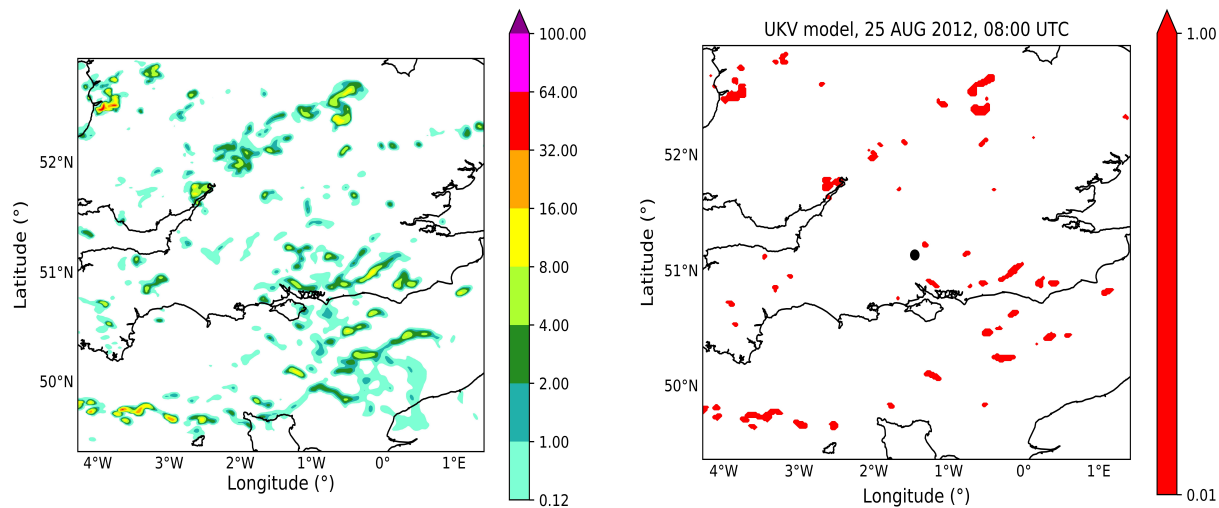


# Beyond $I_{\text{Org}}$ : Advertisement for a suite of more robust organization measures

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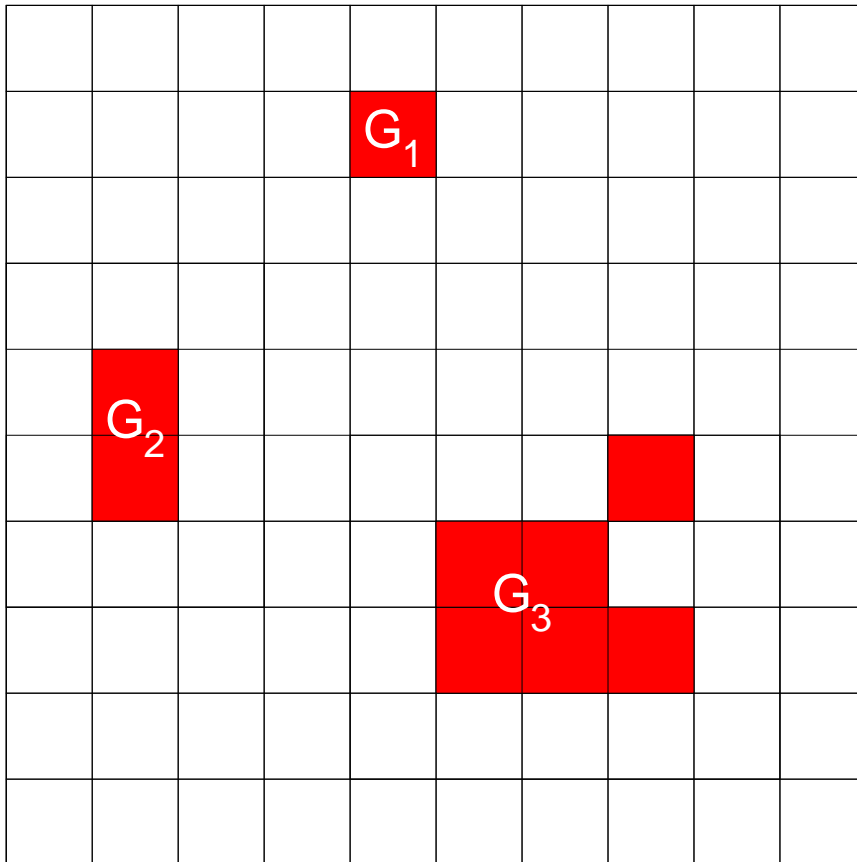
# Measuring Organization



- By **organization**, we will mean **an arrangement of points that can be distinguished from the action of a Poisson process**
- i.e., something other than scattering  $N$  points randomly within the area  $A$
- Fortran code to calculate suite of metrics that test this and characterise the organization with options for
  - cloud definition
  - fixed or singly-periodic or bi-periodic bc's
  - corrections for finite size effects
  - corrections for finite domain effects



# Reduction to a set of points



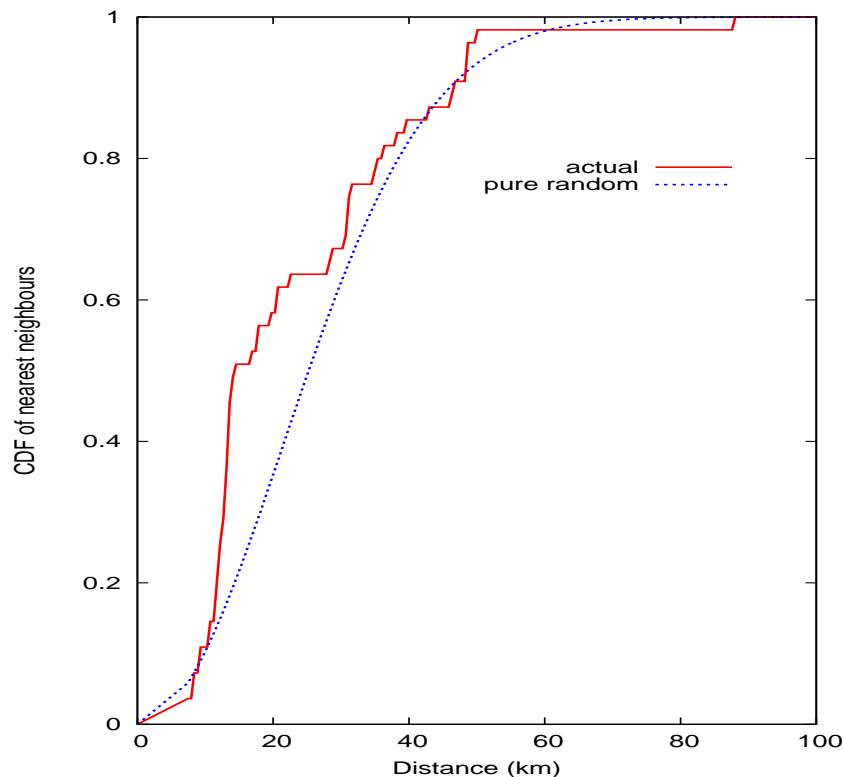
- Apply threshold
- Group the points into cloud objects
- Define a central point

$$\mathbf{x} = \frac{\sum_{i=1}^P \mathbf{x}_i}{P}$$

- Or: apply watershed algorithm (based on local maxima)



# $I_{org}$ (Tompkins and Semie 2017)



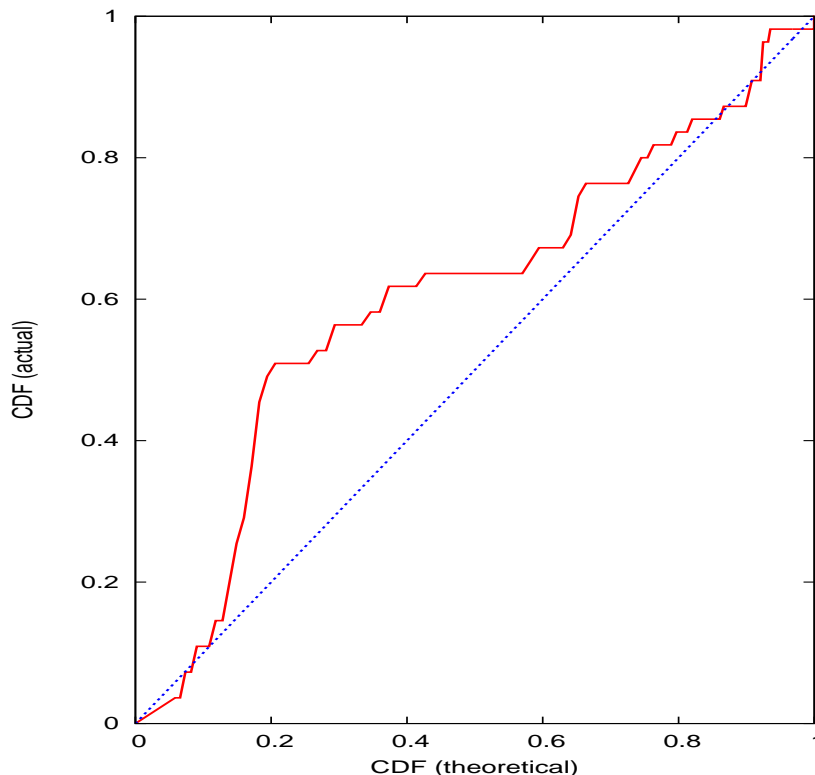
- From each cloud point find distance to its nearest neighbour
- Construct a **CDF**
- Compare to distribution for a **random pattern**

$$1 - \exp(-\lambda\pi r^2)$$

where  $\lambda = N/A$



# $I_{\text{org}}$ (Tompkins and Semie 2017)



- Plot actual ( $A$ ) vs theoretical ( $T$ ) CDF

$$I_{\text{org}} = \int_0^1 A(T) dT$$

- Random if  $I_{\text{org}} = 0.5$  (for  $A = T$ )
- Clustered if  $I_{\text{org}} > 0.5$
- Regular if  $I_{\text{org}} < 0.5$
- In our example,  $I_{\text{org}} = 0.604$



# Practical Issues



- Formula assumes points in infinite domain, not extended objects on finite domain
- Consider a Monte-Carlo approach
  - Relocate the clouds randomly in the area
  - Redraw until they all fit without touching
  - Calculate separation distances
  - Repeat many times (e.g. 1000)
- Sampling approach allows significance tests if required
- In our example, using MC reference increases  $I_{\text{org}}$  from 0.604 to 0.646



# Boundary effects



- $I_{\text{org}}$  now well defined and 0.5 is meaningful
- But for a true best-estimate of the CDF on a finite domain, we must correct for clouds close to fixed boundary edges
- Let distance to nearest boundary be  $b_i$
- If  $b_i < r_i^{\text{nn}}$  there might be a nearer neighbour outside our domain

$$C(r) = \frac{\sum_i I(b_i > r > r_i^{\text{nn}})}{\sum_i I(b_i > r_i^{\text{nn}})}$$

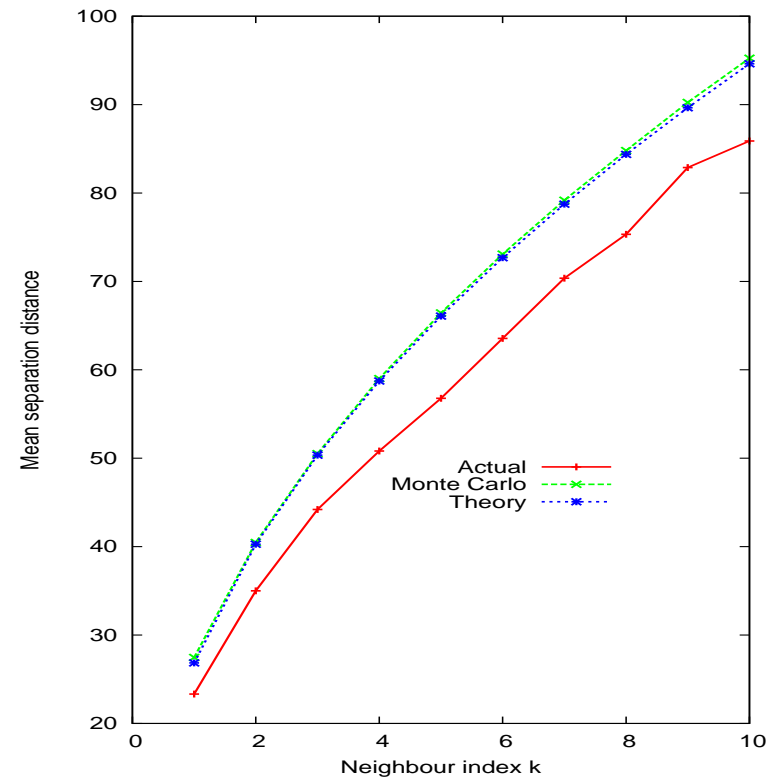
where  $I = 1$  if bracketed condition is true



# Higher order measures



- $I_{\text{org}}$  can be sensitive to cloud definition (small fragments)
- Consider  $k$ th nearest neighbour distance
- Can generalize  $I_{\text{org}}$  in a natural way, using theory and/or Monte Carlo methods with corrections
- Increasing  $k$  focuses on the pattern at larger distances





# Empty space function



- Robust to cloud fragments very near other clouds
- Take  $M$  arbitrary points and measure the distance to the nearest cloud
- The CDF is the probability that a randomly-placed disc of radius  $r$  contains at least one cloud
- The theoretical distribution is the same as the nearest-neighbour distribution
- But better to obtain a Monte Carlo reference as before



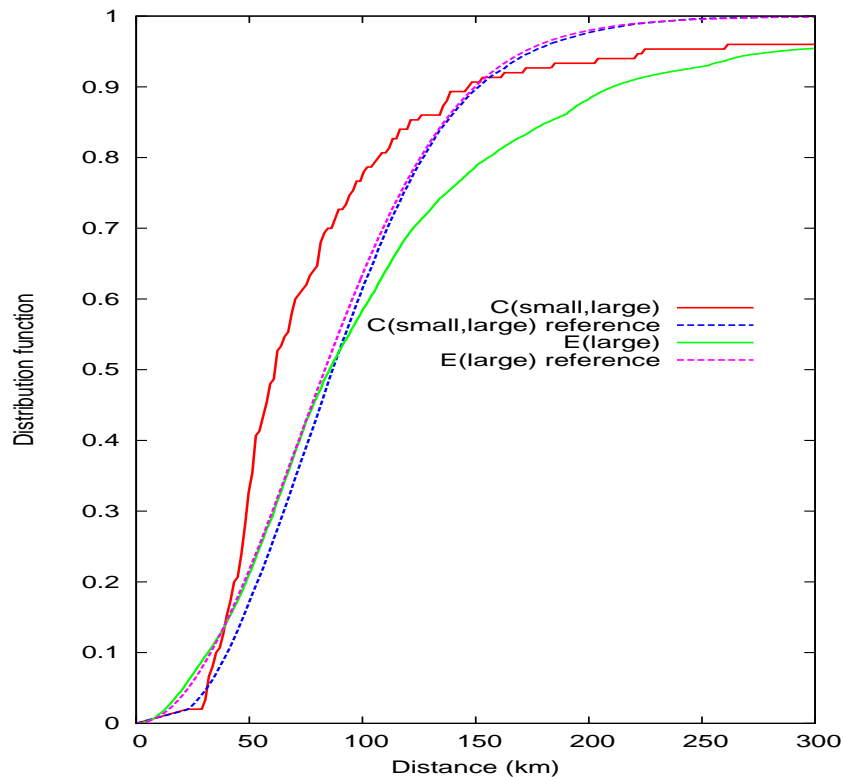
# Example of use



- Divide the clouds into two types (by area, rain rate,  $w...$ )
- We can analyse each type separately
- But we could also ask “starting from a cloud of type 1 what is the distance to the nearest type 2 cloud?”
- Let the corresponding mixed CDF be  $C_{12}(r)$
- Is the pattern of the type 1 clouds related to the pattern of the type 2 clouds?
- If the two types are independent then  $C_{12}(r) = E_2(r)$



# An RCEmip example



- RCEmip snapshot split into **smaller (2/3 of the total)** and **larger** clouds
- $C_{12} \neq E_2 \neq$  **reference**
- Interpretation is that small clouds form clusters that are preferentially within the vicinity of the large ones



# Summary



- A growing number of studies seek to measure organization
- Often using nearest-neighbour distances
- Finite cloud-size effects and finite domain effects can be important which means that a Monte-Carlo approach is needed for the unorganized reference process
- Other related measures are valuable to cross-check and enhance our interpretations (e.g. empty space functions more robust)
- **Happy to share code if you want to explore**

