

Climate extremes *from statistics to society*

Andrea Toreti



Outline

Characterising

EVT, TX90p, SU, FD, ...

Predicting

Z500, PV, AMO, ...

Evaluating

CMIP5, CORDEX,...

Impacts

Characterising

Different approaches, different data sets, different problems

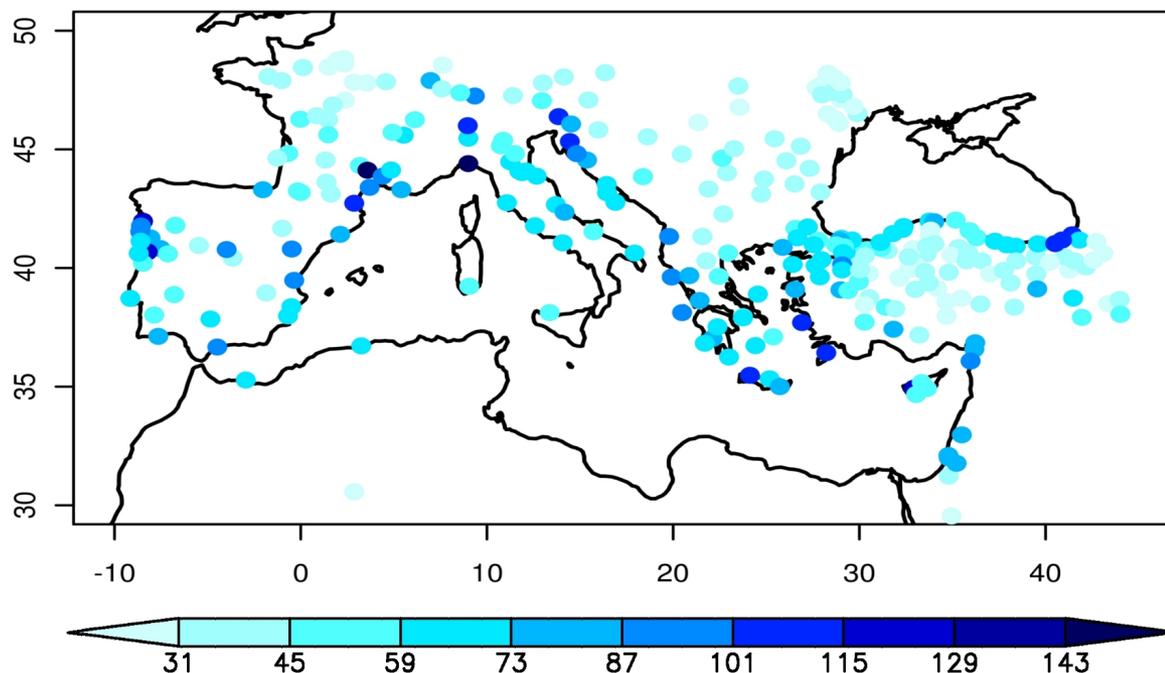
EVT, index-based, others, ...

observations from weather stations, satellite-ground observations, reanalysis, ...

Spatio-temporal scales

Characterising I

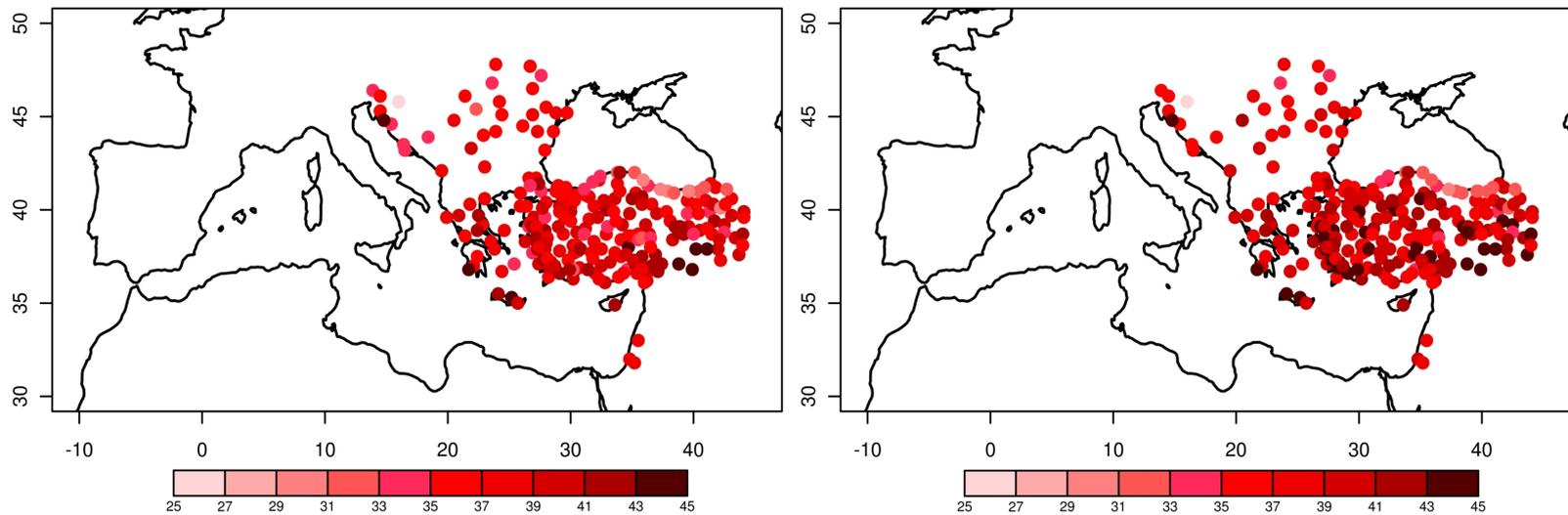
stationary univariate POT approach



Estimated 5-y ret levels of daily precipitation in winter. Data from the last 5 decades. Source: Toreti, 2010

Characterising II

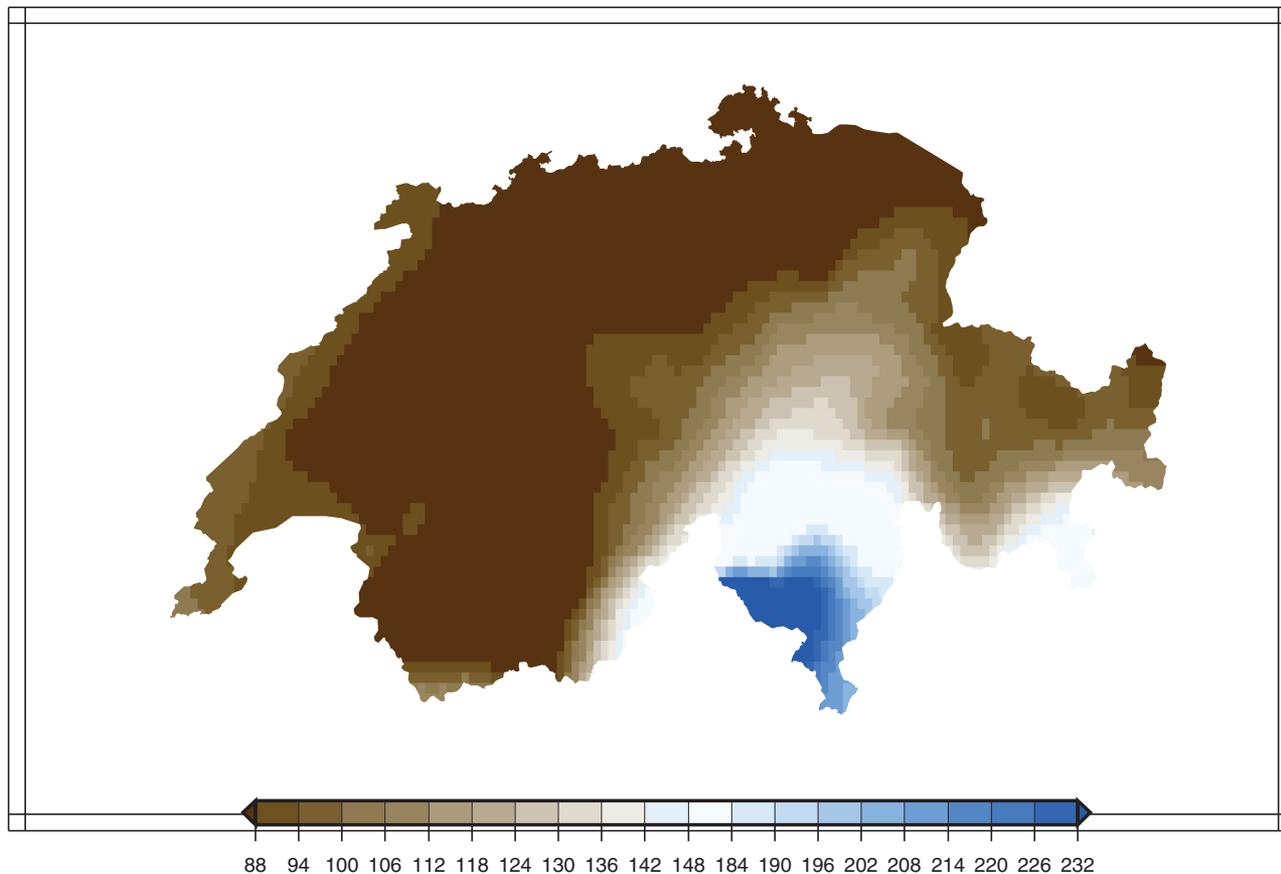
time-dependent location, univariate GEV approach



Estimated 5-y ret levels of daily maximum temperature in summer. Data from the last 5 decades. Source: Toreti, 2010

Characterising III

(partially) non-stationary approach



*Estimated 50-y ret levels of daily precipitation in autumn. Data from 2001-2010.
Source: Naveau et al. 2014, WRR 50*

Characterising III

EVT - GP family

$$\bar{G}_{\sigma,\xi}(y) = 1 - G_{\sigma,\xi}(y) = \begin{cases} (1 + \xi \frac{y}{\sigma})^{-1/\xi} & \text{if } \xi \neq 0, \\ \exp(-\frac{y}{\sigma}) & \text{if } \xi = 0. \end{cases}$$

$\sigma(\mathbf{X})$ and assuming ξ constant

Probability weighted moments

$$\mu_r(\mathbf{X}) = \mathbb{E}[Y(\mathbf{X})\bar{G}_{\sigma(\mathbf{X}),\xi}^r(Y(\mathbf{X}))] \quad \text{where } Y(\mathbf{X}) \text{ follows } GP(\sigma(\mathbf{X}), \xi)$$

$$\mu_r(\mathbf{X}) = \sigma(\mathbf{X}) \frac{1}{(1+r)(1+r-\xi)}$$

Characterising III

$$\mu_r(\mathbf{X}) = \sigma(\mathbf{X})\mathbb{E}[Z\overline{G}_{1,\xi}^r(Z)] \quad Z \sim GP(1, \xi)$$

by using $\mu_0(\mathbf{X})$, $\mu_r(\mathbf{X})$ and $\mu_s(\mathbf{X})$

$$\xi = \frac{(1+s)^2 - (1+r)^2\alpha_{rs}}{(1+s) - (1+r)\alpha_{rs}} \quad \text{and} \quad \sigma(\mathbf{X}) = \mu_0(\mathbf{X})(1 - \xi)$$

$$\alpha_{rs} = \frac{\mathbb{E}[Z\overline{G}_{1,\xi}^r(Z)]}{\mathbb{E}[Z\overline{G}_{1,\xi}^s(Z)]}$$

ξ becomes function of only $\alpha_{r,s}$

Characterising III

Let $\hat{\mu}_0(\mathbf{X})$ and $\hat{\alpha}_{rs}$ be the estimators of $\mu_0(\mathbf{X})$ and α_{rs}

by selecting $r = 1$ and $s = 2$

$$\hat{\xi} = \frac{9 - 4\hat{\alpha}}{3 - 2\hat{\alpha}} \quad \text{and} \quad \hat{\sigma}(\mathbf{X}) = \hat{\mu}_0(\mathbf{X})(1 - \hat{\xi})$$

$$\hat{\mu}_0(\mathbf{X}) = \frac{1}{\sum_i K(\mathbf{X} - \mathbf{X}_i)} \sum_{i=1}^n Y(\mathbf{X}_i) K(\mathbf{X} - \mathbf{X}_i)$$

where K is a Kernel

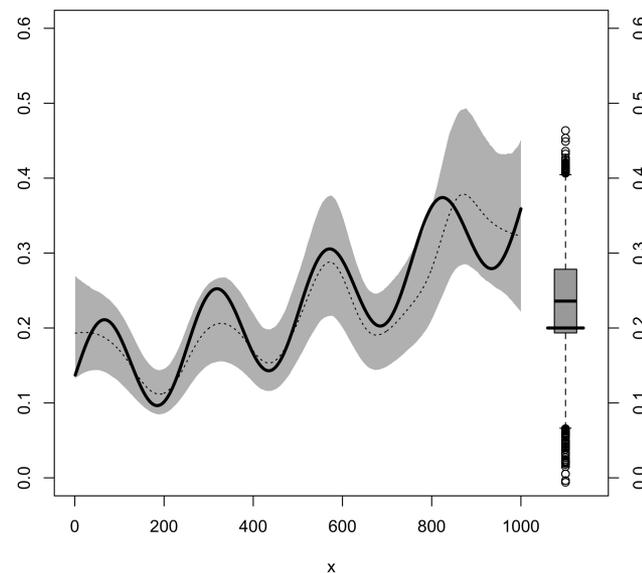
Characterising III

To estimate α_{rs}

$$Z'_i = Y(\mathbf{X}_i) / \hat{\mu}_0(\mathbf{X}_i)$$

use your favourite method (e.g. U-statistic approach) to estimate

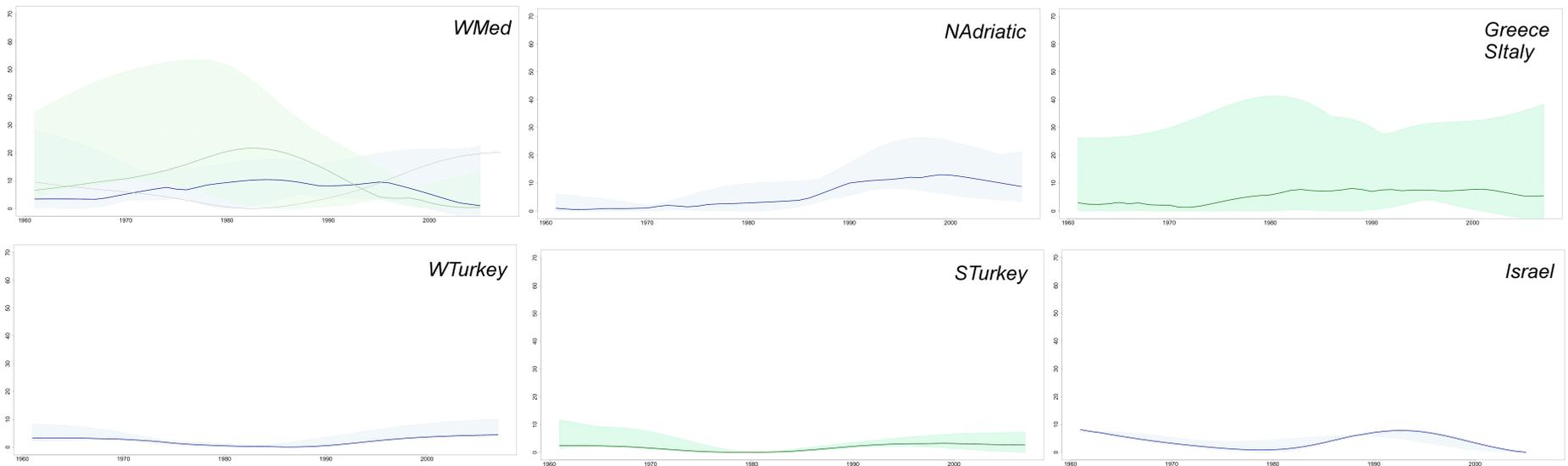
$$\mathbb{E}[Z' \overline{G'}_{1,\xi}^r(Z')] \text{ for } r = 1, 2$$



Source: Naveau et al., 2014. WRR 50

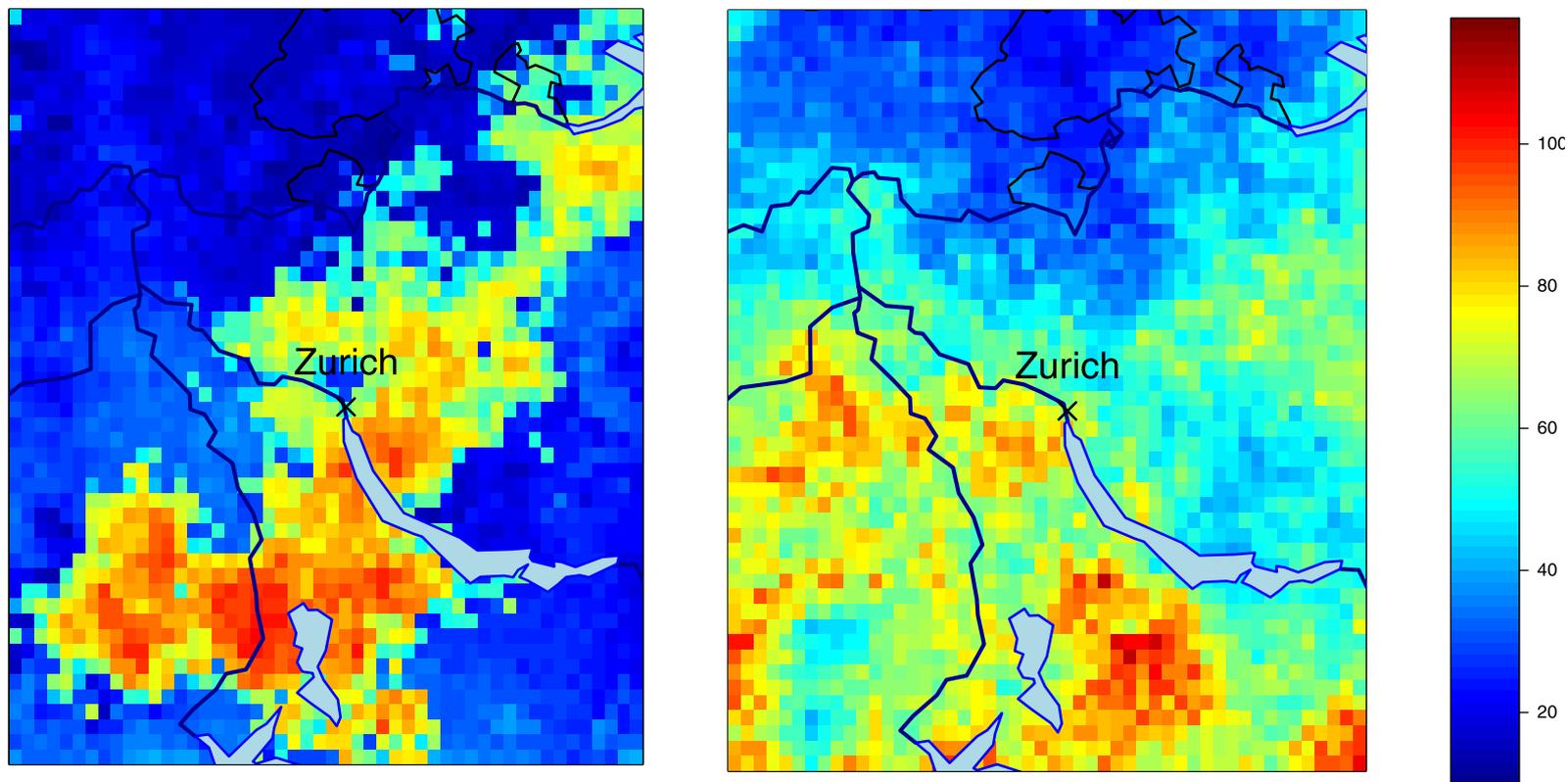
Characterising III

an attempt to model and understand the temporal evolution of precipitation extremes



Characterising IV

Max-stable models



One realisation from the Schlather and Brown Resnick models applied to daily summer precipitation from 1962 to 2008. Source: Davison et al., 2012. Statistical Science 27

Characterising \mathbf{V}

$$Y(\mathbf{s}) \sim GEV(\mu(\mathbf{s}), \sigma(\mathbf{s}), \xi(\mathbf{s}))$$

$$Y(\mathbf{s}) = \mu(\mathbf{s}) + \frac{\sigma(\mathbf{s})}{\xi(\mathbf{s})} [X(\mathbf{s})^{\xi(\mathbf{s})} - 1]$$

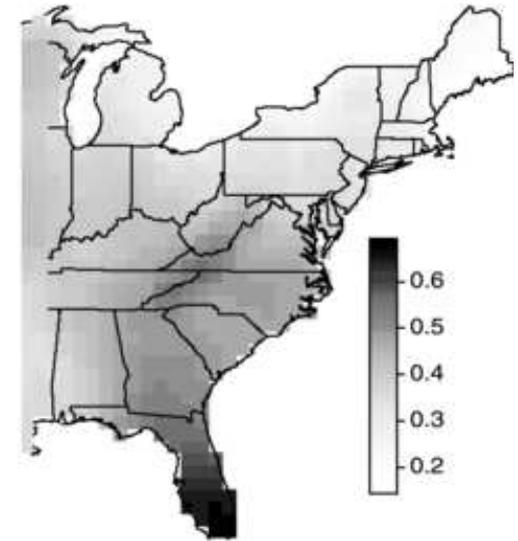
$$X(\mathbf{s}) = U(\mathbf{s})\theta(\mathbf{s}) \text{ with } U(\mathbf{s}) \sim GEV(1, \alpha, \alpha)$$

$$\theta(\mathbf{s}) = \left[\sum_{l=1}^L A_l w_l(\mathbf{s})^{1/\alpha} \right]^\alpha \text{ with } A_l \sim PS(\alpha)$$

$$Y(\mathbf{s}_i) | A_1, \dots, A_2, \dots, A_L \sim_{indep} GEV[\mu^*(\mathbf{s}_i), \sigma^*(\mathbf{s}_i), \xi^*(\mathbf{s}_i)]$$

Hierarchical spatial model for precipitation extremes

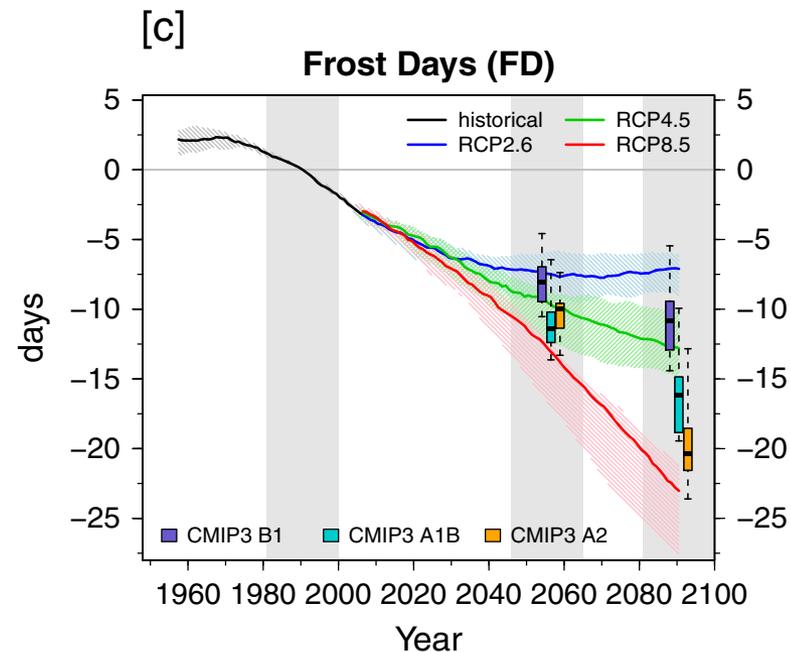
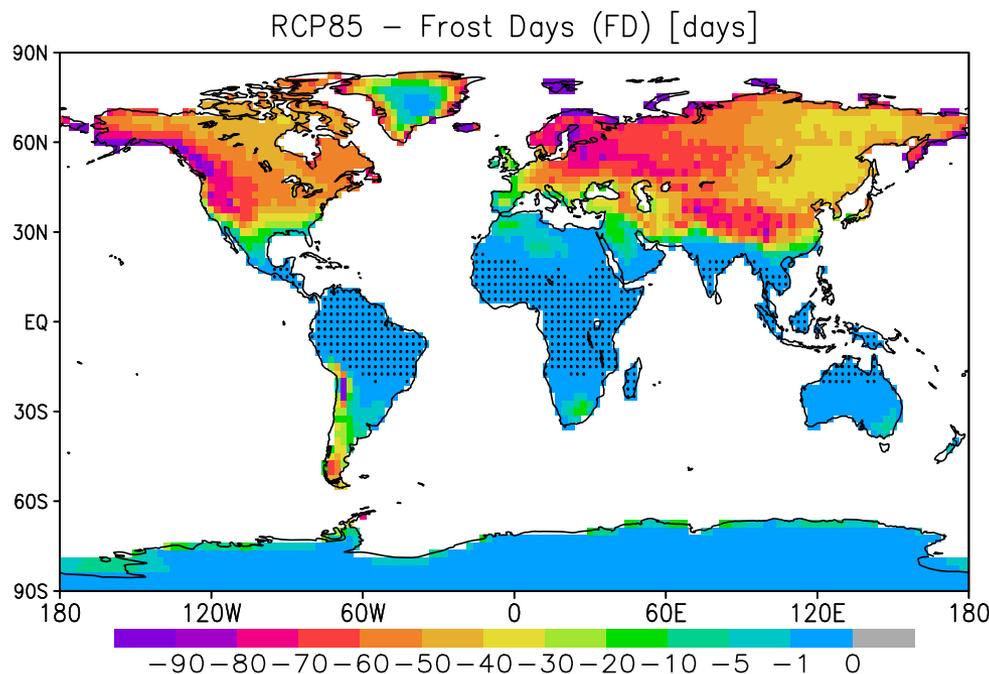
Source: Reich and Shalby 2012. Ann. Appl. Stat. 6



(c) GEV shape, posterior mean

Characterising VI

Index-based approach



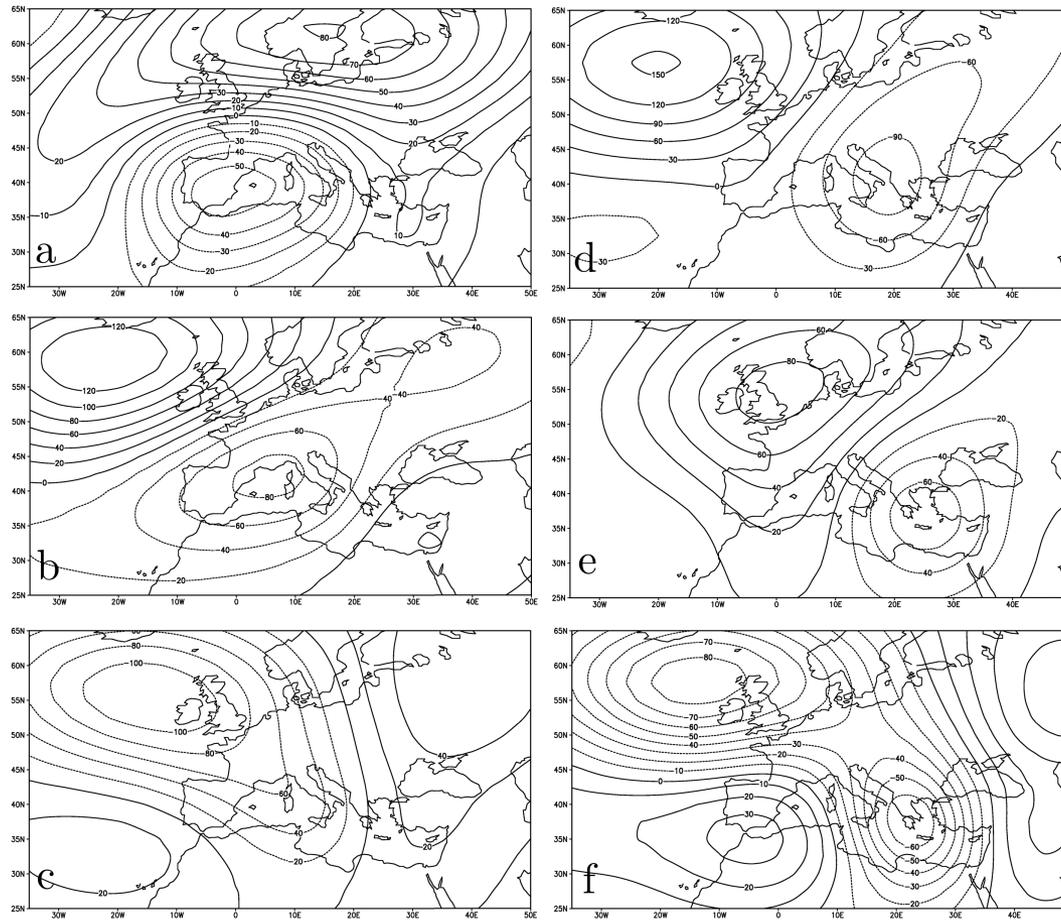
2081-2100 Changes w.r.t. 1981-2000. Source: Sillmann et al. 2013, JGR 118

Characterising

- Erroneous data
- Inhomogeneities
- Missing data
- inhomogeneous spatial coverage
- Complexity of the events is often not captured
- Computational and statistical issues when dealing with large data sets
- Constraints and assumptions for large regions

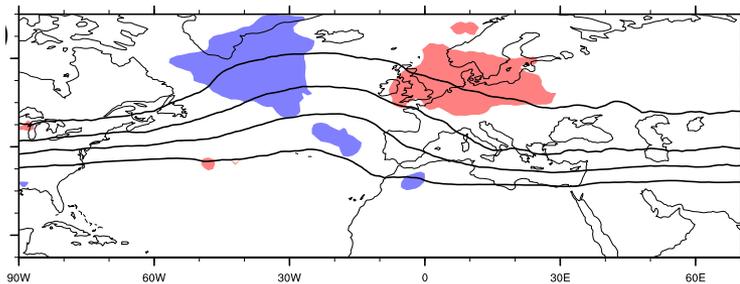
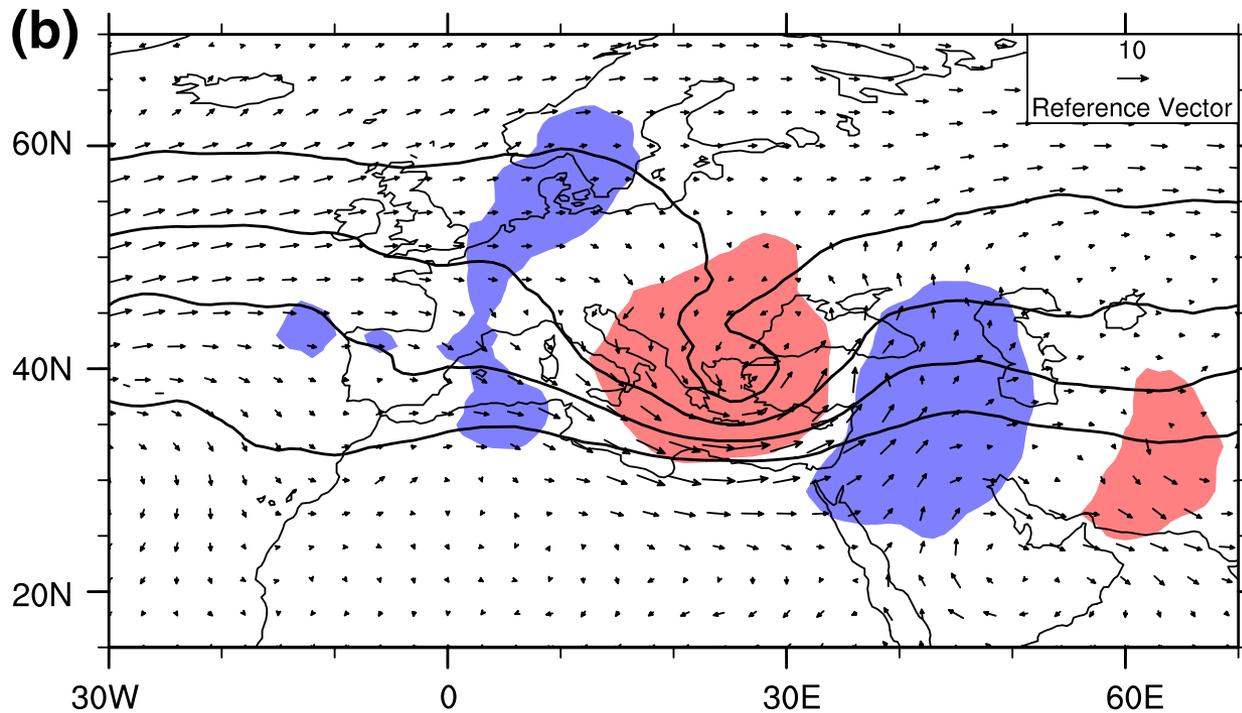
Predicting I

Synoptic patterns associated with extreme preci



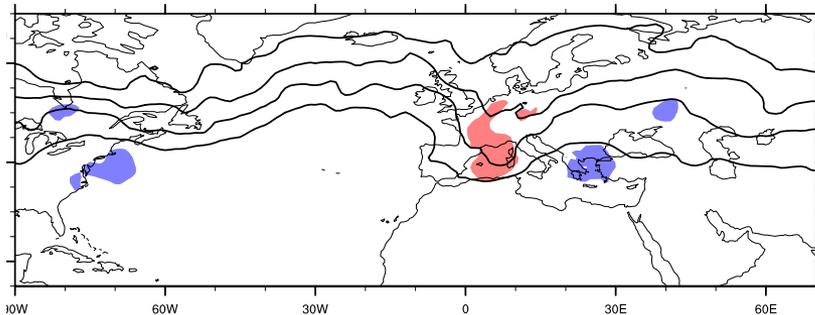
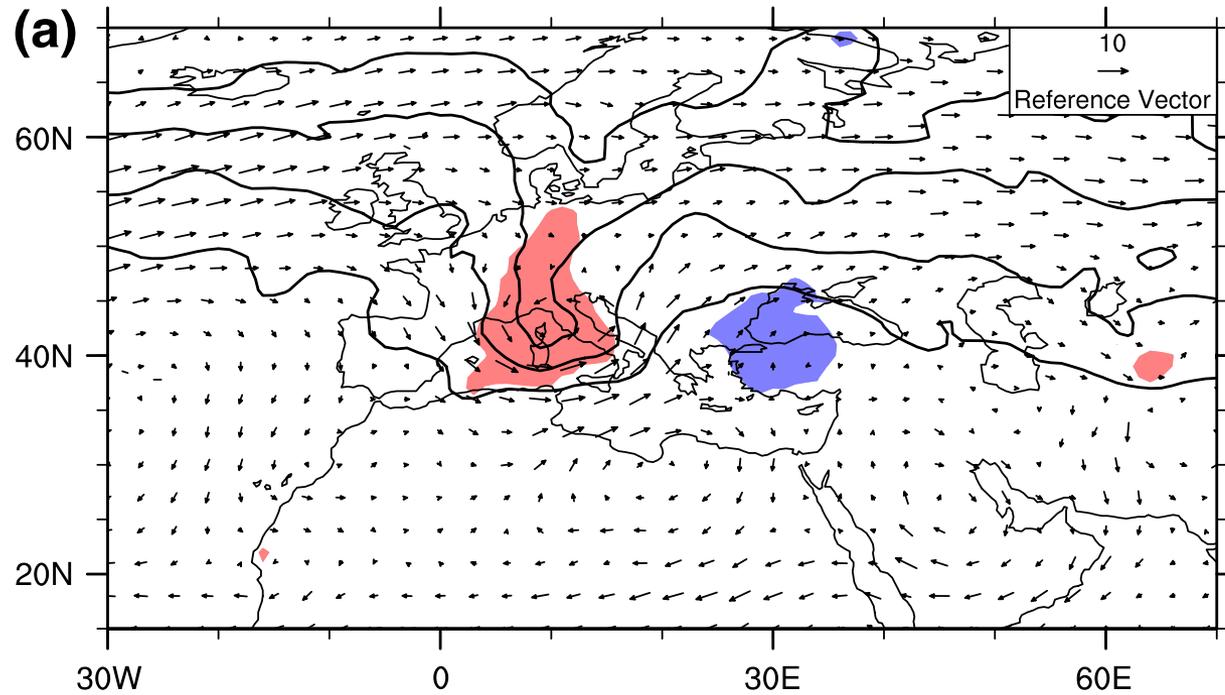
Z500 anomalies associated with precipitation extremes. Source: Toreti et al., 2010. NHESS 10

Predicting I



PV anomalies and wind at 850 hPa DJF associated with precip extremes in Western Turkey. Source: Toreti et al. 2016, Climate Dynamics in press

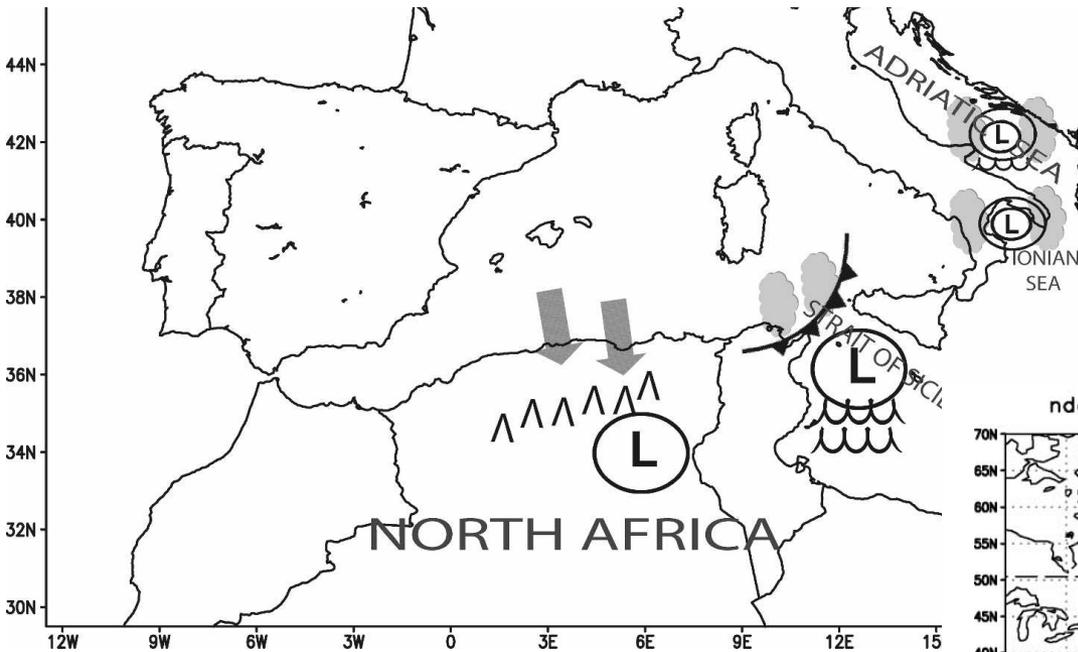
Predicting I



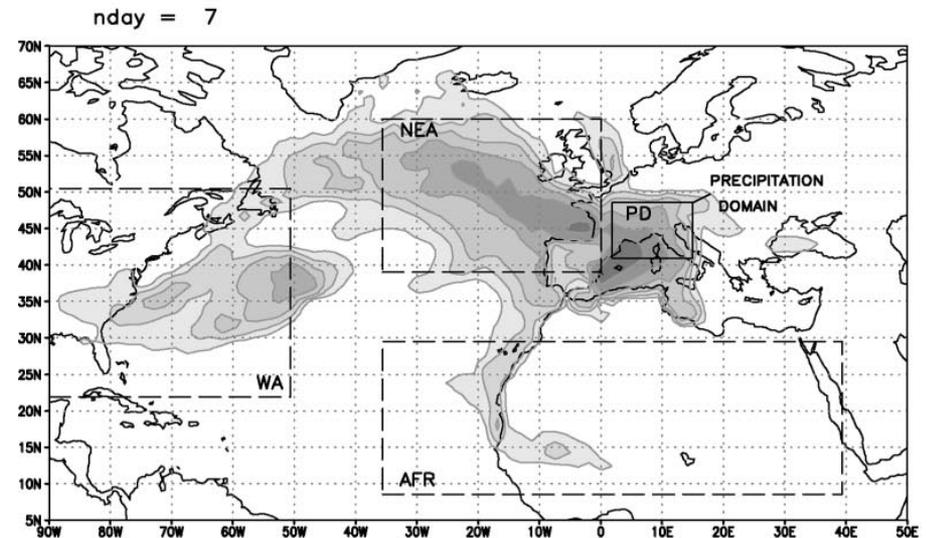
PV anomalies and wind at 850 hPa DJF associated with precip extremes in Po Valley. Source: Toreti et al. 2016, Climate Dynamics in press

Predicting I

Other factors contributing to precipitation extremes

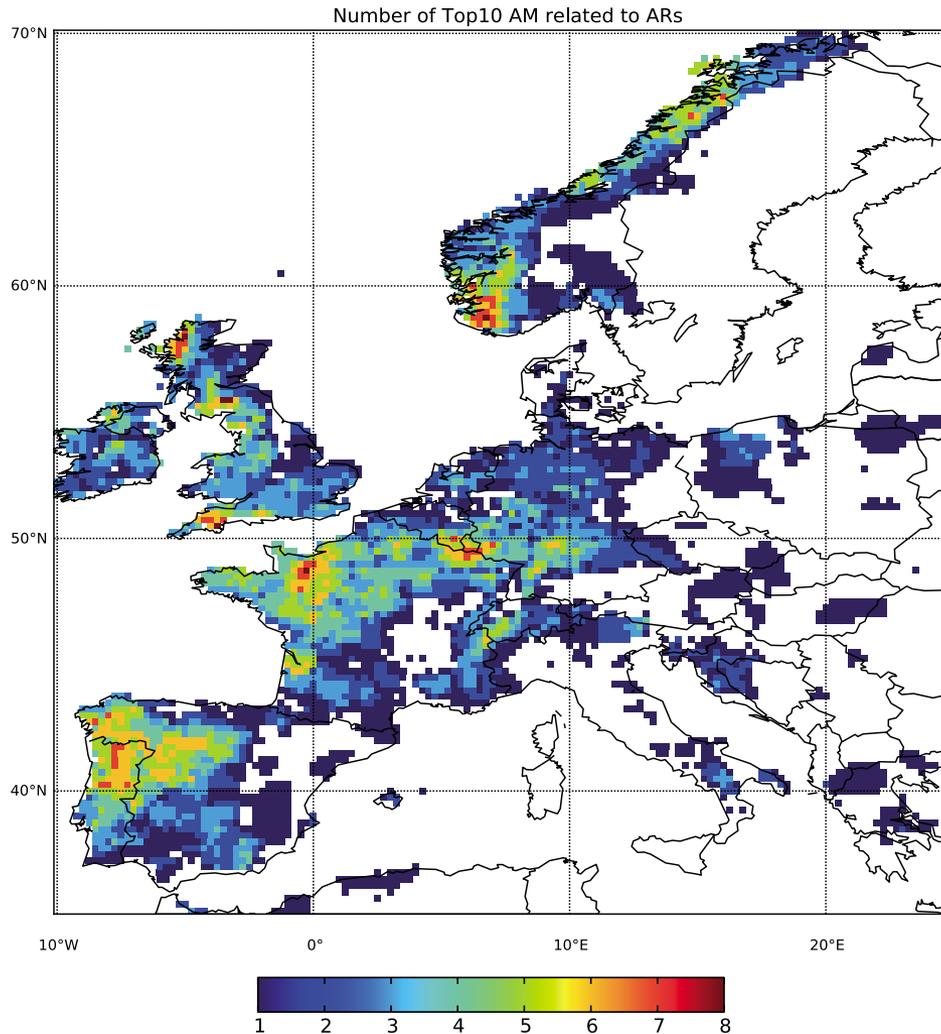


Source: Moscatello et al., 2008. MWR 136



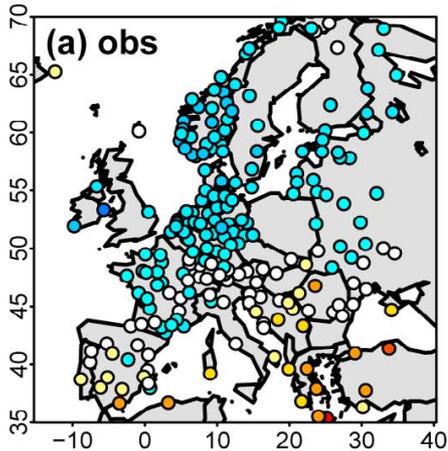
Source: Turato et al., 2004. J Hydromet 5

Predicting I

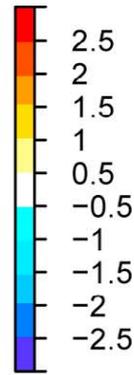


Predicting II

Cold Spell winter 2009-10

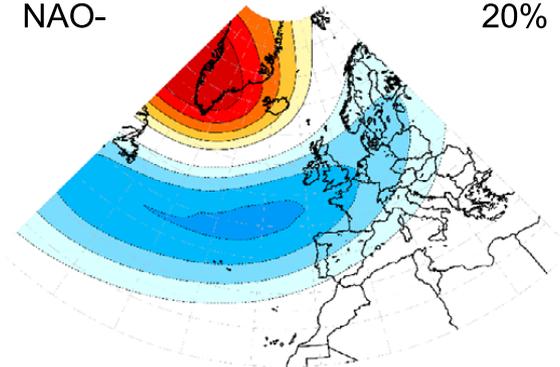


Normalised winter temperature anomalies

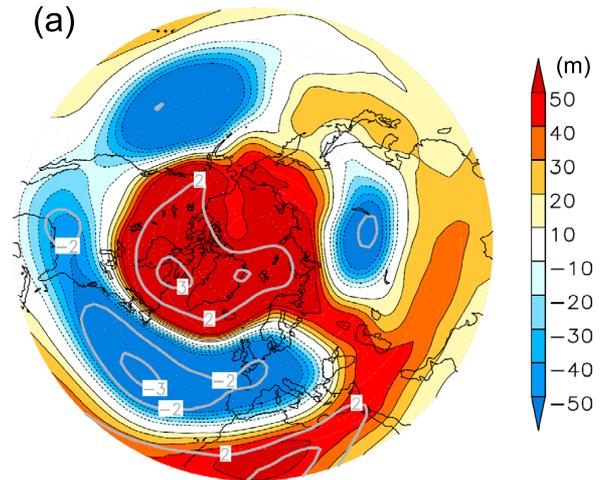


NAO-

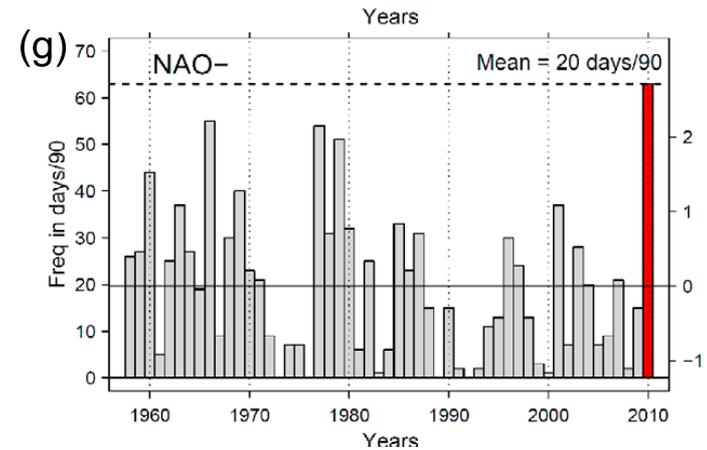
20%



Negative NAO

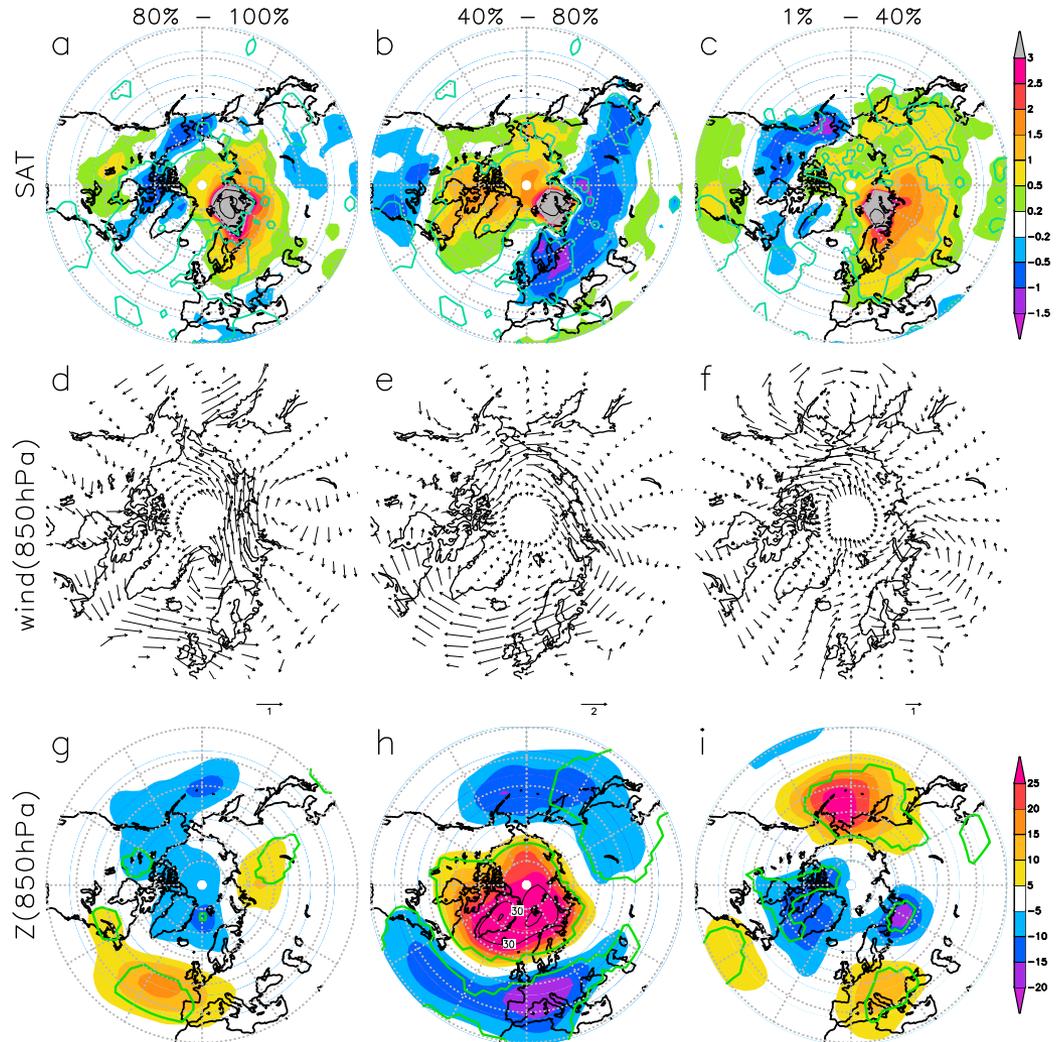
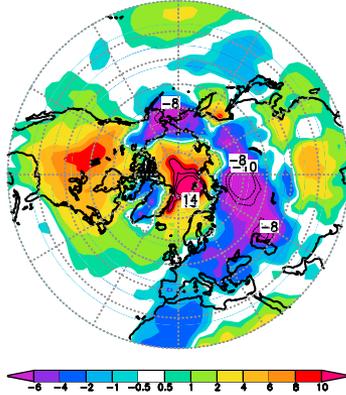


Anomalies of Z500



Predicting II

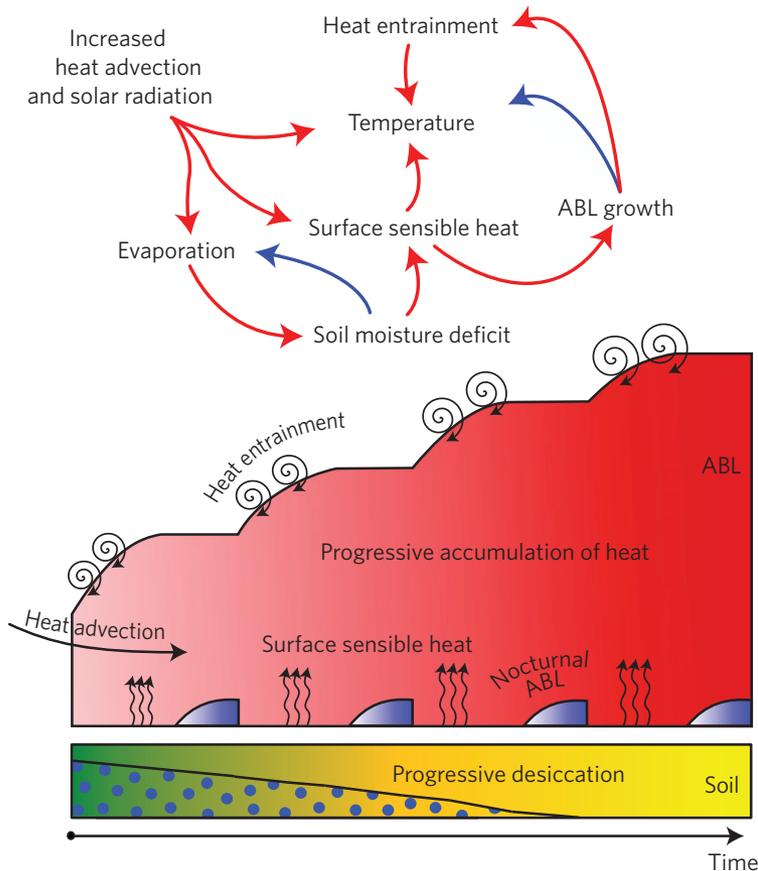
a) Jan 2006 SAT anom.



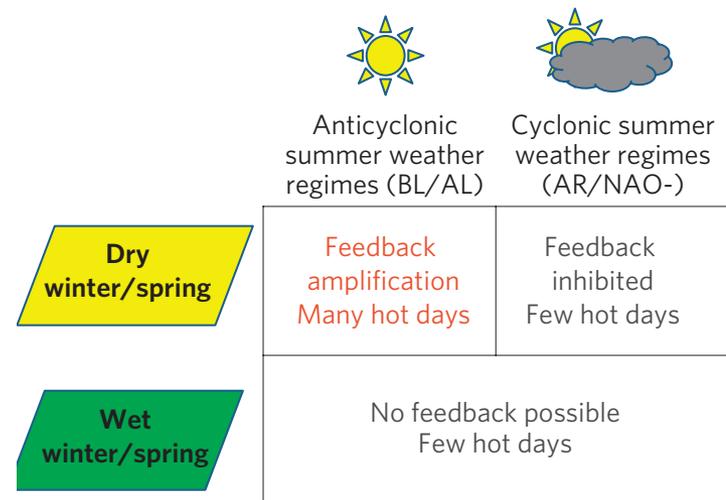
Nonlinear response to decrease in the Barents-Kara sea ice concentration.

Source: Petoukhov and Semenov 2010, JGR 115

Predicting III

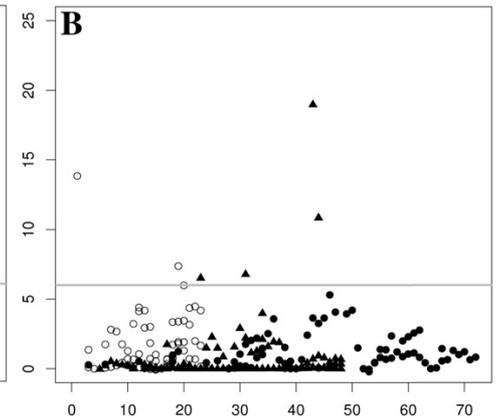
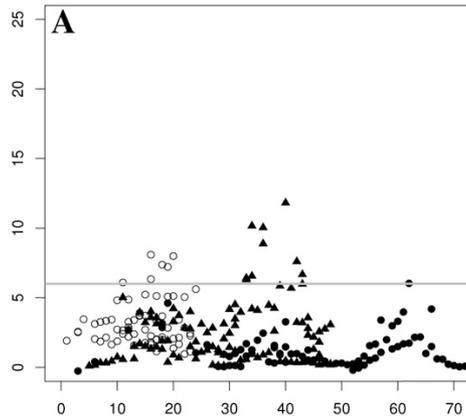
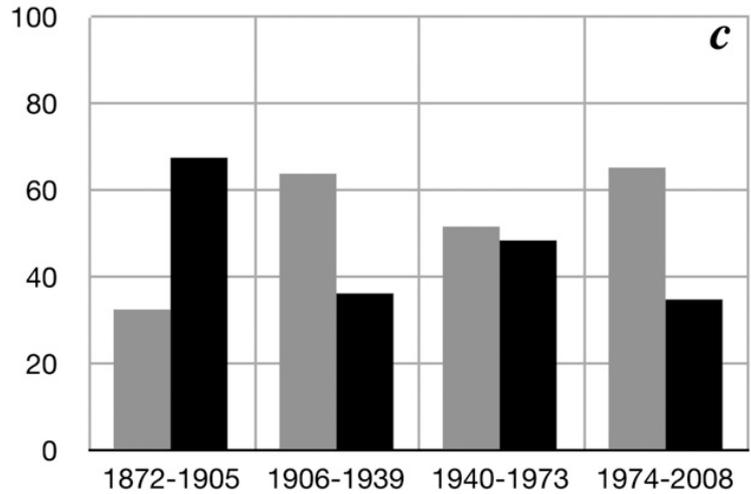
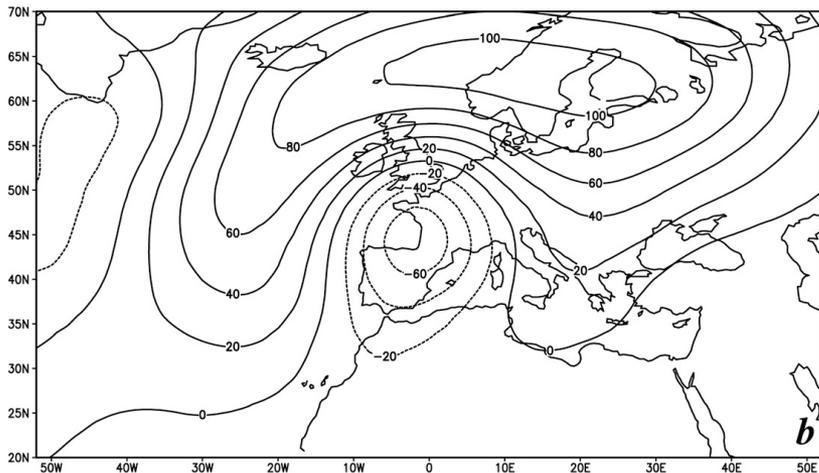
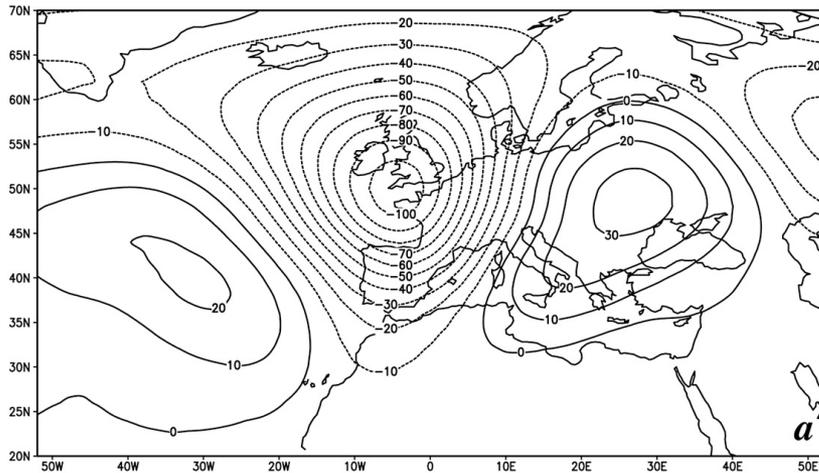


Soil moisture - air temperature interaction in the development of mega heat waves. Source: Miralles et al. 2014, Nature Geoscience 7.



Sensitivity of hot days to summer atm circulation and soil moisture conditions. Source: Quesada et al. 2012, NCC 2.

Predicting IV

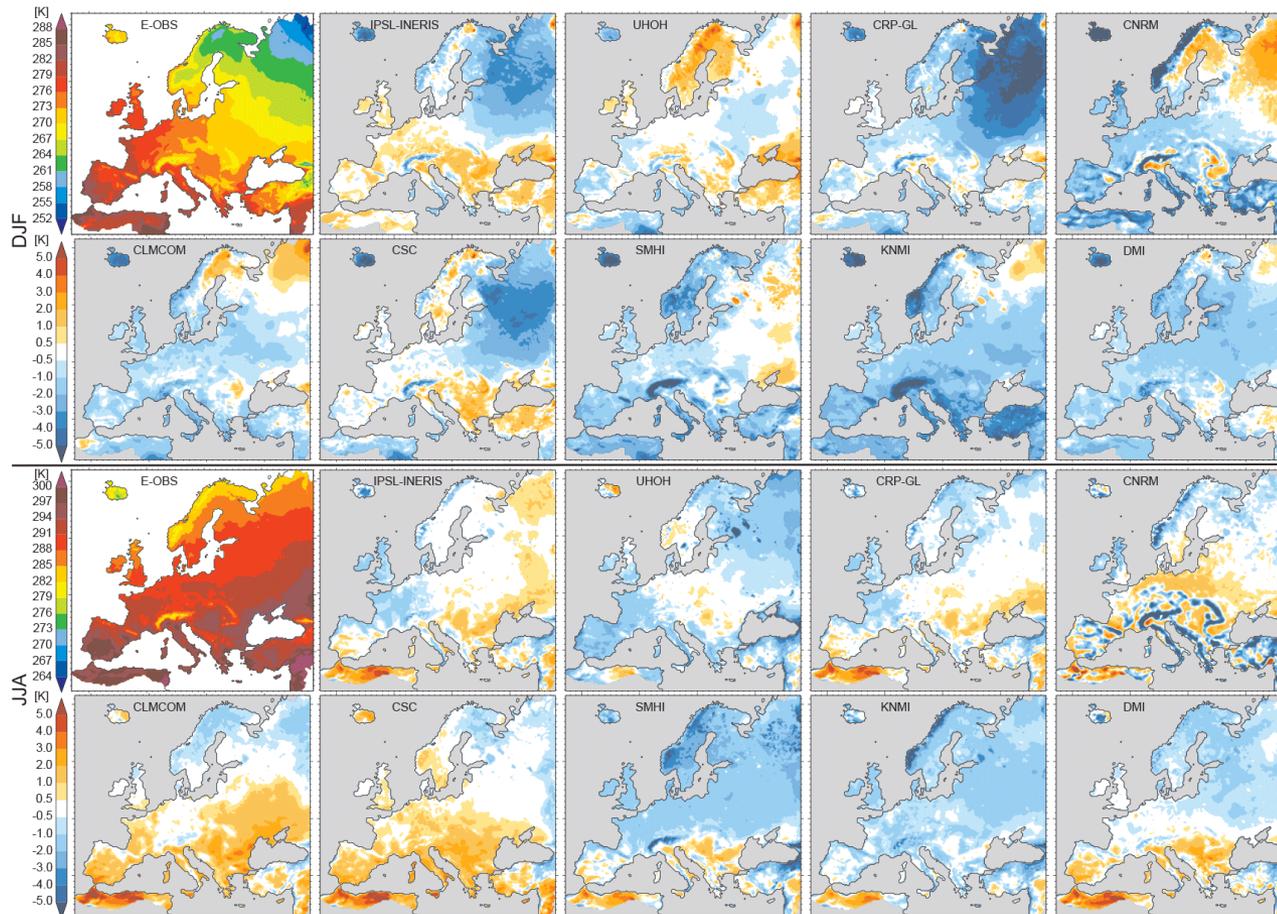


Z500 anomalies associated with Debris Flows in the souther Swiss Alps and convective time scales. Source: Toreti et al. 2013, JAMC 52

Predicting

- Interactions of different phenomena acting at different spatio-temporal scales
- coarse resolution of available reanalyses
- nonlinear processes

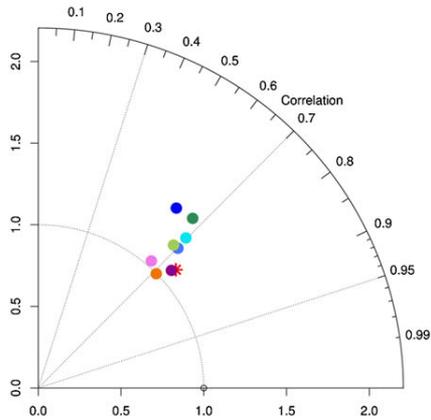
Evaluating I



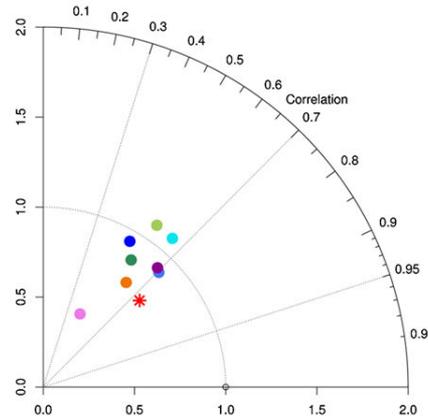
*Bias in mean seasonal temperature of EUR-11 cordex runs 1989-2008.
Source: Kotlarski et al. 2014, Geosci. Model Dev., 7*

Evaluating II

c



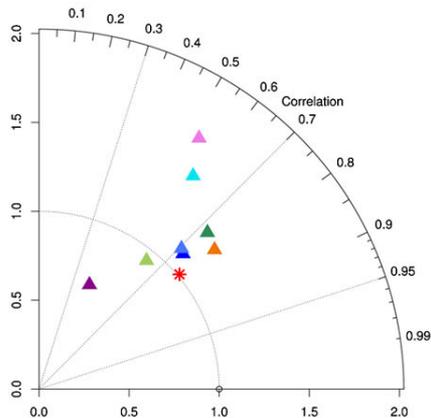
e



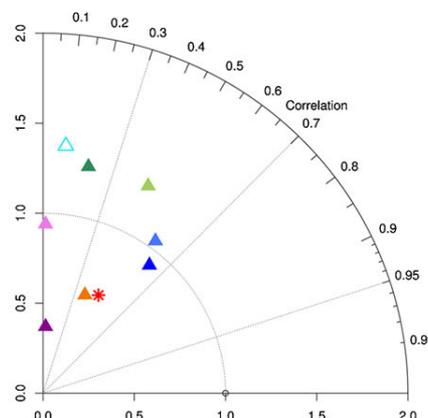
- CMCC
- HadES
- MIROC
- CNRM
- INM
- MRI
- * ensemble mean
- HadCC
- IPSL

Taylor diagrams for estimated 50 year return levels in winter and summer over (c, d) northern Eurasia and (e, f) North America . Source: Toreti et al. 2013, GRL 40.

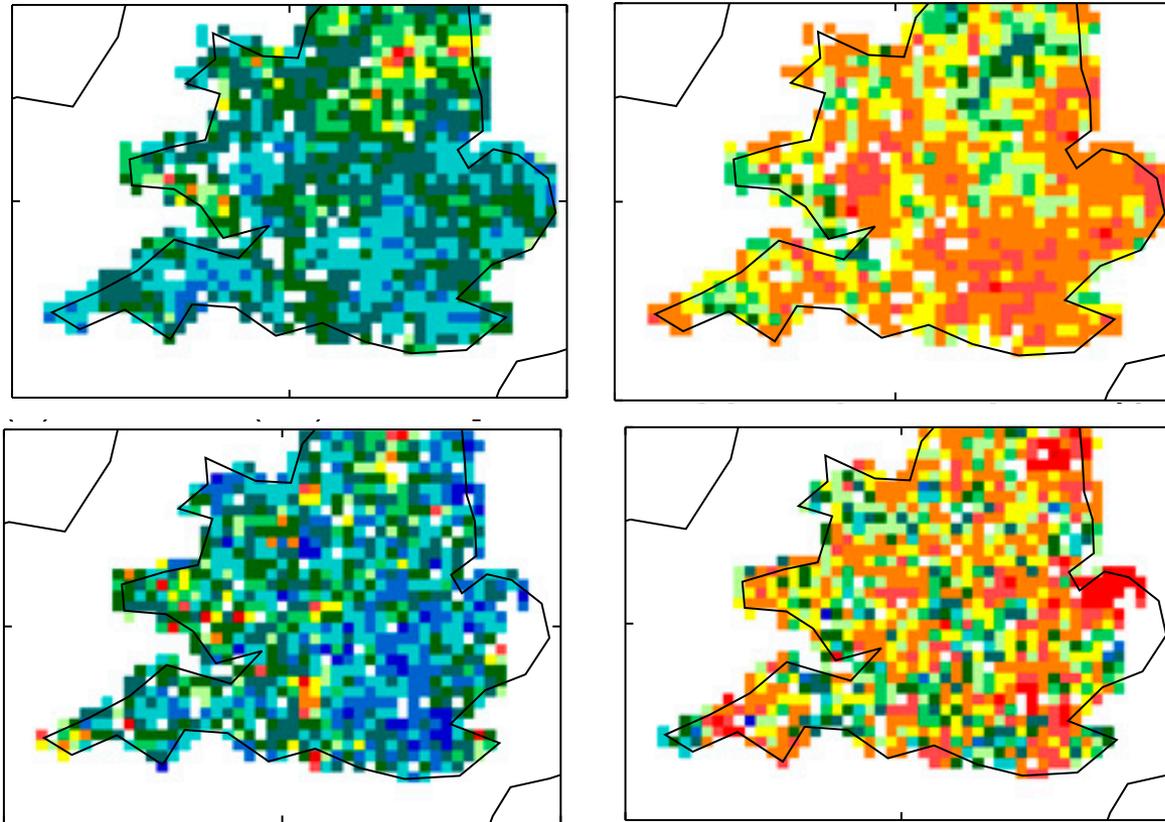
d



f



Evaluating III



*Estimated GPD parameters for winter (DJF) precipitation extremes. Model w.r.t. observations
Source: Chan et al. 2014, J Clim 27*

Evaluating IV

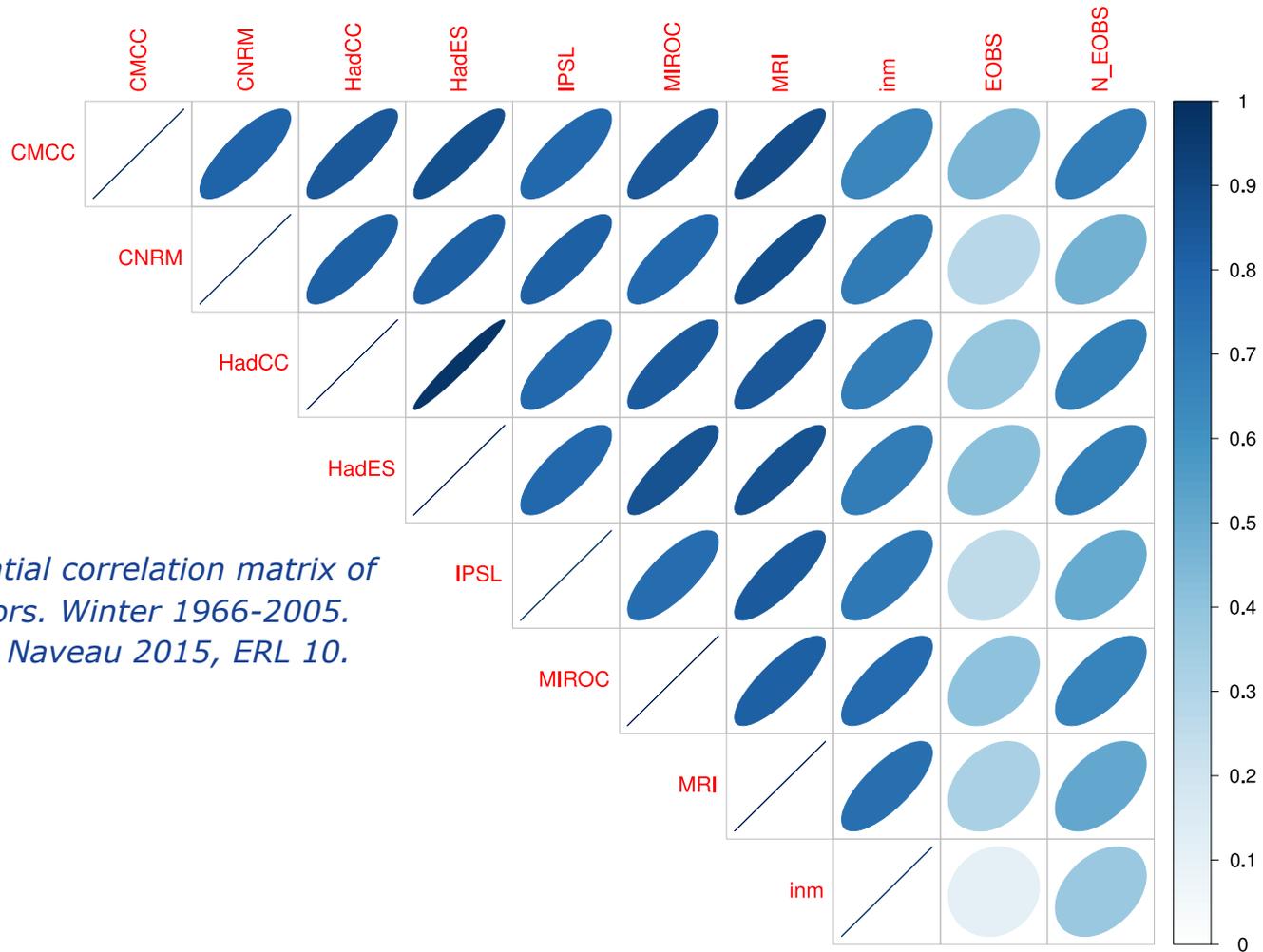
non-parametric approach

$$A = \frac{mn}{N} \int_{-\infty}^{\infty} \frac{(\bar{G}_m - \bar{F}_n)^2}{(\bar{H}_N)} dH_N$$
$$\bar{H}_N = \frac{n\bar{F}_n + m\bar{G}_m}{N}; \quad N = n + m$$

$$T = (A - \mathbb{E}(A))\sigma_A^{-1}$$

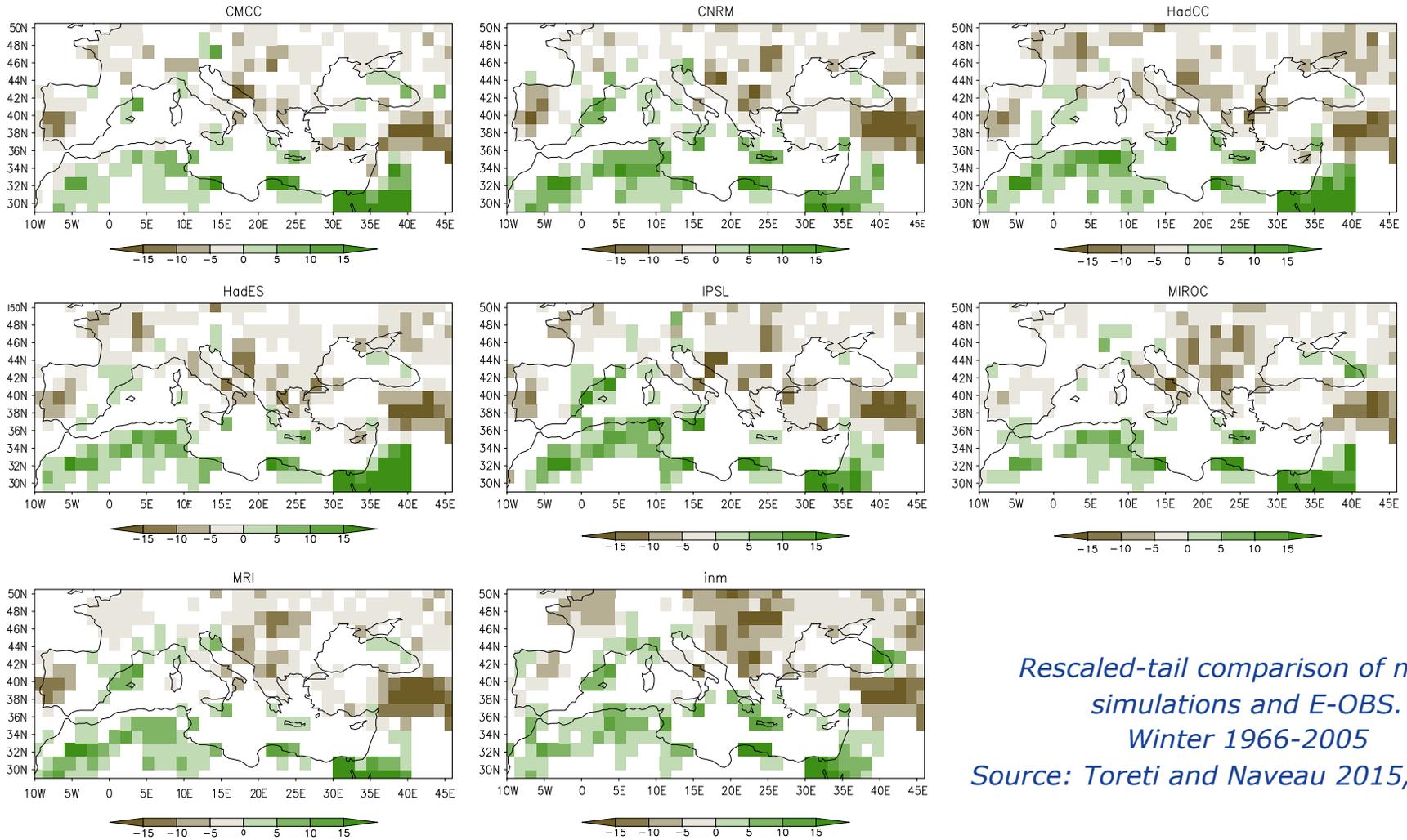
$$I(f_e; g_e) = \mathbb{E}_{f_e} \left\{ \log \left(\frac{f_e(X_e/\mu_0^{X_e})}{g_e(X_e/\mu_0^{X_e})} \right) \right\}$$

Evaluating IV



Spearman-based spatial correlation matrix of the tail scaling factors. Winter 1966-2005. Source: Toreti and Naveau 2015, ERL 10.

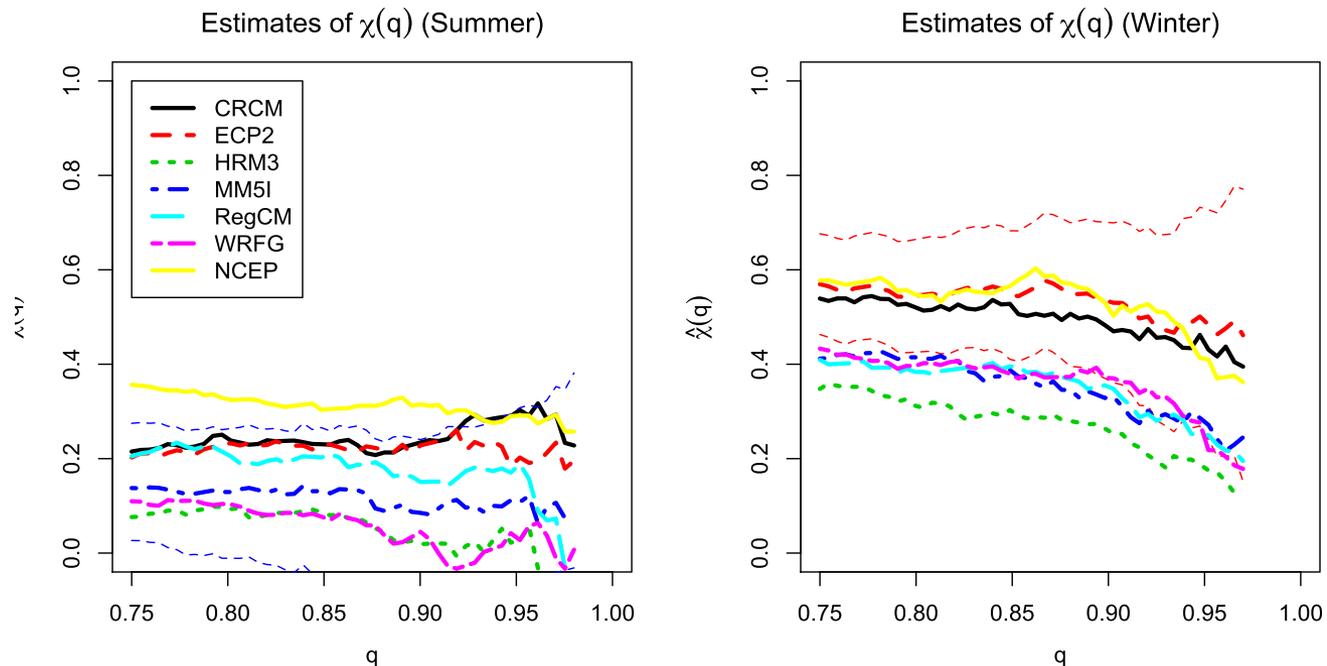
Evaluating IV



*Rescaled-tail comparison of model
simulations and E-OBS.
Winter 1966-2005
Source: Toreti and Naveau 2015, ERL 10.*

Evaluating V

$$\chi(q) = 2 - \frac{\log \mathbb{P}(Z_2 < z_q, Z_1 < z_q)}{\log \mathbb{P}(Z_2 < z_q)}$$

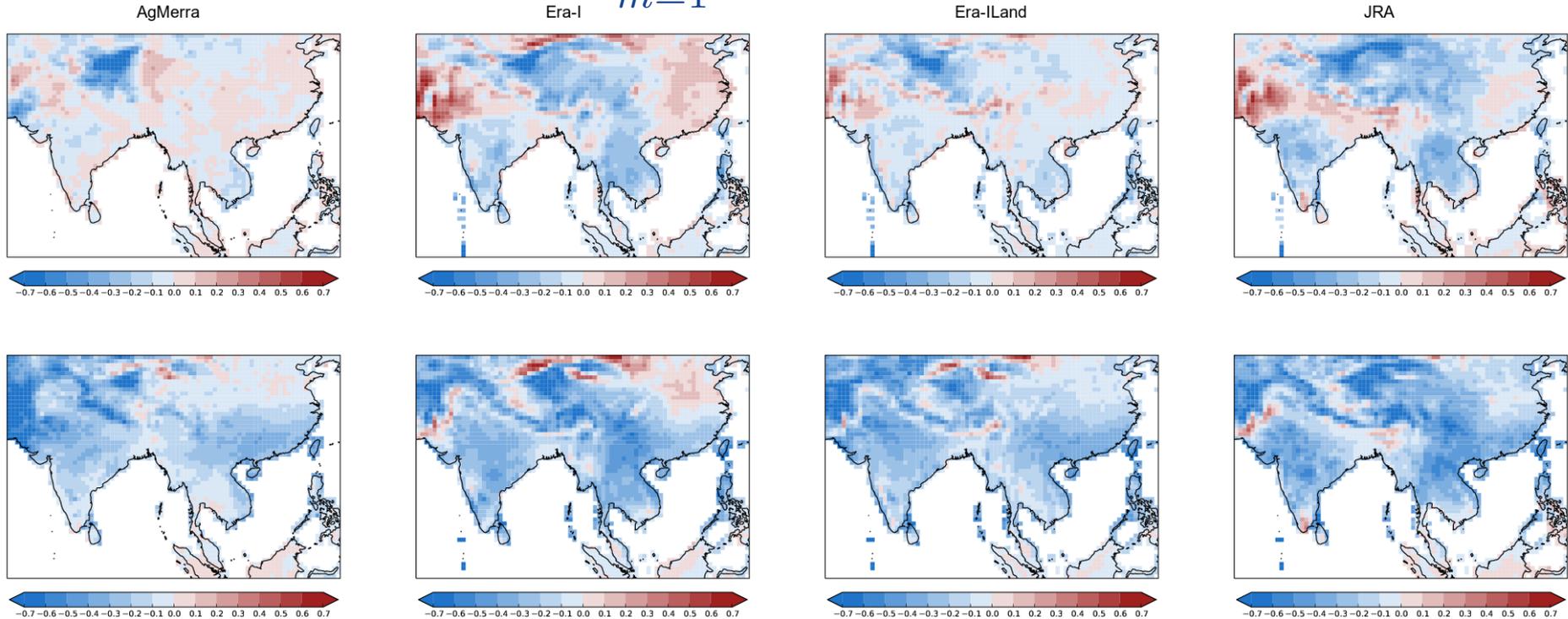


Tail dependence of NARCCAP RCMs w.r.t NCEP reanalysis over Pacific region.

Source: Weller et al. 2013, JGR 118.

Evaluating VI

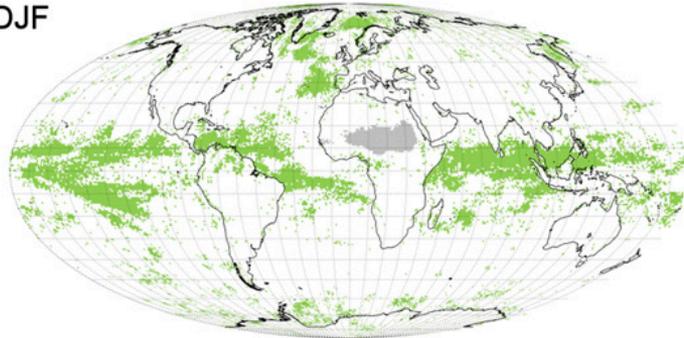
$$D(x) = \sum_{m=1}^{12} p_m(x) \log_2 (12 p_m(x))$$



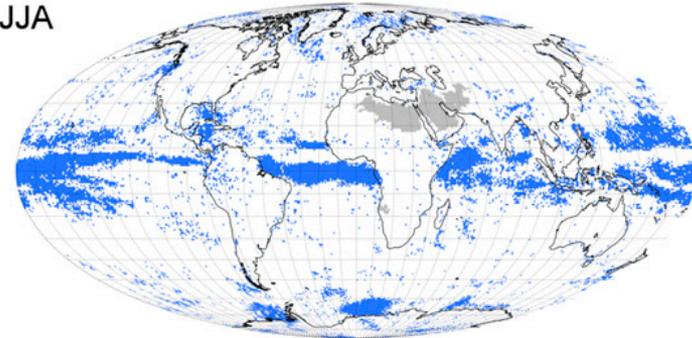
Comparison of estimated precipitation seasonality of reanalyses with Aphrodite and Chirps. Source: Ceglar et al. 2016, submitted.

Evaluating VII

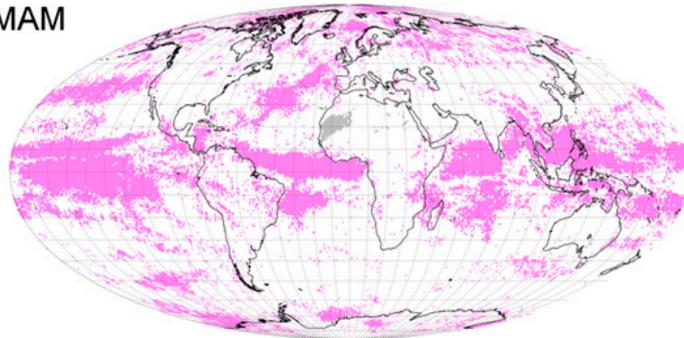
DJF



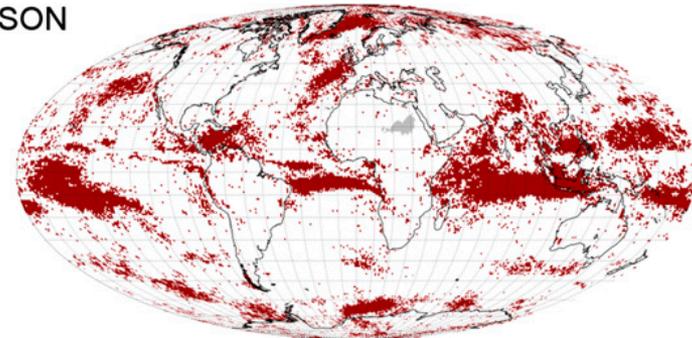
JJA



MAM

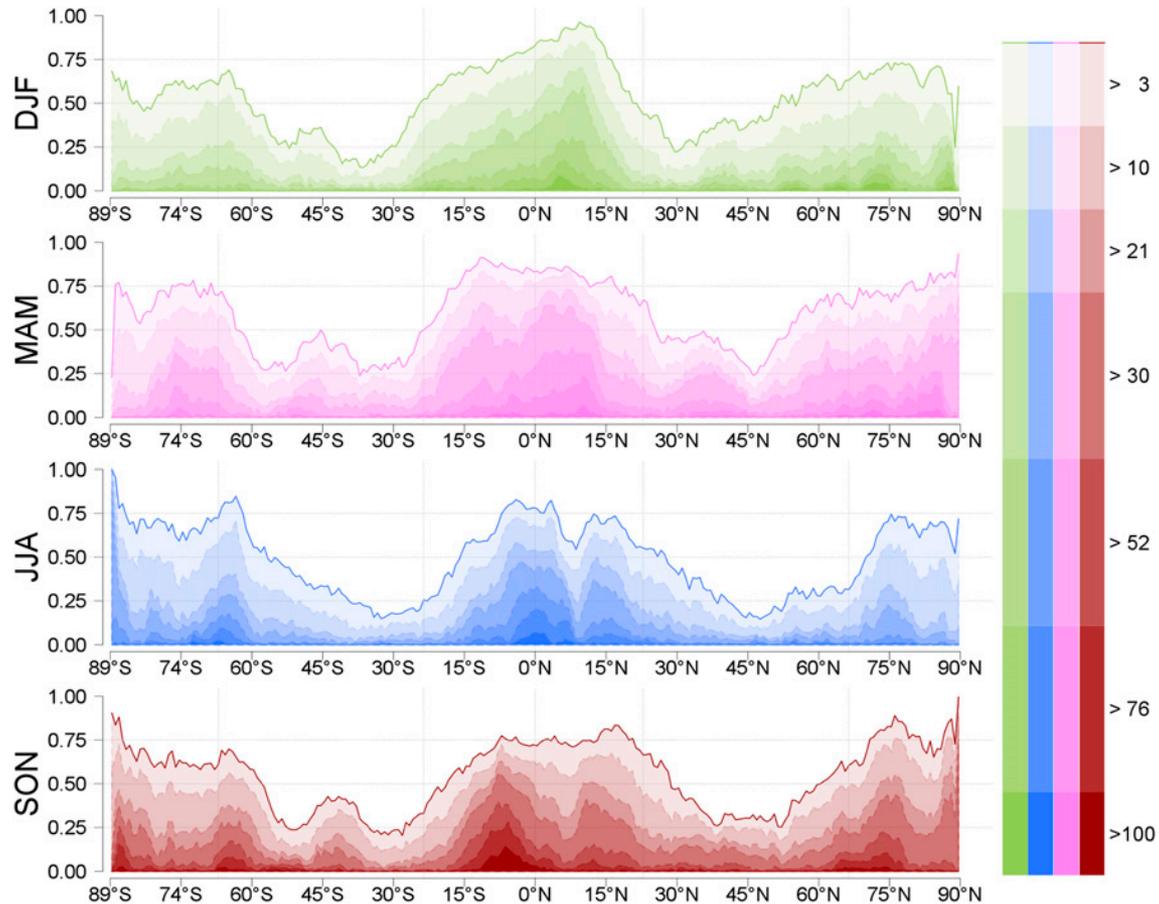


SON



Seasonal maps of points having a minimal length to estimate internal variability longer than 30 years. Source: Schindler et al. 2015, J Climate 28

Evaluating VII

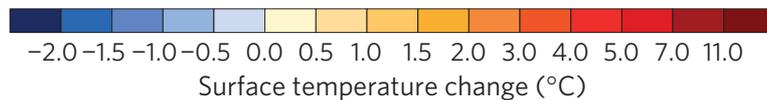
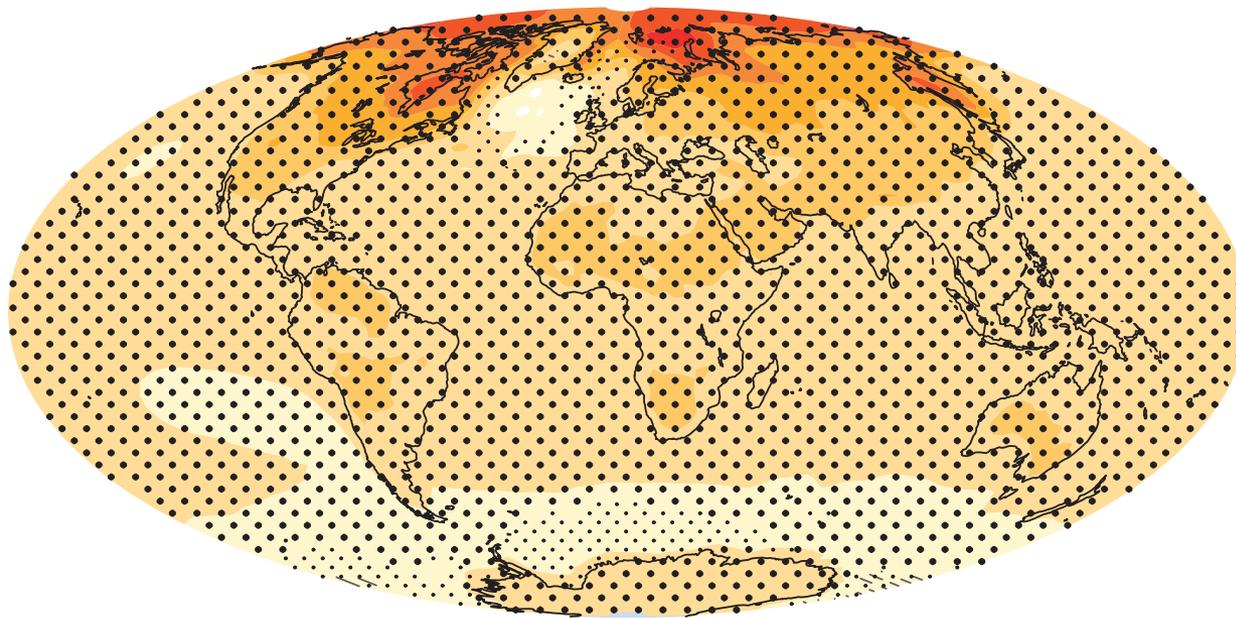


Zonal plots of % of points having minimum length longer than some fixed thresholds. Source: Schindler et al. 2015, J Climate 28

Evaluating VIII

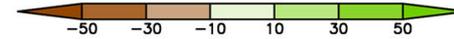
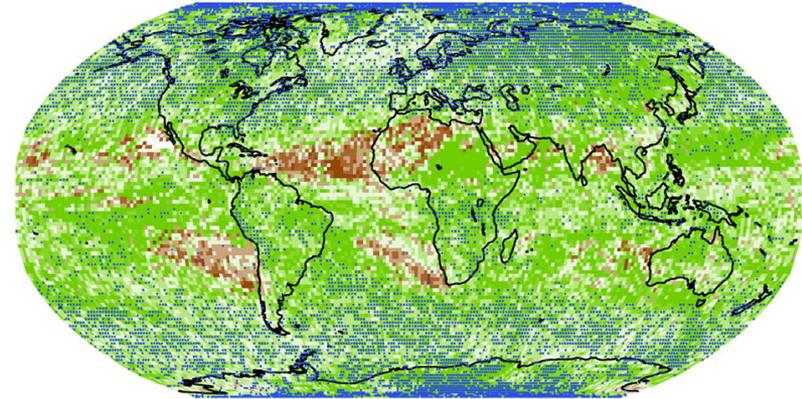
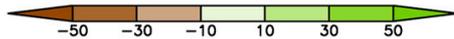
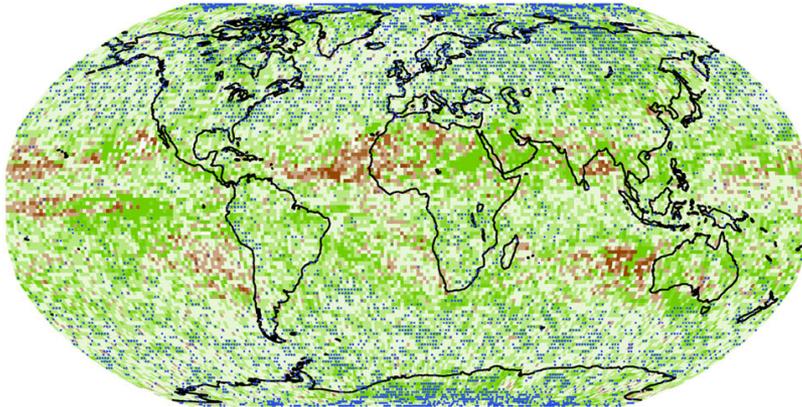
RCP85: 2016–2035

DJF

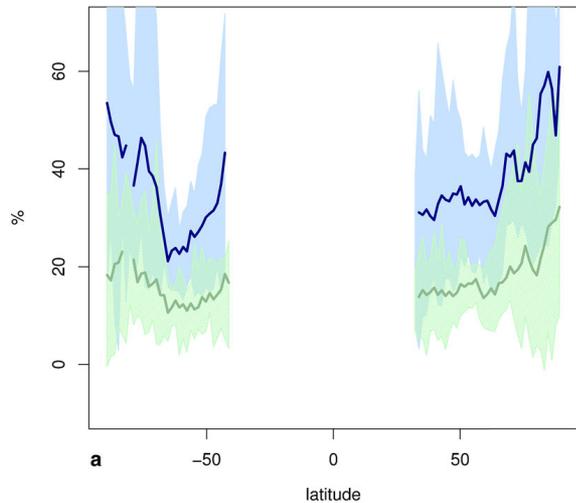


Temperature changes w.r.t. 1986-2005. Source: Knutti and Sedlacek 2013, NCC1716

Evaluating IX



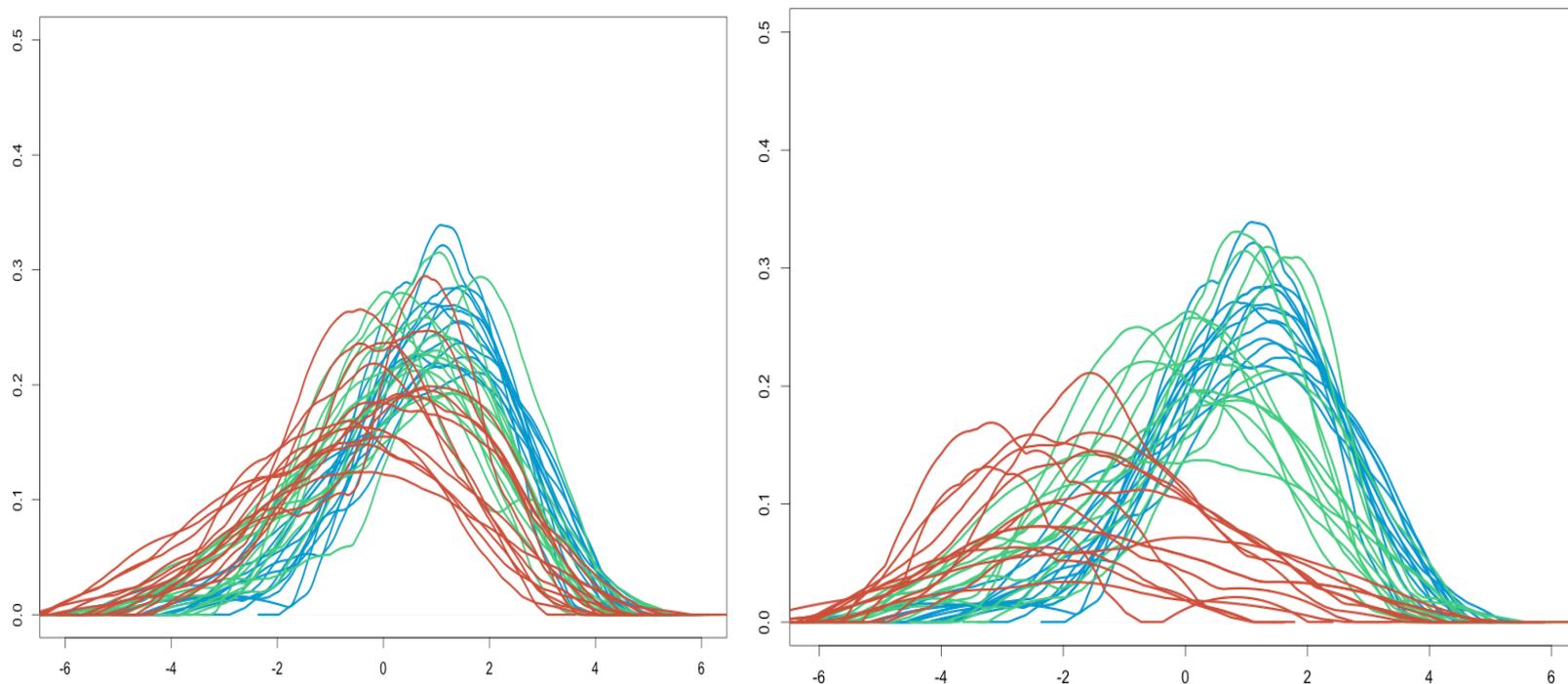
Ensemble mean changes under RCP8.5. 2020-2059 and 2060-2099 w.r.t. 1986-2005 during winter. Source: Toreti et al., 2013



Changes in 50-y ret. levels of extreme precipitation

Zonal mean changes under RCP8.5. 2020-2059 (green) and 2060-2099 (blue) w.r.t. 1986-2005 during winter. Source: Toreti et al. 2013, GRL 40

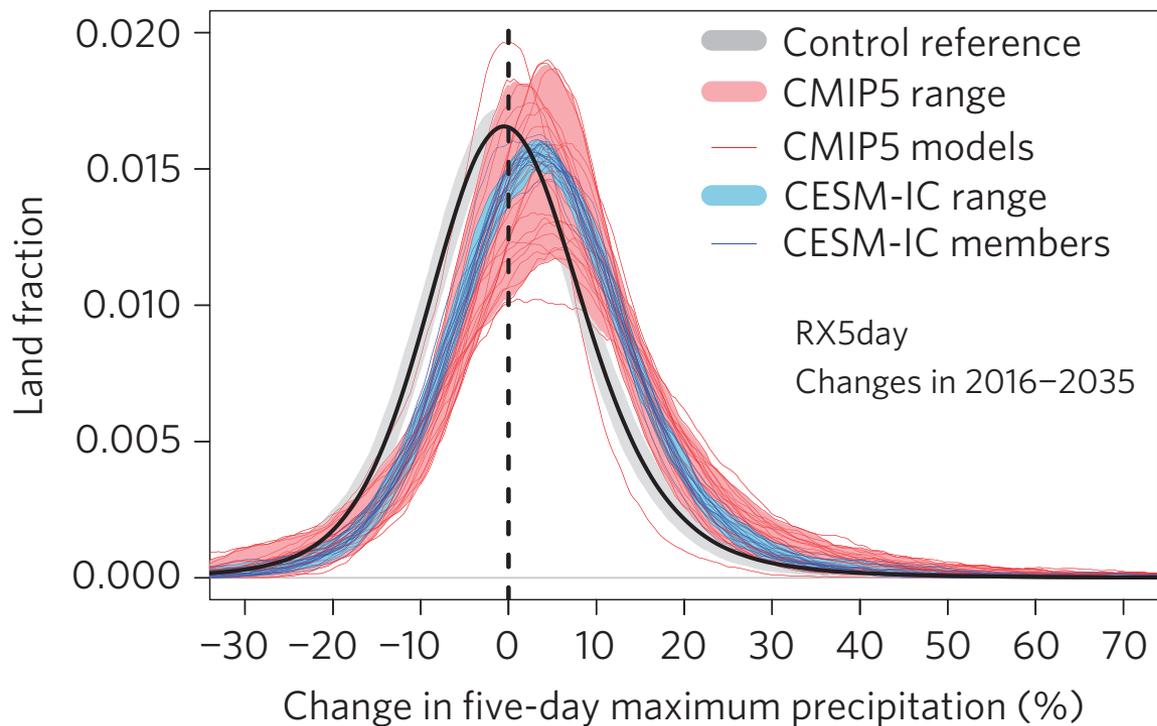
Evaluating X



Estimated PDFs of the (transformed) spatial extension of early cold spells over Europe. Blue lines are associated with the historical simulation (1976-2005), green lines with the mid-century (2020-2049) and red lines with the end of the century (2070-2099). Left Panel: RCP4.5. Right Panel: RCP8.5.

Source: Toreti et al., in preparation

Evaluating XI



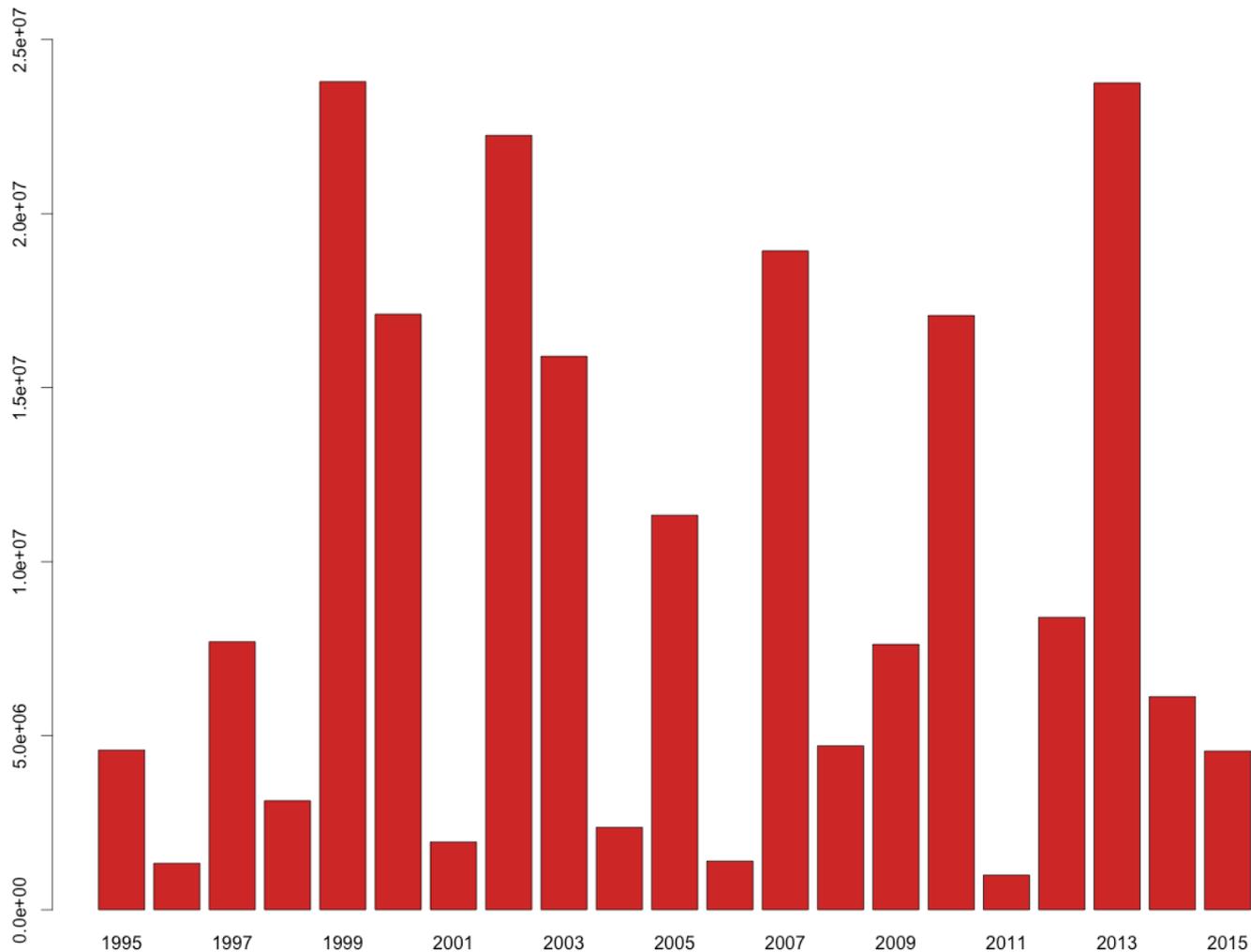
Spatial pdfs of changes in land points 66° N - 66° S

Source: Fischer et al. 2013, NCC 3

Evaluating

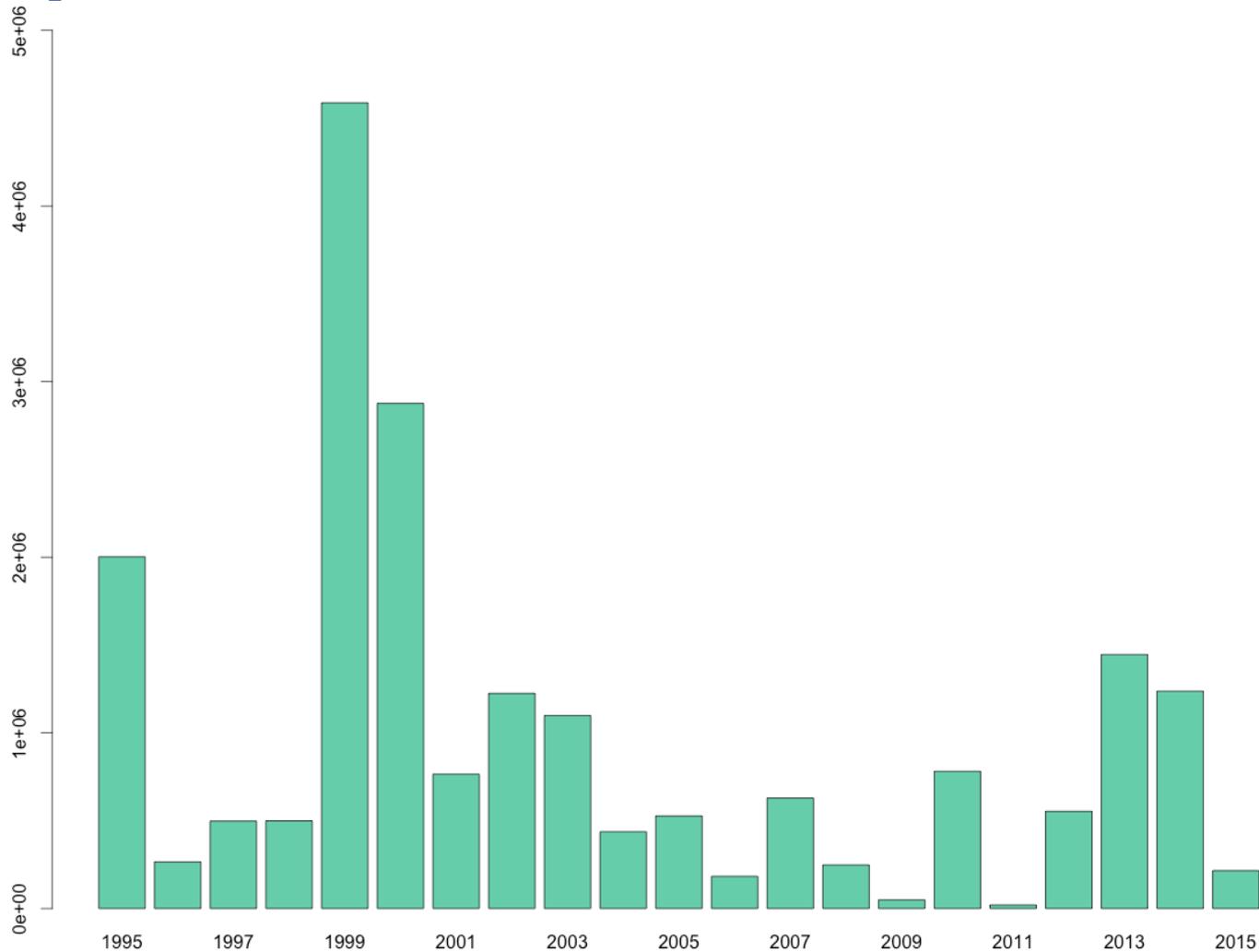
- Grid based evaluations
- complexity of extreme events not really captured
- Gridded observations not anymore available at the current resolution of regional climate models

Impacts I



Total damages (10³ \$) caused by extremes events (floods, drought, extreme temp, storms) in Europe. Data from: EM-DAT, The CRED/OFDA International Disaster Database 2016.

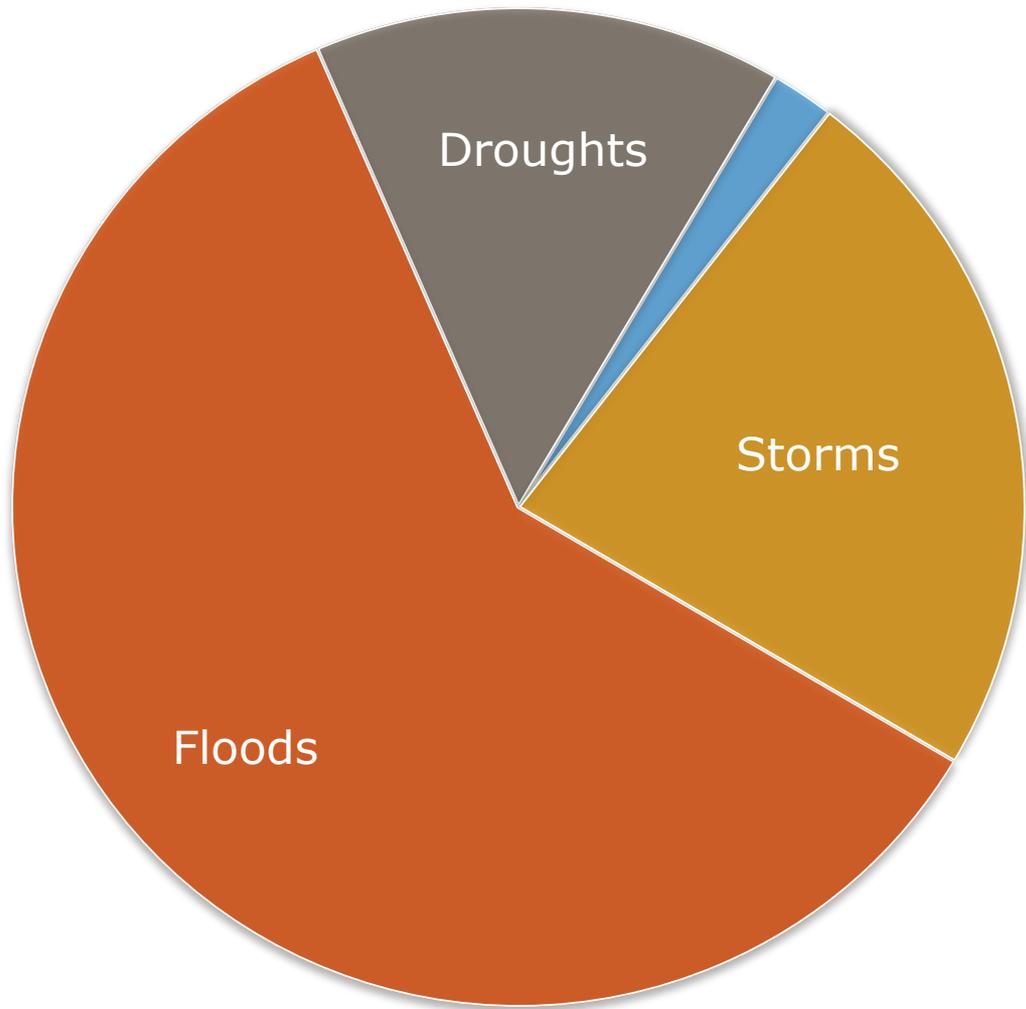
Impacts I



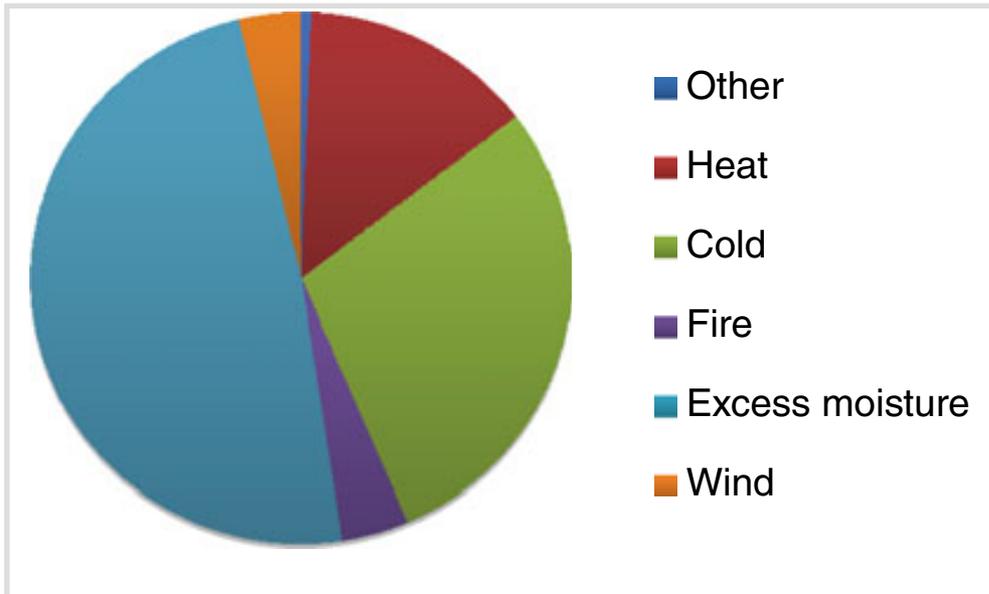
Total number of people affected by extremes events (floods, drought, extreme temp, storms) in Europe. Data from: EM-DAT, The CRED/OFDA International Disaster Database 2016.

Impacts II

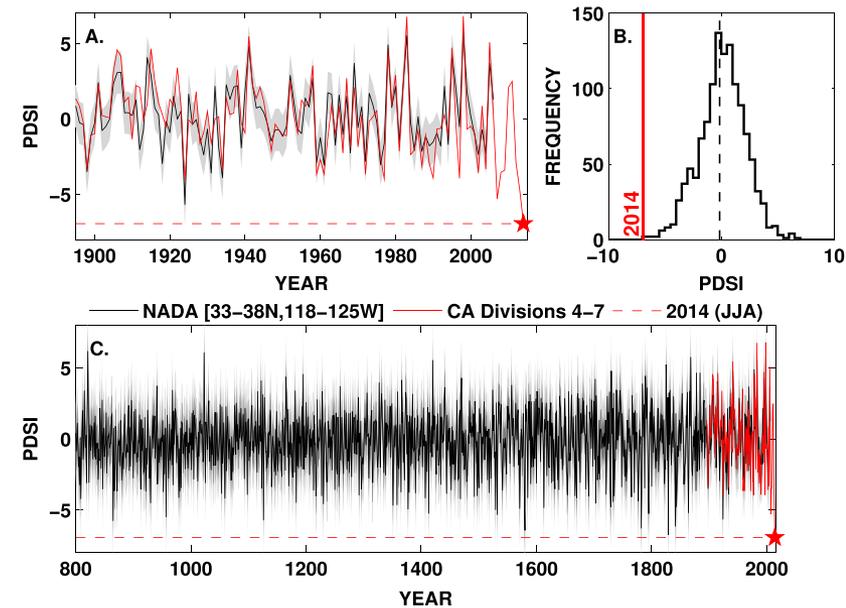
2003-2013, Developing Countries. Average percentage share of damage and loss to crops by type of hazard. Adapted from FAO, 2015.



Impacts II

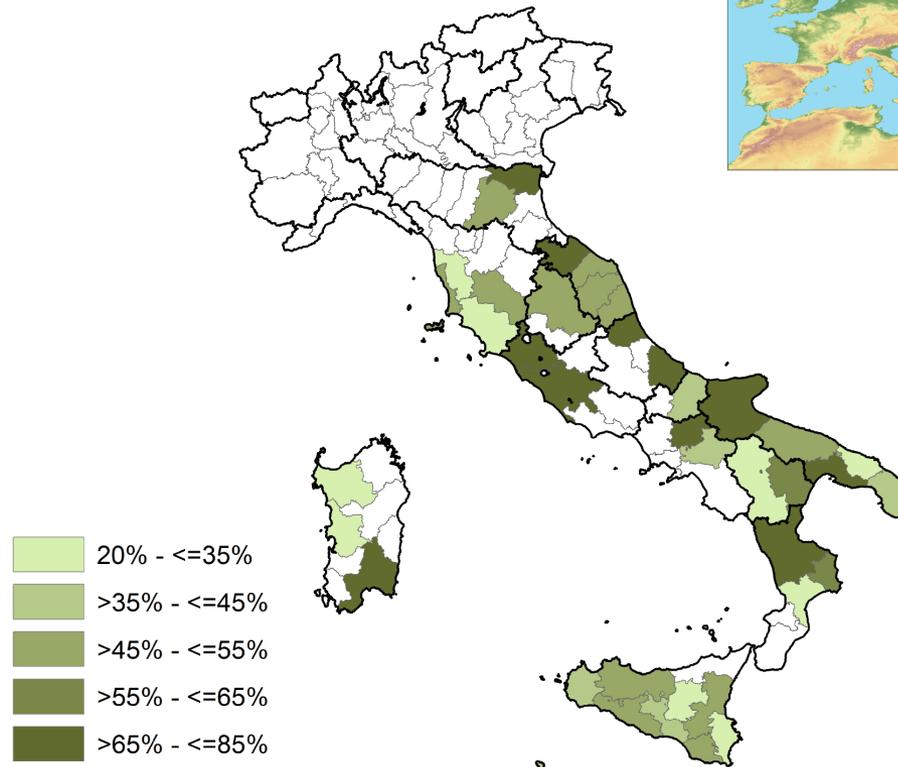


Estimated total losses from disasters attributable to different extremes (1993-2007) - California agriculture. Source: Lobell et al., 2011



The 2012-2015 California drought. 2014 PDSI value w.r.t. the historical and reconstructed values. Source: Griffin and Anchukaitis, 2014.

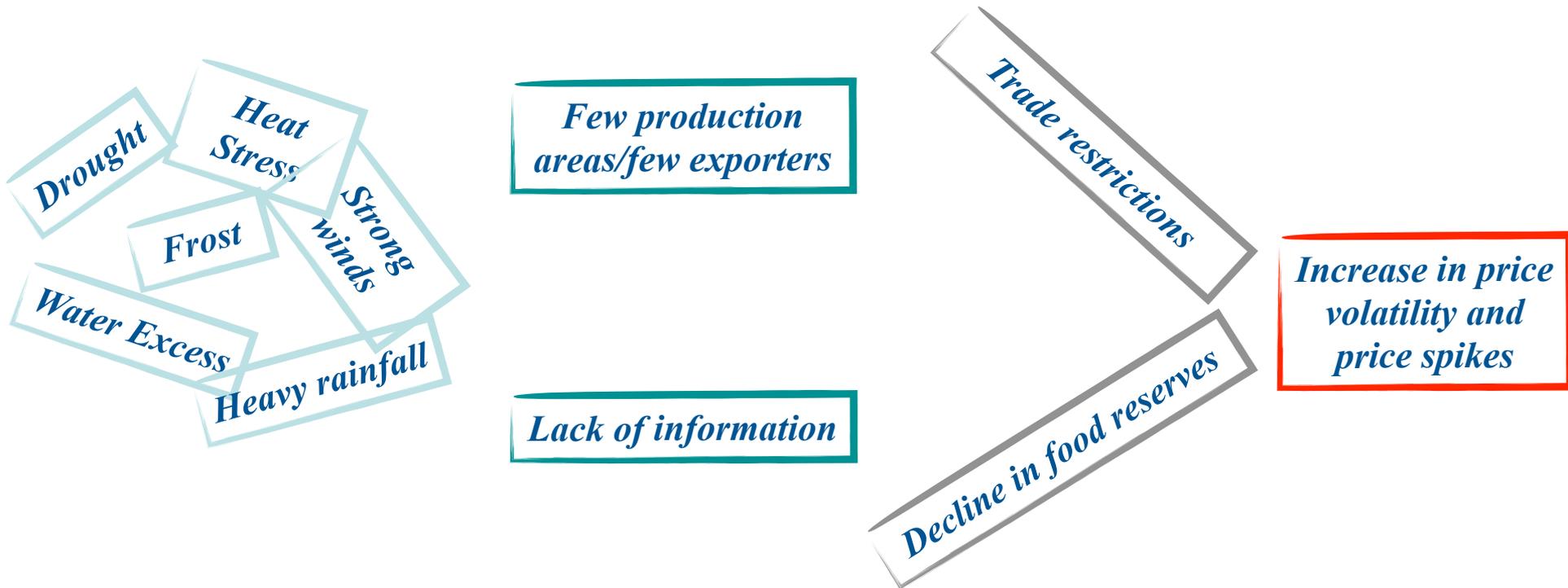
Impacts II



Number of concurrent early heat waves and significant negative yield anomalies of durum wheat in the period 1995–2013 (% w.r.t. the total number of year with significant negative yield anomalies). Source: Fontana et al. 2015, NHESS 15

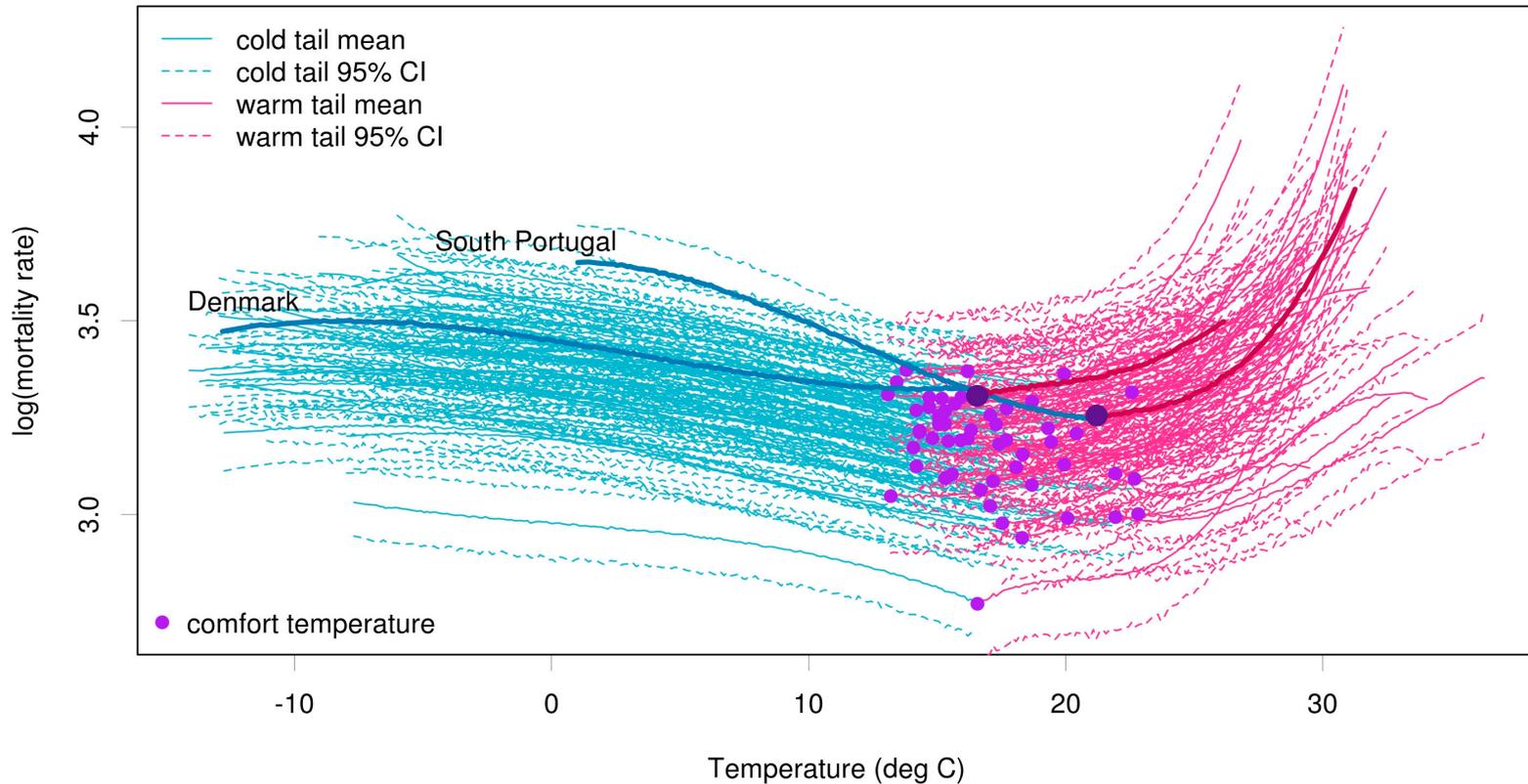
Impacts II

cascade of events



Adapted from von Braun and Tadesse, 2012

Impacts III



*Estimated mortality and comfort zone in 54 European regions, 1998-2003.
Source: Lowe, 2015*

Impacts

- Interaction of different extremes as well as occurrence of favourable/triggering conditions
- non-climatic factors
- many variables are needed
- Bias in model outputs

Open issues, new challenges,...

- Modelling multivariate extremes having different spatio-temporal scales
- Understanding of past and current changes in extremes still limited
- Process/event oriented evaluation of models
- lack of high resolution gridded observations
- Gap between impact community and climate community

Thank you!

The views here expressed are purely those of the author and may not in any circumstances be regarded as stating an official position of the European Commission.



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