The 21st century decline in damaging European windstorms

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Why have windstorm losses declined since 2000?



Notes: Spikes in losses correspond to wind storms in 1990 and flooding in 2007.

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The 21st century decline in damaging European windstorms

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Footprint database: 1979/80-2013/14

Windstorm footprint: Maximum 3-second wind-gust speed at each location in a 72 hour period covering the passage of the storm, centred on the time at which the maximum wind speed over land occurs

6301 storms identified in extended winters (October-March) using objective tracking algorithm (Hodges, 1995)

Then footprints created for each storm by dynamically downscaling ERA-Interim using the Met Office 25km resolution North Atlantic-European operational NWP model. Example: Footprint for windstorm Daria (24th - 26th January 1990)



Met Office

Damaging windstorm examples: 1990 & 2013





Area A₂₀ is a good predictor of severity



Klawa, M. and Ulbrich, U.: A model for the estimation of storm losses and the identification of severe winter storms in Germany, Nat. Hazards Earth Syst. Sci., 3, 725-732, 2003.

Indices defined as

$$L_{98} = \sum_{v_j > v_{98j}}^{J} d_j \left(\frac{v_j}{v_{98j}} - 1\right)^3$$
$$A_{20} = \sum_{v_j > 20}^{J} 1$$

where for grid point j, v_j is gust speed, v_{98j} is 98^{th} percentile of gust speeds, and d_j is population density.

\rightarrow Area A20 strongly related to L98 and a better classifier for largest loss storms

Roberts, J et al. (2014): The XWS open access catalogue of extreme European windstorms from 1979-2012, Natural Hazards and Earth System Sciences Discussions 2 (3), 2011-2048.

Recent trends: 1979/80-2013/14



← The number of European storms in Oct-Mar has become more volatile and has shown an increasing trend. No explanation yet for this recent 2% change which is not consistent with long term decrease expected from climate change projections. Early period is less volatile than chance i.e. Poisson standard deviation of sqrt(175)=13.



 \rightarrow 12% drop in average area which has led to a 10% drop in total area



- \rightarrow statistically significant decreases in frequencies over most of Europe.
- \rightarrow increase in 21st century over Iberian peninsula.

Change in gust speed distribution



Figure 5. Quantile-quantile plot of footprint wind gust speeds in the grid cell closest to Paris for events in the two comparative periods: winters in the 20th century (1979/1980–1999/2000) and winters in the 21st century (2000/2001–2013/2014). The thick solid black line shows where y = x, the dashed black lines show the 20 ms⁻¹ damage threshold and the thin solid black lines show the 95% confidence interval, based on the asymptotic sampling distribution of the order statistics.

Are the changes related to climate mode variability?



North Atlantic Oscillation (NAO) index for Oct-Mar winters 1900-2017 (Hurrell PC index)



→ NAO is known to modulate extreme storms (e.g. Economou et al. 2014)
→ NAO has been more positive during 1979-1999 than from 2000-2014

So can NAO changes account for the changes noted in footprint area?

Relationship between A₂₀ and Sea Level Pressure

Correlation between winter A20 and mean SLP Oct-Mar 1979-2014



A20 versus mean NAO index

 \rightarrow Correlation of A₂₀ with SLP shows NAO-like pattern (shifted eastwards slightly)

 \rightarrow Strong significant correlation (0.72) between A₂₀ and NAO indices

However, note than NAO doesn't explain everything e.g. 89/90 and 13/14 differences!

NAO as a loss driver on longer time scales



Fig. 2 Histogram of values representing the sum of storm loss index each year, and the corresponding running 10-year mean (*blue line*) for historical wind storms in the Netherlands



→ Additional confirmation of NAO being a potential driver of EUWS loss trends



Figure 14.16: Summary of multi-model ensemble simulations of wintertime (Dec-Feb) mean NAO, NAM and SAM sea-level pressure indices for historical and RCP4.5 scenarios produced by 39 climate models participating in CMIP5. Panels a-c) show time series of the ensemble mean (black line) and inter-quartile range (grey shading) of the mean index for each model.

→ NAO/NAM is *likely* to become slightly more positive (on average) SAM positive trend is *likely* to weaken as ozone depletion recovers

→ Medium confidence that projected changes in NAO and SAM are sensitive to boundary processes (stratosphere-troposphere interaction, ozone chemistry, response to Arctic sea ice loss), which are not yet well represented in many climate models

Climate Phenomena and their Relevance for Future Regional Climate Change



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Summary

- Area of wind gusts >20m/s is a good storm severity index for classifying high loss storms;
- Total area of wind gusts >20m/s has decreased by 10% in winters from 2000-2013 compared to winters 1979-1999;
- The decrease in total area is due to a significant reduction in wind speeds over much of Europe (apart from Iberian peninsula). Storm numbers have not shown any decrease;
- This wind speed change can be accounted for by a decrease in the North Atlantic Oscillation. Total area of wind gusts is highly correlated with winter mean NAO;
- The decrease in NAO is contrary to what one expects due to global warming so change is most likely natural variability. Indication that we are now heading back into a more positive phase of NAO.

Additional slides for questions etc. ...

Future directions: use of big data

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Figure 1: The Passage of the Great Storm

Windstorm footprint estimation typically only uses standard meteorological observations which can have poor spatial coverage (e.g. at urban postcode scales);

There are many other less conventional data sources that may be useful for inferring high gust speeds e.g.

- Amateur observer network (WOW);
- Newspaper reports;
- Social media feeds e.g. twitter;
- Webcams and CCTV imagery.

Can this data be mined to improve footprint estimation? The stochastic model is a way of integrating these diverse data sources.

NERC Highlight Proposal BIGFOOT (£1.5 million funding – 4 postdocs – 3 years).

Let me know if you have any ideas about this or would like to be involved!

The Storm (1704) by Daniel Defoe



Projected CMIP5 ensemble mean changes RCP4.5 scenario Difference in time means over 2070-99 and 1976-2005

1.5

0.9

0.3

-0.3

-0.9

-1.5



Summer (JJA)

Track density (no. of storms/month)







Wind speed intensity (m/s)





0.6

0.2

-0.2

-0.6

-1





Precipitation intensity (mm/hour)





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