Development of a hydrometeor classifier for the Met Office Met Office C-band weather radar network Steven Best¹, Dawn Harrison¹, Ben Pickering^{2,} Ryan Neely²

1. Introduction

The UK Met Office has been trialling a version of the hydrometeor classifier described by Al-Sakka et al. (2013) to process real-time data from the UK C-band weather radar network. The trial has initially proven successful, and work is now underway to develop and improve the algorithm. Here we describe developmental changes made to the algorithm, and present case studies to illustrate the resulting gains in performance.

2. The "Baseline" Algorithm

The classifier described by Al-Sakka et al. (2013), developed at Metéo France, is a Fuzzy Logic-type classifier. The membership functions (MFs) therein are defined for the seven hydrometeor types (rain, wet snow, dry snow, ice, small hail, medium hail, large hail) and for each of the radar / environmental variables considered (Z_H , Z_{DR} , K_{DP} , ρ_{HV} , and T (temperature)). The MFs are derived from datasets built from many individual hand-picked radar observations of 'unambiguous' hydrometeor types. MFs of both onedimensional trapezoidal (F(x)), and two-dimensional Gaussian (F(x, y)) form are used. The inference rule used to combine the individual membership values has a combined additive/multiplicative form:

$P_{i} = F_{i}(Z_{H}) * F_{i}(T) * F_{i}(BB) * [F_{i}(Z_{H}, Z_{DR}) + F_{i}(Z_{H}, K_{DP}) +$ $F_{i}(Z_{H}, \rho_{HV})],$

Here, P_i is the overall probability of occurrence of hydrometeor type *i*, and $F_i(\ldots)$ is the 'membership' of that type based on the just given input parameter(s). $F_i(BB)$ acts as a 'bright band' MF, and is used as a 'safety net' to exclude certain hydrometeor types within a certain region of the 0°C isotherm (which is provided by NWP data from the Met Office unified model).

3. Case Study Example

In the example to the right, the Metéo France classifier (described above) was used to process radar data collected during a storm system which swept across northern England on the 1st July 2015. Figure 1 (a) shows the number of counts of hail registered by the classifier during the storm period, while Fig 1 (b) shows the ground reports of hail - from both traditional observations and social media reports – recorded during the same period. The correlation between the two sets of results is encouraging.

As part of ongoing developmental work, some simple changes have been made to the Metéo France algorithm with the aim of further improving its performance:

 $P_i = F_i(T) * [F_i(Z_H, Z_{DR}) * F_i(Z_H, \rho_{HV})]$

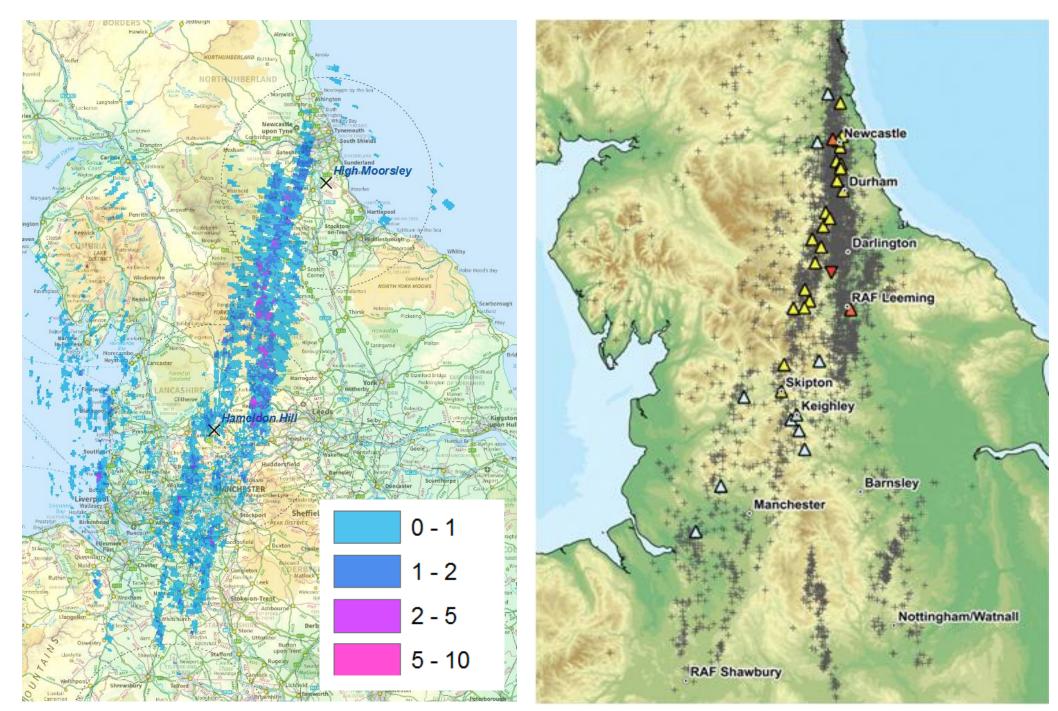


Figure 1 a) Hail classification count using 0.5° elevation scan radar data from Hameldon Hill radar. UK

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4. Algorithm Developments

• Use of the K_{DP} membership function (MF) has been omitted. As K_{DP} is not a raw radar variable, and its calculation method can produce noisy / unreliable results, the classifier is likely to perform better without it.

• The Z_H dimensions of the two-dimensional MFs have been premultiplied by the one-dimensional trapezoidal $Z_{\mu}MF$ – this is a fairly ineffectual change, but simplifies the inference rule slightly. • The 'bright band' MF has been omitted. This change improves the appearance of wet snow classifications, which previously showed sharp-looking, regular boundaries.

• The one-dimensional trapezoidal temperature MF has been replaced with a one-dimensional beta function, as described by Dolan et al. (2013). This produces smoother, more naturallooking boundaries between regions of hydrometeors at increasing elevations.

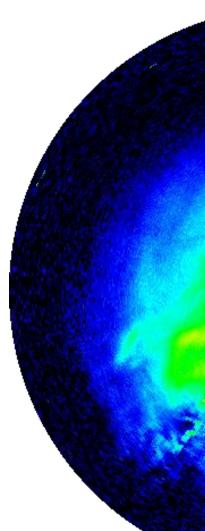
• A convection-permitting exception is activated when a reflectivity of 30 dBZ or more is detected 1 km or more above the 0 °C isotherm. i.e., under these circumstances, classification of rain is not suppressed above the melting layer.

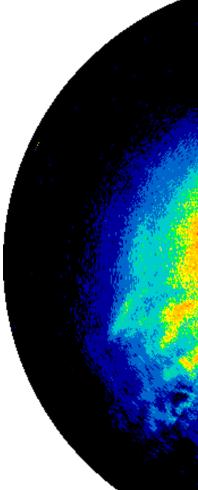
• The three hail classes have been combined into one. In the UK, where the occurrence of hail is relatively rare, the definition of three different sizes of hail is not necessary.

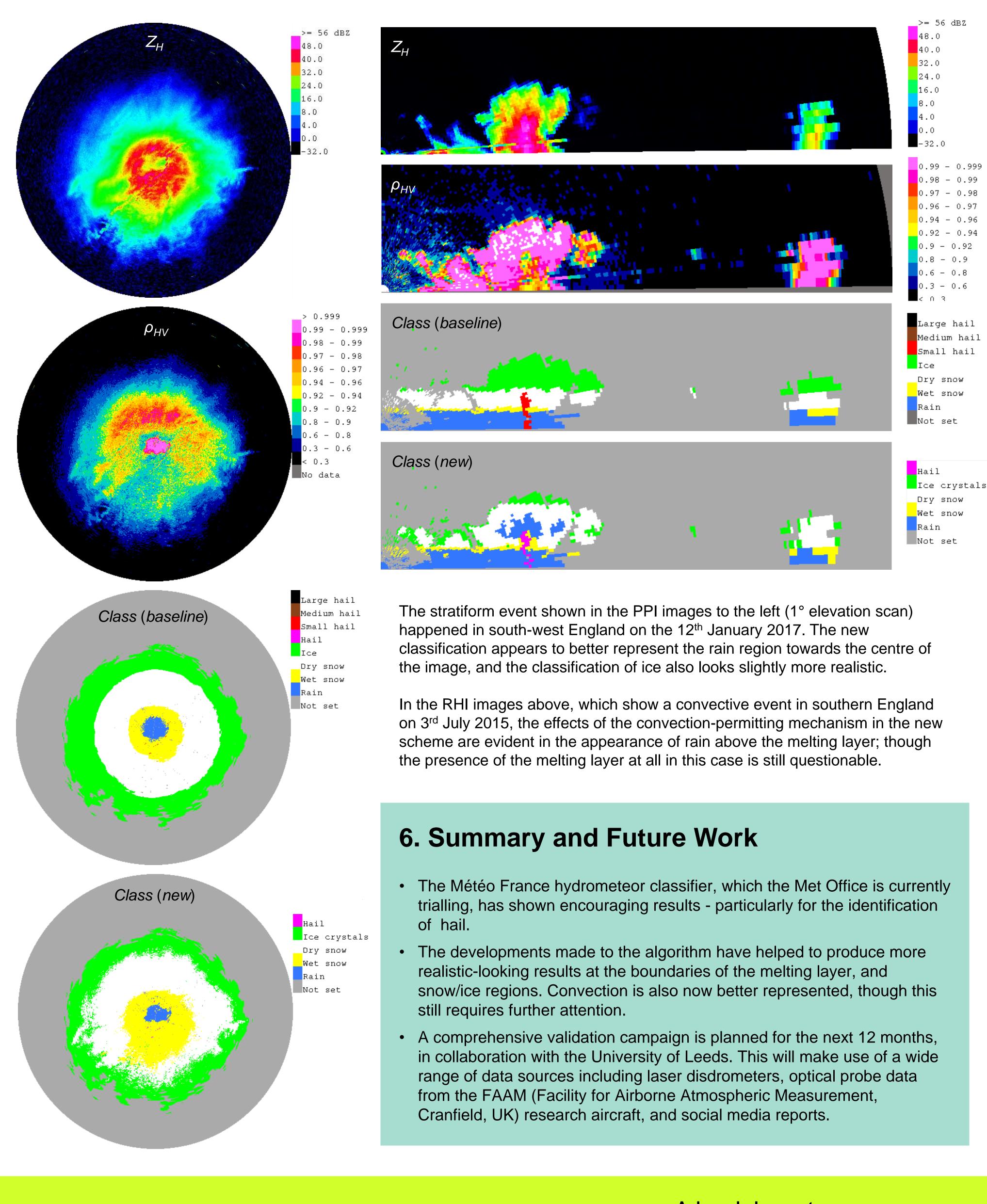
• The two-dimensional $F_i(Z_H, \rho_{HV})$ MF is used multiplicatively, instead of additively. This improves the appearance of the rain / wet snow boundary below the melting layer, at the cost of reducing the classification confidence to zero in regions where ρ_{HV} is affected by reduced signal-to-noise ratio. • The new inference rule thus looks as follows:

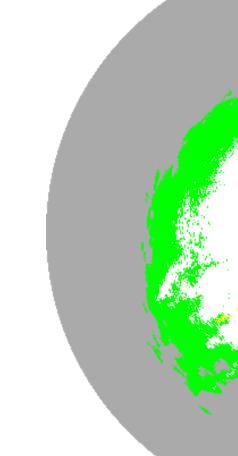
b) Hail (Δ) and lightning (+) reports for 01 July 2015 (from Lewis and Silkstone (2017))

5. Results: Classifications of data from a stratiform event (left) and a convective event (right)









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References:

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