

Global Flash Flood Forecasting from the ECMWF Ensemble

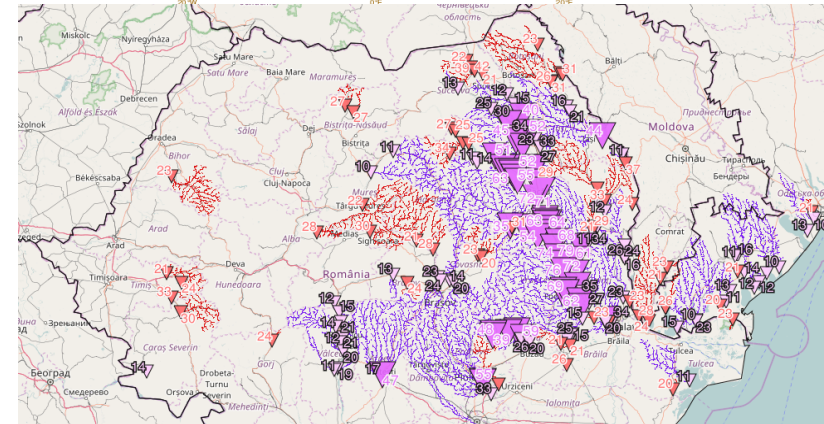
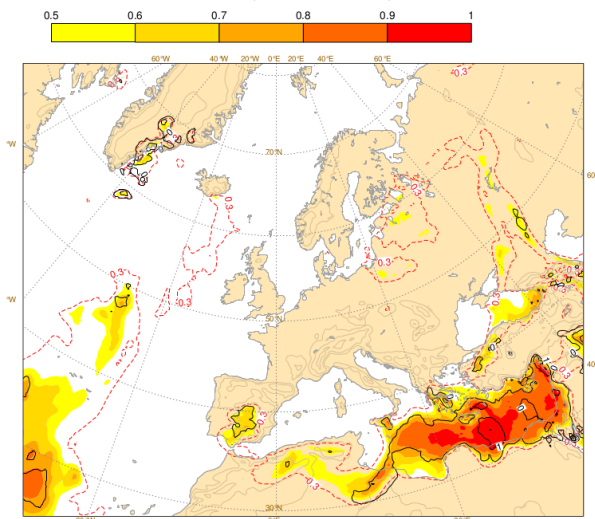
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Building a Global FF System

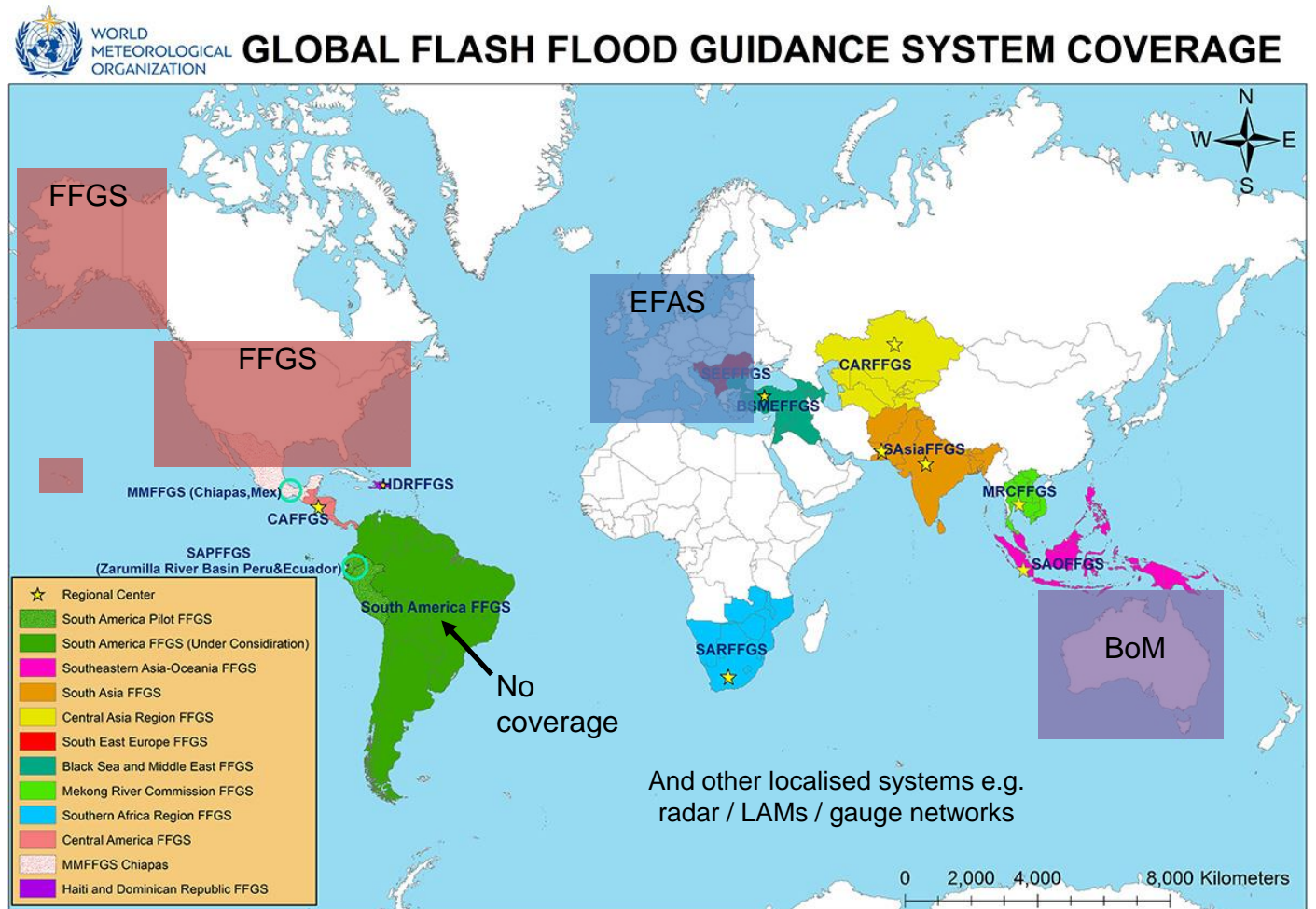
1. Current flash flood forecasting globally
2. Flash flood forecasting from ECMWF ENS
 - Extreme Forecast Index (EFI)
3. Verification
4. Enhancing forecasts with population exposure

Sun 06 Sep 2015 12UTC ©ECMWF 1-12-36h VT: Mon 07 Sep 2015 00UTC - Tue 08 Sep 2015 00UTC
Extreme forecast index and Shift of Tails (black contours 0,1,2,5,8) for CAPE



Current Flash Flood Systems Globally

- Single global system not present
- Instead a piecemeal approach regionally / locally – important?
 - Opportunity for users to tailor their own systems
 - Or contributes to high risk associated with flash floods?

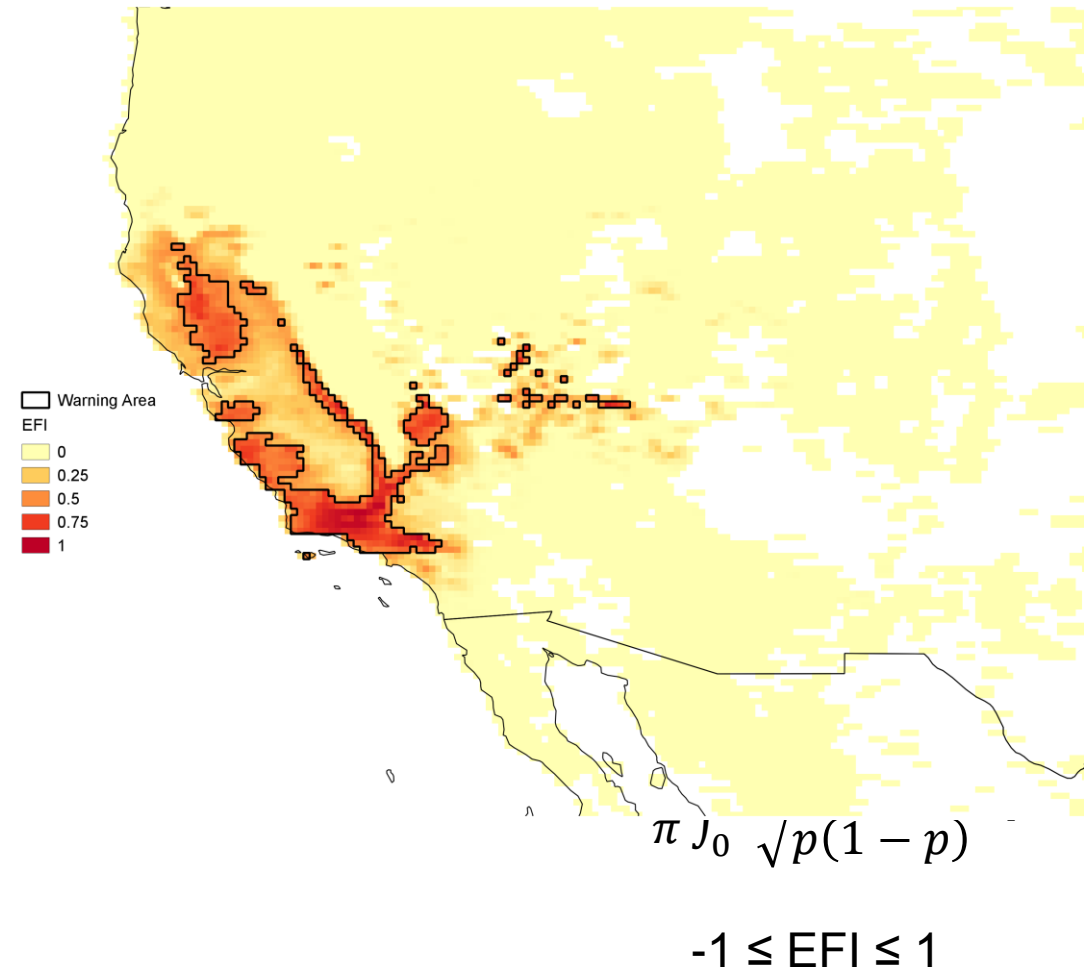


Adapted from: http://www.wmo.int/pages/prog/hwrp/flood/ffgs/images/FFGS-global-coverage14_12_2016-full.jpg

Converting Forecasts to Warnings

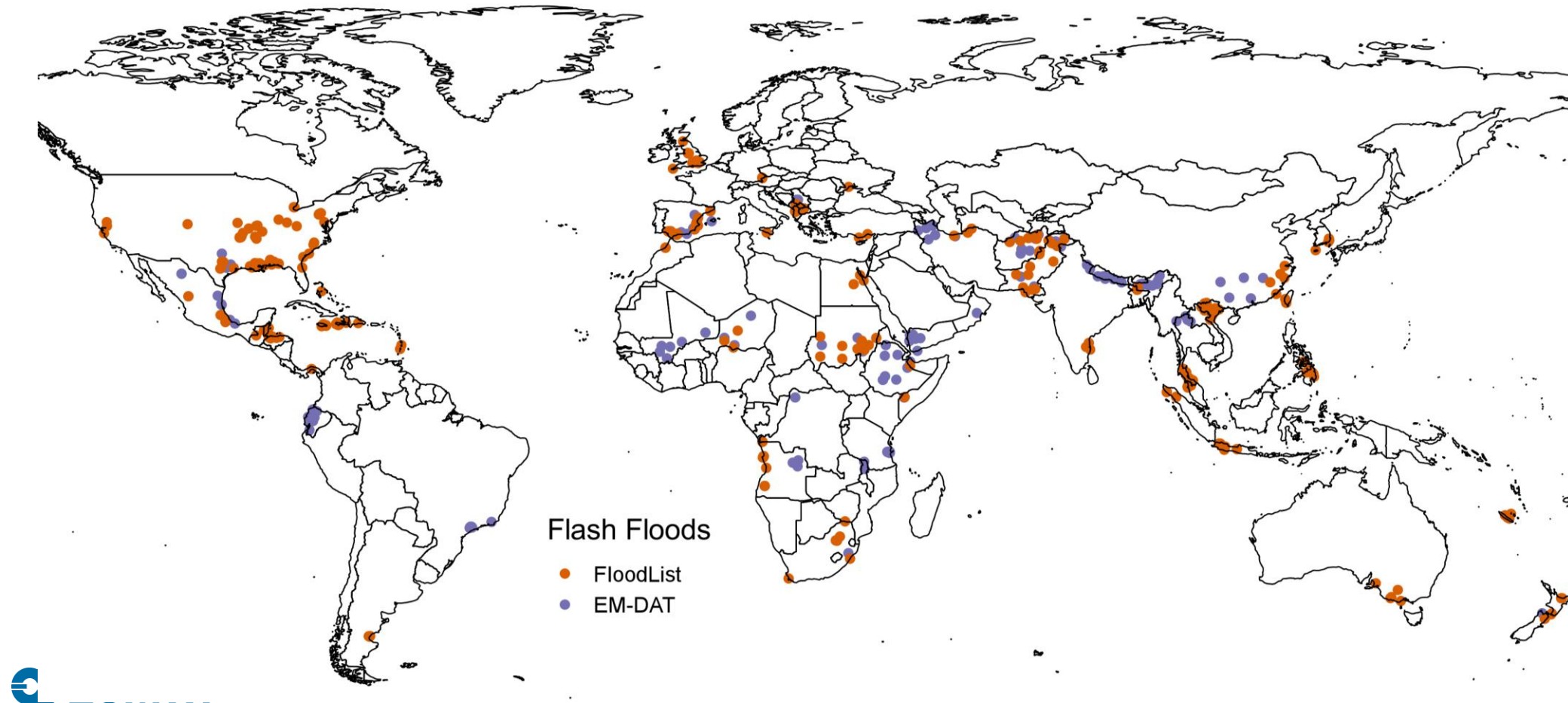
Extreme Forecast Index (EFI):

- Integration of difference between model forecast and model climatology (20 years reforecasts)
 - $EFI > \sim 0.5$ = severe event
- Convert to warning areas based on:
 - Minimum EFI threshold



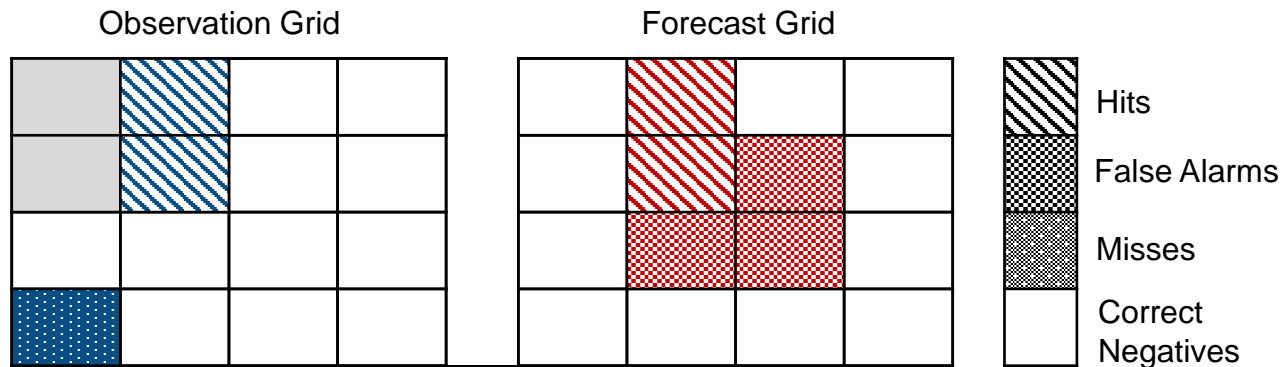
Verification Experiment: March 2016 – March 2017

- Calculate total precipitation & surface runoff EFI daily at 00 UTC, 6 hourly out to 120 h lead time
- Create warning areas using EFI thresholds 0.0 – 1.0 (increments = 0.05)
- Compare against 'flash flood' observations from EM-DAT (161) & FloodList.com (238)

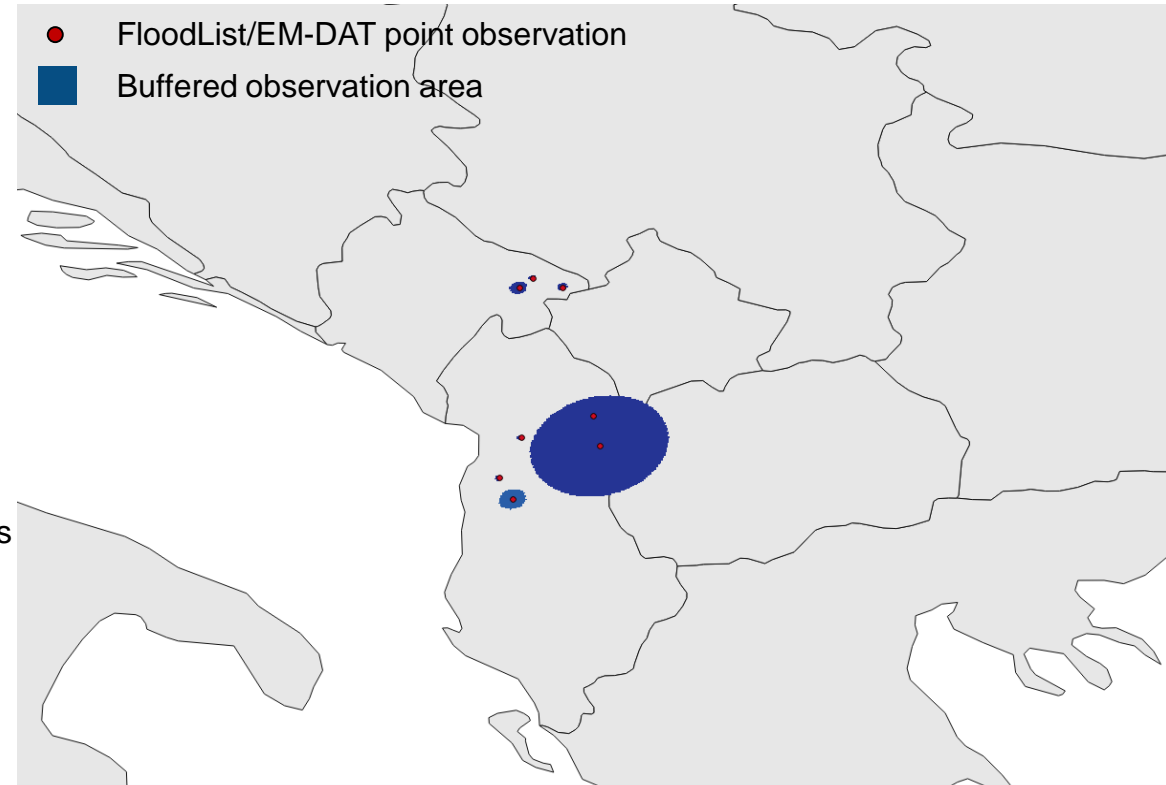


Verification: Methodology

- Convert point observations to 18km ENS grid:
 - Buffer by spatial uncertainty (1 to 100's km)
- Create 1 observation grid for each verification date
- Compute the contingency table
- Summarise over the whole verification year

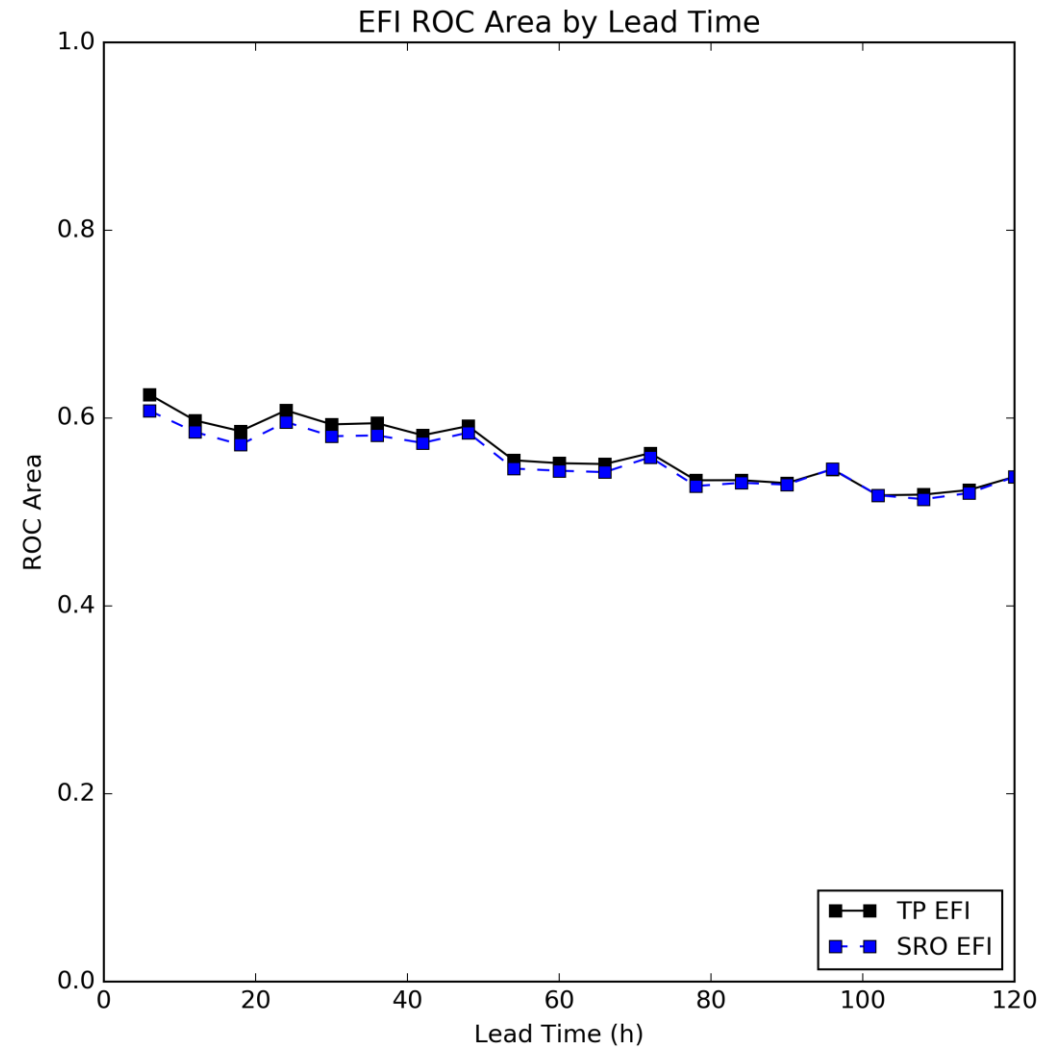
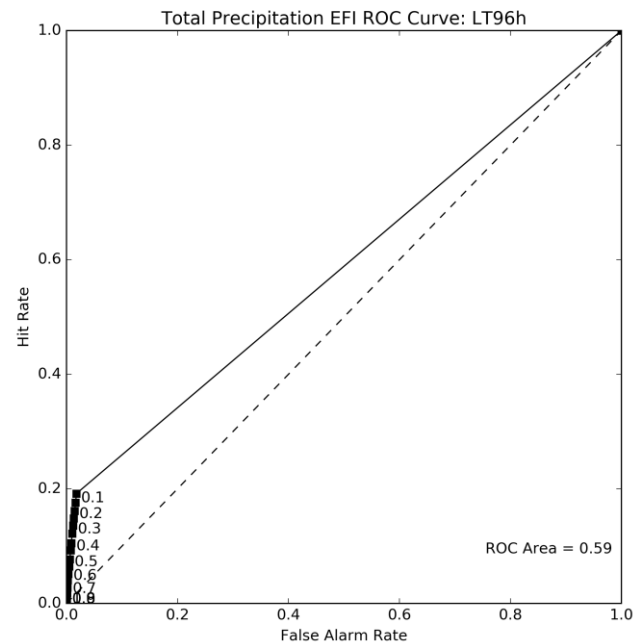
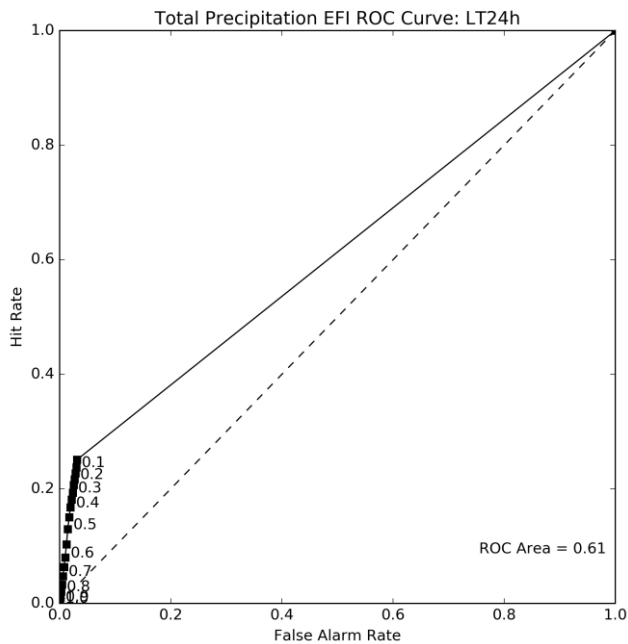


		Observed?	
		Yes	No
Forecasted?	Yes	2	3
	No	1	8



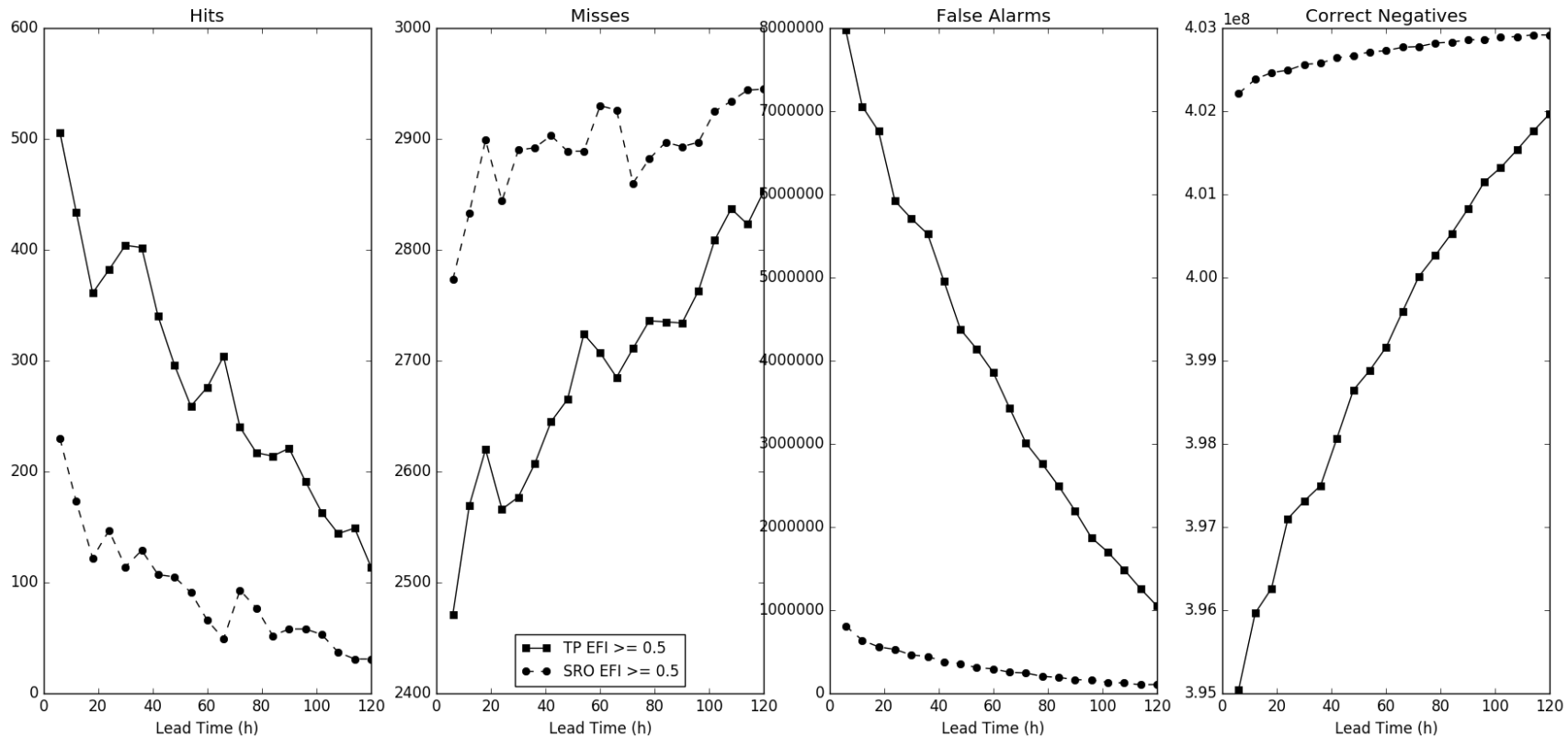
Verification Results: ROC Area

- ROC Area ~0.6
 - Only shallow gradient with lead time – really??
- ROC curves actually clustered in bottom left corner
 - Really low hit and false alarm rates – the latter doesn't really vary



Verification Results: Raw Contingency Table Results

- Surface runoff is less active than total precipitation
- Misses $\sim \times 9$ greater than hits = low hit rate
- Correct negatives $\sim \times 1000$ greater than false alarms = v. low false alarm rate
- Skews the subsequent analysis



Verification Results: Problems with Global Analysis

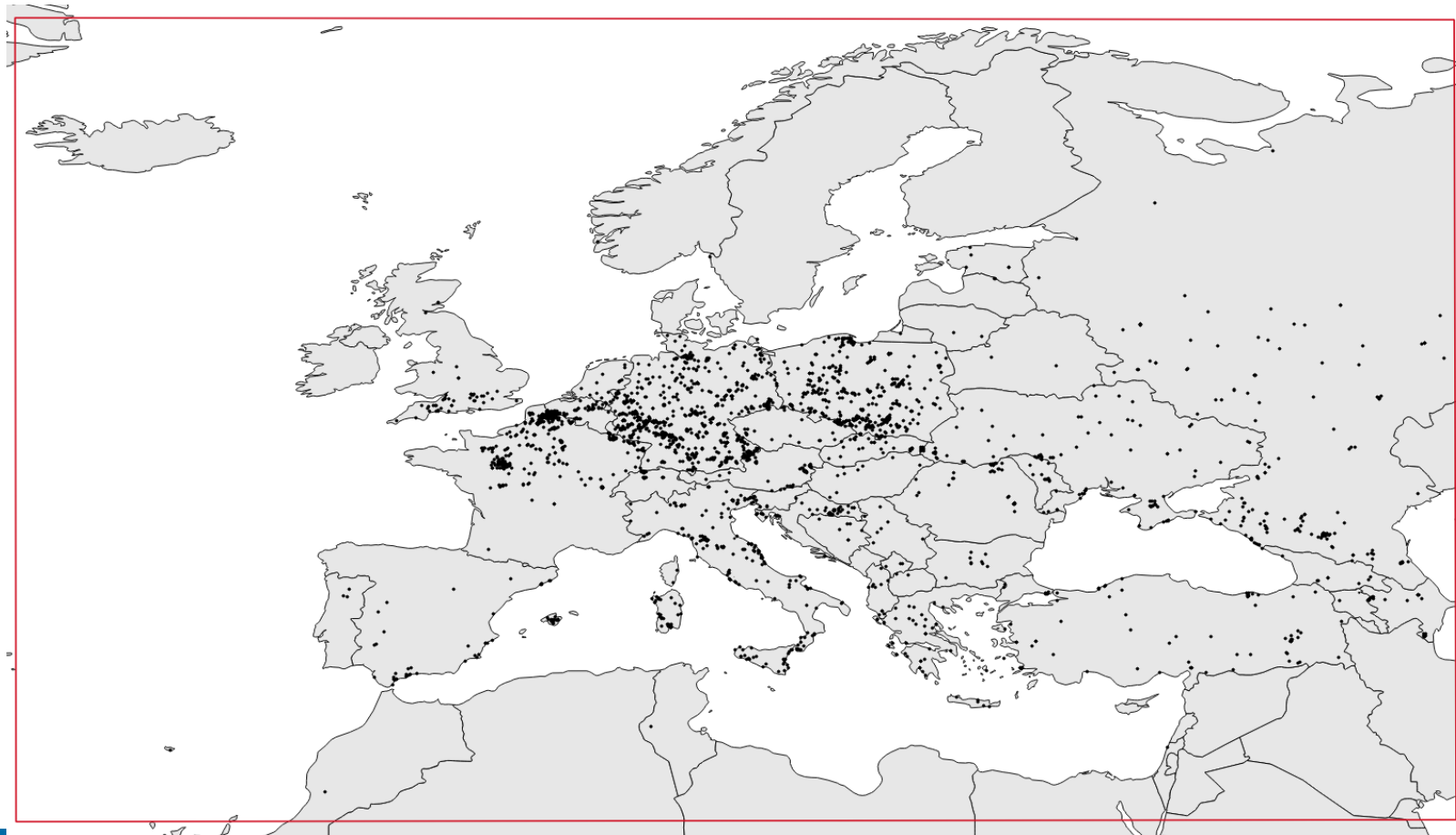
- Only actually have 199 observations (w. spatial uncertainty $\leq 18\text{km}$)
 - = 995 observation pixels over whole verification period
- 1,116,414 land pixels * 365 days
- Consequently the contingency table is unevenly distributed
 - Need to be careful not to repeat the Finley affair (1884)

$$\text{Finley skill} = \frac{\text{Hits} + \text{Correct Negatives}}{\text{Total}} = \frac{0 + (1,116,414 * 365)}{0 + 995 + 0 + (1,116,414 * 365)} = 99.99\%$$

- Cannot robustly verify the system at global scale
- Need to focus on smaller area with higher density observations >> Europe

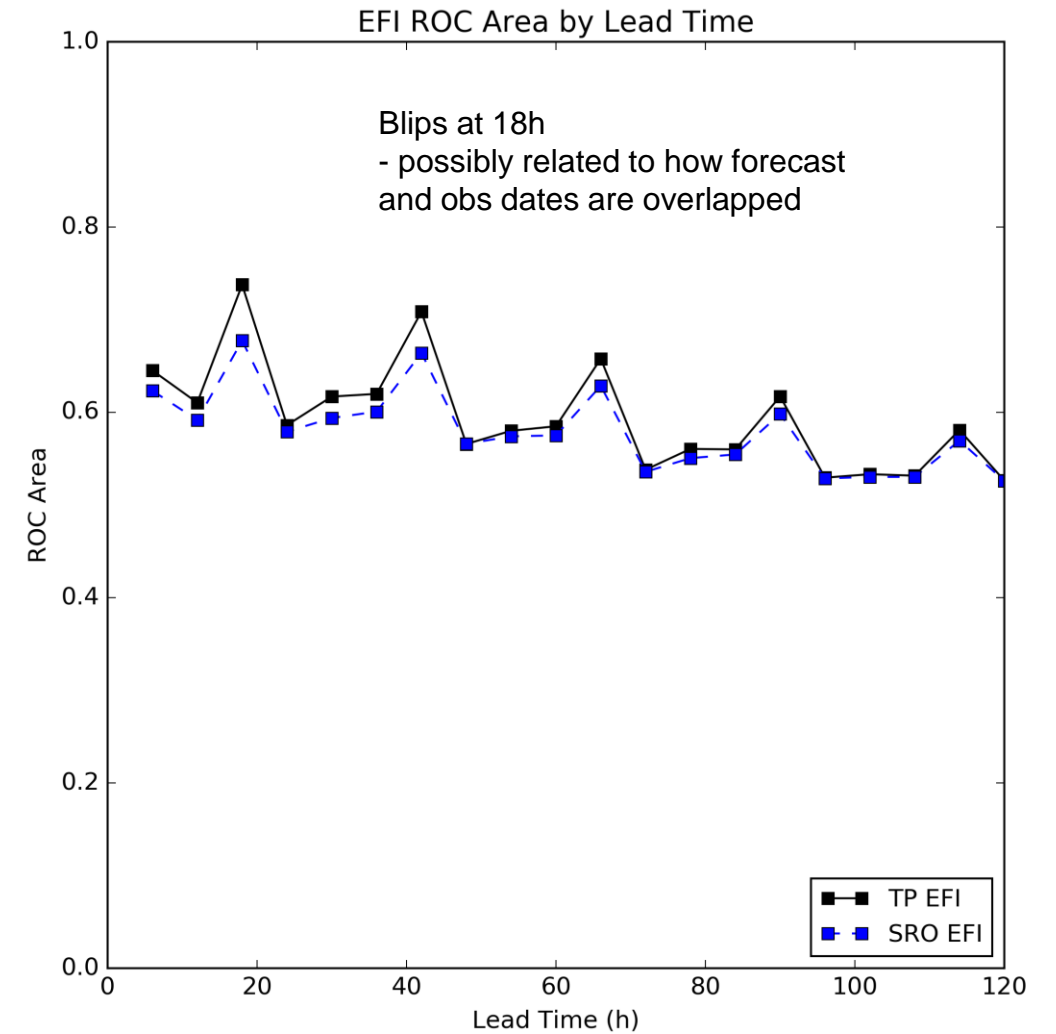
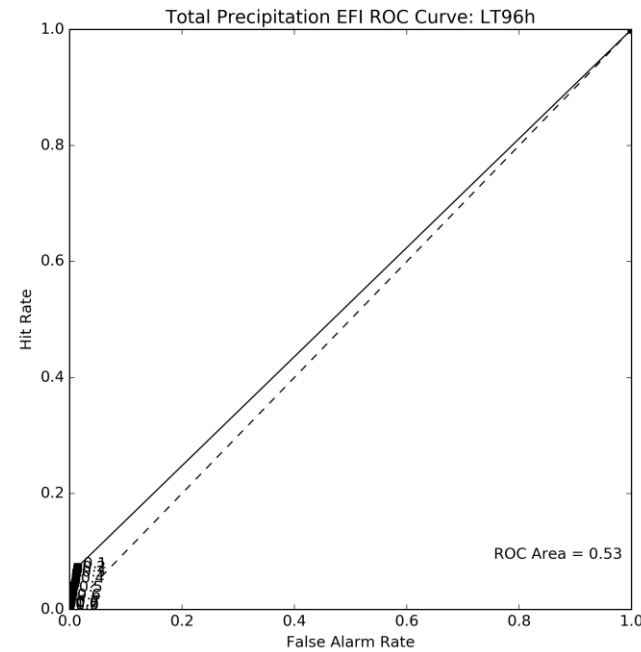
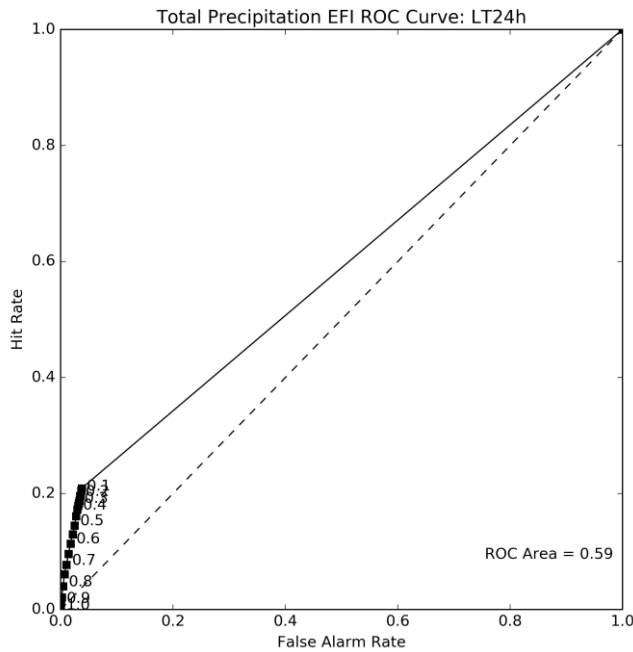
European Verification

- European Severe Weather Database (ESWD) logs 2,544 heavy rainfall reports which mention flooding
- Repeat previous analysis: spatial uncertainty < 18km, timing uncertainty < 6 hours



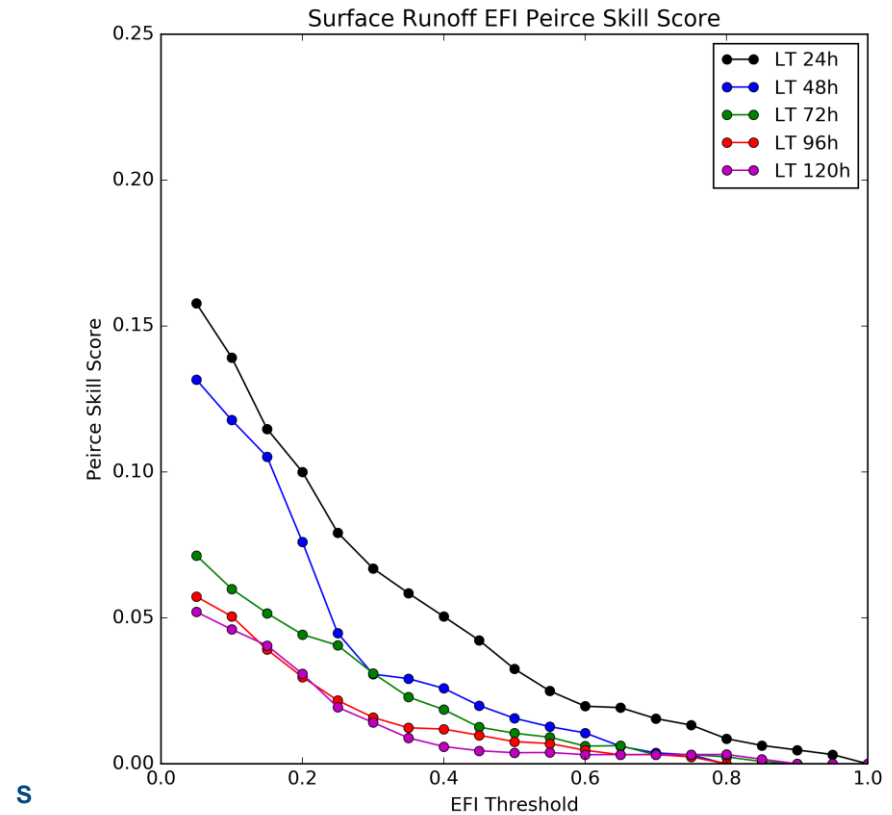
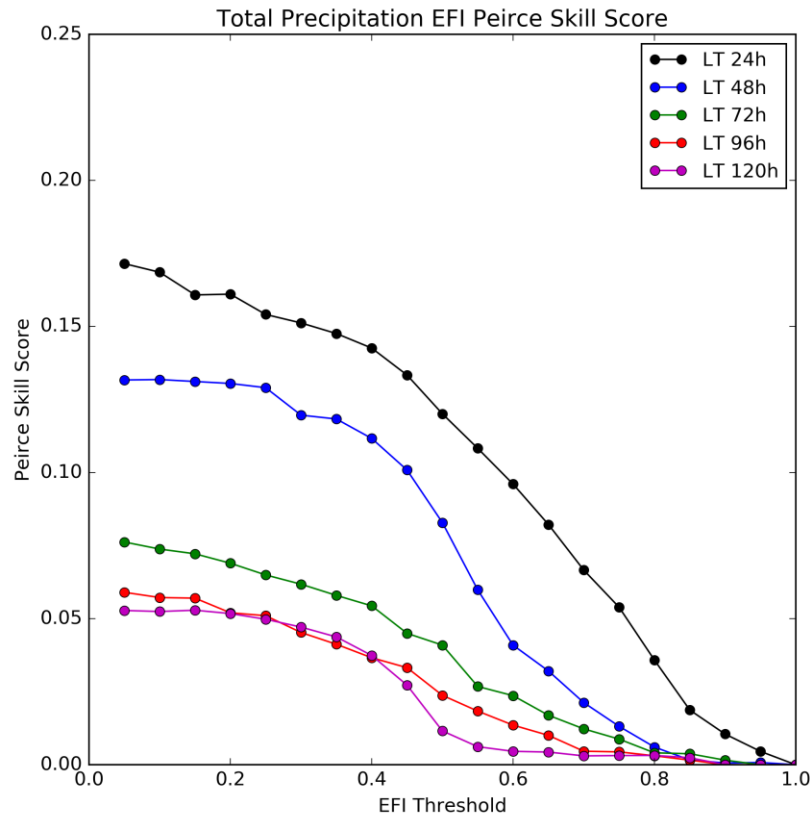
European Verification: Results

- ROC score still ~0.6
- ROC curves clustered in left hand corner
- Correct negatives and false alarms still dominating results



European Verification: Peirce Skill Score

- = Hit Rate – False Alarm Rate
 - How well can the forecast distinguish ‘yes’ events from ‘no’ events
- Hit rate marginally better than false alarm rate
 - Surface runoff EFI skill drops off rapidly with increased EFI threshold



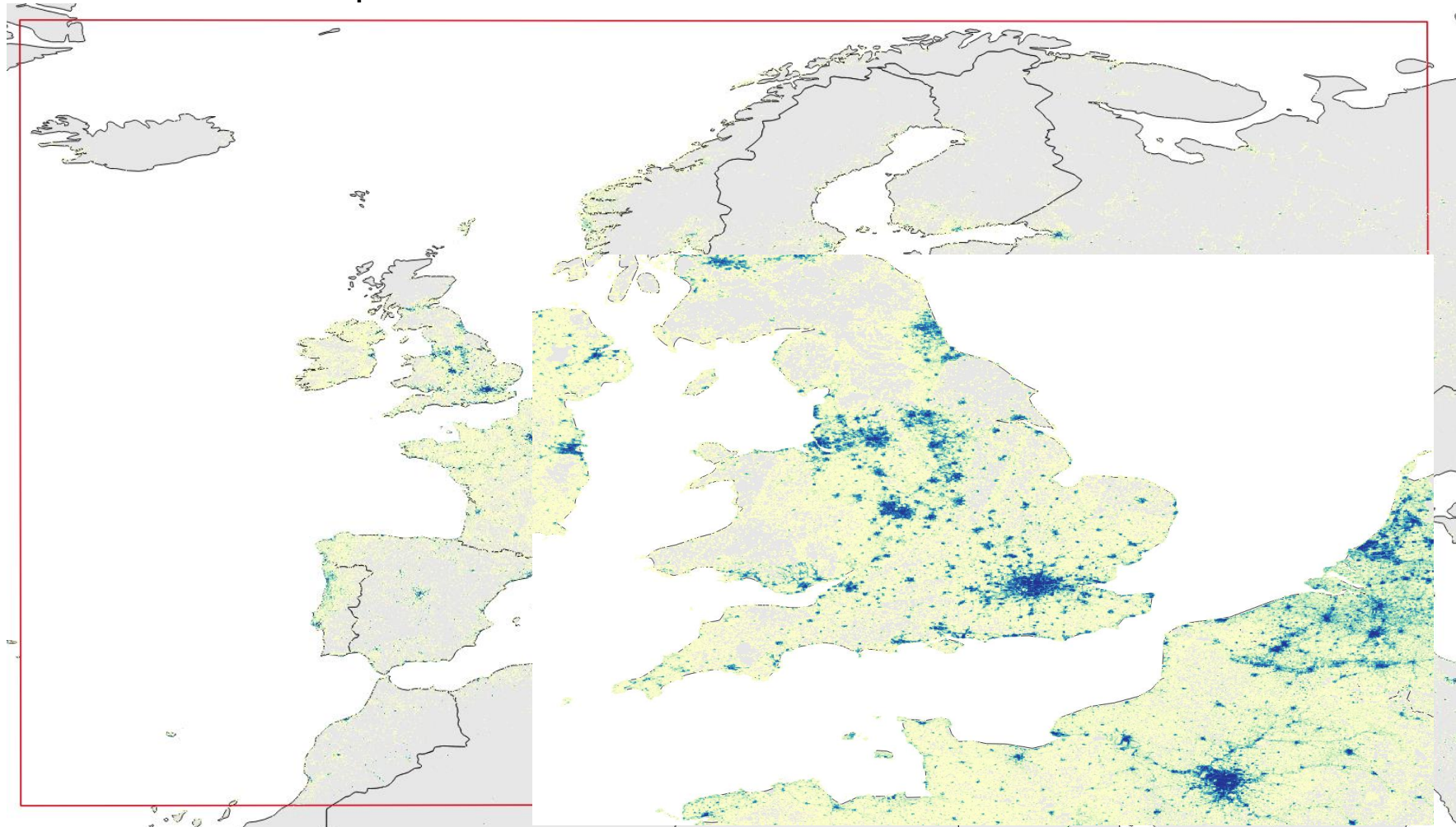
Verification so far...

- On balance have a slight semblance of skill at shorter lead times
 - Total precipitation edges surface runoff but at cost of more false alarms
- Metrics are being hindered by uneven distribution of contingency table values
 - Correct negatives are being awarded where there could never be a (reported) event e.g. Sahara desert
 - Therefore refine verification only to areas where we can expect reports i.e. populated areas...

Inclusion of Population Exposure: GHSL

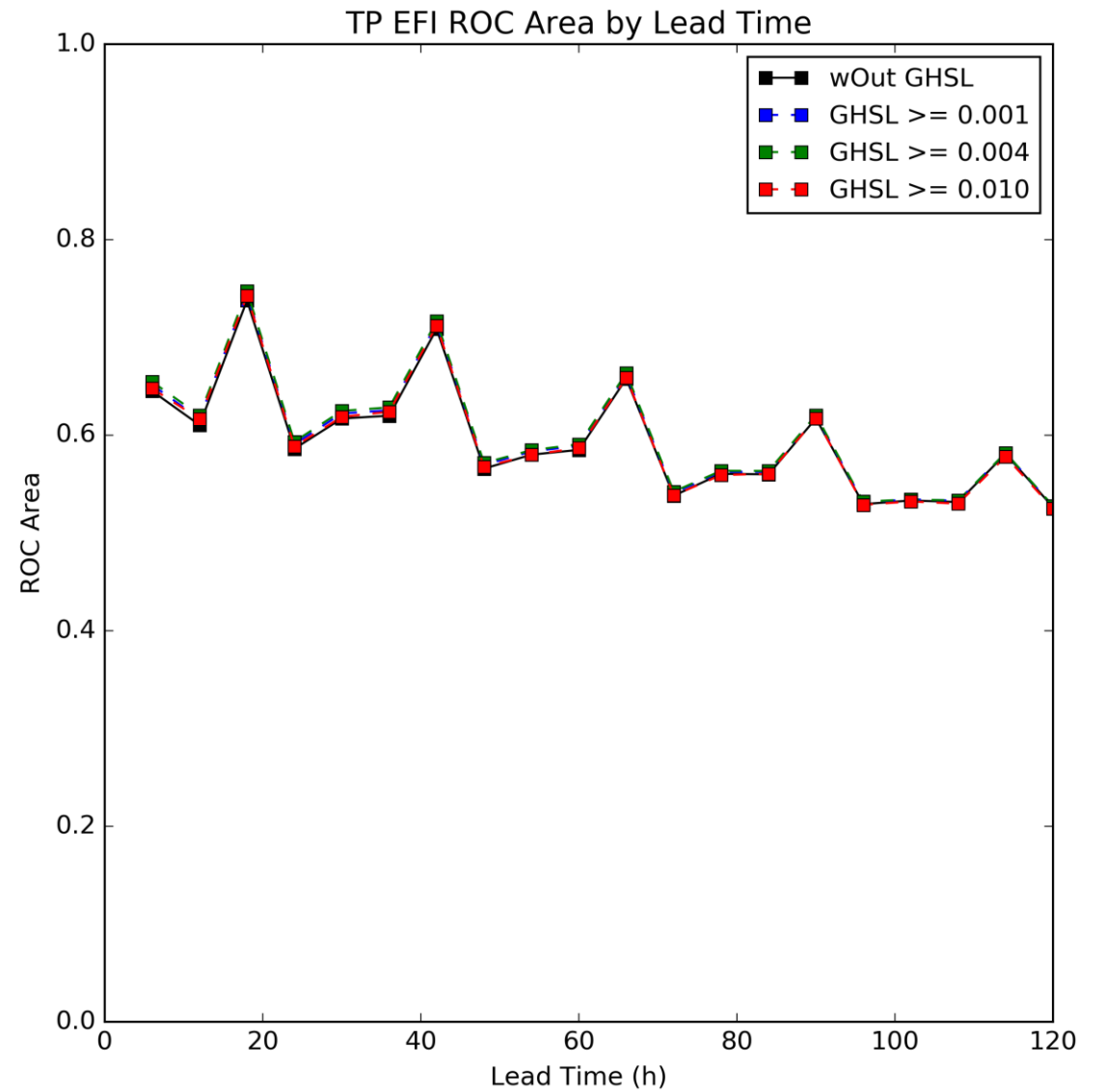
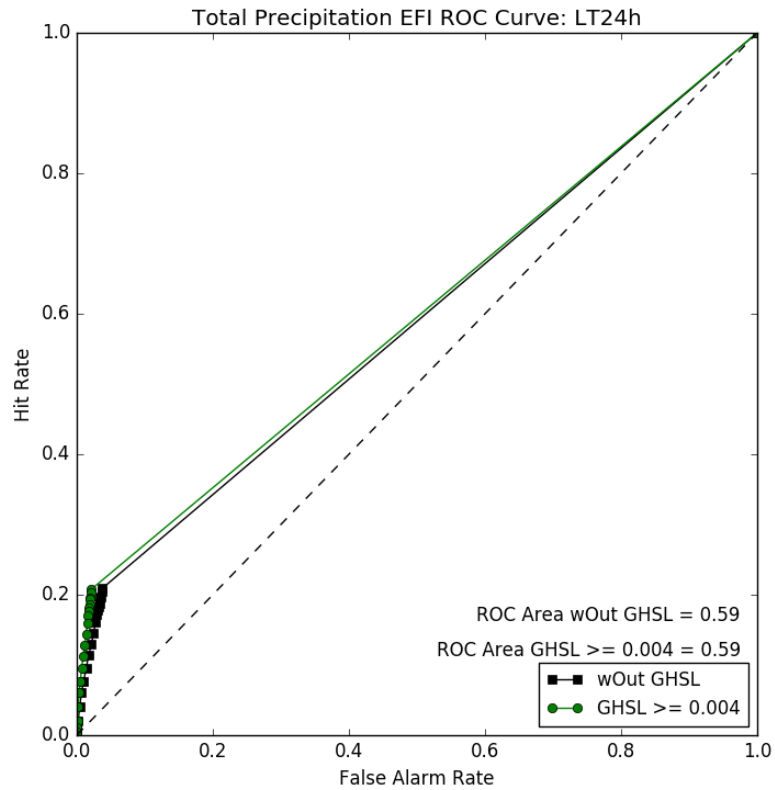
- Global Human Settlement Layer – developed at JRC
- % urbanised cover in each 1km pixel

Apply threshold to
remove unpopulated
areas from analysis



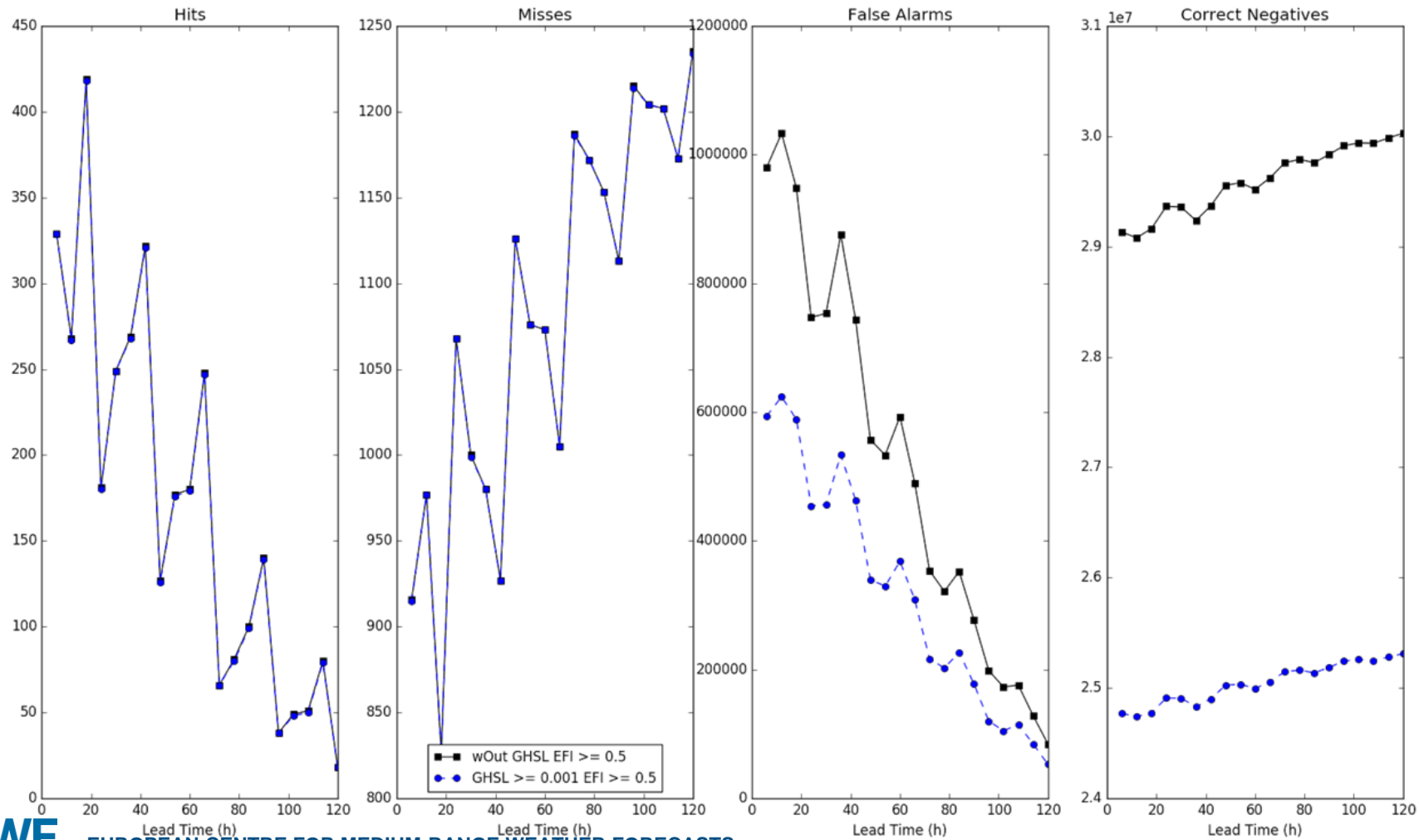
Verification Results: with GHSL

- ROC scores show little difference ...



Verification Results: with GHSL

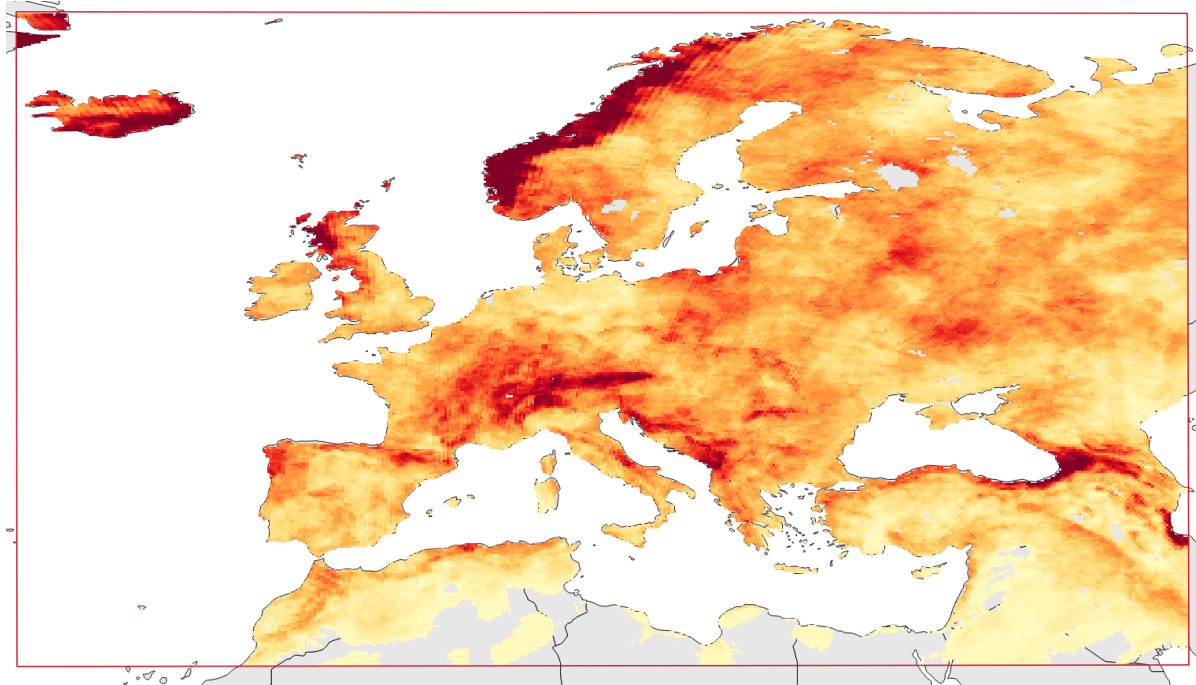
- ...large reduction in number of false alarms and correct negatives
- No difference in hits and misses



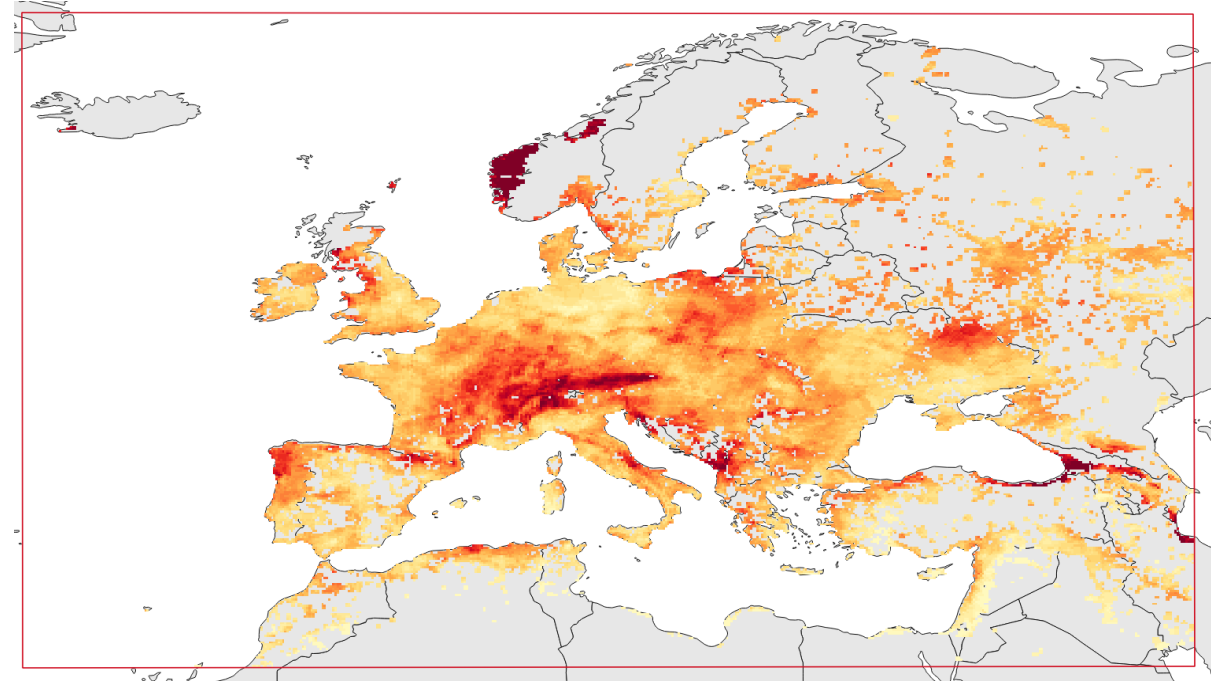
Verification Results: with GHSL

- Total Precipitation EFI: False Alarms
- Reduction in Scandinavia and eastern areas

Without GHSL



GHSL ≥ 0.004

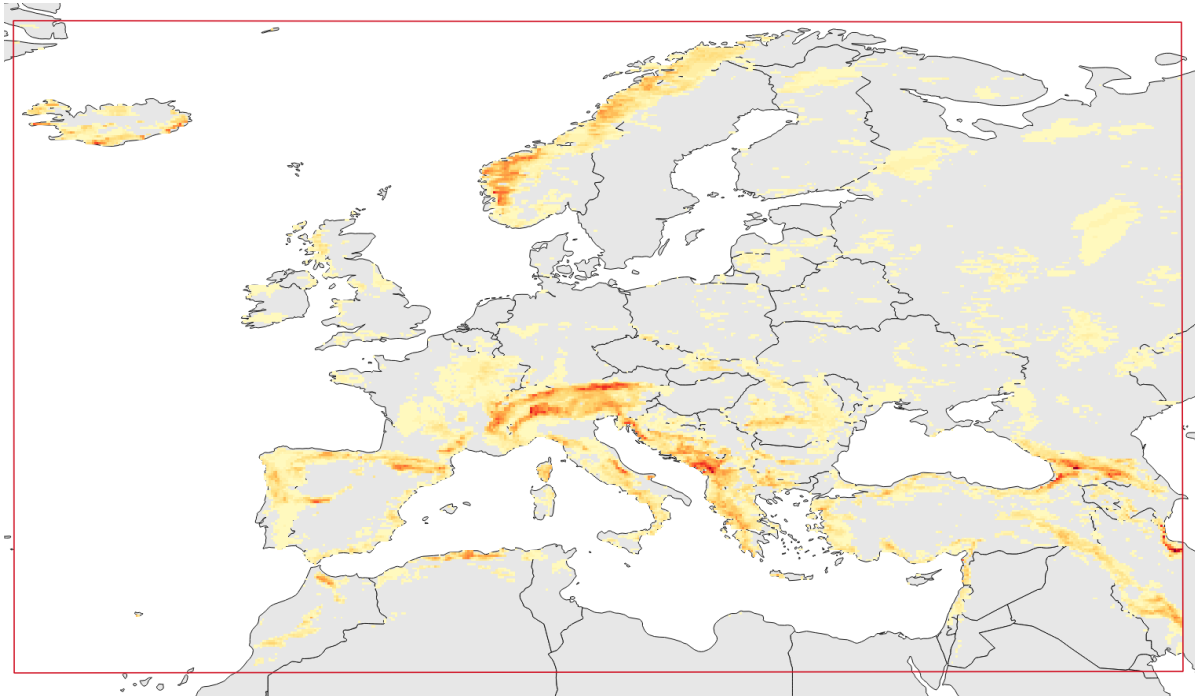


EFI ≥ 0.50
Lead Time 24h

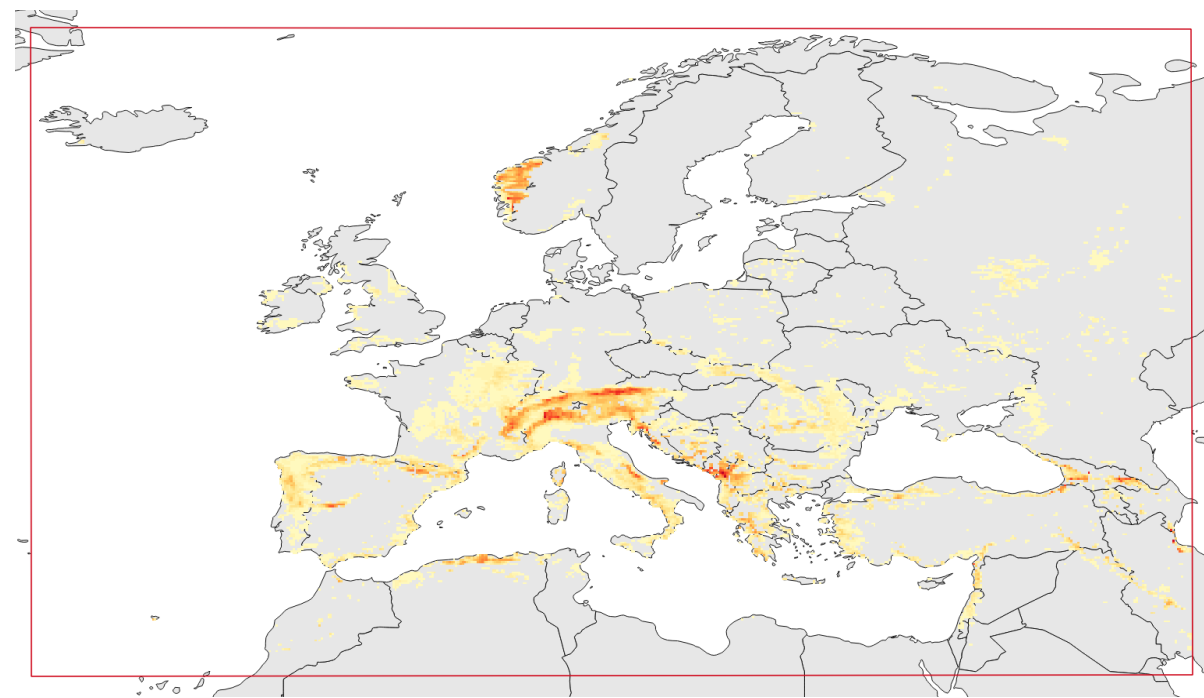
Verification Results: with GHSL

- Surface Runoff EFI: False Alarms

Without GHSL



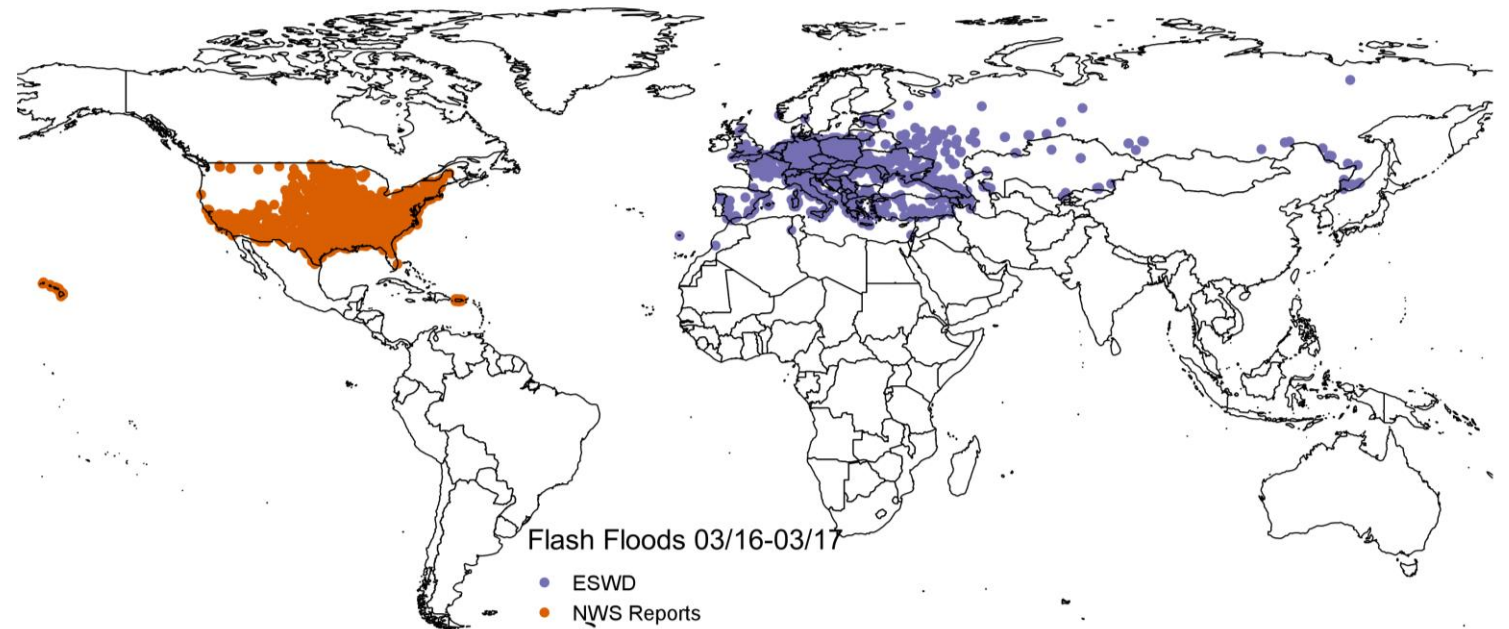
GHSL ≥ 0.004



EFI ≥ 0.50
Lead Time 24h

Future Work

- Repeat analysis in USA using NWS storm reports
- River routing function – simple MC approach?
- Other exposure variables?
 - Transport networks
 - Infrastructure



Final Conclusions

- No global flash flood forecasting system – leaves areas with no capability
- Global ECMWF total precipitation/surface runoff forecasts give limited predictability of flash floods
 - Issue of scale: misses could be localised convection
 - Not enough observations to do robust verification
- Focussing verification upon Europe showed:
 - Large number of false alarms
 - Precipitation (more false alerts) and surface runoff (more misses) forecasts showed similar skill
- Thresholding by population exposure helps to reduce false alarms
 - Though little change in skill score

Thank You

Questions?

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