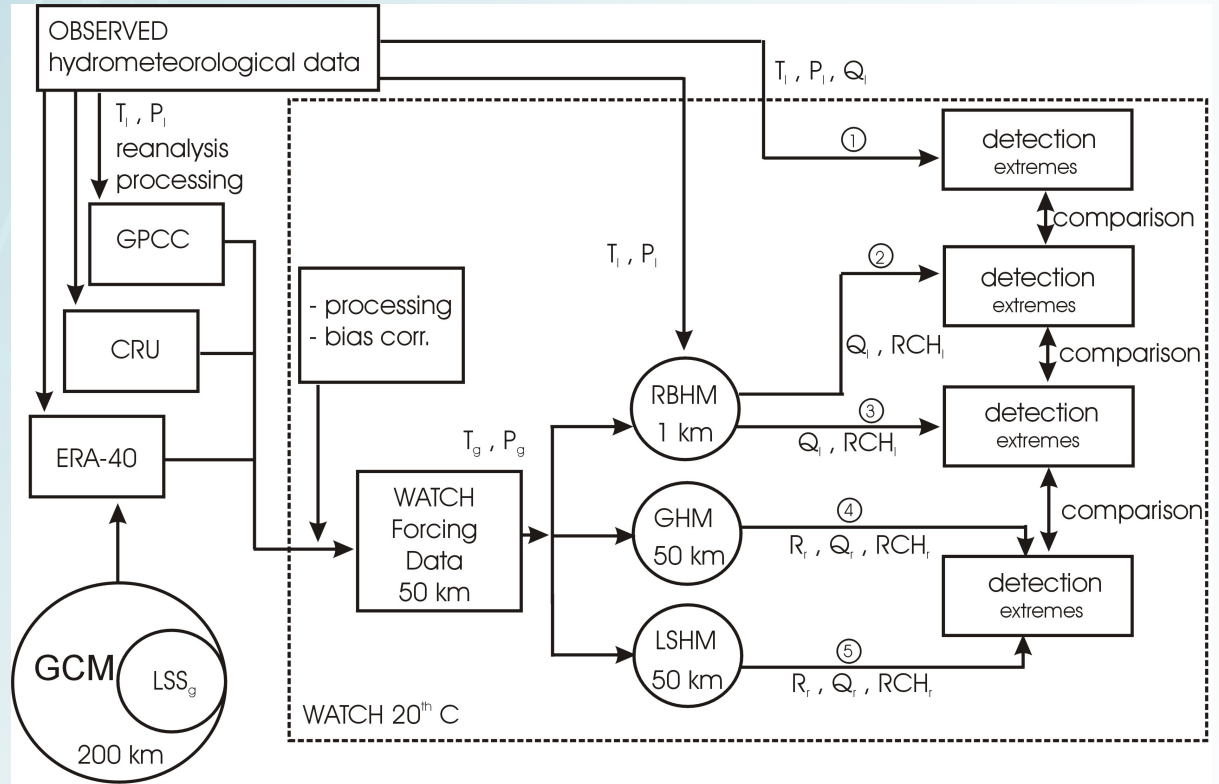
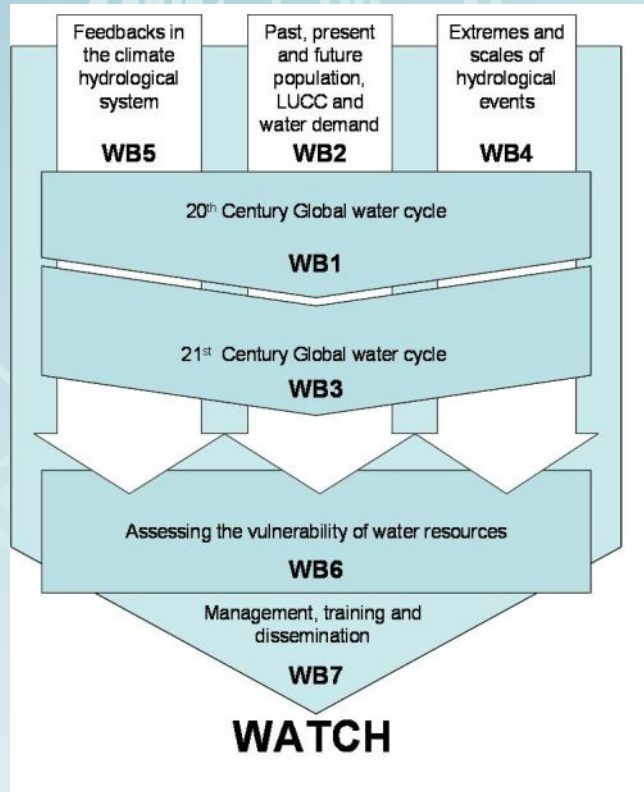


Methodologies for the spatial and temporal analysis of European droughts

Geraldine Wong and Lena M. Tallaksen
Department of Geosciences, University of Oslo

EU-WATCH background



EU-WATCH background

NAME	Energy Balance	Evapo-transpiration	Runoff-scheme	Snow scheme
H08	YES	Bulk formula	Saturation excess/Beta function	Energy balance
HTESSEL	YES	Penman-Montieth	Variable infiltration capacity/Darcy	Energy balance
JULES	YES	Penman-Montieth	Infiltration excess/Darcy	Energy balance
LPJ	NO	Priestley-Tailor	Saturation excess	Degree day
MPI-HM	NO	Thornwaite	Saturation excess	Degree day
WaterGAP	NO	Priestley-Tailor	Beta function	Degree day
GWAVA	NO	Penman-Montieth	Saturation excess/Beta function	Degree day
ENSEMBLE	Mean of all models			

Aim

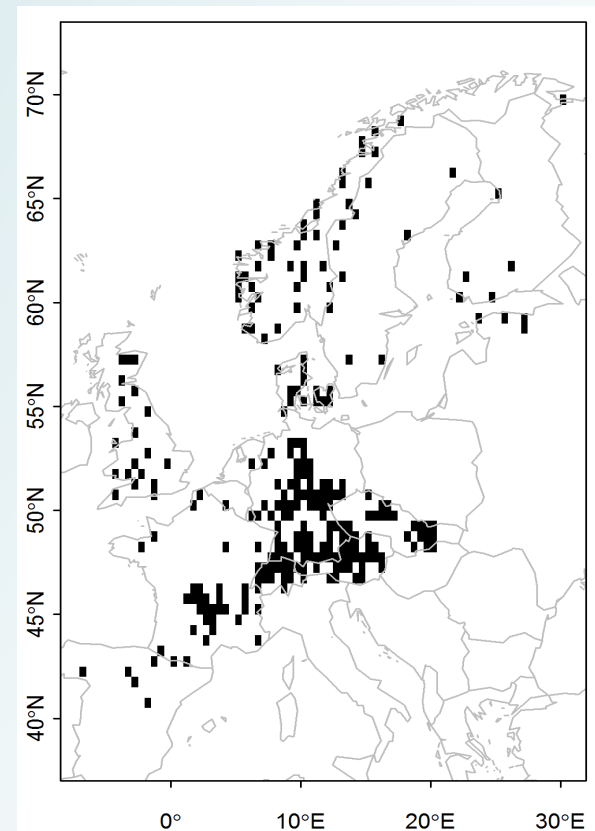
- To explore to what extent large-scale models (GHM and LSHMs) capture characteristics of historic drought (e.g. frequency, duration, scale, severity)
- To examine and analyse the spatial distribution of European drought over time using different clustering algorithms
- To provide an insight into return periods of extreme events using severity-area-frequency (SAF) curves

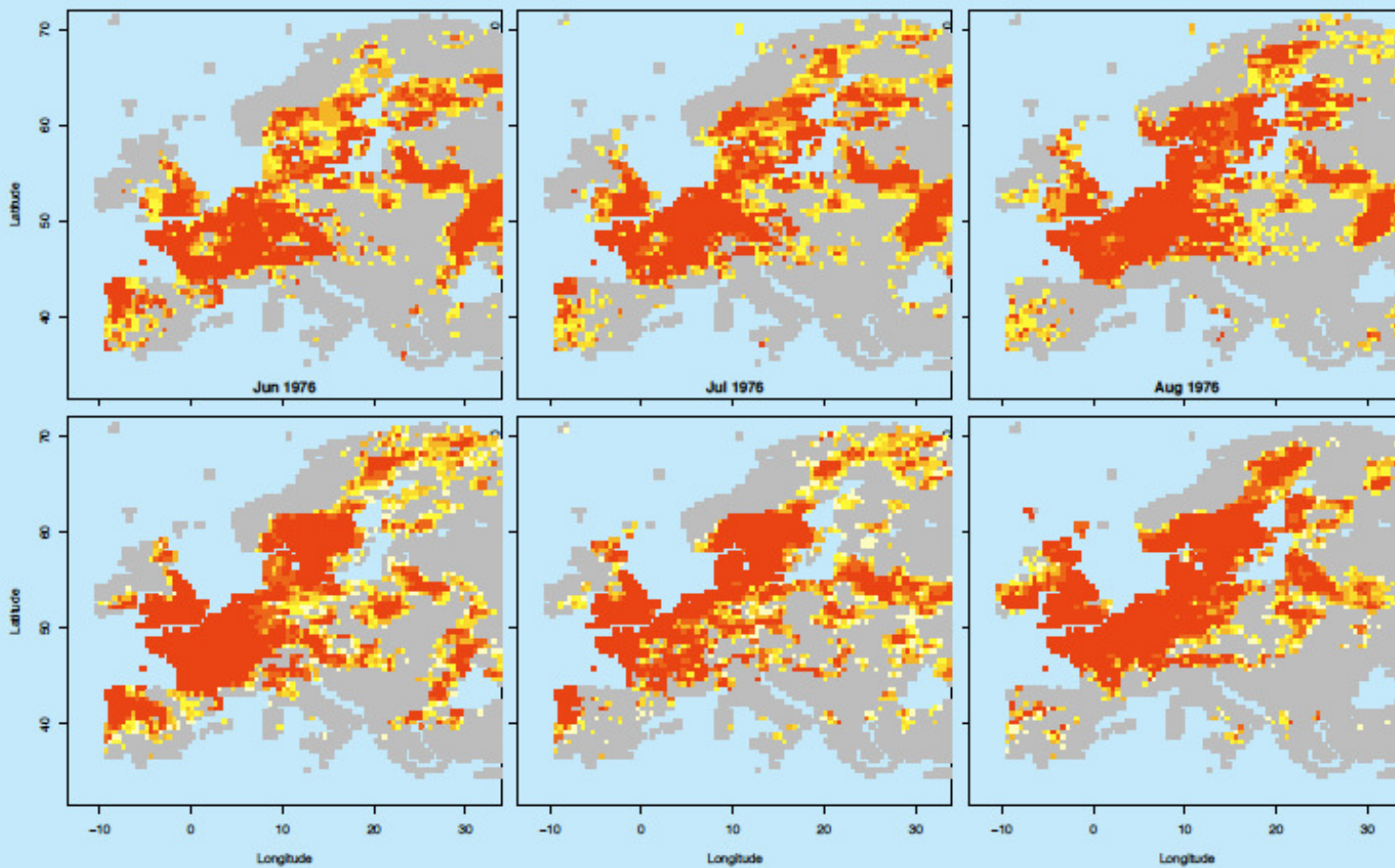
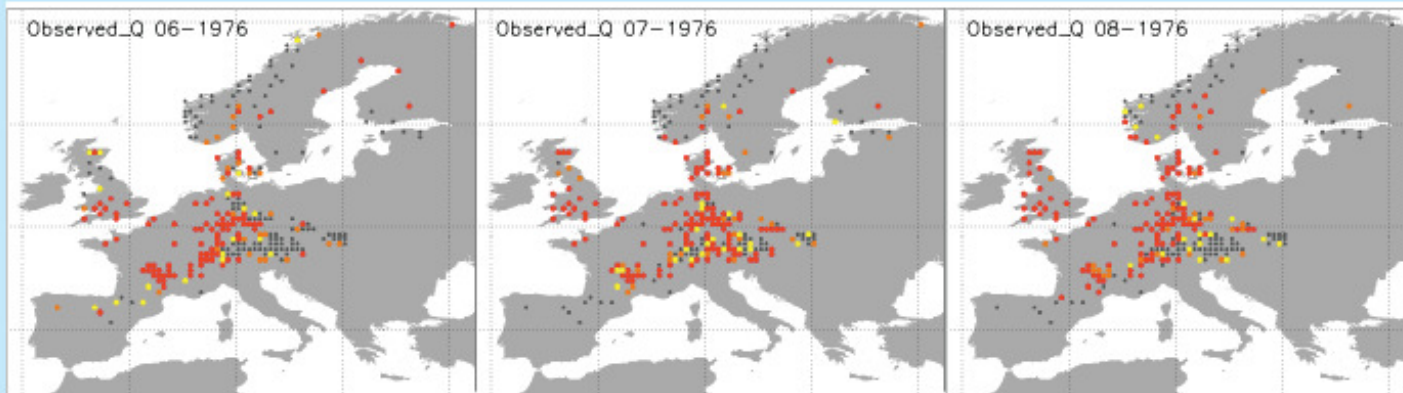
WB4: Hydrological Extremes – DROUGHT

Task 4.1.2 Spatial and temporal scales and severity of droughts

WATCH Multi-Model evaluation of Extremes

- Grid cells with observations: 263
- Grid cell $\sim 2\,500\text{ km}^2$
- Catchment $\sim 306\text{ km}^2$
- Monthly sum of subsurface (Q_{sb}) and surface (Q_s) flow covering the period 1958-2001
- Models used: GWAVA and Htessel
- Droughts defined by 20th quantile, for each pixel with respect to their month





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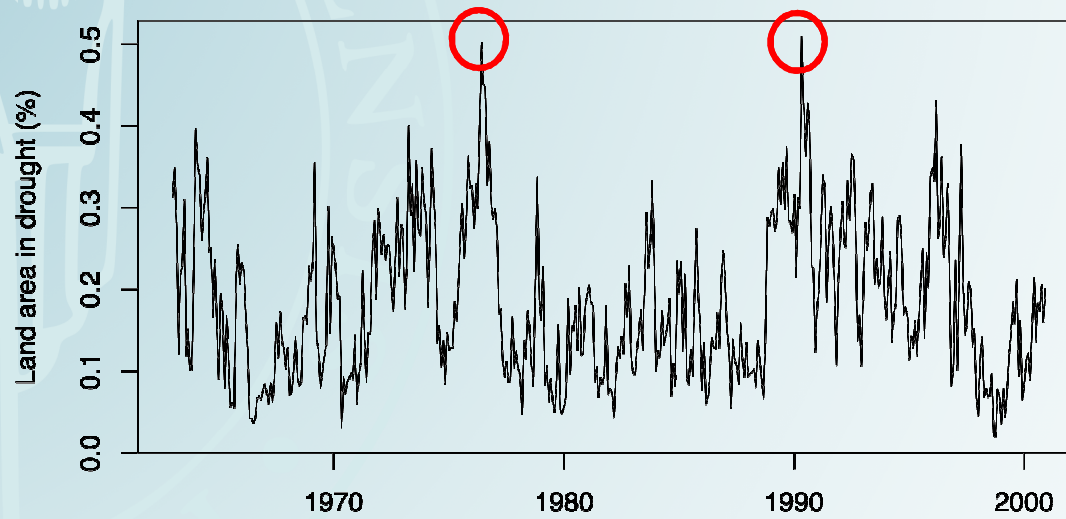
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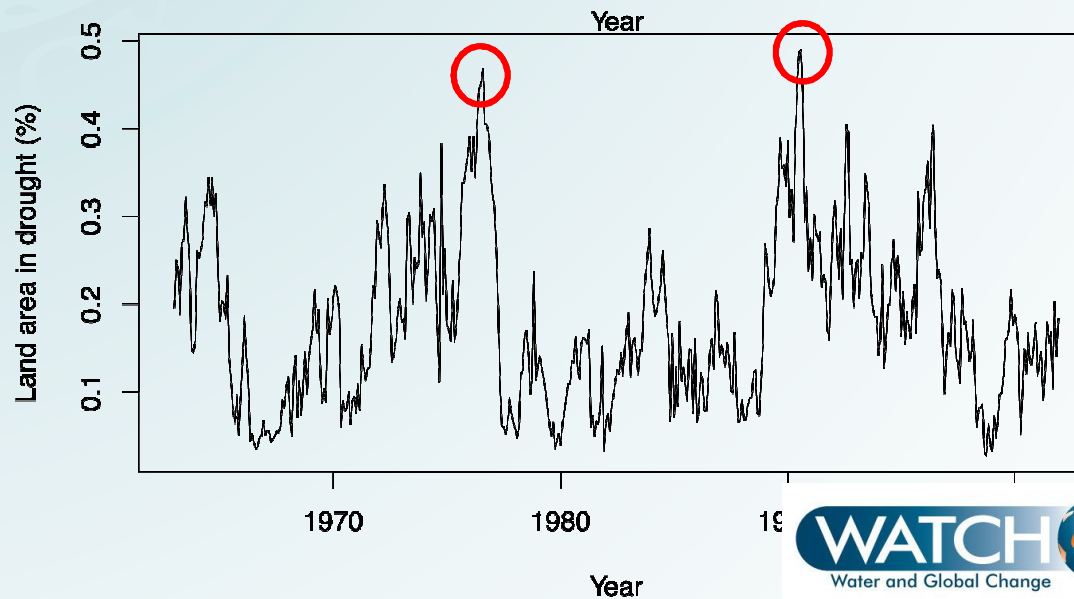
1976 summer drought:

Basic statistics

GWAVA



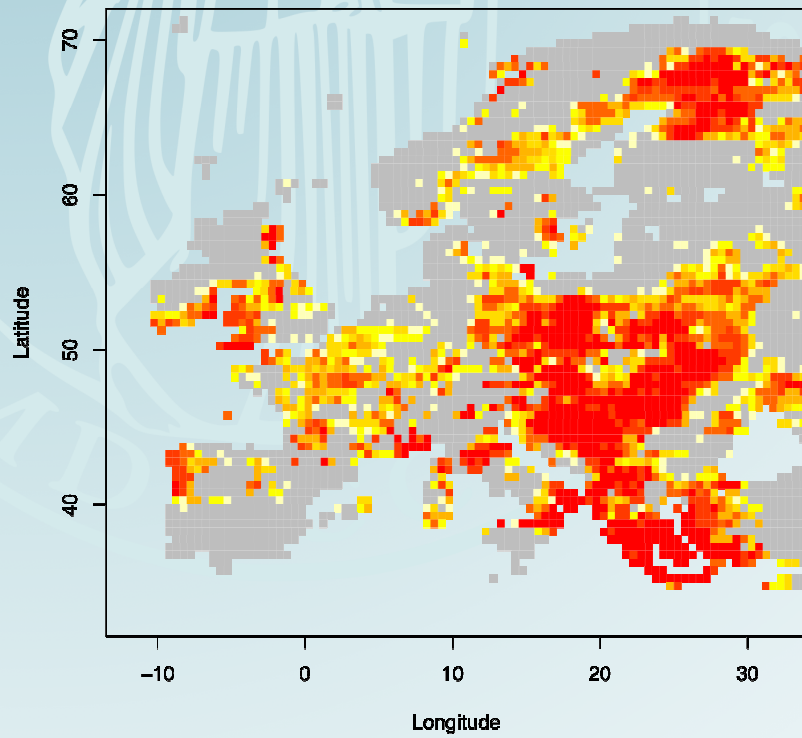
Htessel



Spatial analysis

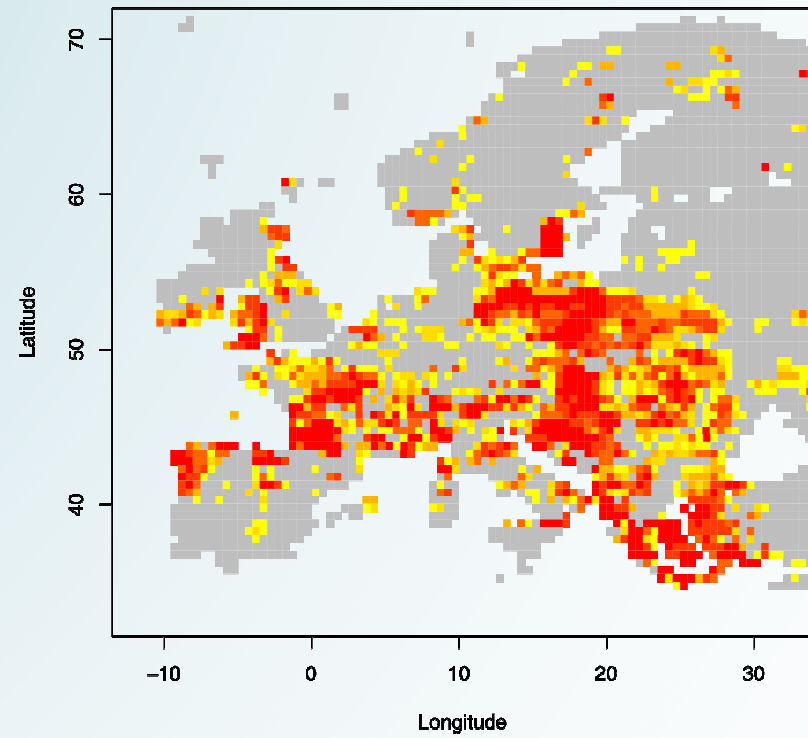
GWAVA

May 1990



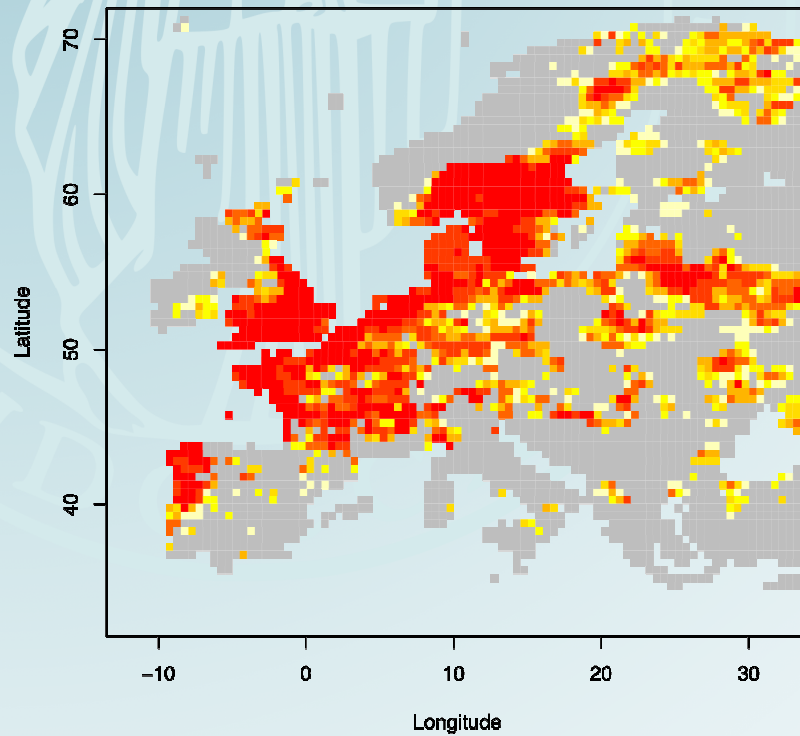
Htessel

May 1990

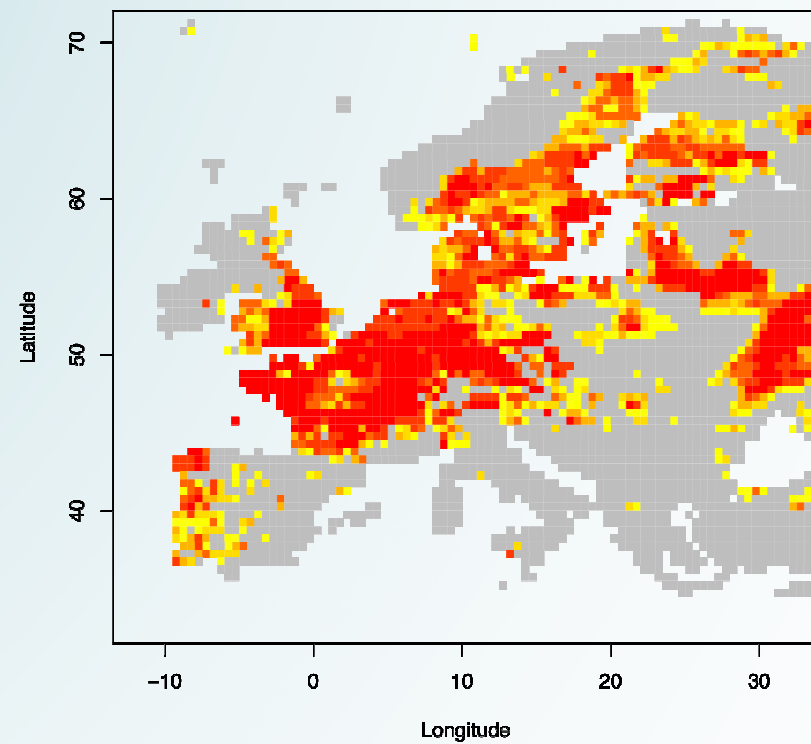


1976 drought

GWAVA
Jul 1976

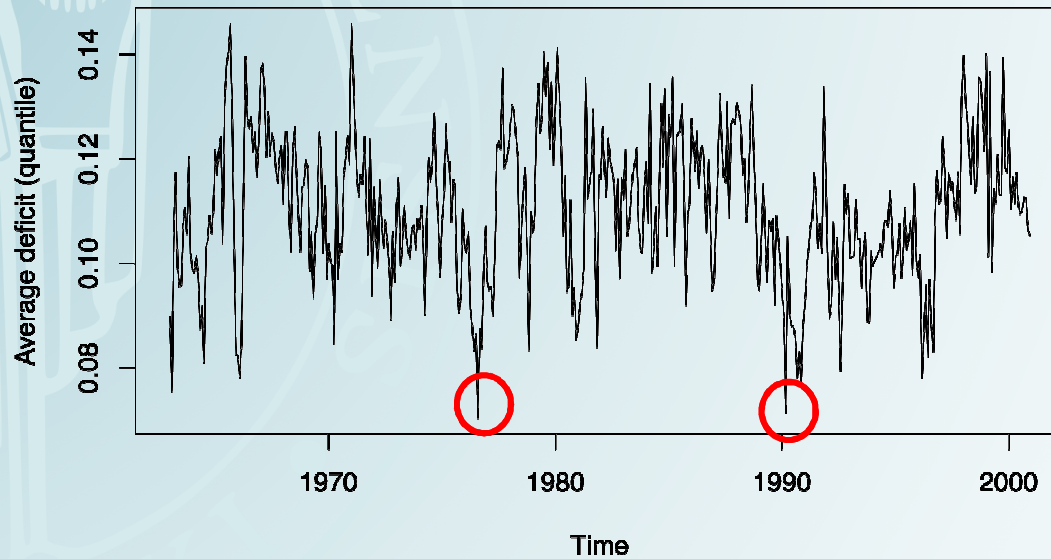


Htessel
Jul 1976

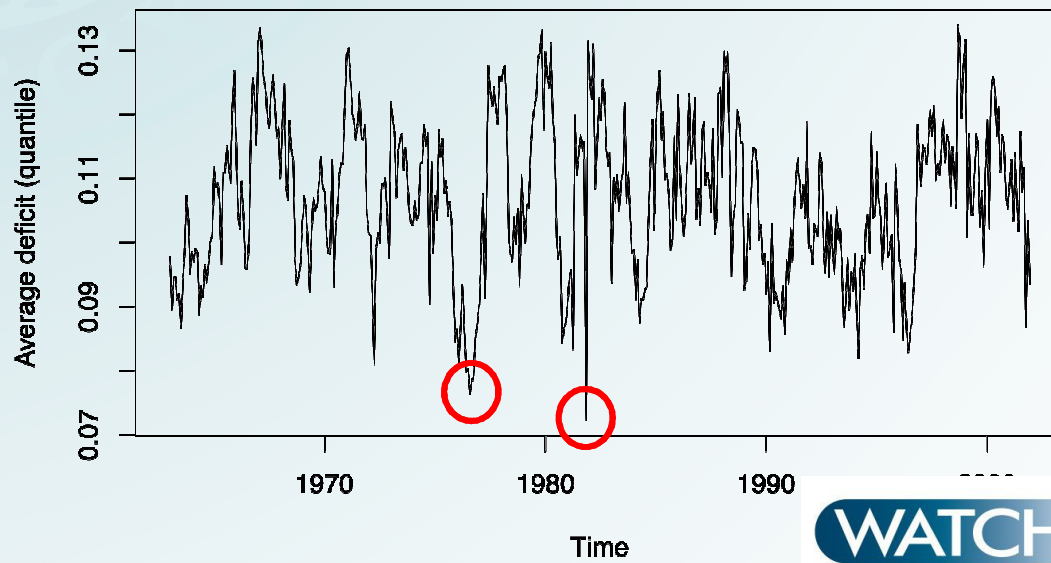


Drought severity

GWAVA

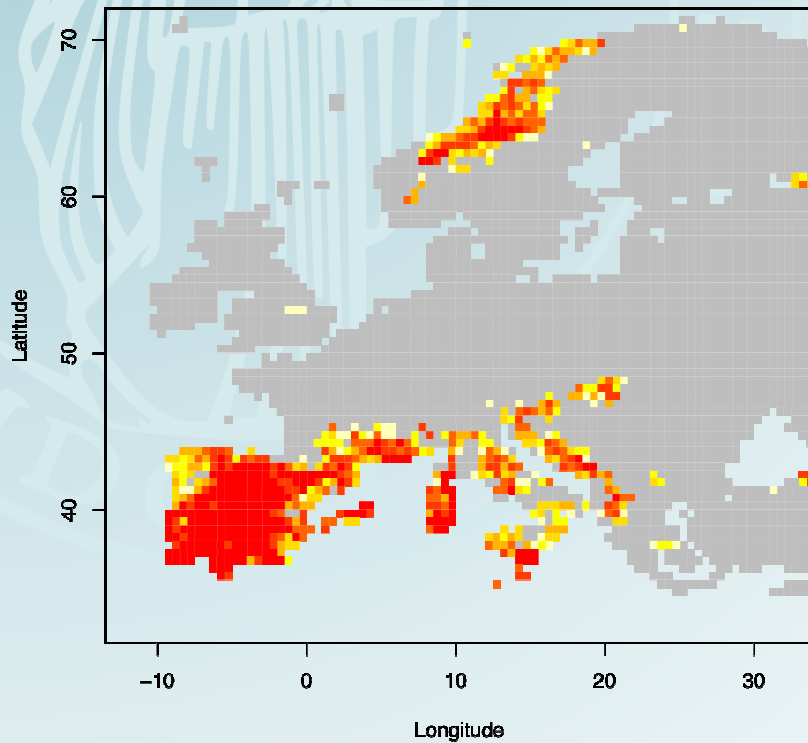


Htessel

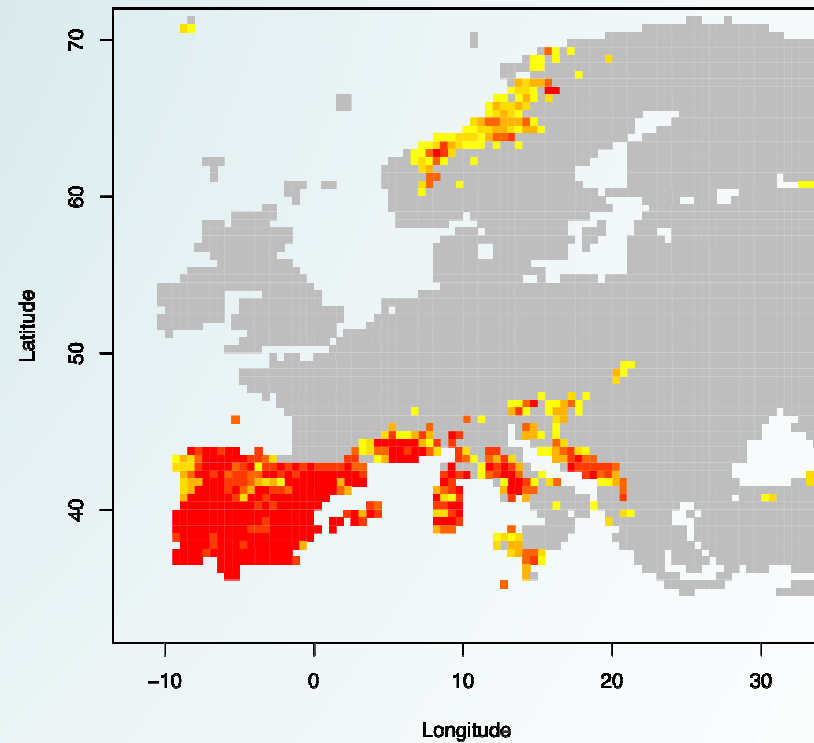


Corresponding drought figures

GWAVA
Nov 1981



Htessel
Nov 1981

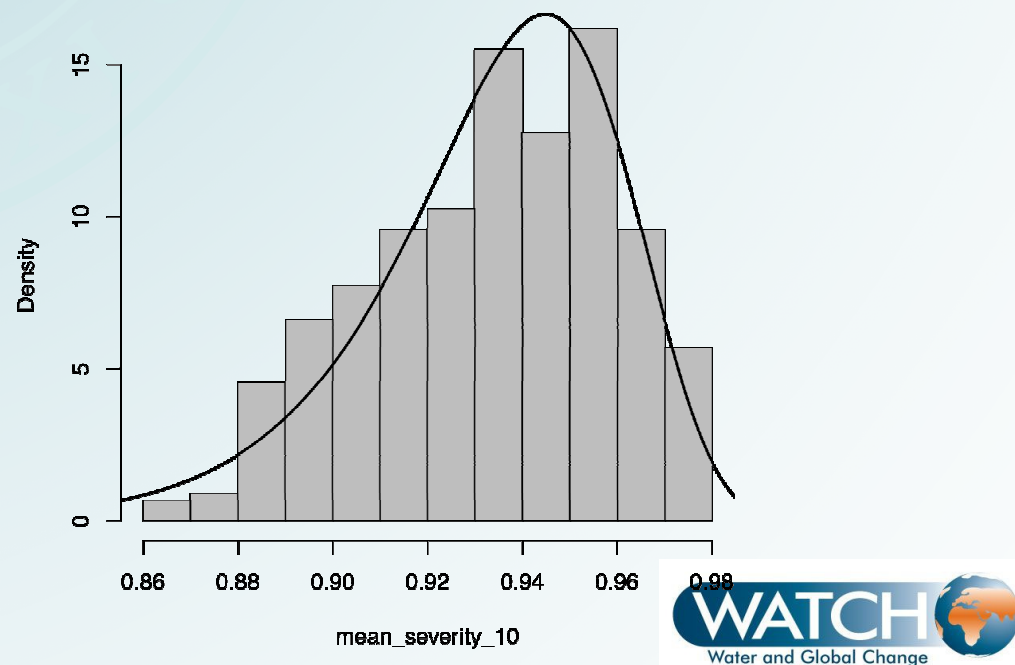


SAF curves – Method 1

- Does not take into account any spatial relationships
- Number of pixels (n) associated with each percentage of land area (p) affected by drought is calculated
- For month t , severity of all drought land pixels are sorted in descending manner and most severe pixels associated with n are chosen
- Mean severity of these pixels are calculated
- For month t , if no. of drought affected pixels $< n$, then NA is recorded

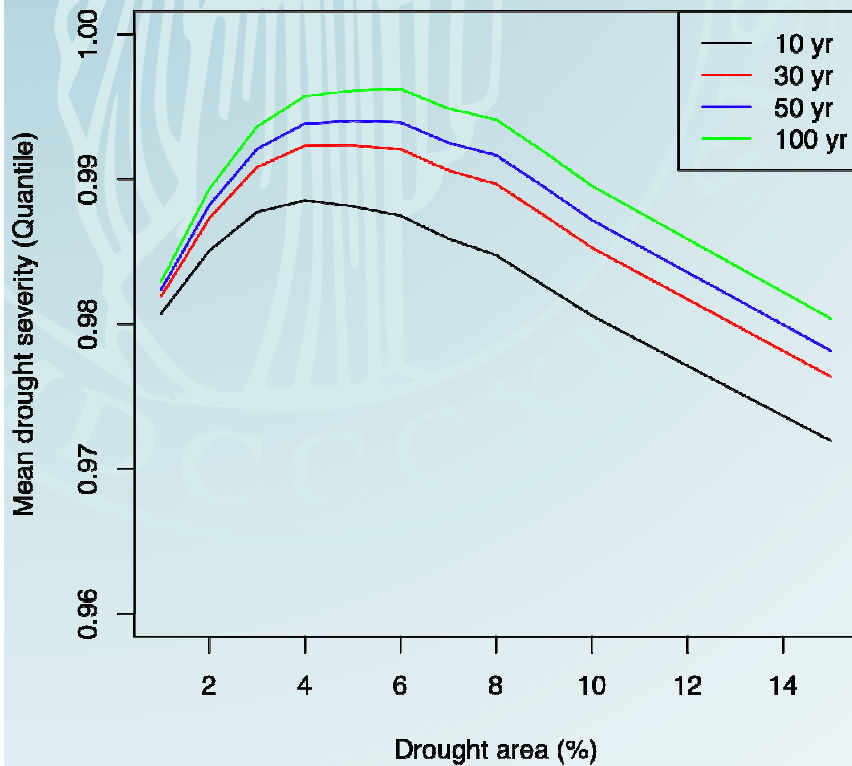
Producing SAF curves

- Suitable distributions are fitted to mean severities associated to each $p\%$
- ARI are calculated for return periods of 10,30, 50 and 100 years

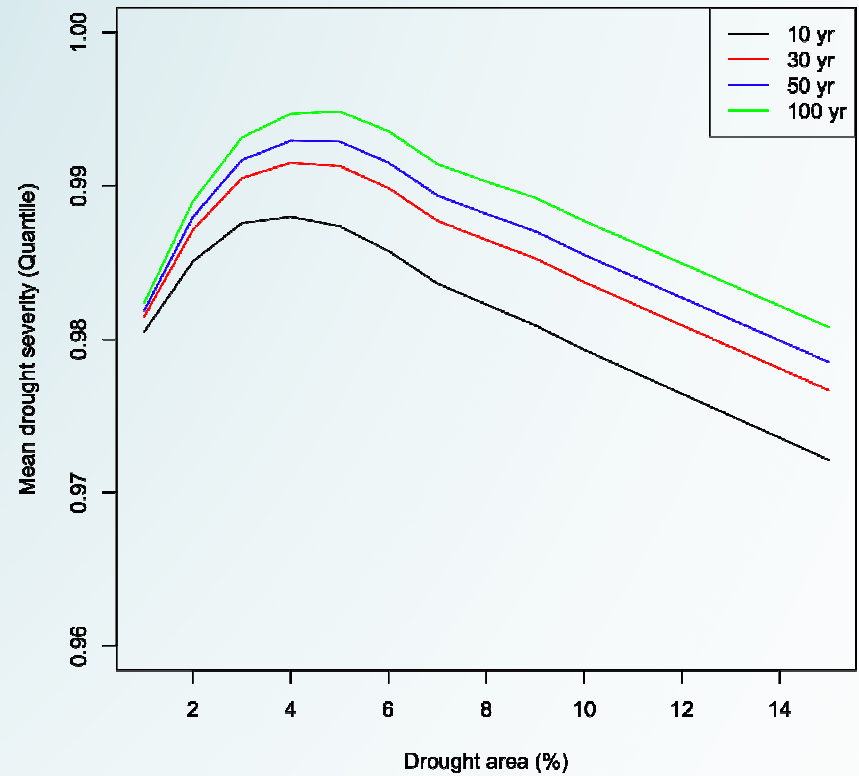


Results – Method 1

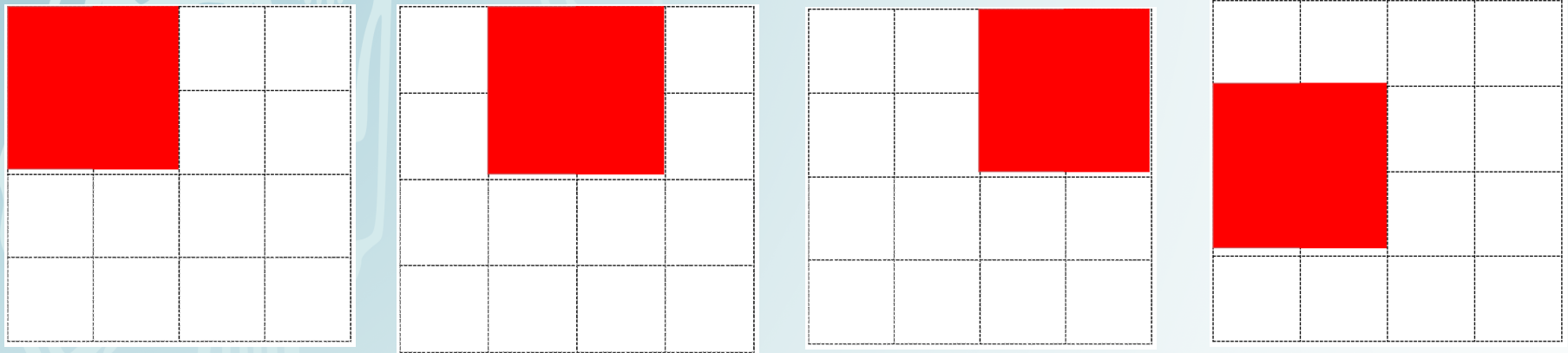
GWAVA



Htessel

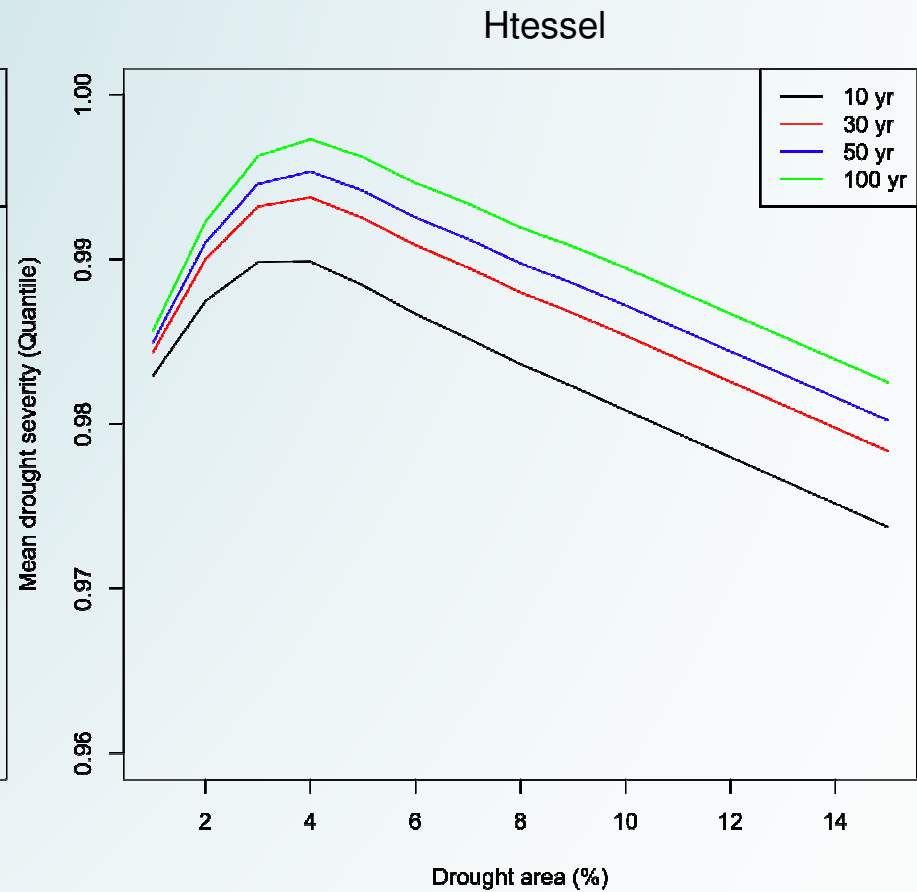
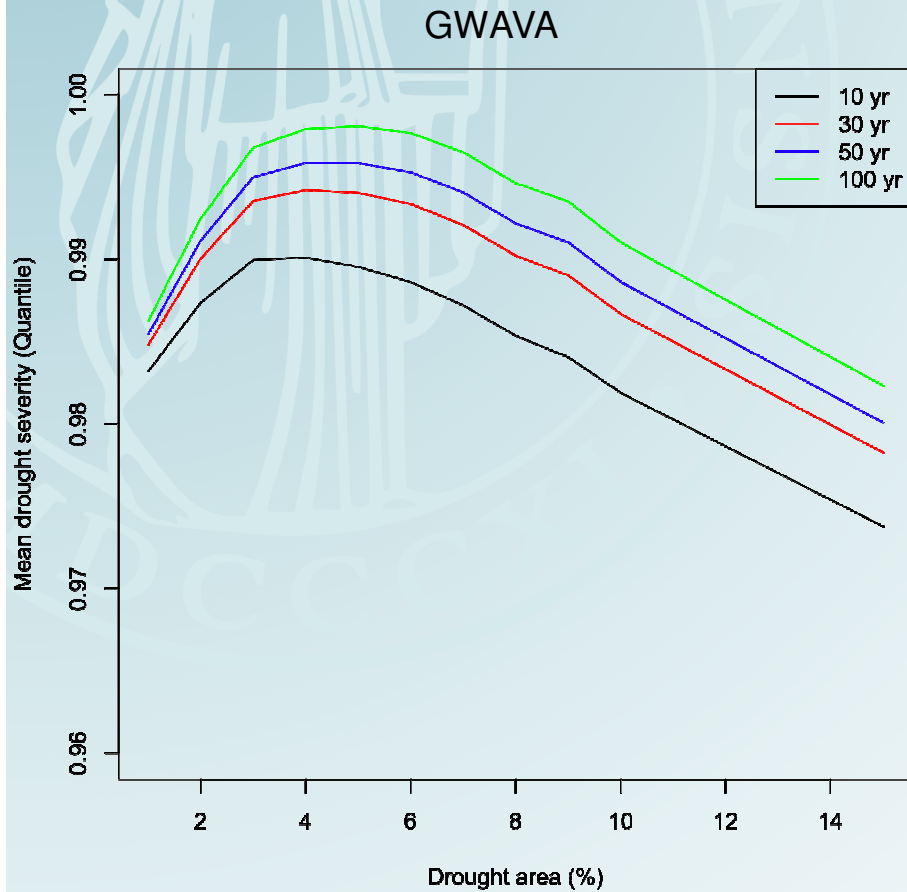


Method 2 – Spatial smoothing

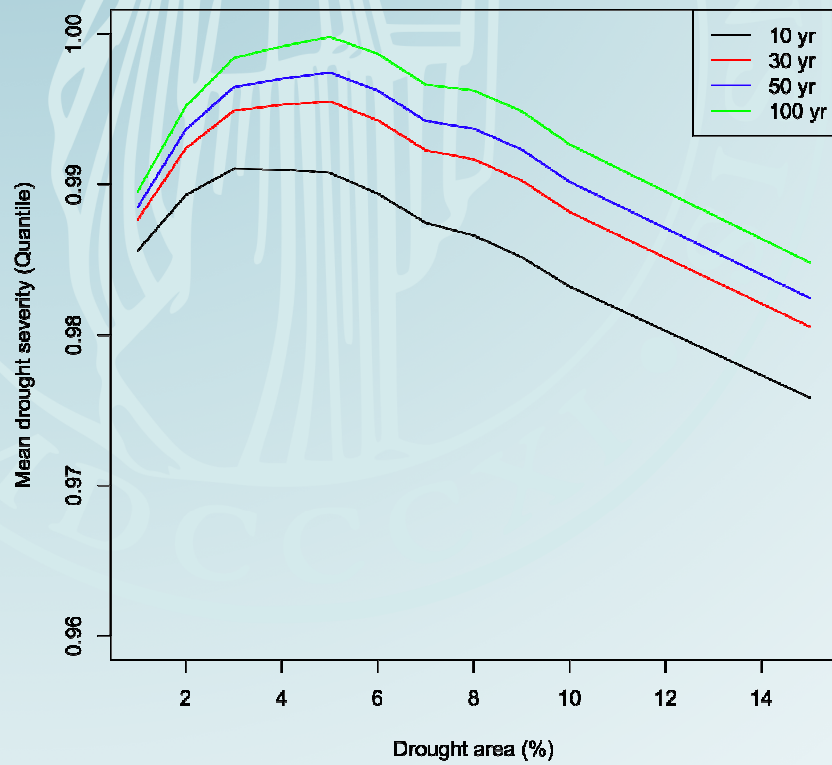


- Spatially smoothing of monthly Q_s+Q_{sb} for n by n pixels
- Quantiles for each pixel are now calculated relative to the spatially averaged pixels
- Continue as Method 1

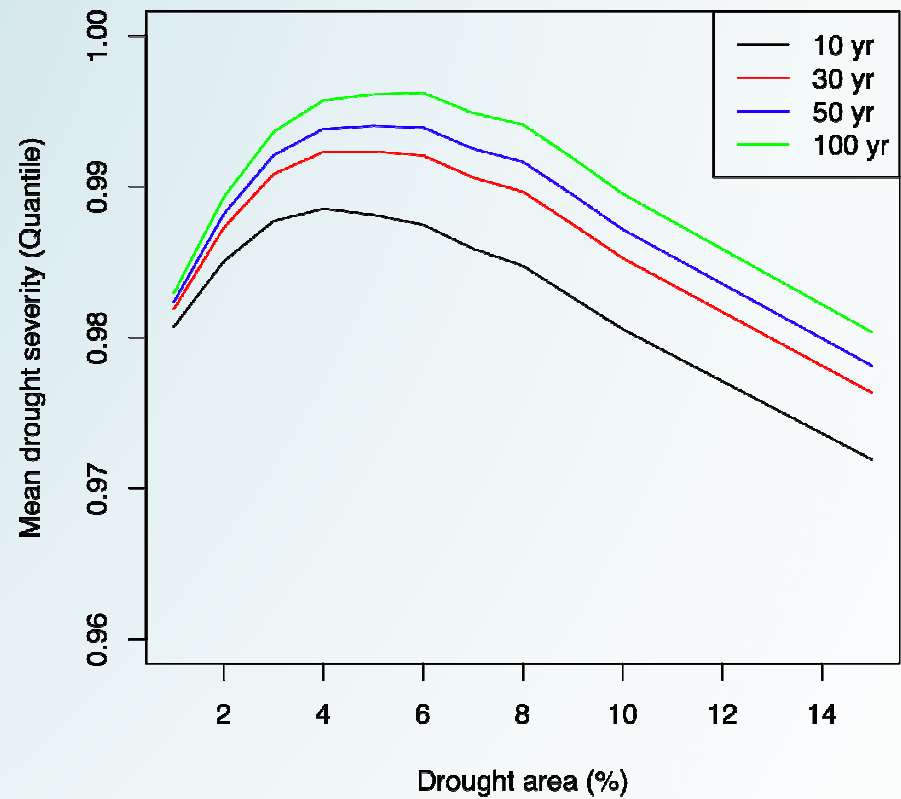
Results – Method 2 (2 by 2)



Results – Method 2 (3 by 3)

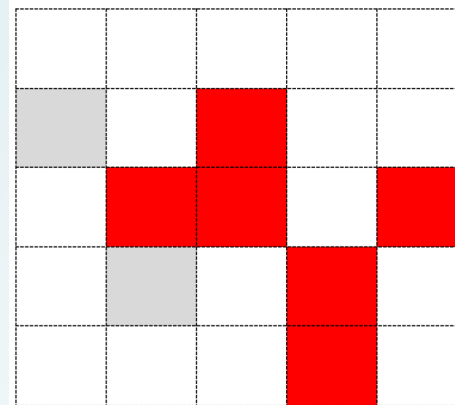
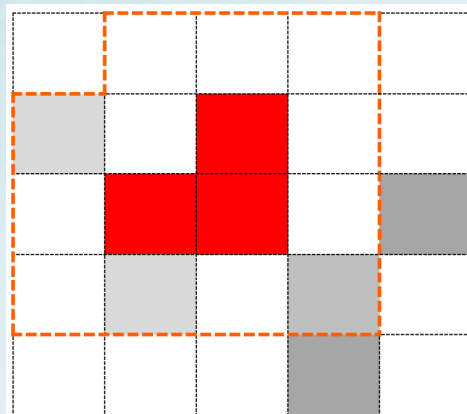
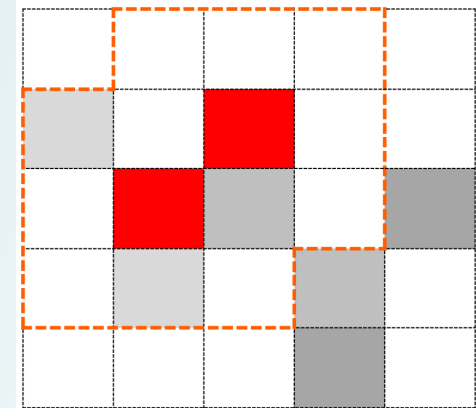
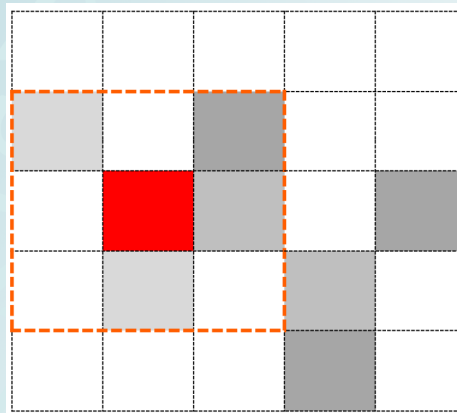
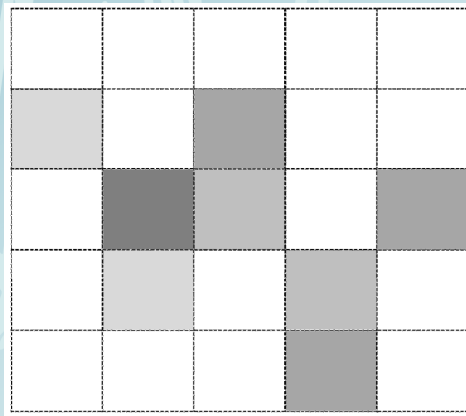


Method 1



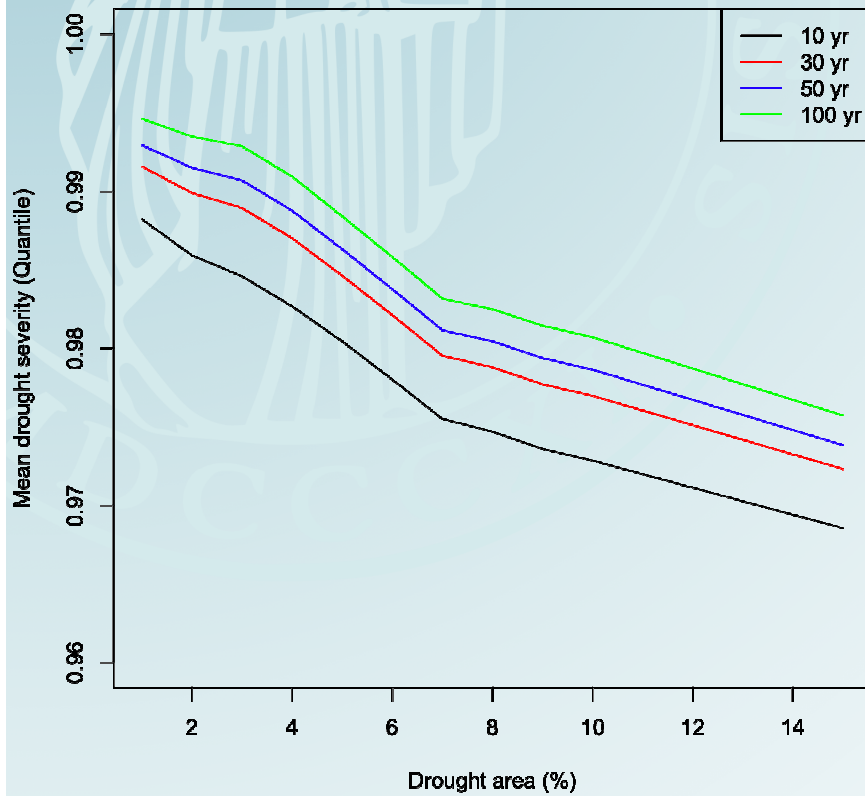
GWAVA

Method 3 – Drought clustering

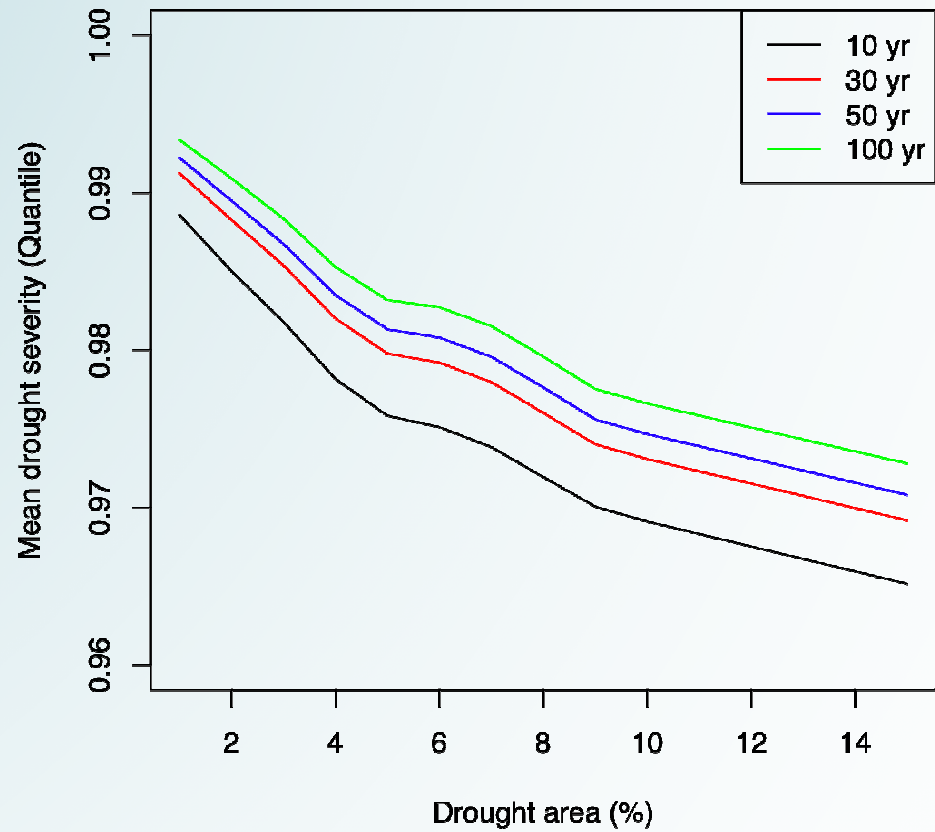


Results – Method 3

GWAVA

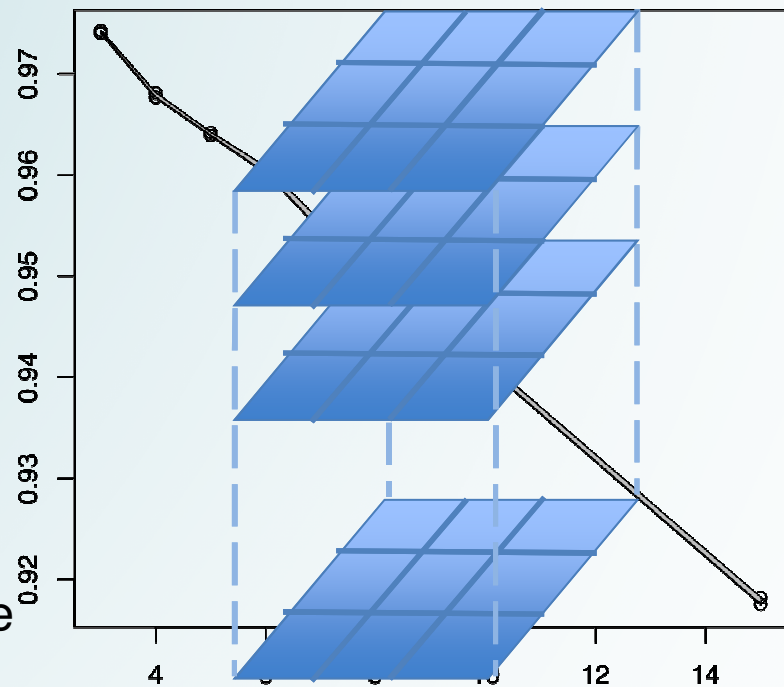


Htessel



Spatial and temporal independence

- For each land grid cell, randomly select from the collection of months.
- Each grid cell is randomly sampled and are independent.
- Choose most severe $n\%$ grid cells and calculate mean severity.
- This is repeated for 100 runs, to produce simulation bounds.



Conclusions

- More severe drought for smaller drought area.
- Higher sensitivity to change in mean drought severity at larger areas for Method 1 and 2 but smaller areas for Method 3 (drought clustering)
- Increased spatial smoothing produces slight increase in severity
- Spatial and temporal dependence is significant.

Ongoing work

- Comparisons across other models
- Modelling spatial-temporal dependence structure as multivariate distribution