरेन्द्र कुमार)
नोडल पदाधिकारी,
कृषि मौसम सेवा
बिहार कृषि विश्वविद्यालय, सबौर, भागलपुर

S2S, A1: “Sub-seasonal prediction of the Indian monsoon: Case study over Bihar”, A. Robertson

S2D, B3: "Evaluating a new calibration method for Seasonal Probabilistic Prediction for Indian Summer Monsoon"- Nachiketa Acharya

Seasonal-to-Decadal: 2-20 yrs out



Decadal Predictions: Temperature

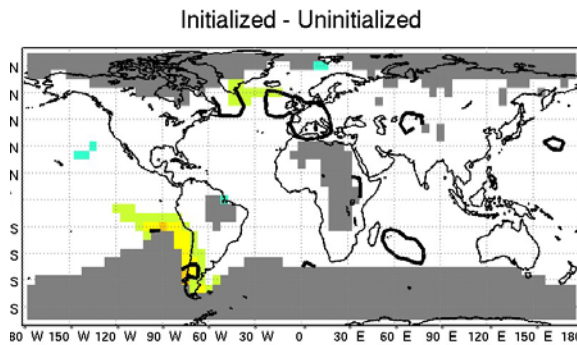
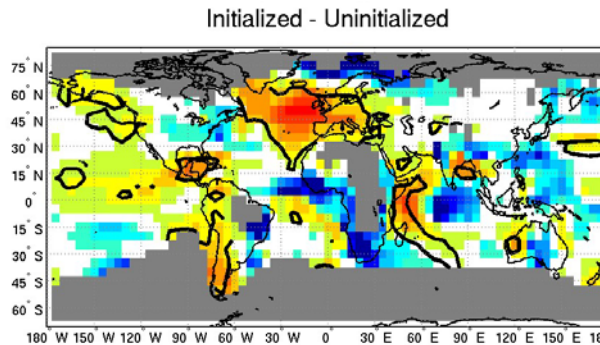
Multi-model Ensemble (12 models: Equal Weighting) – Decadal Average

Mean Error
Skill Score

Correlation

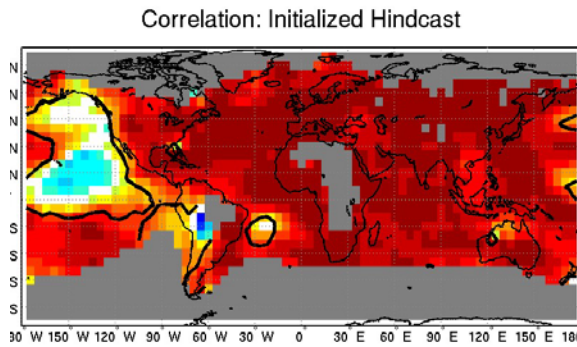
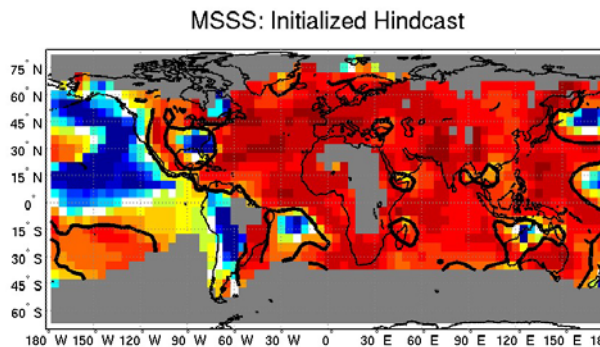
DIFFERENCE

DIFFERENCE



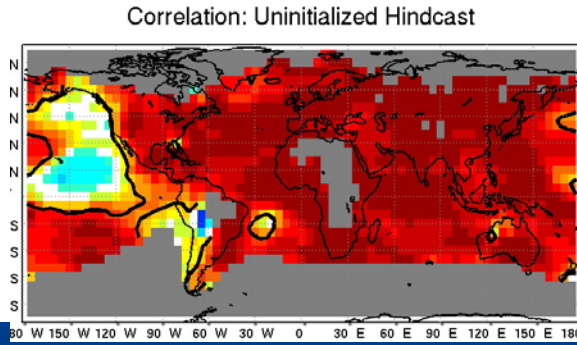
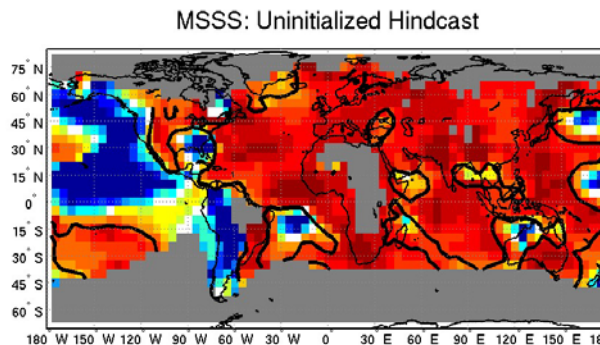
Decadal
Prediction

Decadal
Prediction



Climate
Change
Projection

Climate
Change
Projection



(Blue = Bad; Red = Good):



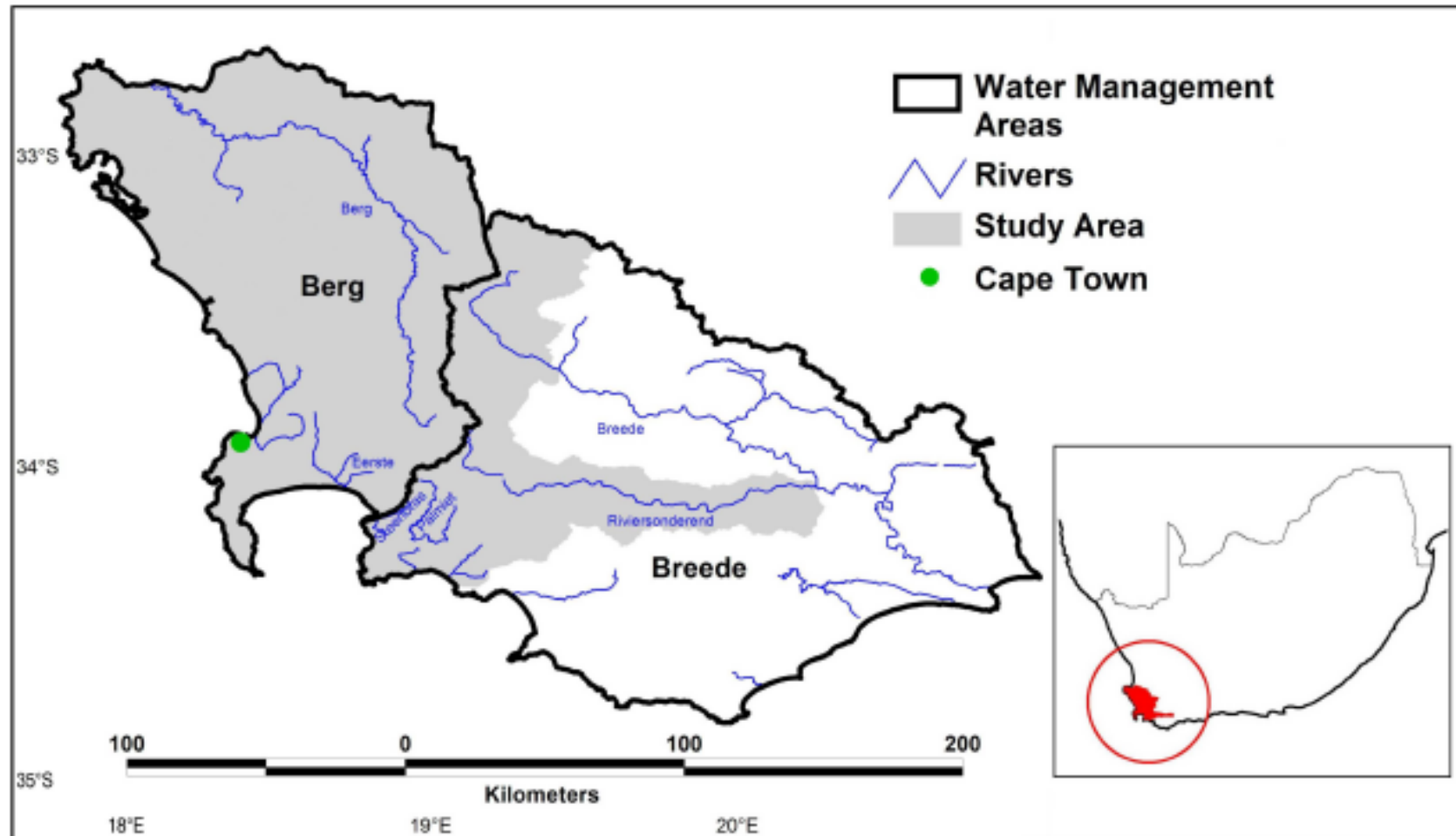
FUTURE AHEAD?

What information can be provided in places where there is no signal from decadal phenomena, or where the models have poor performance??



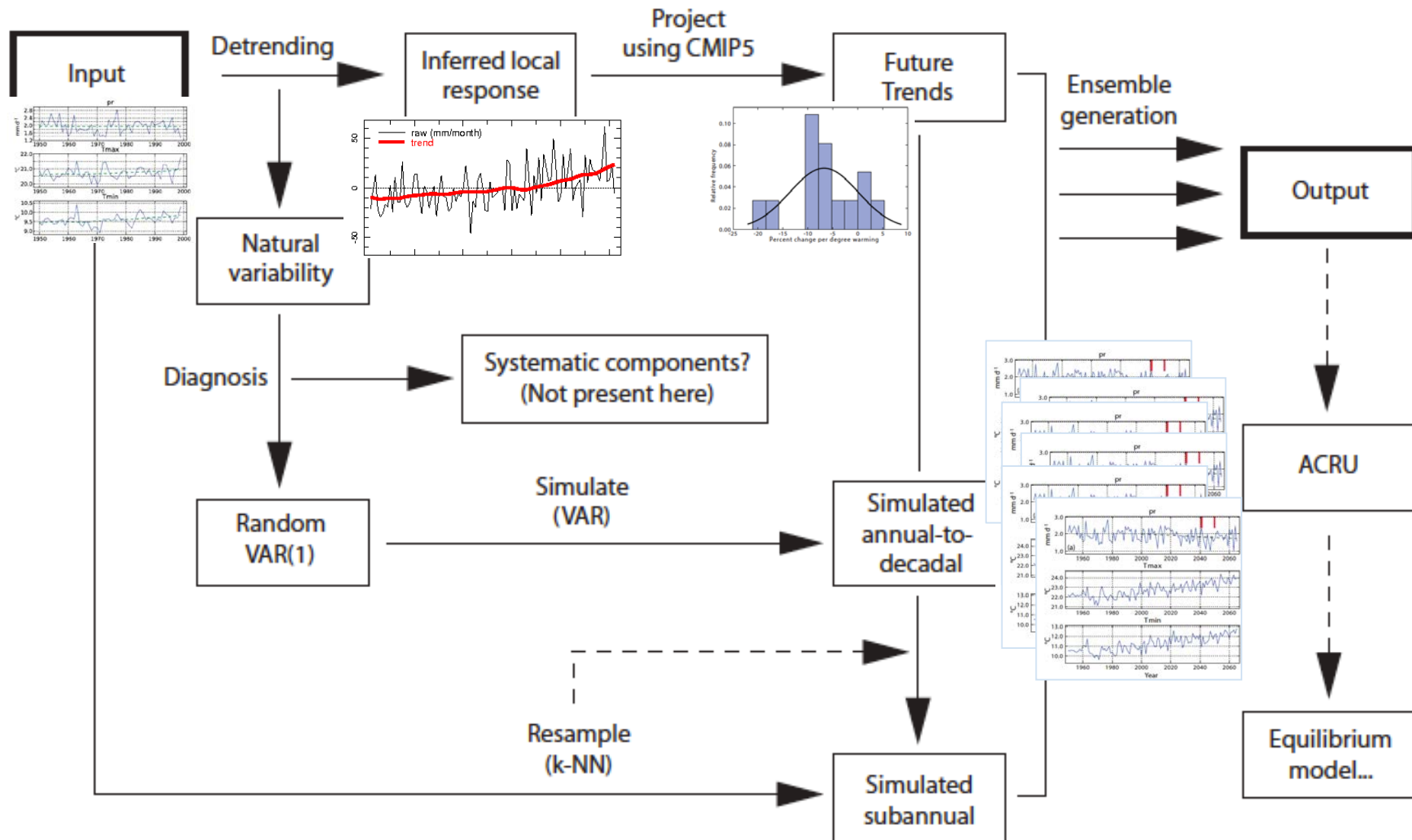
Resiliency of Water Resources and Mgmt

Berg and Breede Water Mgmt Areas



STOCHASTIC SIMULATIONS:

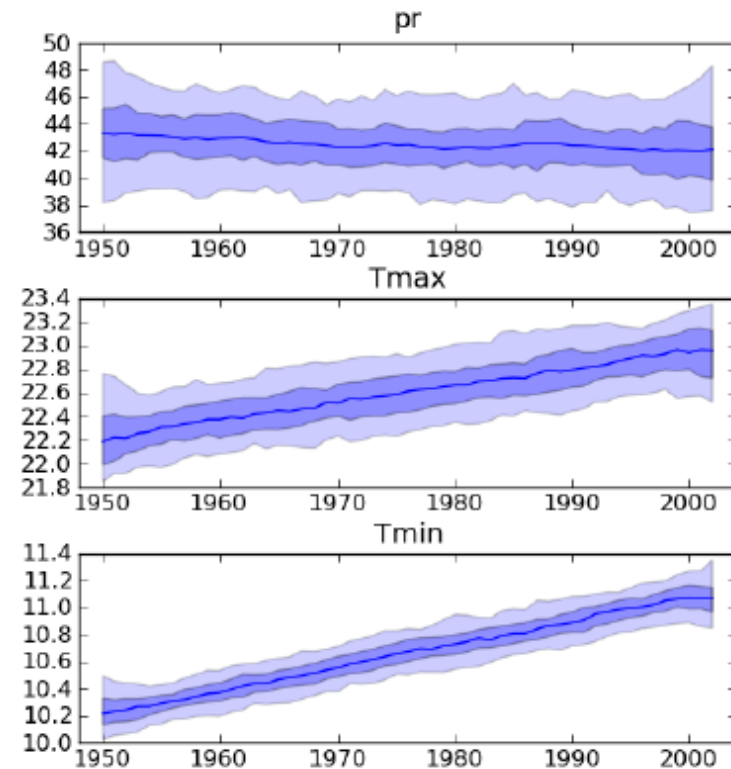
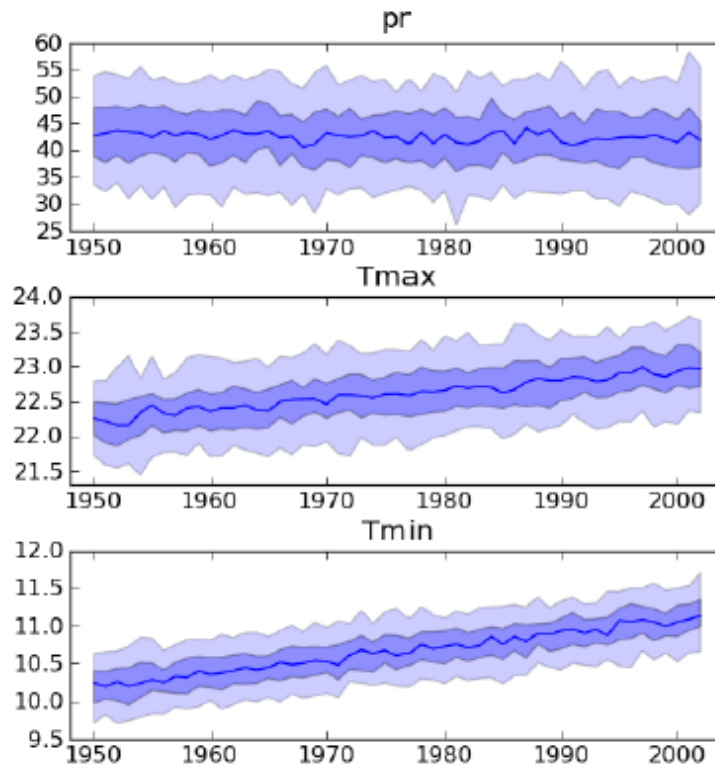
(Greene, et al. 2012)



Dec-I

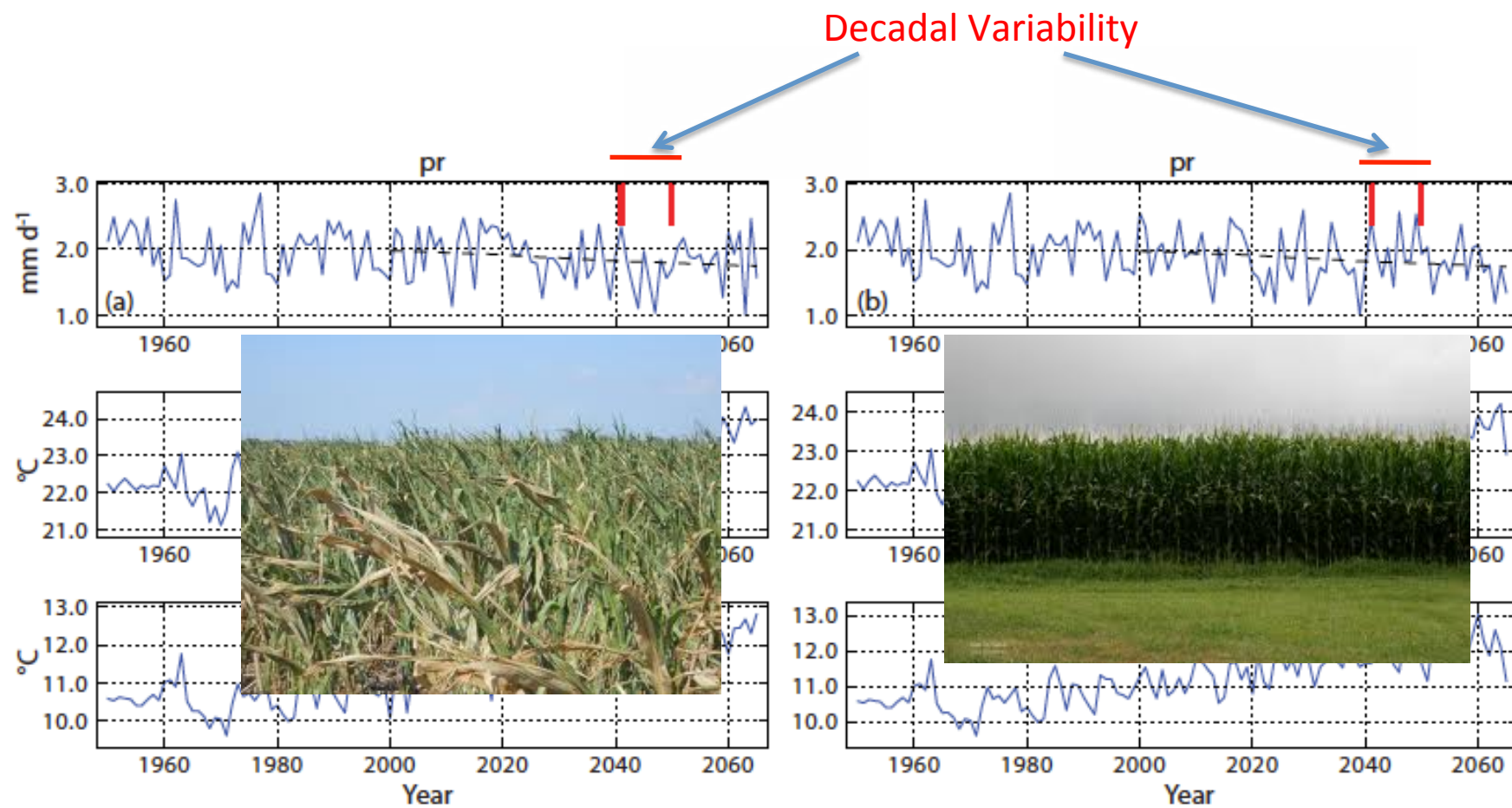
STOCHASTIC SIMULATIONS:

... the envelope please



- Left: Unsmoothed simulations
- IQR and 0.05-0.95 quantile shown, 100 simulations.
- Right: Similar, but for smoothed simulations (9-year unweighted running mean)

STOCHASTIC SIMULATIONS: 2 Ensemble Members



Dec-I

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There are no answers, only choices.

Stanislaw Lem/Steven Soderbergh (Solaris)



Thank You



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IRI Prediction Talks

S2S, P-A1: *Andrew Robertson:* “Sub-seasonal prediction of the Indian monsoon: Case study over Bihar”

S2S, A3: *Ángel Muñoz:* “How Much Can M.O.S. Improve Sub-seasonal Prediction Skill”

S2S, P-A3: *Nicolas Vigaud:* “Wintertime Weather Regimes over North America and Their Predictability from Sub-Monthly Reforecasts”

S2S, P-B3: *Nicolas Vigaud:* “North American Summer Heat Waves and Modulation from North Atlantic Simulated by an AGCM”

S2S, P-A4: *Colin Kelley:* “Differences in Timescales of Rainfall Predictability for Six Countries, within Agricultural Context”

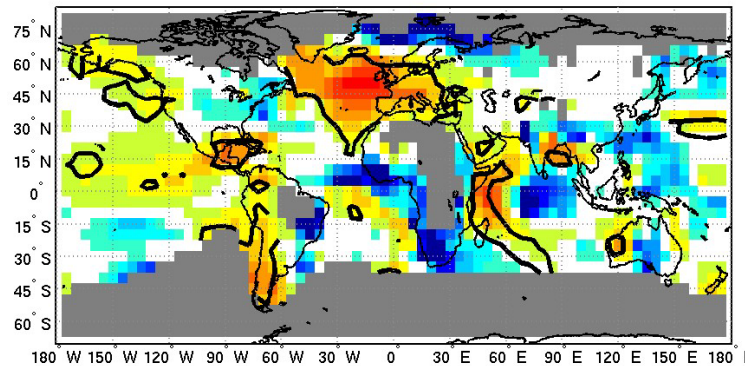
S2D, B3: *Nachiketa Acharya:* “Evaluating a new calibration method for Seasonal Probabilistic Prediction for Indian Summer Monsoon”

Predict the Future: Decadal (skill)

Decadal Predictions: *Skill still to be demonstrated*

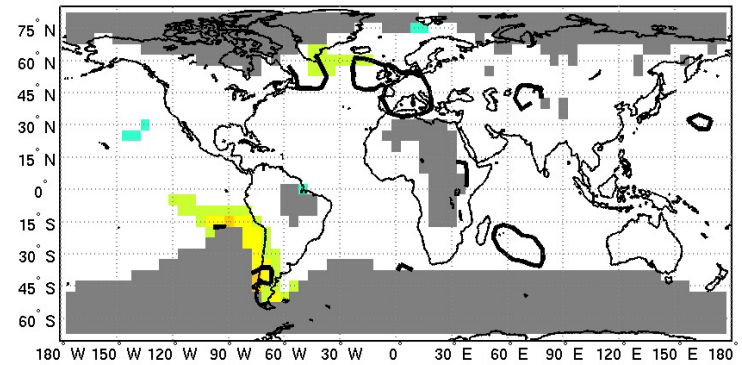
Relative Accuracy

MME temp MSSS: year 2-9 ann
Initialized - Uninitialized



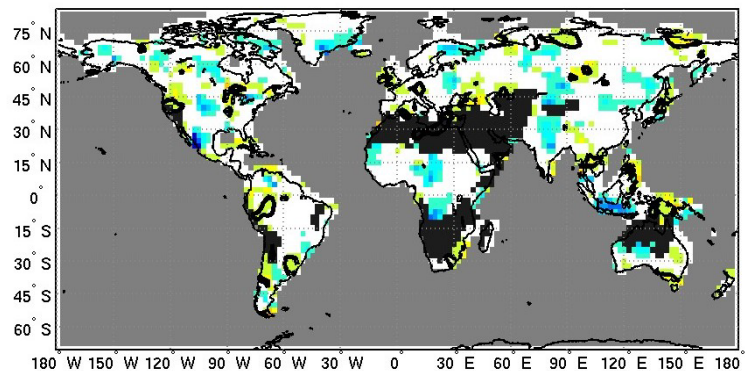
Correlation

MME temp Correlation: year 2-9 ann
Initialized - Uninitialized

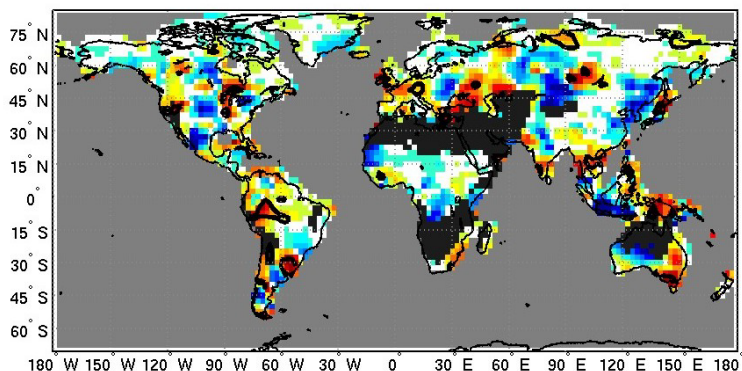


Annual Temp.

MME prcp MSSS: year 2-9 JAS
Initialized - Uninitialized



MME prcp Correlation: year 2-9 JAS
Initialized - Uninitialized



Jul-Aug-Sep
Rainfall

(based on Goddard et al. 2012, *Climate Dynamics*)