Quantifying drought risk through multiple large-ensembles: the case of Cape Town 2015-17 and Central America 2015-19 multiyear droughts

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Theewaterskloof Dam, April 2018

#### Case 1: the Cape Town "Day Zero" water crisis



Cape Town, South Africa, Feb. 8, 2018 (Image credit: fivepointsix/iStock)

- Extensive economic impacts (37k jobs lost in WC Province, 50k people pushed into poverty)
- Agriculture: 13-20% drop in exports; Tourism: 10% drop
- Public health concerns
- Stringent water restrictions: ban of outdoor and non-essential water use; consumption restricted to about 50 gallons per person in February 2018.

### The Cape Town "Day Zero" drought

- Multi-year meteorological drought (2015-17) unprecedented in the centennial record (Wolski 2018; Otto et al. 2018) over most of Southwestern South Africa (SSA)
- April-September rainfall totals at 35%-50% below average in most of SSA. Rainfall deficit during the **shoulder season**
- Hydrological drought: dams supplying Cape Town ≈ 20% in austral fall 2017/2018
- If below 13.5% "Day Zero": disconnect much of the municipal water supplies



1. To what extent did anthropogenic global warming make the Day Zero drought more likely (event attribution)?

2. How will the probability of occurrence of another similar or worse meteorological drought change in the **coming decades**?

#### How to attribute extremes to anthropogenic climate change?



Otto F.E.L., Ann. Rev. Env. Res.. 2017. 42:627-46

Probabilistic event attribution: Risk Ratio

#### Large ensembles suite

- Large Ensemble simulations from the Seamless System for Prediction and Earth System Research (SPEAR\_MED, 2020): 0.5 degree resolution, (Delworth et al., 2020, <u>https://www.gfdl.noaa.gov/spear</u>)
- Additional large ensembles at same or coarser resolution to test model uncertainties:
  - **SPEAR\_LO: 1 degree resolution** (Delworth et al., 2020)
  - FLOR, FLOR\_FA: 0.5 degree resolution (Vecchi et al., 2014)
  - CESM1: 1.3 x 0.9 degree resolution (Kay et al., 2015)
  - MPI-GE: 1.9 degree resolution (Mahler et al., 2019)

# The SPEAR\_MED large ensemble

- **CTRL**: forcing kept constant at 1850 (pre-industrial) levels: **3000** yrs
- ALLFORC, 30 ensemble members: 1921-2100 SPEAR\_MED:
  - Historical forcing up to 2014
  - SSP5-8.5 (high-emission) and SSP2-4.5 (intermediate) scenarios 2015-2100
- NATURAL, 30 ensemble members: natural historical forcing (solar+volcanic) until 2014, idealized solar after 2015

### Estimation of PDFs

- 3-yr Winter Rainfall Index anomaly (relative to 1921-70) for 2015-2017 ≈ -12 mm/month
- **CTRL's PDF** (3yr mean anomalies) : randomly select non-overlapping 50yr and a 3yr time windows and take the mean difference (repeat it 10,000 times)



#### **Estimation of PDFs**

- 3-yr Winter Rainfall Index anomaly (relative to 1921-70) for 2015-2017 ≈ -11.5 mm/month
- Decadal PDF (3yr mean anomalies) for each 20-yr window (18x30=540 different sequences)



#### How likely event\_1519 in the next decades?



#### **Risk ratios**



### Case 2: the 2015-19 Central American megadrought





THE SCIENCES MIND HEALTH TECH SUSTAINABILITY VIDEO

😨 Eye of the Storm

#### **Fifth Straight Year of Central American Drought Helping Drive Migration**

Recent rains have helped, but long-term climate change is likely to significantly increase migration

- Prolonged droughts in Honduras, El Salvador, Guatemala, ٠ Nicaragua led to massive loss of major crops (beans, coffee, corn), which are a fundamental part of the livelihoods of the population in the region.
- 2014, 2015, 2016, 2018, 2019 were dry. The 2015-2019 ٠ mean exceptionally dry

#### The 2015-19 event



### RR for event\_1519



Risk Ratio



## Backup slides





#### Definition of event\_1517



# Weather and climatic extremes: present and future

#### Climate change is already affecting every inhabited region across the globe with human influence contributing to many observed changes in weather and climate extremes



c) Synthesis of assessment of observed change in agricultural and ecological drought and confidence in human contribution to the observed changes in the world's regions

#### AR6, Summary for policymakers, IPCC, 2021

Small

Islands

### Confidence in attributing different events



#### **Enough to estimate tails?**

SPEAR med

10.00 p 1.00 Probability (%) ............ 540 yrs 0.10 0.01 500 1000 1500 2000 2500

N sample

#### **SPEAR:** global mean surface temperature evaluation





#### 500 Geop. Height



#### 1951-2017 linear trends (AMJJAS precip) Individual members



Range of AMJJAS trends consistent (i.e., within the range of natural variability) with observed trends over most of SA



#### Are observed trends attributable to anthropogenic forcing?

To estimate natural variability range: Trends calculated over a 67-yr period picked up randomly 30 times to form the ensemble mean (this process is then repeated 3000 times).



# Method

#### How exceptional the mean 2015-17 drought was?

Randomly select non-overlapping 50yr and a 3yr time windows..



# Method

#### How exceptional the mean 2015-17 drought was?

Randomly select 3 consecutive years



Repeat 20 times, for each ensemble member (20x30=600)

#### **Other Large Ensembles?**



Fraction of WRI reduction (2071-2100 vs 1921-1970)



Ensemble mean difference MSLP (2070-2100 vs.1921-2000)

#### Model evaluation of SSA precipitation



#### How likely event\_1519 in the next decades?



#### Grid point view

Probability of exceeding the 1st percentile from the CTRL probability distribution of the three-year winter rainfall anomalies.

# Approach to event attribution: high-resolution LE

Large Ensemble simulations from the newly developed **S**eamless System for **P**rediction and **Ea**rth System **R**esearch at 0.5 degree resolution: SPEAR\_MED

SPEAR: The Next Generation GFDL Modeling System

for Seasonal to Multidecadal Prediction



**F** 



and Projection

#### **RESEARCH ARTICLE**

10.1029/2019MS001895

#### **Special Section:**

Geophysical Fluid Dynamics Laboratory CMIP6 Models

Rich Gudgel retired at the end of March 2019. Shian-Jiann Lin retired in May 2019.

#### **Key Points:**

- · Development and performance of the next generation GFDL seasonal to decadal prediction model is documented
- The response of this model to realistic radiative forcing changes is shown via a large ensemble of climate simulations for 1921-2100

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https://www.gfdl.noaa.gov/spear

- Large decadal/interdecadal variability of winter SSA rains (Dieppois et al. 2016, Reason et al., 2002; Philippon et al.,  $(2012) \rightarrow$  Large Ensembles powerful method to isolate internal natural variability from forced signal
- Additional large ensembles at same or coarser resolution to test model uncertainties: SPEAR LO, FLOR, FLOR FA ۲ (Vecchi et al., 2014), CESM1 (Kay et al., 2015), MPI-GE (Mahler et al., 2019) 35

# SPEAR: The next generation GFDL modeling System Built from component models of GFDL CM4: AM4 (atmosphere), MOM6 (ocean), LM4 (land surface), SIS2 (sea ice)

- Ocean: 1 degree; Atmosphere 0.25 (SPEAR\_HI), 0.5 (SPEAR\_MED), 1 (SPEAR\_LOW) degree: thought for regional climate and extremes



#### **Decadal MSLP forced anomalies**



#### **Duration of large AMJJAS rainfall anomalies**



probability change (%)

### Why did this happen?

- Prolonged rainfall deficit the main factor (Otto et al., 2018; Sousa et al., 2018); poor water management co-factor (Muller 2018).
- It's been suggested a southward shift of extratropical cyclones (Sousa et al., 2018; Mahlela et al., 2018)
- Others found no anomaly in # front, but higher post-frontal MSLP  $\rightarrow$  Less rainy days, shorter duration of wet spells (Burls et al., 2019)
- Hadley Cell expansion?



(a) SLP and 850hPa wind (AMJJAS) for 2015-2017 drought minus the 1979-2017 mean



### Time of detectability

Decadal precipitation differences relative to 1921-1970 Non-stippled: signal detectable from internal natural variability

Anthropogenic signal in the **mean clearly emerges** after 2020-30 in the Western Cape province.





### What is the role of humans in extremes

#### nature

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Published: 02 December 2004

# Human contribution to the European heatwave of 2003

Peter A. Stott 🖂, D. A. Stone & M. R. Allen

*Nature* **432**, 610–614 (2004) Cite this article

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#### How do we study extreme climate event risk?



Climate models can be used to generate sufficiently large datasets (i.e., 100s to 1000s to 10,000s of years) to explore extreme event probabilities for different moment in times)

## Did Anthropogenic Global Warming make it worse?

- 3-yr mean precipitation (observations, AGCM ensembles, GCM ensembles) fitted to a Gaussian or GEV whose parameters scale with temperature
- **1.4 to 6.4 times** more likely at +1 deg of GW



but..

- Older generation models (e.g., CMIP3/5) have systematic biases in SH jet stream position (Curtis et al., 2020)
- Small extent of the SSA region & difficulty to get regional features (e.g., orography)
- Dynamical mechanisms ?

### Other Large Ensemble experiments

Experiment	Years	Members	Atm. res.	Description
SPEAR_LO CTRL	5000	1	1°×1°	Preindustrial (1850) forcing
SPEAR_LO NATURAL	1921-2100	30	1°×1°	Natural historical forcing before 2014; solar variability only after 2014.
SPEAR_LO ALLFORC8.5	1921-2100	30	1°×1°	All historical forcing before 2014; SSP5-8.5 afterwards.
FLOR_FA CTRL	3500	1	$0.5^\circ  imes 0.5^\circ$	Preindustrial (1860) forcing
FLOR_FA NATURAL	1941-2050	30	$0.5^\circ  imes 0.5^\circ$	Natural historical forcing before 2005; solar variability only after 2005;
FLOR_FA ALLFORC4.5	1941-2050	30	$0.5^{\circ}  imes 0.5^{\circ}$	All historical forcing before 2005; RCP4.5 afterwards.
FLOR CTRL	2200	1	$0.5^\circ  imes 0.5^\circ$	Preindustrial forcing (1860)
FLOR NATURAL	1921-2100	30	$0.5^\circ  imes 0.5^\circ$	Natural historical forcing before 2005; solar variability only after 2005.
FLOR ALLFORC8.5	1921-2100	30	$0.5^\circ  imes 0.5^\circ$	All historical forcing before 2005; <b>RCP8.5</b> afterwards.
CESM-LENS CTRL	1800	1	$1.3^{\circ} imes 0.9^{\circ}$	Preindustrial forcing
CESM-LENS ALLFORC8.5	1921-2100	40	$1.3^\circ  imes 0.9^\circ$	All historical forcing before 2005; <b>RCP8.5</b> afterwards.
MPI-GE CTRL	2000	1	1.9°×1.9°	Preindustrial forcing (1850)
MPI-GE ALLFORC8.5	1850-2100	100	$1.9^\circ \times 1.9^\circ$	All historical forcing before 2005; RCP8.5 afterwards.
MPI-GE ALLFORC4.5	1850-2100	100	$1.9^{\circ} \times 1.9^{\circ}$	Same as MPI-ESM1 ALLFORC8.5 before 2005; RCP4.5 afterwards.
MPI-GE ALLFORC2.6	1850-2100	100	1.9°×1.9°	Same as MPI-ESM1 ALLFORC8.5 before 2005; RCP2.6 afterwards.

### The Cape Town "Day Zero" hydrological drought

- Dams supplying Cape Town ≈ 20% in austral fall 2017/2018
- If below 13.5% "Day Zero": disconnect much of the municipal water supplies





#### Extreme events' attribution

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#### Link to large scale circulation shifts?





- **Higher sea-level pressure in 2015-17** invoked as the cause of a southward shift of the moisture corridors, contributing to winter rainfall (Sousa et al., 2018).
- Burls et al. (2019): no significant regional trends over the last 40 y in the number of cold fronts making landfall over SSA; shorter duration of rainfall events due to larger sea-level pressure during postfrontal days.
- Hadley Cell expansion? For the SH, indication forced signal in Hadley Cell emerging around 2020 (Amaya et al., 2018; Grise et al., 2019). If CO<sub>2</sub> keeps increasing, not a matter of if, but of when. But how seasonally?

#### **Decadal MSLP forced anomalies**



Similarities, but challenging to discern forced signal in the observational record (Staten et al., 2018)



### **Other Large Ensembles?**





#### Large scale anomalies: the Caribbean Low-Level Jet



The stronger the CLLJ, the drier CA (CLLJ index: Wang et al., 2007, JCLIM)

### Large scale anomalies: the Atlantic-Pacific SST difference



Persistent –ve TNATNP index (TNA colder than TNP)

Warmer TNP and colder TNA→ drier over Central America (Taylor et al., 2002; Fuentes-Franco et al., 2015)

#### Seasonality of the anomaly/change



#### Poleward shift of the SH jet stream?



Ensemble mean difference in MSLP and EKE300 (2070-2100 vs.1921-2000)

The most evident forced signals in April-March and August–September (AS)

### CLLJ and TNATNP projections (SPEAR\_MED, SSP5-8.5)



Already known and discussed (e.g., Rauscher et al., 2008, 2011; Fuentes-Franco et al., 2015)

Are these configurations already more likely because of ACC?

# Changes in CLLJ and TNATNP attributable?



### How do we study extreme climate event risk?

- **Observations alone:** using observations alone can lead to high uncertainty for historic events (LEFT).
  - observations have limited records
  - statistical methods fill in gaps to create uncertainty range
- Climate models+Statistical methods to estimate extreme values: increased data (2000yrs) from climate models can be used to reduce the uncertainty (CENTRE)
- **Climate models+Direct calculations of probabilities:** with a sufficient number of years, probabilities can be calculated directly, reducing uncertainty added by statistical methods (RIGHT)



#### Seasonal Rainfall Accumulation Anomaly by pentad

2020 season May - Aug

(May pentad 1 thru Aug pentad 6) - Average (1981-2010)





# Weather and climatic extremes: present and future

#### Climate change is already affecting every inhabited region across the globe with human influence contributing to many observed changes in weather and climate extremes

a) Synthesis of assessment of observed change in **hot extremes** and confidence in human contribution to the observed changes in the world's regions



#### AR6, Summary for policymakers, IPCC, 2021

# Weather and climatic extremes: present and future

#### Climate change is already affecting every inhabited region across the globe with human influence contributing to many observed changes in weather and climate extremes

b) Synthesis of assessment of observed change in **heavy precipitation** and confidence in human contribution to the observed changes in the world's regions



#### AR6, Summary for policymakers, IPCC, 2021

## Future?



1. To what extent did anthropogenic global warming make 2015-19 Central America rainfall deficit more likely (**event attribution**)?

2. How will the probability of occurrence of another similar or worse meteorological drought change in the **coming decades**?

### Conclusions (I): "Day Zero" drought

- 1. The PDF has **already shifted**: AGW made it ≈**5.5 times more likely (CI [4.5-8])**. Further constrain the risk ratio of SSA drought at and above the original [1.4, 6.4] by Otto et al. (2018)..
- RR ~110 (30) by the end of 21<sup>st</sup> century in SSP5-8.5 (SSP2-4.5). RR>80 in any high emiss. scenarios, >30 in RCP4.5 (MPI), ~12 in RCP2.6 (MPI)
- 3. Shift due to increasing MSLP and southward storm track shift: regional Hadley cell expansion? Robust indication for late fall/early winter and spring.
- 4. 2015-17 conditions may be a glimpse of what the future will look like in SSA



# Conclusions (II): Central American drought

- Convincing evidence that the 2015-19 rainfall deficit has been made more likely by ACC by a factor 4
- % of similar or worse megadrought increasing rapidly without actions to reduce GHGs

### **Risk ratios**



Ref. 5: Otto et al., 2019

### Conclusions (I): "Day Zero" drought

- 1. The PDF has **already shifted**: AGW made it ≈**5.5 times more likely (CI [4.5-8])**. Further constrain the risk ratio of SSA drought at and above the original [1.4, 6.4] by Otto et al. (2018)..
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