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The link between catchment precipitation forecast skill and spread to that of downstream ensemble hydrological forecasts

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# Outline

- 1. Project aims, methodology and data sets
- 2. Results
  - $\circ\,$  Sensitivity to observation type
  - $_{\odot}\,$  National and regional variations in precipitation and river flow forecast attributes
  - o Spatial variations in skill
  - Mapping precipitation skill onto river flow skill by catchment
- 3. Conclusions

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# Project outline and data sets

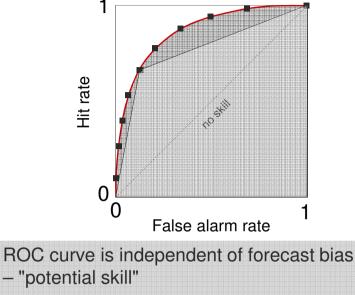
Table 2.1 Metrics to be used for NWP and Hydrological forecast verification.

Common between NWP and Hydrology	NWP only	
<ul> <li>Continuous Rank Probability Score (CRPS) with decomposition</li> <li>Brier Score (BS) with decomposition</li> <li>Continuous Rank Probability Skill Score (CRPSS)</li> <li>Brier Skill Score (BSS) with decomposition</li> <li>Reliability Diagram</li> <li>Relative Operating Characteristic Diagram and Area Under Curve Skill Score (ROCSS)</li> <li>Relative Economic Value (REV)</li> <li>Rank Histogram</li> </ul>	<ul> <li>Mean Error (ME, a measure of bias) of areal mean precipitation per member</li> <li>Root Mean Squared Error (RMSE) of areal mean precipitation per member</li> </ul>	

- Aim to develop a verification framework for catchmentscale *ensemble* precipitation which drives a river-flow *ensemble*.
- Understand how *uncertainty* (affecting skill) is propagated from precipitation to river flow.
- Understand the *impact of truth type* and *accumulation length* on precipitation verification.

Deterministic	Probabilistic (suggestions)	Visual aid
Mean bias	Reliability term of BS $\frac{1}{N}\sum_{k=1}^{K} n_k (p_k - \overline{o}_k)^2$	o <sup>ŏ</sup>
RMS error	Brier score (square root) $\sqrt{BS} = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (p_i - o_i)^2}$	Pk
Mean absolute error	CRPS $\int (P_{fcst}(x) - P_{obs}(x))^2 dx$	CDF 0 x





- Area under curve ("ROC area") is a useful summary measure of forecast skill
- Perfect:1;No skill: 0.5



Measures how well the ensemble spread of the forecast represents the true variability (uncertainty) of the observations

Count where the verifying observation falls with respect to the ensemble forecast data, which is arranged in increasing order at each grid point.

Perfect = flat; u-shape=under spread, many obs falling outside bounds of ensemble



# Selative value score

Measures the relative improvement in economic value as a function of the cost/loss ratio C/L for taking action based on a forecast as opposed to climatology

if  $C/L < \overline{O}$ 

if  $C/L > \overline{O}$ 

 $V = (1-F) - \left(\frac{1-C/L}{C/L}\right) \left(\frac{\overline{o}}{1-\overline{o}}\right) (1-H)$ 

 $V = H - \left(\frac{C/L}{1 - C/L}\right) \left(\frac{1 - \overline{o}}{\overline{o}}\right) F$ 

where H is the hit rate and F is the false alarm rate

- The relative value is a skill score of expected expense, with climatology as the reference forecast.
  - Range:  $-\infty$  to 1. Perfect score: 1
  - Plot V vs C/L for various probability thresholds. The envelope describes the potential value for the probabilistic forecasts.





# Data sets



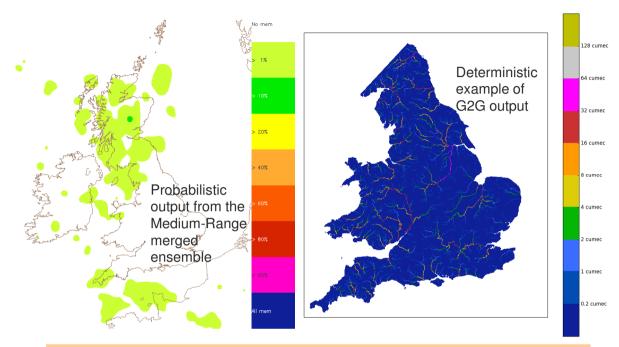
Fig. 1 The UK weather radar network

# Second state54°N52°N50°N6°W6°W2°W0°52°N6°W70070

= Met Office
 = Environment Agency

# Radar-only (gauge-adjusted) analysesGridded gauge-only analyses

Two periods considered: winter Nov-Dec 2015 & summer May-Jun 2016.



- Merged nowcast (STEPS t+7h) and MOGREPS-UK (2.2 km –  $\sim$ t+32h) and MOGREPS-G (32 km –  $\sim$ t+144h) precipitation ensemble out to 6 days, on a 1 km grid. Output is at 15 min resolution.
- River-flow ensemble (G2G) output at 15 min intervals

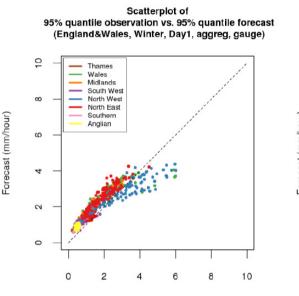
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## Met Office College & Hydrology Radar v gauge

- Only showing results for Day 1.
- Precipitation was evaluated at the hourly and daily scale, aggregated over 4 runs per day

Centiles in good general agreement.
Radar shows more large totals, which may be due to the detailed spatial sampling radar can provide. [Interpolation can not "create" large totals between gauges, i.e. the largest totals may be missed if they don't occur precisely over a gauge. More acute problem in convective conditions]
Tendency for model to produce too much light

rain but under-estimating larger totals.



Observation (mm/hour)

3.0

2.5

2.0

1.5

0

0.5

0

C

0.0

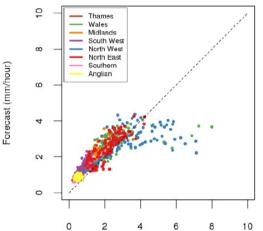
<sup>-</sup>orecast (mm/hour)

Scatterplot of

95% guantile observation vs. 95% guantile forecast

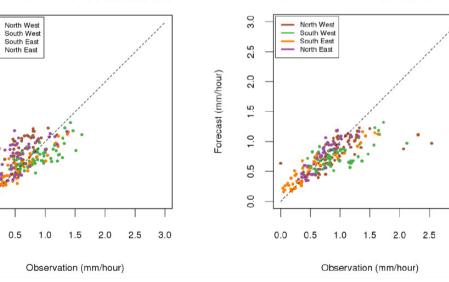
(Scotland, Summer, Day1, aggreg, gauge)



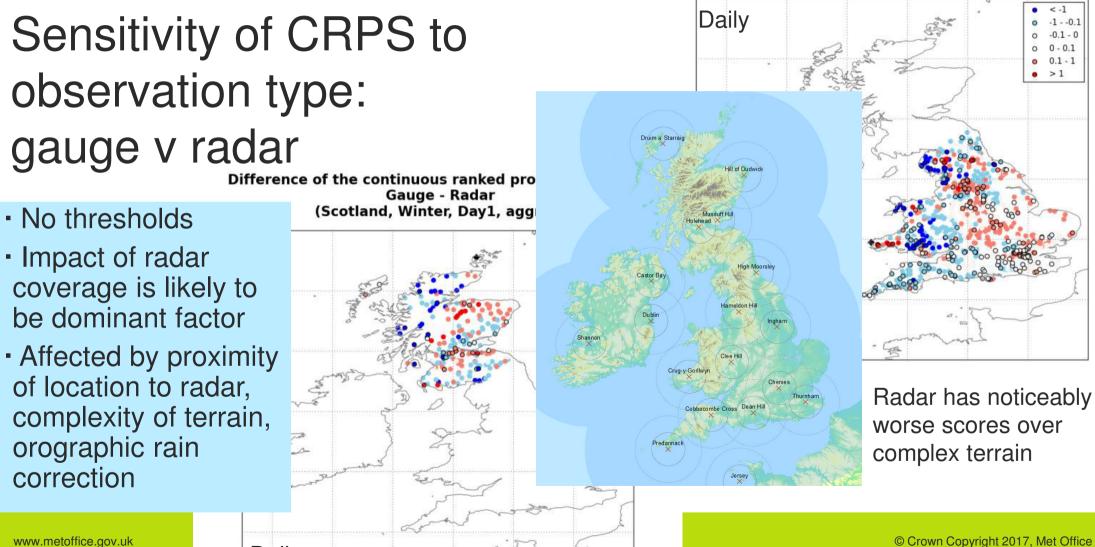


Observation (mm/hour) Scatterplot of 95% quantile observation vs. 95% quantile forecast (Scotland, Summer, Day1, aggreg, radar)

30



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Difference of the continuous ranked probability scores

Gauge - Radar (England&Wales, Winter, Day1, aggreg)

Daily

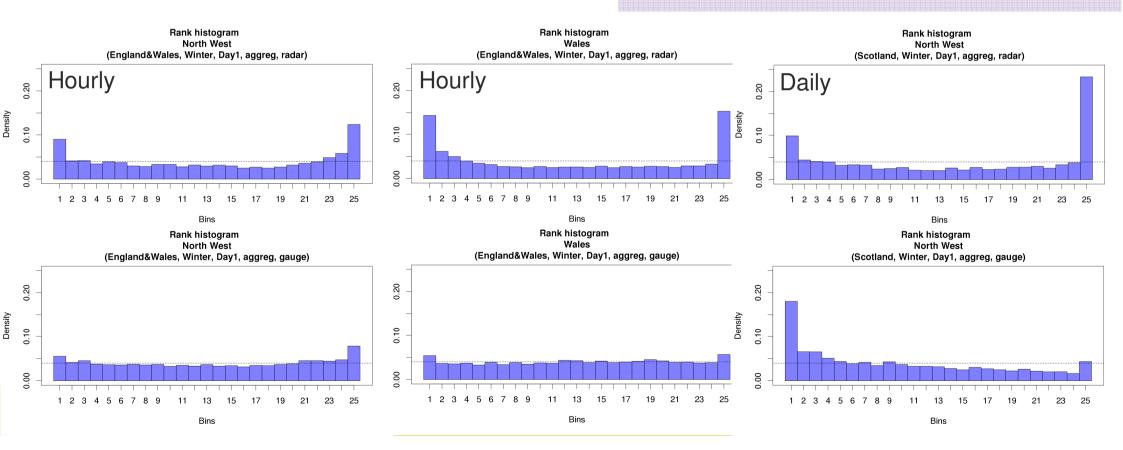
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# Sensitivity of rank histogram to observation type: gauge v radar

Rank Histograms are found to be very sensitive to the underlying distribution, which affects interpretation of ensemble spread.

 Strong regional variations which are also be related to radar coverage and the spatial sampling capabilities of radar compared to the point measurements from the gauges.



#### **Day 1 forecasts, England & Wales**

ROC

21 23 25

0.0

0.4

1 3 5 Pielae alarminate

13 15 17 19

Bins

6.6

#### Centre for Ecology & Hydrology Met Office National scale variation

0.6 0.8 1.0

¥0

80

0.0

0.2

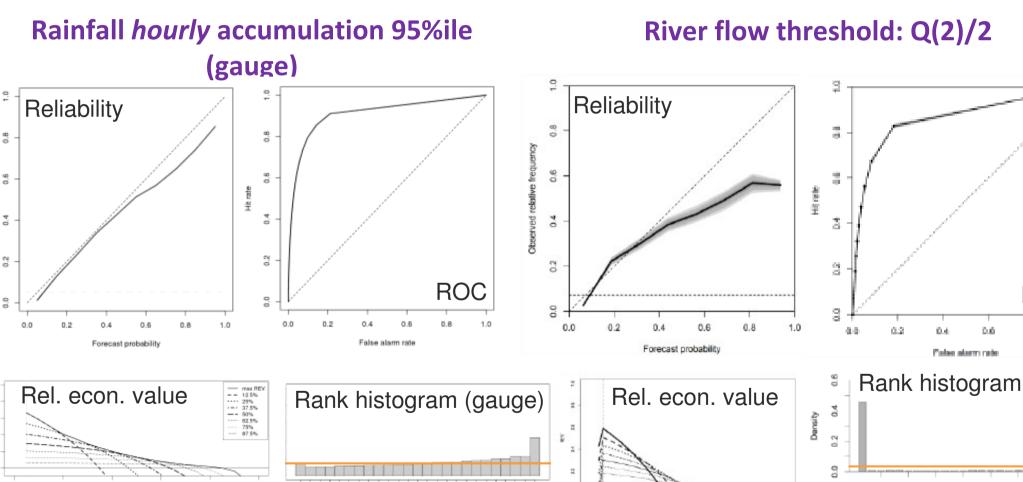
0.4

Cost/Loss

0.8

1.0

1.2.3



92

24

Constraint, and

20

0.8

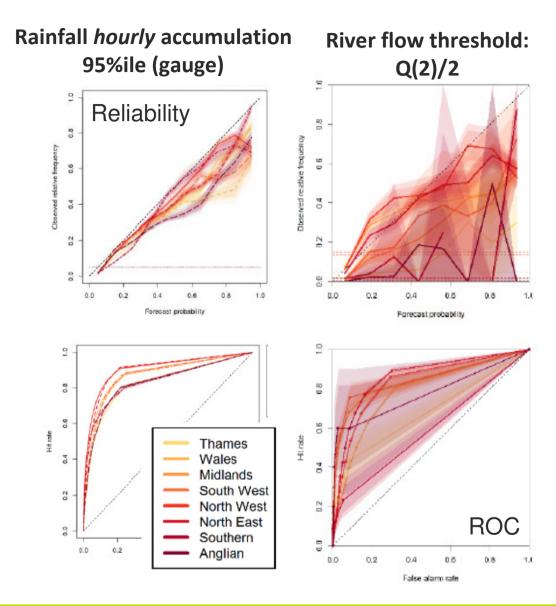
0.9

#### River flow threshold: Q(2)/2

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# Regional scale variation

- Day 1 results for England and Wales winter.
- River flow forecasts generally overconfident.
- Sampling issues for the river flow even for very modest thresholds, leading to very large variations.
- Precipitation forecasts also overconfident but fairly reliable with good potential skill (Area under ROC).





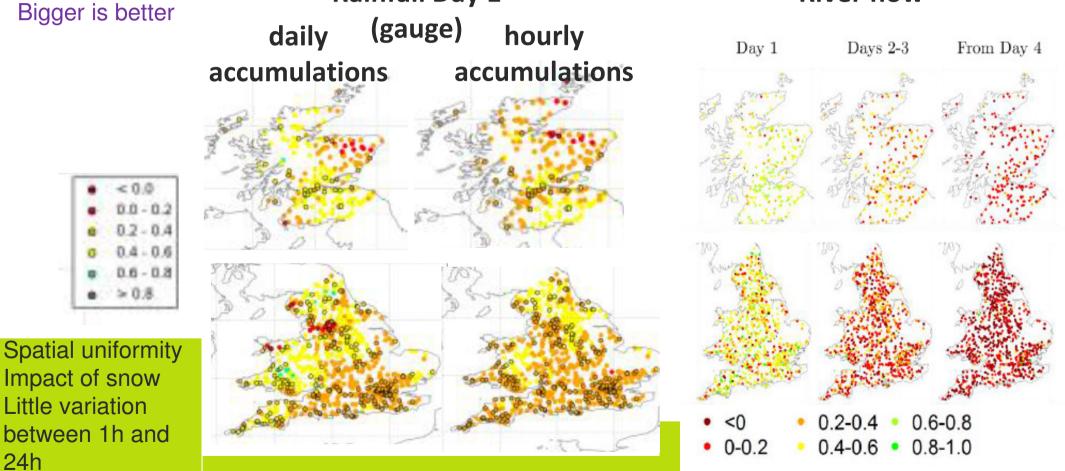
Use sample climatology

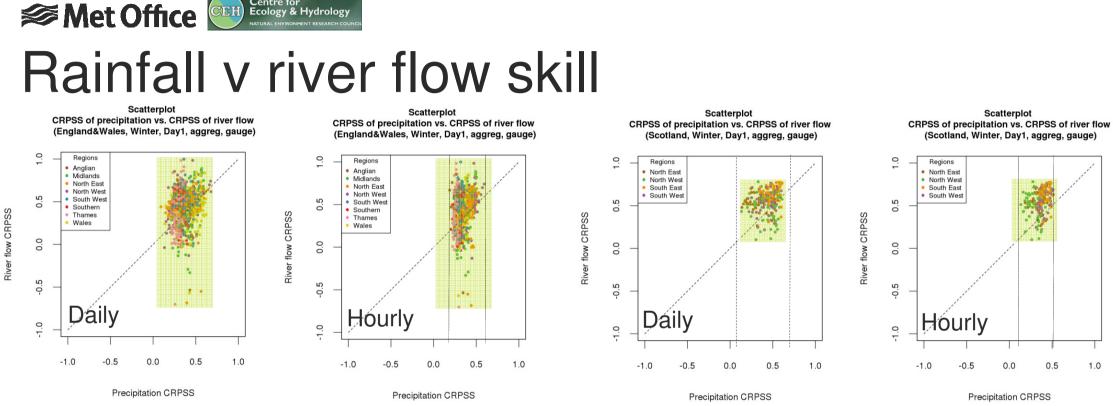
# Spatial distribution of the CRPSS

**Rainfall Day 1** 

Spatially noisy. Large decrease in skill from day 1 Very little skill by day 4

#### **River flow**





- · Larger spread in river flow scores, but a greater proportion of river flow CRPSS are higher than the rainfall CRPSS for the same catchment.
- Strong regional dependence/clustering.

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•Reduction in spread of rainfall scores for hourly accumulations.



# Conclusions

- **Sampling uncertainty** is the major consideration to obtain robust, meaningful verification information for ensembles in particular.
- **Observation uncertainty** can have a noticeable impact on the verification, especially on the rank histogram and the evaluation of ensemble spread.
- Probabilistic forecasts derived from both river flow and precipitation accumulation ensembles tended to be over-confident, with over-confidence increasing with forecast probability, threshold and lead-time. This is true particularly for river flow forecasts.
- The river flow ensemble was found to be severely under-spread according to the Rank Histogram. This suggests that unaccounted-for uncertainties in the hydrological modelling process may be important for forecast accuracy.
- Both river flow and precipitation ensembles showed good potential skill.
- Threshold-based verification scores were found to be regionally dependant.
- Daily and hourly precipitation accumulations lead to similar overall conclusions





# Questions?

Work commissioned by: Flood Forecasting Centre (EA) Scottish Flood Forecasting Service (SEPA)

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