

Flood inundation risk under climate and social change

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18 May 2012

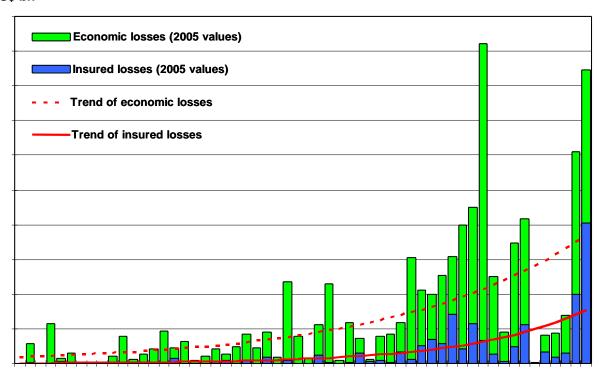


Natural disasters trends

GREAT Natural Disasters 1950 - 2005

Economic and insured losses

US\$ bn





Bristol hazard research



- earthquake engineering
- volcanoes
- floods
- landslides
- disease (animals/crops)
- evolutionary biology
- statistics and uncertainty analysis
- process mathematical models
- climate change risk
 (e.g. sea level, more storms)

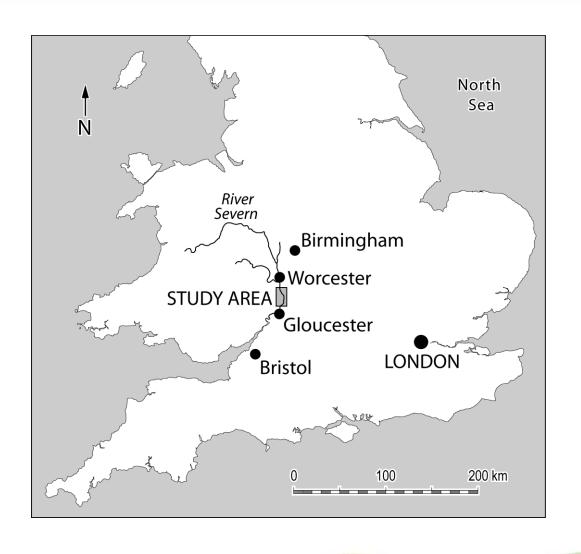


Flood risk research

- Major strength at Bristol for over a decade
- Key contributions
 - Demonstrated the critical importance of terrain in flood risk mapping
 - Developed new tools for hyper-efficient 2D modelling to take advantage of LiDAR data
 - Championed rigorous validation and benchmarking of 2D models using field data
 - Developed new methods for dealing with uncertainty

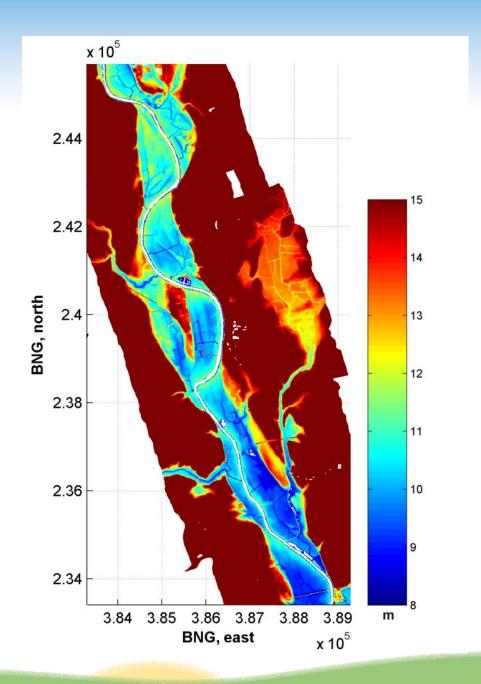


Example: LISFLOOD-FP model





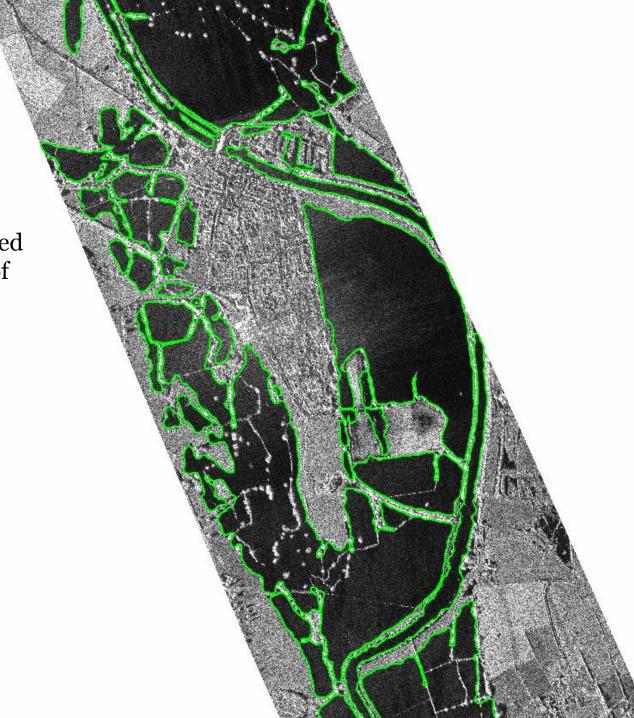
Environment Agency LiDAR digital elevation model (DEM) at 3m resolution.





Living with global uncertainty

Airborne Synthetic Aperture Radar classified at a spatial resolution of 1m, using a statistically active contour (Snake).



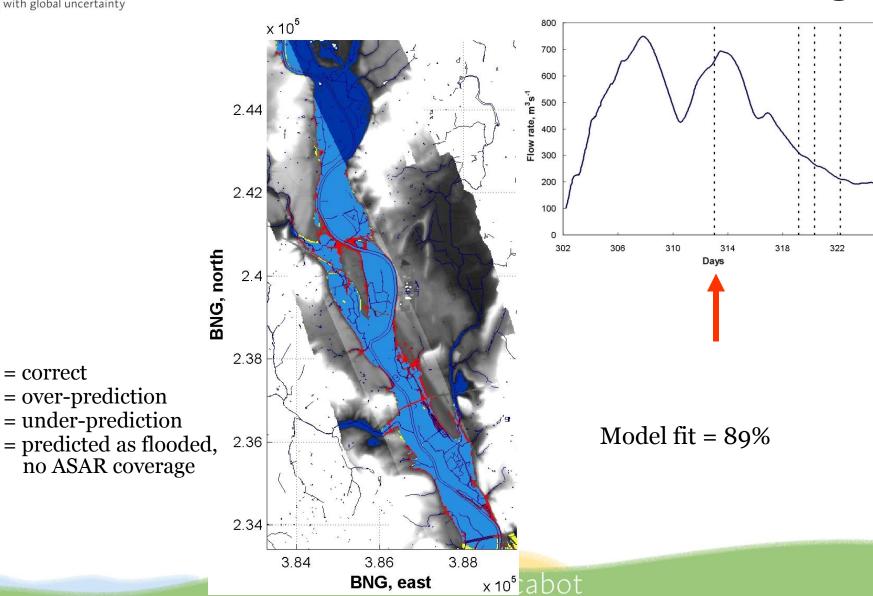


LISFLOOD-FP modelling



= correct

= over-prediction = under-prediction





New LISFLOOD-FP formulation

Continuity Equation

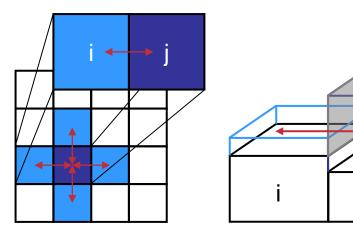
Continuity equation relating flow fluxes and change in cell depth

$$\frac{\Delta h^{i,j}}{\Delta t} = \frac{Q_x^{i-1,j} - Q_x^{i,j} + Q_y^{i,j-1} - Q_y^{i,j}}{\Delta x^2}$$

Momentum Equation

Flow between two cells now calculated using:

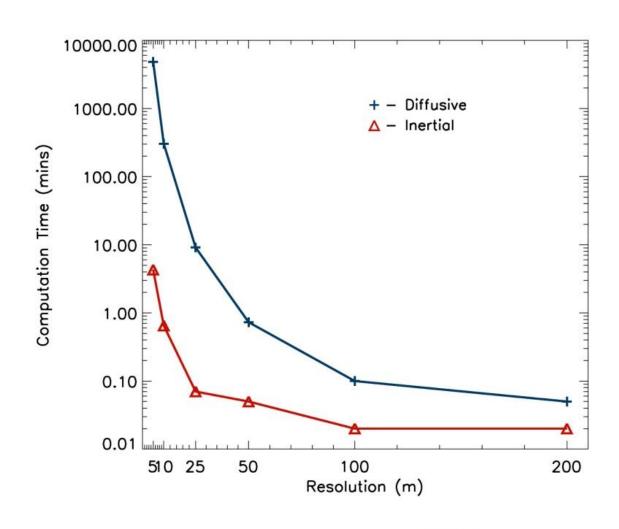
$$Q = \frac{q - gh_{flow}\Delta t \frac{\Delta (z + z)}{\Delta x}}{(+gh_{flow}\Delta t n^2 q / h_{flow}^{10/3})} \Delta x$$



Representation of flow between cells in LISFLOOD-FP

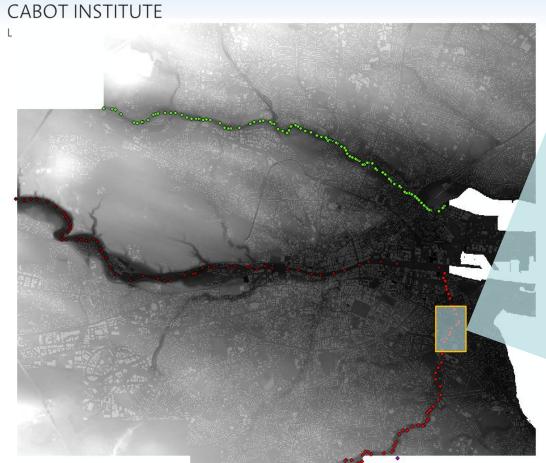


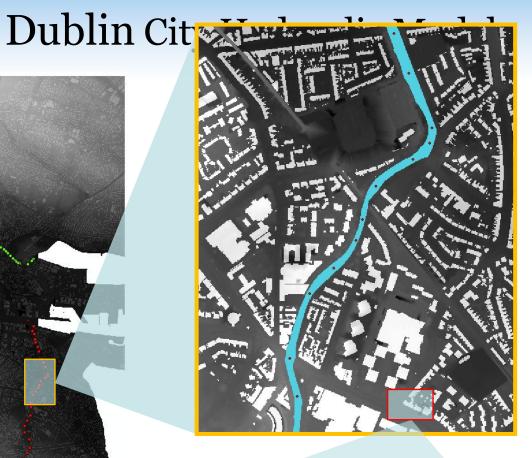
New equation speed up





Dubl



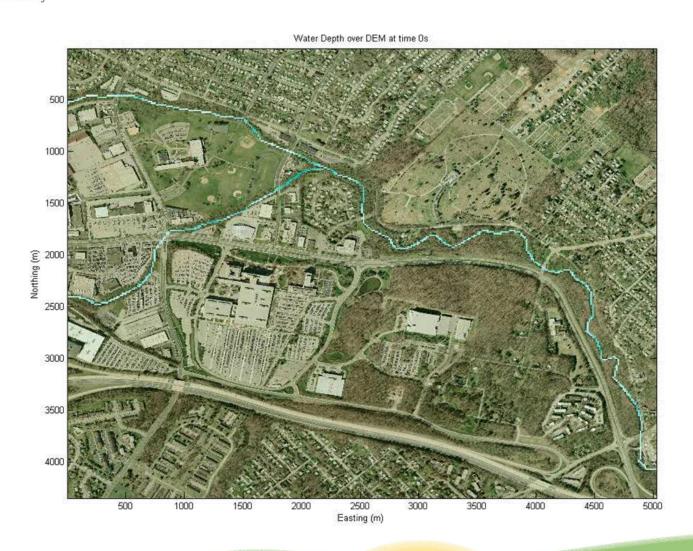


2 m





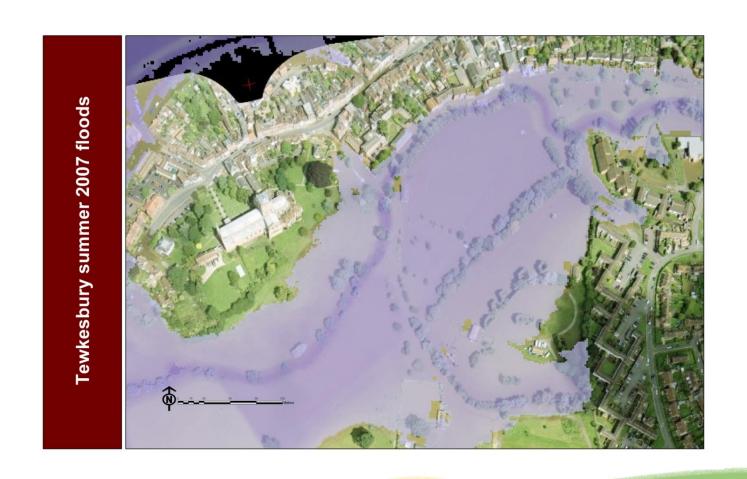
Baltimore, USA





Does it work?

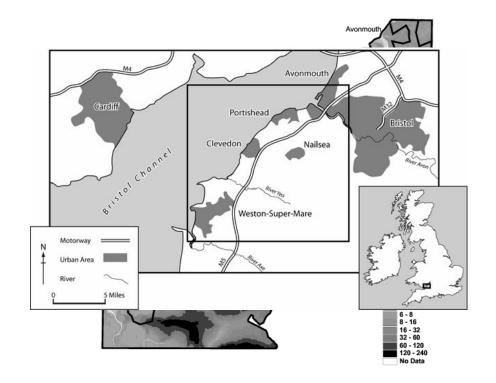
• Whole city flood modelling at 2m resolution – Tewkesbury, UK summer 2007





Risk analysis under uncertainty

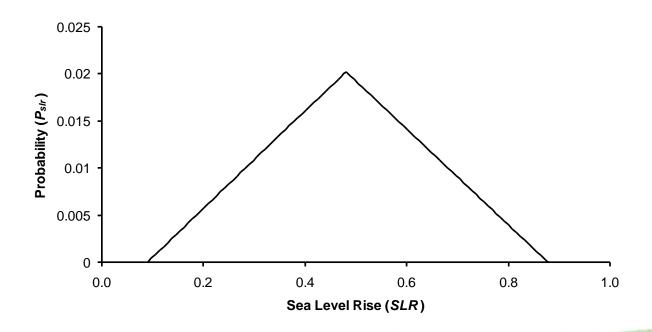
- IPCC Third Assessment Report estimates sea level by 2100 will be between 9 and 81cm higher than today
- Attempt to evaluate the impact of this uncertainty for a future 1 in 200 year flood event on the Somerset coast in the UK





Uncertain future flooding

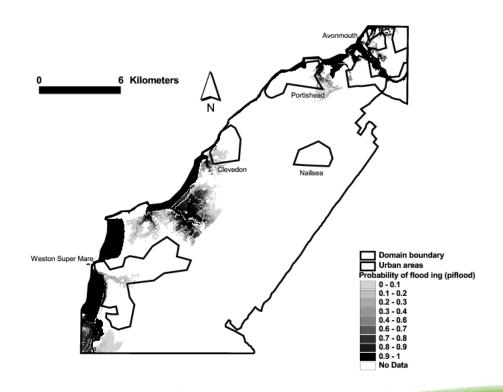
• Convert IPCC estimate to a probability distribution and add to water levels for a 1 in 200 year extreme tidal event





Uncertain future flooding

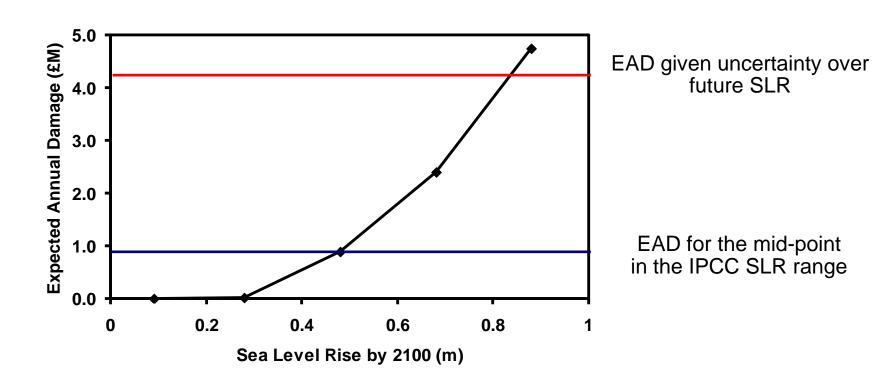
 Run an MC analysis to calculate the probability of flooding in each cell weighted by uncertainty over future SLR





Deterministic vs. uncertainty weighted modelling

Calculate expected annual damages for different scenarios















DEMON

<u>Developing Enhanced impact MO</u>dels for integration with <u>Next generation NWP and climate outputs</u>







Shrewsbury, Shropshire – future flooding under climate change?

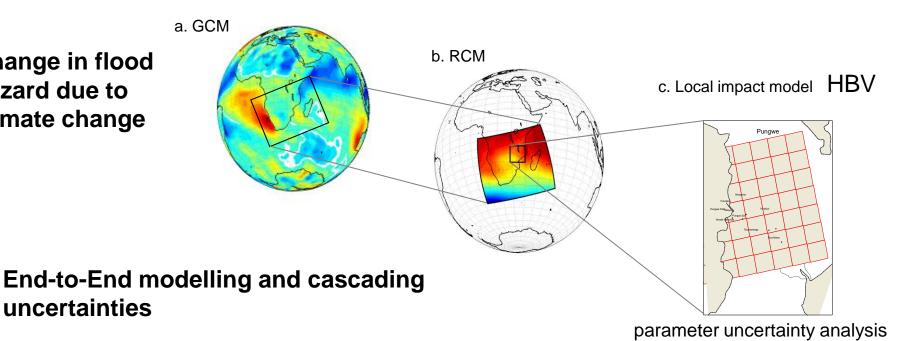








Change in flood hazard due to climate change



"grand" ensemble:

uncertainties

- multi-model RCMs (ENSEMBLES) cascaded directly into hydrology impact model
- 2. ensemble of single RCM (UKCP09) cascaded directly into hydrology impact model
- 3. Perturbed Physics Ensemble future rainfall overlain on impact model response surfaces



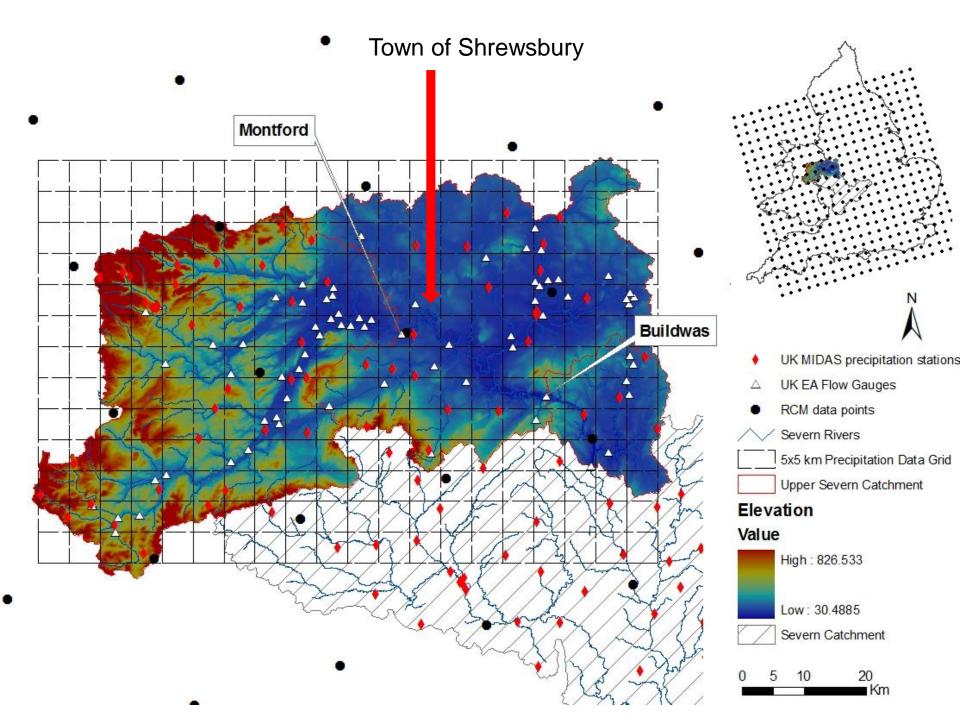
16 RCMs



11 members

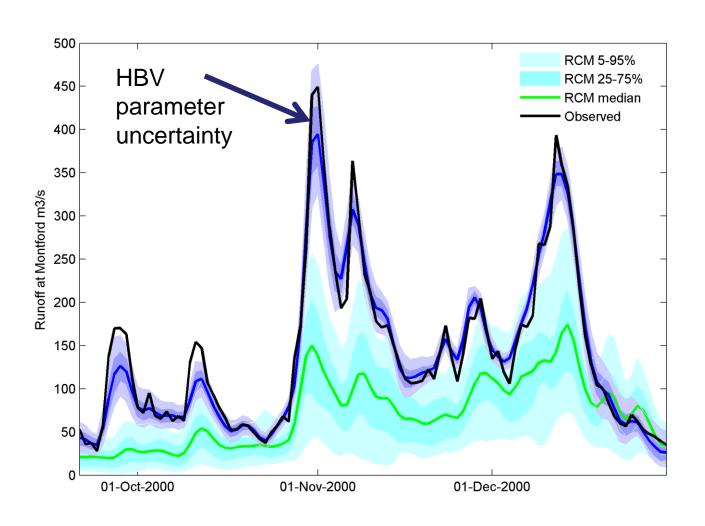
Harris et al (2010)

10000 paired preciptemp samples from Joint Probability Distribution

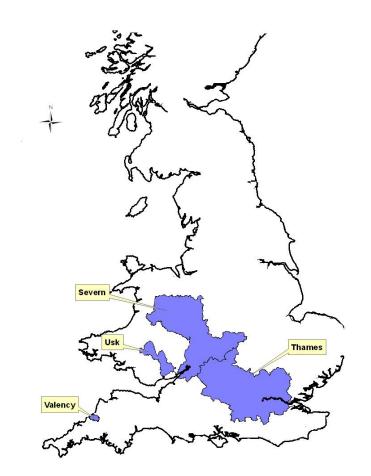




HBV model – using 'raw' UKCPo9 and ENSEMBLES

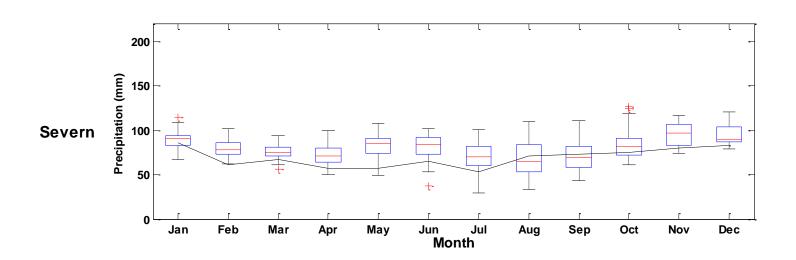






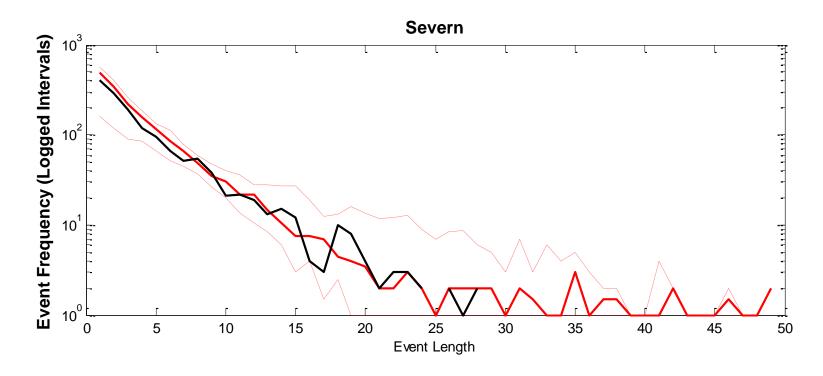
Ensemble#	Country	Driving GCM	RCM	Resolution	E-Scenario
1	SWEDEN	ECHAM5-R3	RCA	25 km	A1B
2	SWEDEN	всм	RCA	25 km	A1B
3	GERMANY	ECHAM5-R3	REMO	25 km	A1B
4	NETHERLANDS	ECHAM5-R4	RACMO	25 km	A1B
5	ITALY	ECHAM5-R5	RegCM	25 km	A1B
6	FRANCE	ARPEGE	HIRHAM	25 km	A1B
7	U.K	HADCM3Q16	HADRM3Q16	25 km	A1B
8	U.K	HADCM3Q3	HADRM3Q3	25 km	A1B
9	U.K	HADCM3Q0	HADRM3Q0	25 km	A1B
10	SWITZERLAND	HadCM3Q0	CLM	25 km	A1B
11	U.K	HADCM3	HadRM3Qk	25 km	A1B
12	U.K	HADCM3	HadRM3Q16	25 km	A1B
13	U.K	HADCM3	HadRM3Q14	25 km	A1B
14	U.K	HADCM3	HadRM3Q13	25 km	A1B
15	U.K	HADCM3	HadRM3Q11	25 km	A1B
16	U.K	HADCM3	HadRM3Q9	25 km	A1B
17	U.K	HADCM3	HadRM3Q6	25 km	A1B
18	U.K	HADCM3	HadRM3Q4	25 km	A1B
19	U.K	HADCM3	HadRM3Q3	25 km	A1B
20	U.K	HADCM3	HadRM3Q0	25 km	A1B





RCM ensemble skill for monthly precipitation 1970-1999

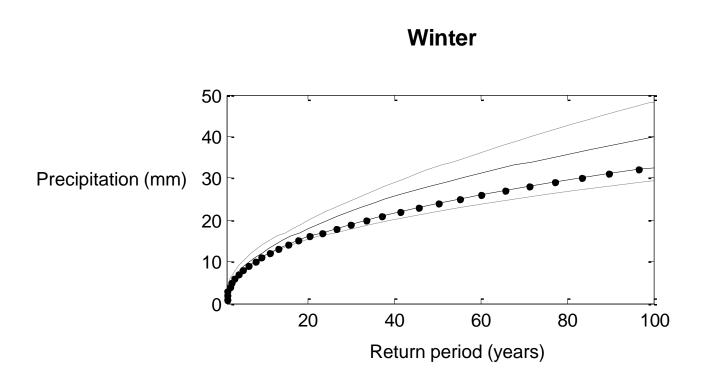




Frequency characteristics of RCM precipitation (red), against observed (black).

Dashed lines indicate the 5th and 95th percentiles of RCM output.





Daily rainfall extreme value distributions for RCMs vs Observed (dotted).

Solid and dashed lines indicate RCM mean and 5th and 95th percentiles respectively.

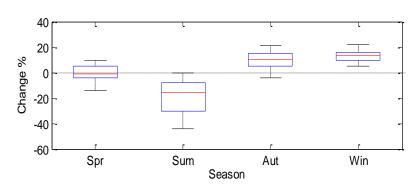


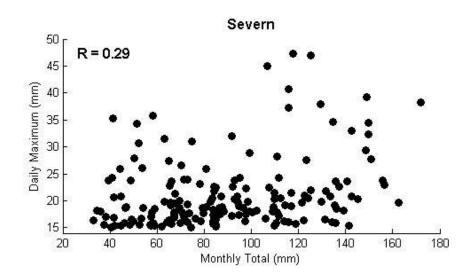
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Living with global uncertainty

Monthly Mean

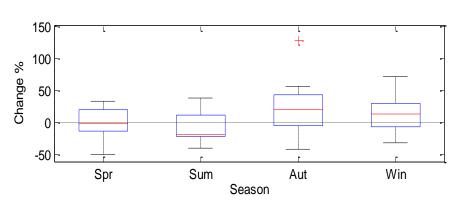
Severn



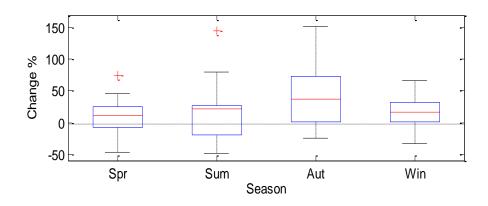




Maximum 5-10 Day Event



Maximum Daily Rainfall





Conclusions

- Hydraulic models can be developed that resolve individual buildings over whole urban areas
- Such schemes also open the way to probabilistic flood risk analysis
- A major application for such techniques is in predicting uncertain future change
- This is possible for sea level rise where the uncertainty in future change can be quantified
- However this is difficult for rainfall generating floods because climate models do not capture extreme rainfall well



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Uncertain model-data comparison

