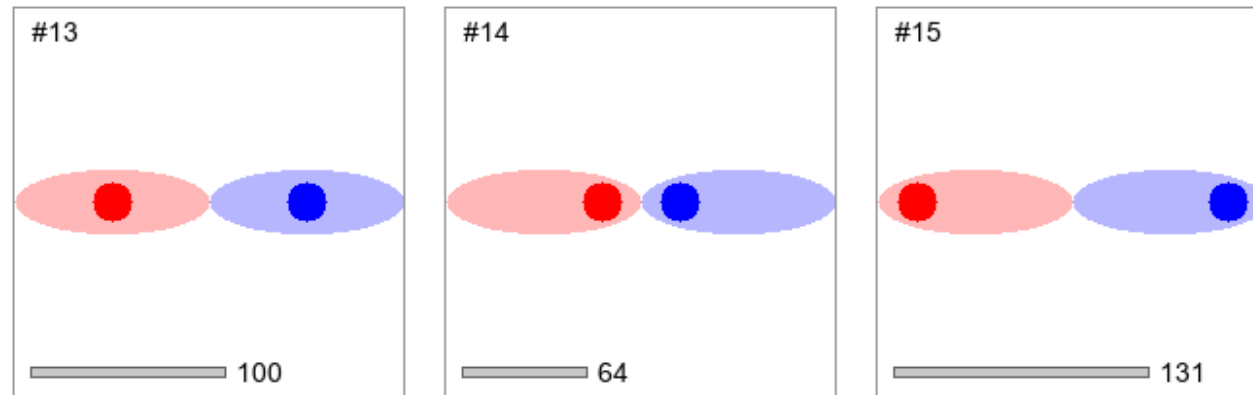


A new spatial displacement measure for continuous fields



Gregor Skok

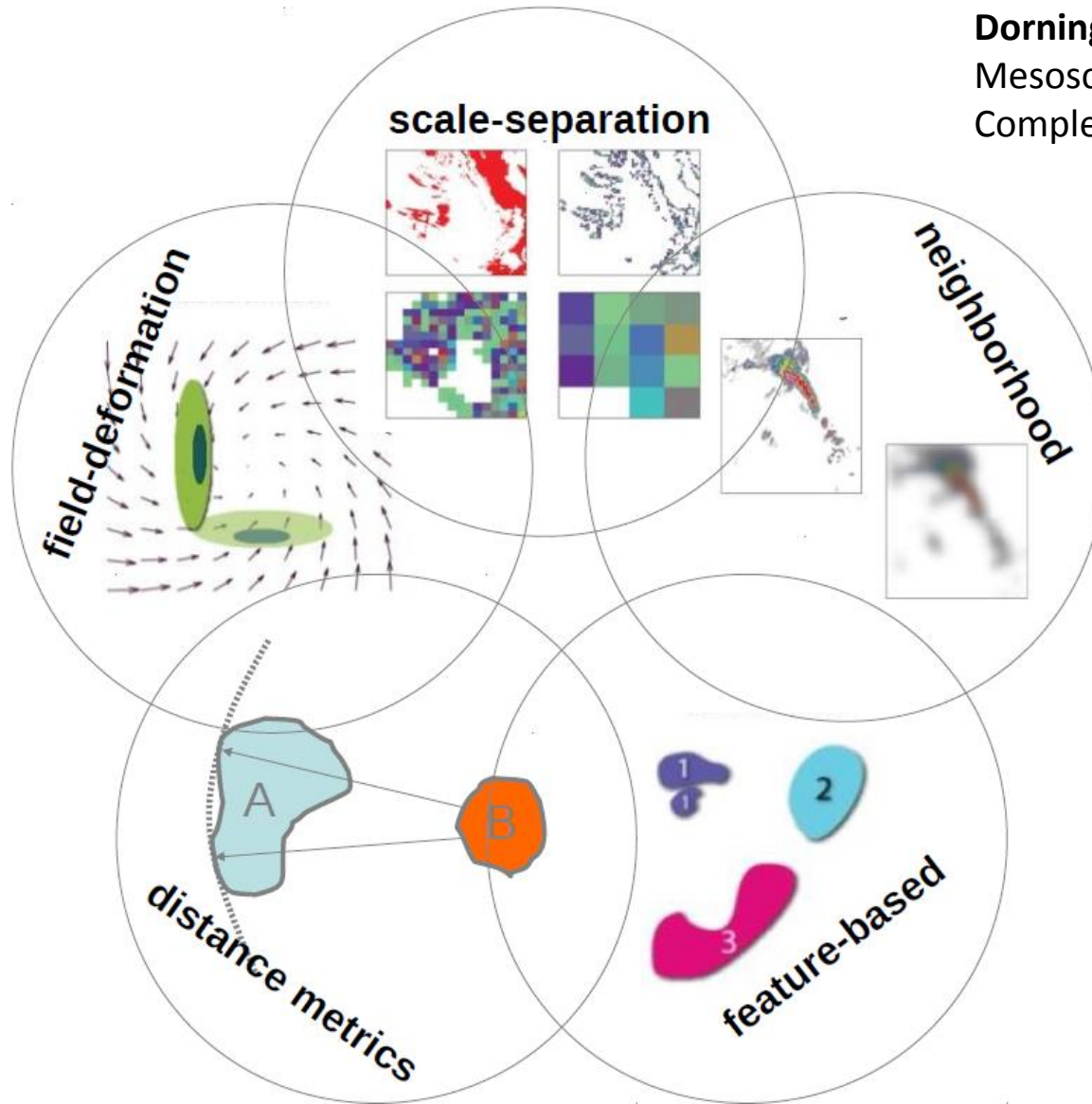
Faculty of Mathematics and Physics, University of Ljubljana, Ljubljana, Slovenia





Classes of spatial verification methods

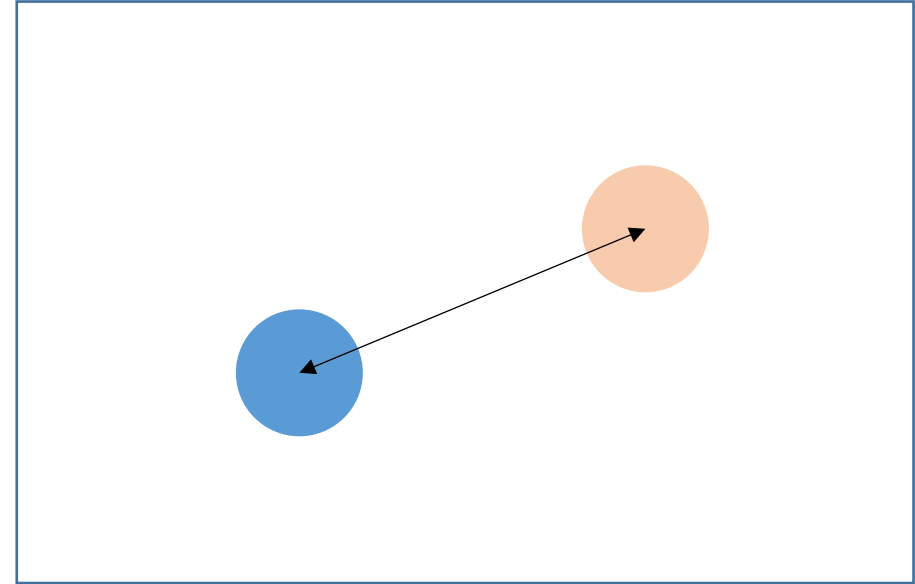
Dorninger et. al, 2018: The set-up of the Mesoscale Verification Inter-Comparison over Complex Terrain (MesoVICT) Project. BAMS.





Distance measures

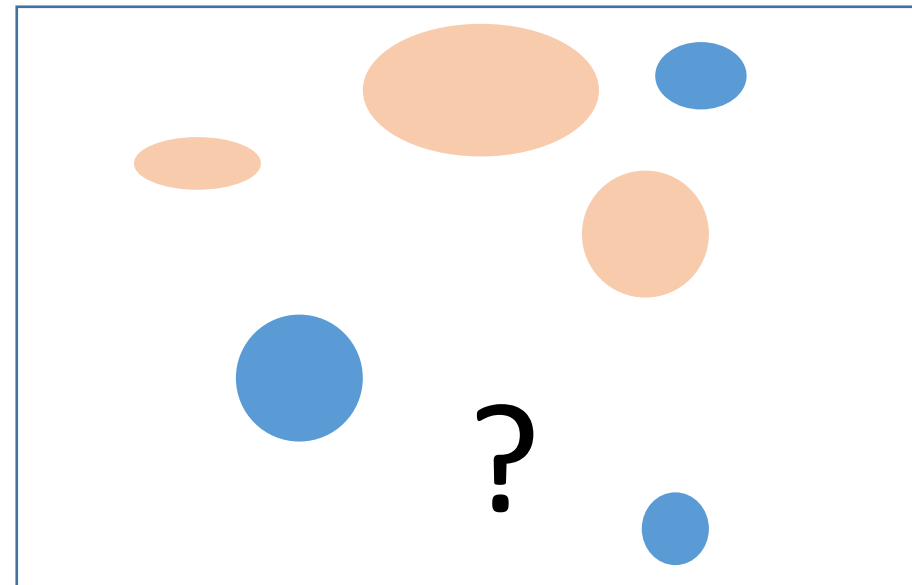
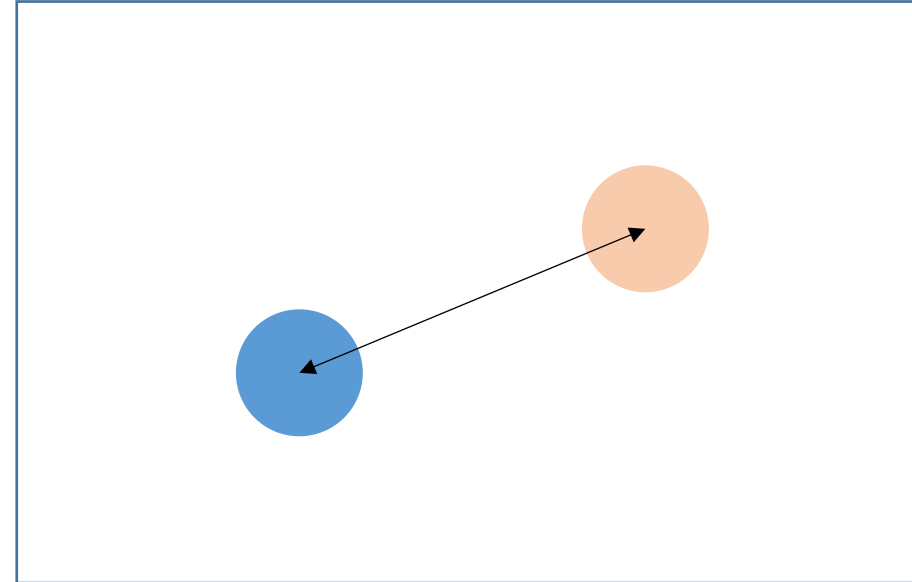
- The result is provided in terms of **spatial distance**
- Some methods try to provide an estimate of **spatial displacement** - **appealing for forecast interpretation**





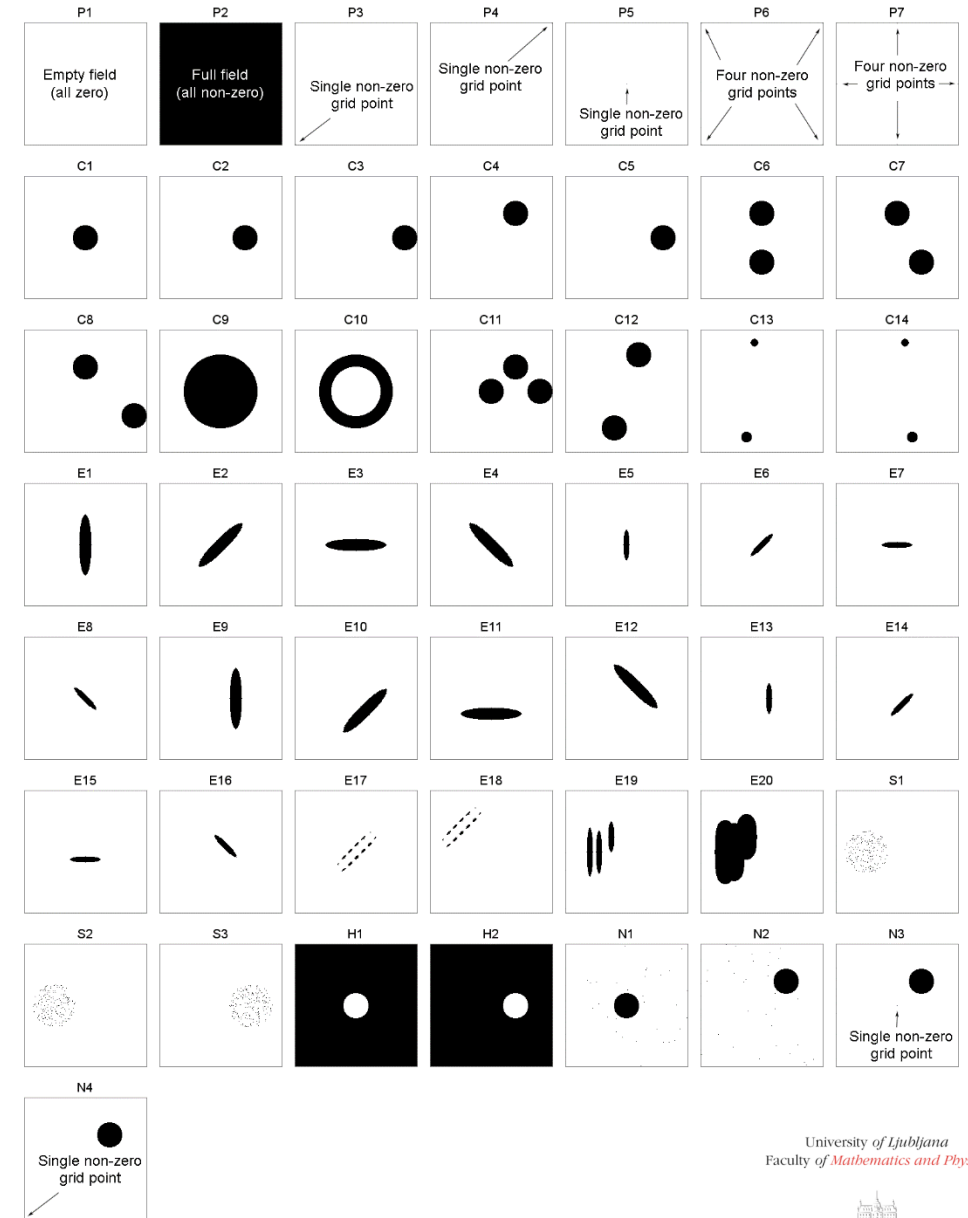
Distance measures

- The result is provided in terms of **spatial distance**
- Some methods try to provide an estimate of **spatial displacement** - **appealing for forecast interpretation**
- It is **not always clear** what the correct spatial displacement should be
- It is important to **evaluate** and **compare results** of different measures



Gilleland et al., 2020, MWR paper

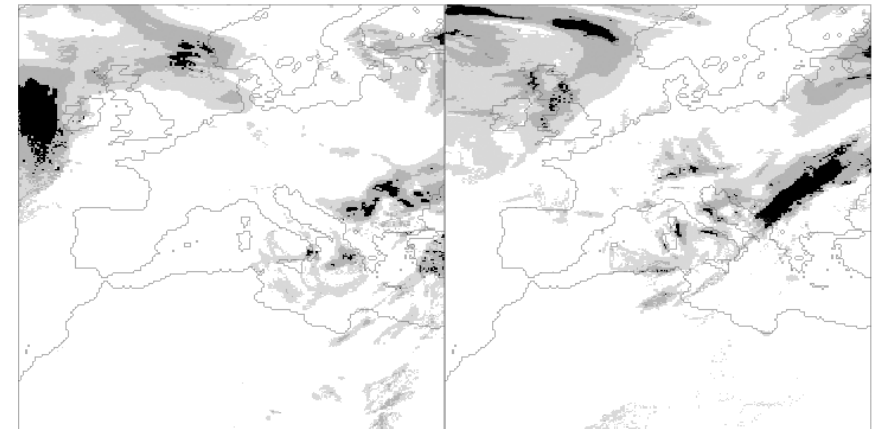
- Intercomparison of **6 distance measures**
- A set of **50 idealized binary fields**
- Gain **basic understanding** of how the **measures behave**





Motivation

- All the distance measures tested by Gilleland et al., 2020, **require binary fields** as input.
- These are usually obtained via **thresholding**:
 - thresholding **removes** some **information**
 - the results are **sensitive** on the choice of threshold
 - the problem of **too low** or **too high** thresholds
 - all values above the threshold are **treated equally**



Goal

- To design a score that uses similar concepts as the **Fraction Skill Score distance (dFSS)**,
- but would be able to **analyse continuous fields**, and thus avoid problems related to thresholding.



The dNSS measure

- Input fields are **smoothed** using a **square neighbourhood** (n)
- A new score: Neighbourhood Skill Score (NSS)

$$NSS(n) = 1 - \frac{\sum_{i,j} |f_A(n)(i,j) - f_B(n)(i,j)|}{\sum_{i,j} f_A(n)(i,j) + \sum_{i,j} f_B(n)(i,j)}$$

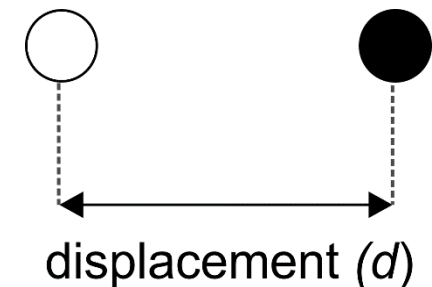
values between **0 (worst)**
and **1 (best)**

- A simple **idealized setup** with a **single displaced event**
- The displacement can be estimated as

Neighbourhood Skill
Score displacement

$$dNSS = \frac{n'}{2}$$

neighbourhood size where $NSS=0.5$





The dNSS measure

- Although this result is based on a **very simple idealized setup** it works quite well **also in other situations**
- The fields **need to be unbiased** prior to analysis. A **special adjustment** needs to be made if events are **overlapping**.

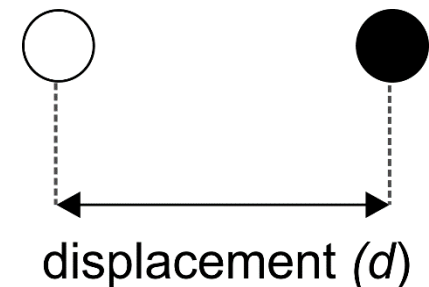
between 0 (worst) and 1 (best)

$$\Delta_{i,j} = \Delta_{i,j}^A(\theta) - \Delta_{i,j}^B(\theta)$$

- A simple idealized setup with a **single displaced event**
- The displacement can be estimated as

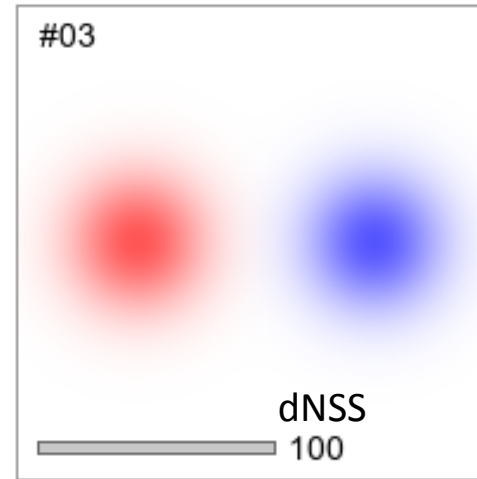
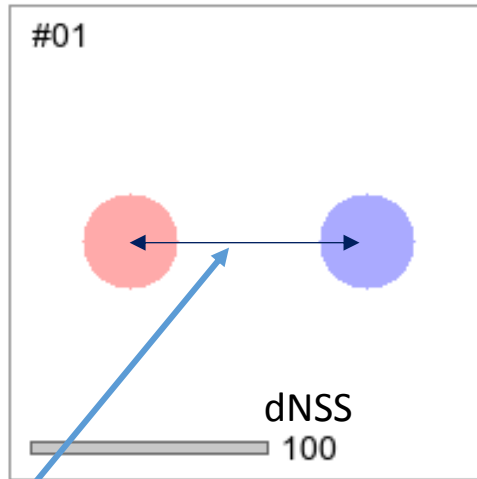
$$dNSS = \frac{n'}{2}$$

neighbourhood size where NSS=0.5



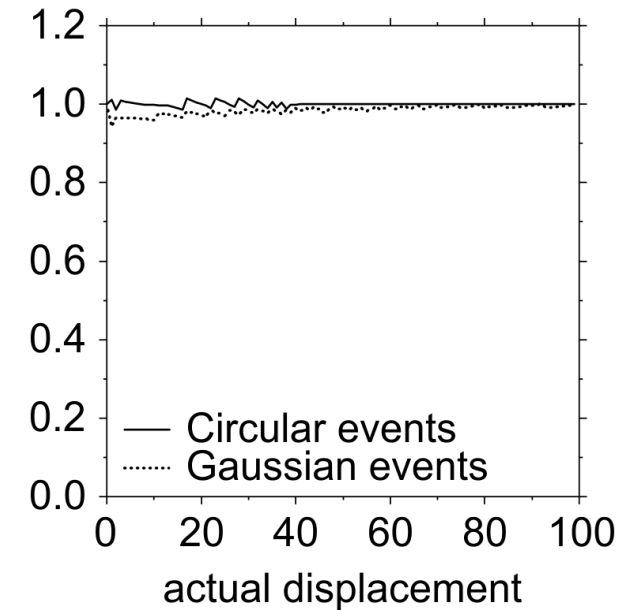


Binary vs. continuous events



Actual displacement
is 100 points

ratio dNSS/(actual displacement)



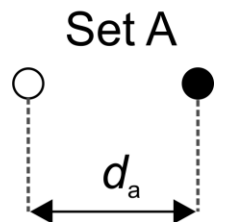
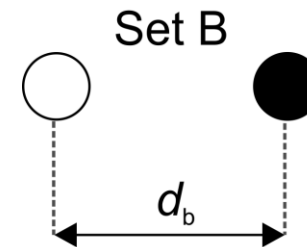


Effect of magnitude of events

- **Magnitude**: the **sum of values** inside an event
- **two sets of events** with different magnitude

$$dN_{SS} = \frac{d_a + x \cdot d_b}{1 + x}$$

- x = ratio between magnitudes of both sets
- **weighted average** of both displacements
- the dependence on magnitude **is linear**

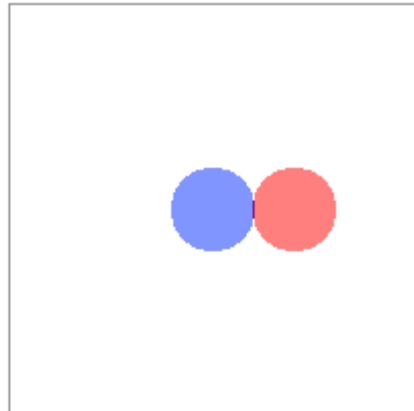




Idealized examples

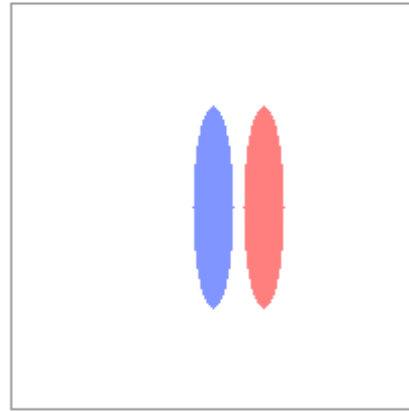
Geometric cases from Gilleland et al., 2020, MWR

C1C2



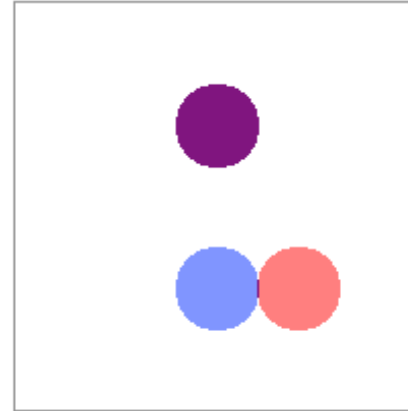
CDST=40
BDEL=29
H=40
MED=22
rMED=22
ZHU=37
rZHU=37
dFSS=34
dNSS=40

E1E9



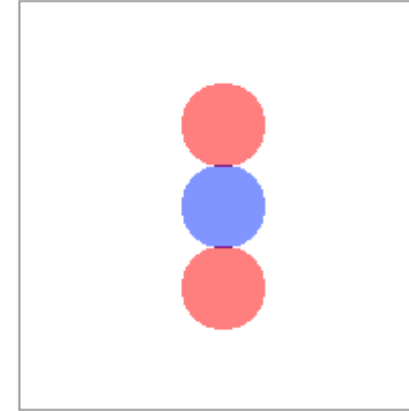
CDST=25
BDEL=23
H=25
MED=17
rMED=17
ZHU=36
rZHU=36
dFSS=22
dNSS=25

C6C7



CDST=20
BDEL=22
H=40
MED=11
rMED=11
ZHU=31
rZHU=31
dFSS=17
dNSS=20

C1C6



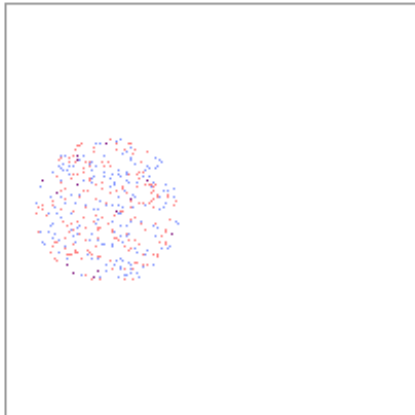
CDST=0
BDEL=24
H=40
MED=22
rMED=13
ZHU=38
rZHU=43
dFSS=#
dNSS=31



Idealized examples

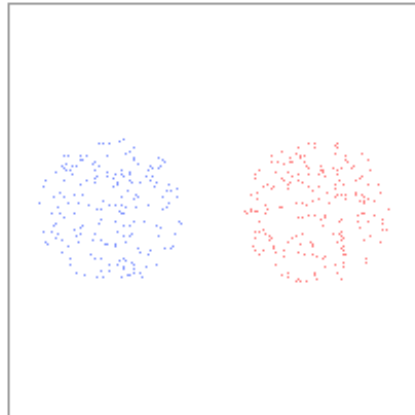
Geometric cases from Gilleland et al., 2020, MWR

S1S2



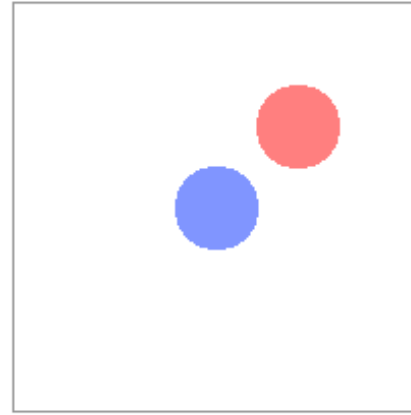
| CDST=1.3
 | BDEL=1.9
 ■ H=9.1
 | MED=2.4
 | rMED=2.6
 ■ ZHU=11
 ■ rZHU=10
 | dFSS=2.8
 | dNSS=2.8

S1S3



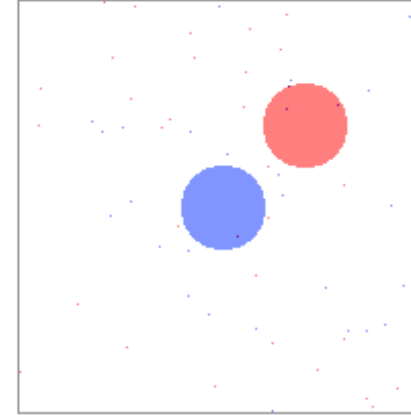
■ CDST=99
 ■ BDEL=63
 ■ H=104
 ■ MED=70
 ■ rMED=71
 ■ ZHU=45
 ■ rZHU=45
 ■ dFSS=80
 ■ dNSS=83

C1C4



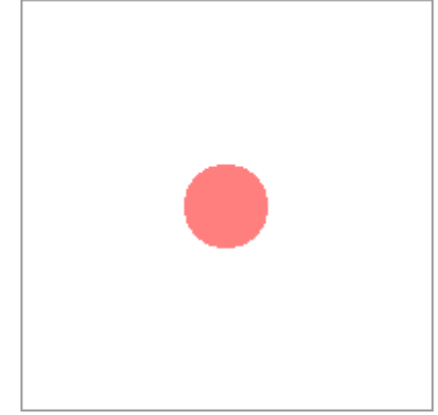
■ CDST=57
 ■ BDEL=41
 ■ H=57
 ■ MED=38
 ■ rMED=38
 ■ ZHU=45
 ■ rZHU=45
 ■ dFSS=54
 ■ dNSS=63

N1N2



■ CDST=55
 ■ BDEL=22
 ■ H=94
 ■ MED=13
 ■ rMED=15
 ■ ZHU=34
 ■ rZHU=32
 ■ dFSS=53
 ■ dNSS=62

P1C1

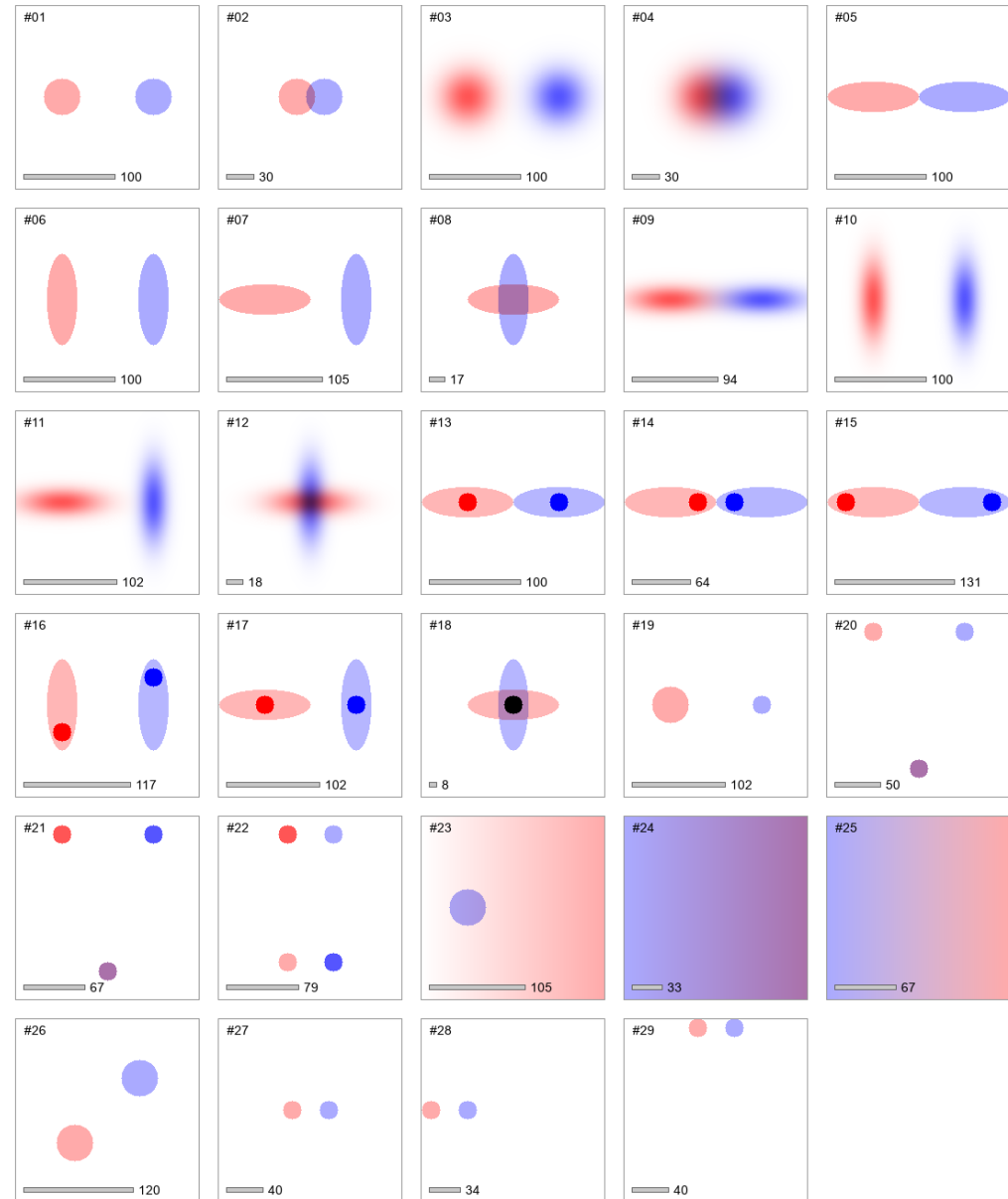


CDST=#
 ■ BDEL=342
 ■ H=400
 ■ MED=400
 rMED=#
 ■ ZHU=118
 ■ rZHU=118
 dFSS=#
 dNSS=#



Idealized examples

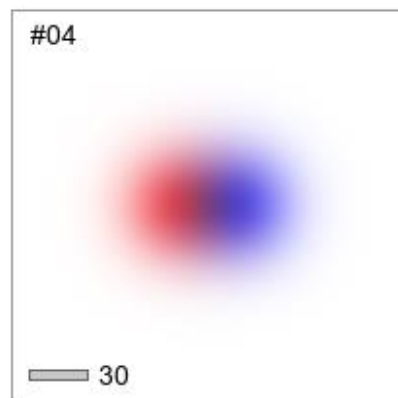
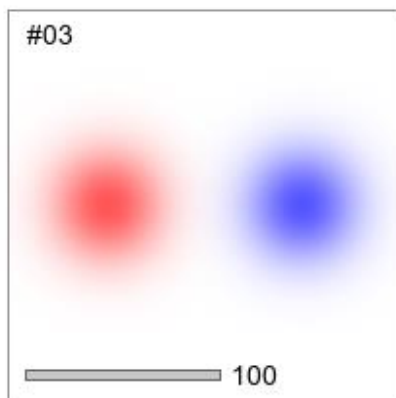
New geometric cases





Idealized examples

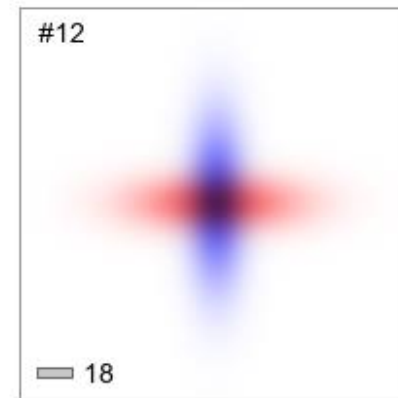
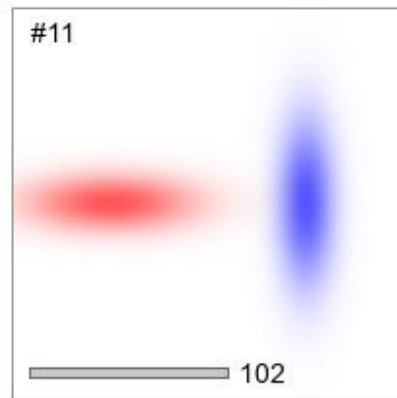
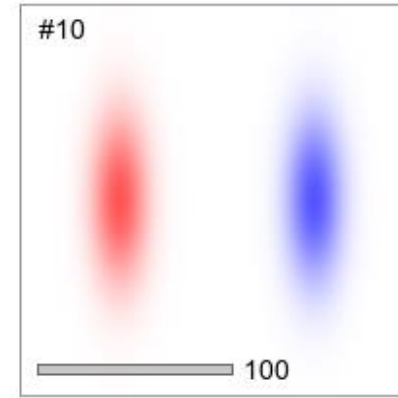
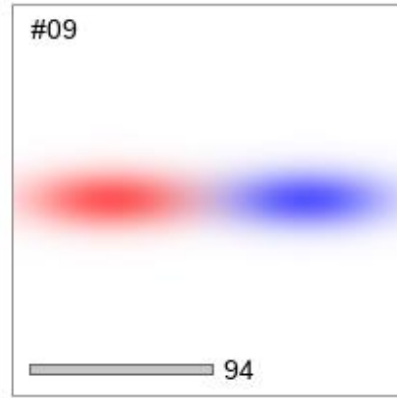
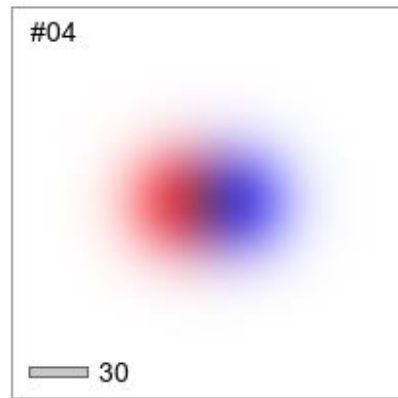
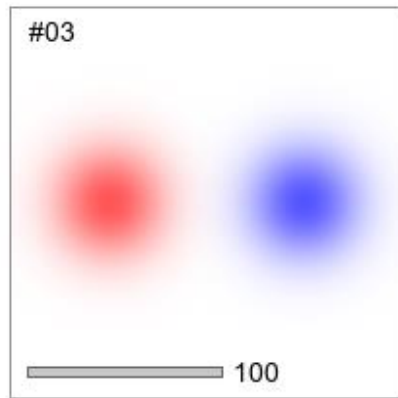
New idealized cases





Idealized examples

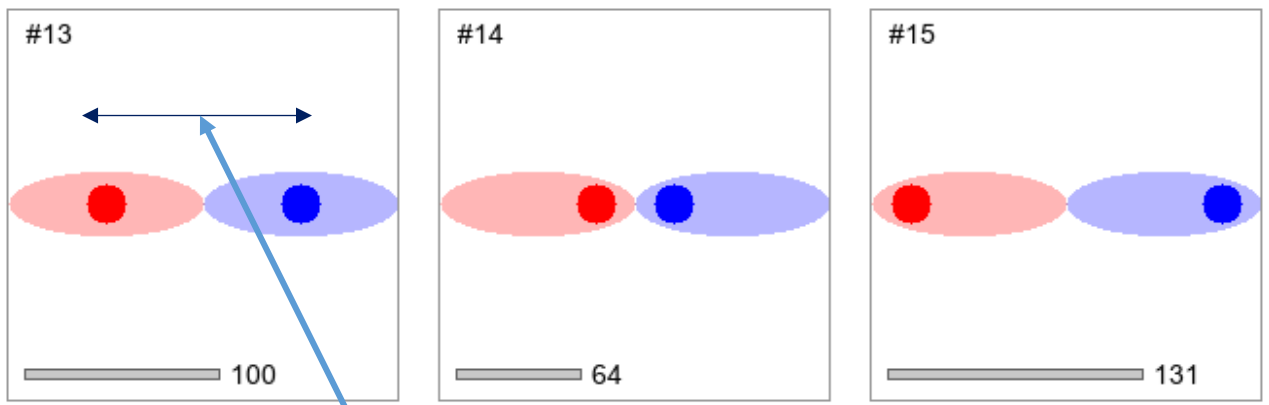
New idealized cases





Idealized examples

New idealized cases

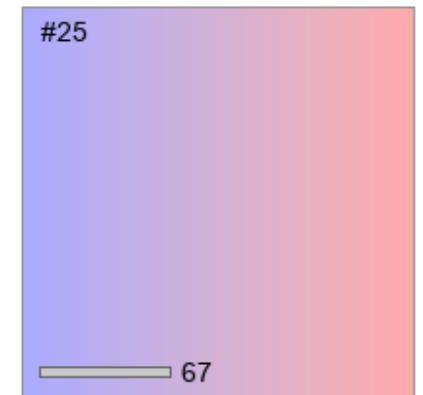
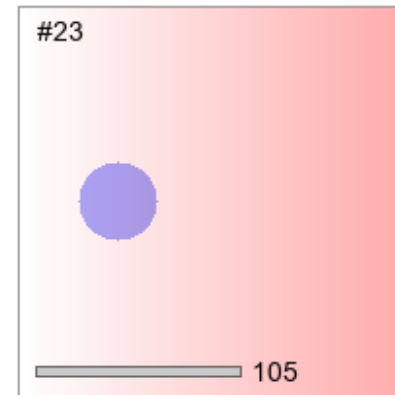
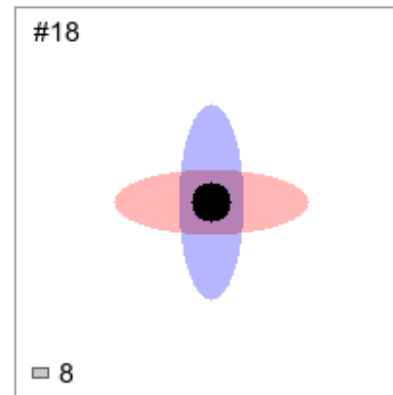
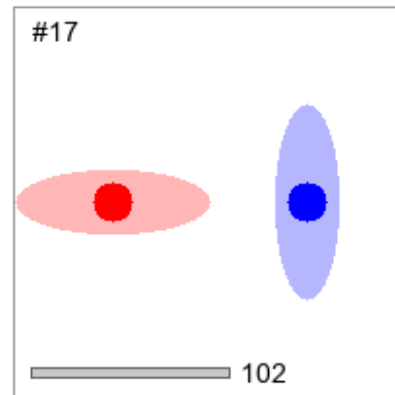
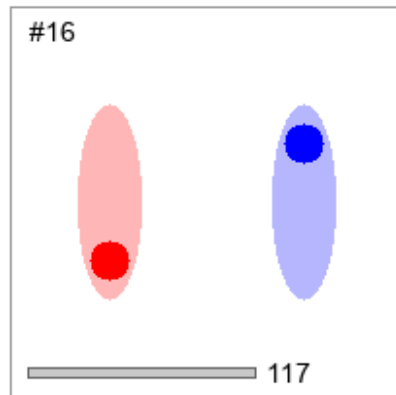
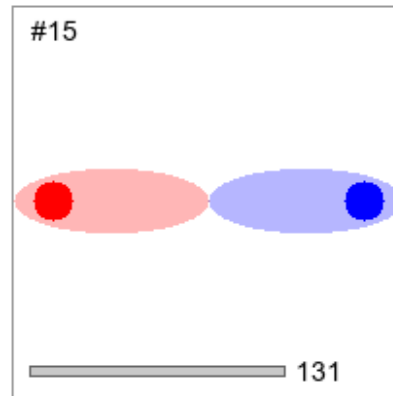
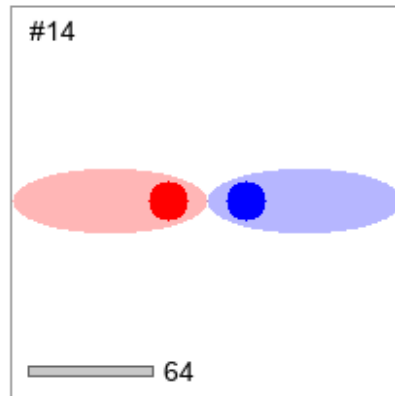
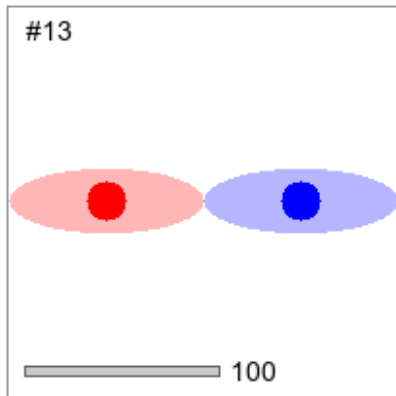


Actual displacement
is 100 points



Idealized examples

New idealized cases

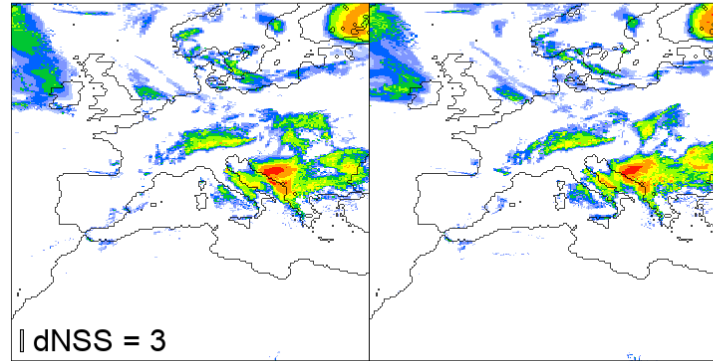




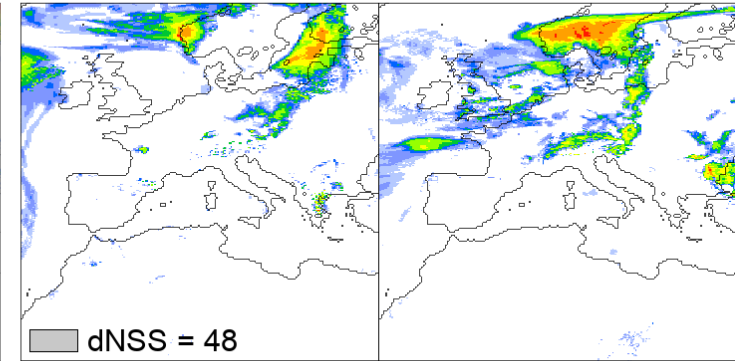
Real examples

ECMWF forecast
6-hourly precipitation
0.125° resolution

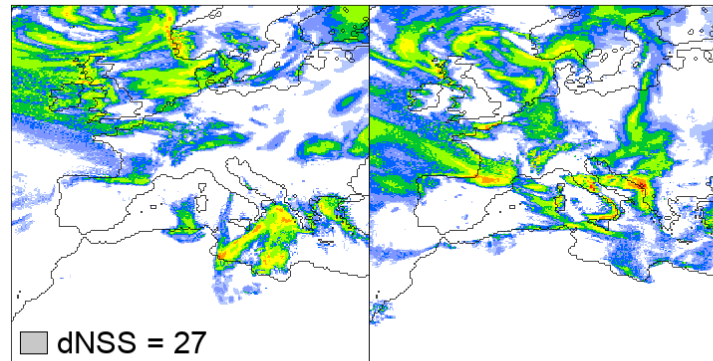
(a) 2014-05-14 1 day forecast $R = 0.95$, $Q = 0.75$



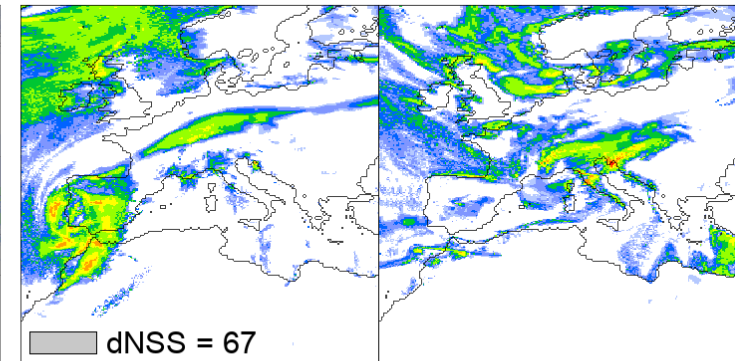
(b) 2014-06-12 9 day forecast $R = 0.62$, $Q = 0.15$



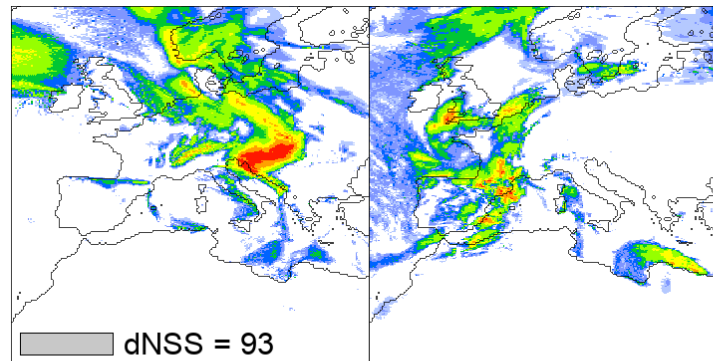
(c) 2014-12-11 9 day forecast $R = 0.97$, $Q = 0.3$



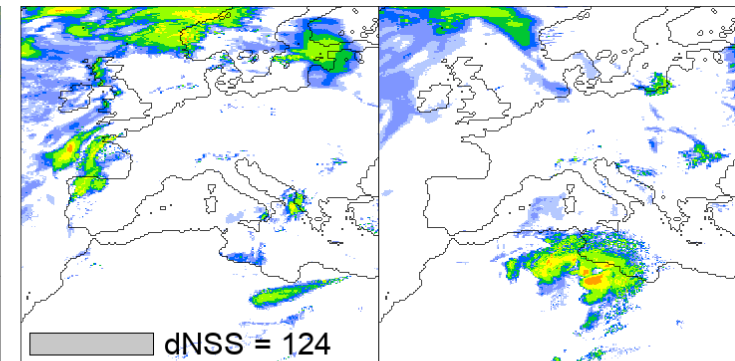
(d) 2014-12-14 9 day forecast $R = 1.33$, $Q = 0.28$



(e) 2014-10-22 9 day forecast $R = 1.59$, $Q = 0.12$



(f) 2010-04-29 9 day forecast $R = 1.27$, $Q = 0.17$

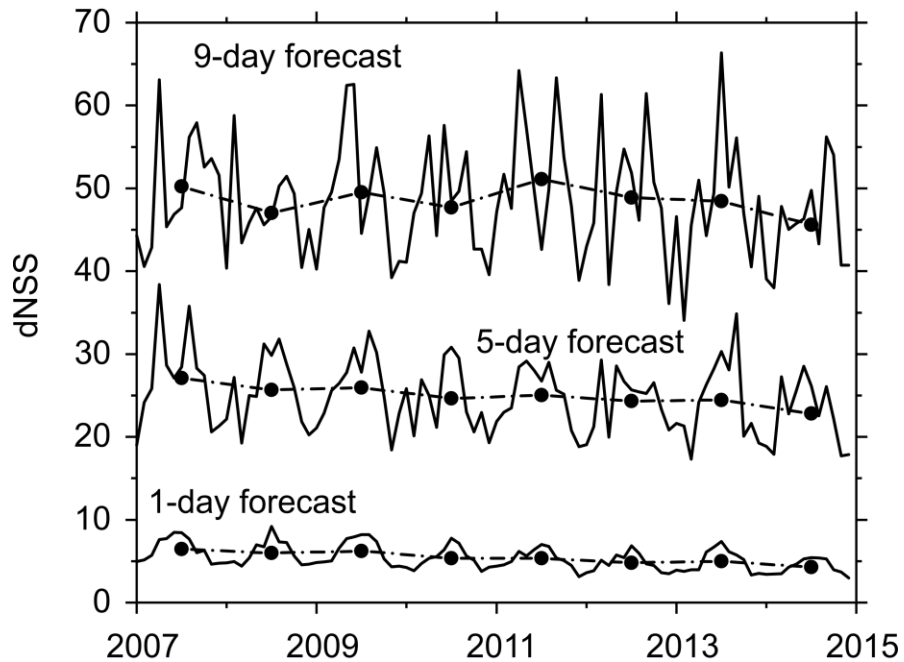




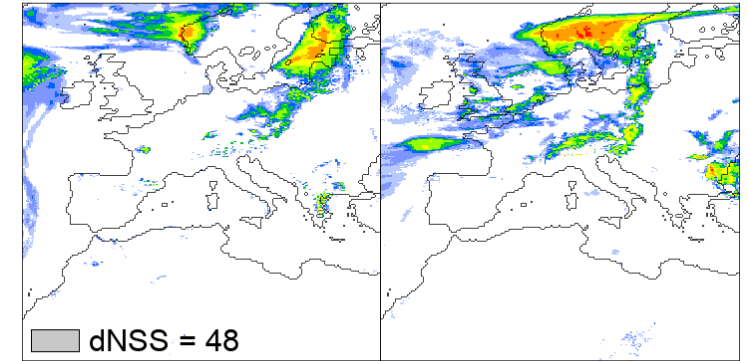
Real examples

ECMWF forecast 6-hourly precipitation

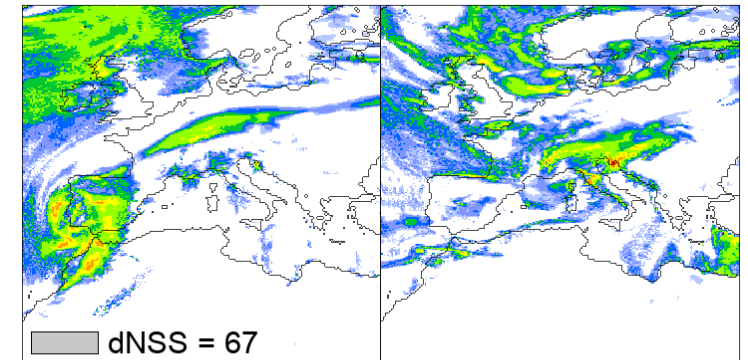
(a) dNSS of ECMWF 6-hourly precipitation



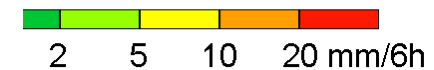
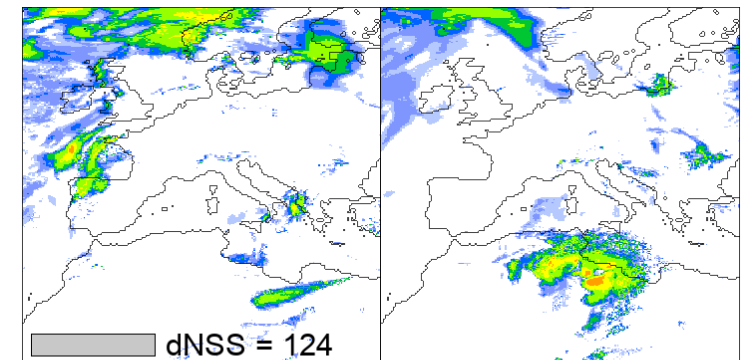
(b) 2014-06-12 9 day forecast $R = 0.62$, $Q = 0.15$



(d) 2014-12-14 9 day forecast $R = 1.33$, $Q = 0.28$



(f) 2010-04-29 9 day forecast $R = 1.27$, $Q = 0.17$

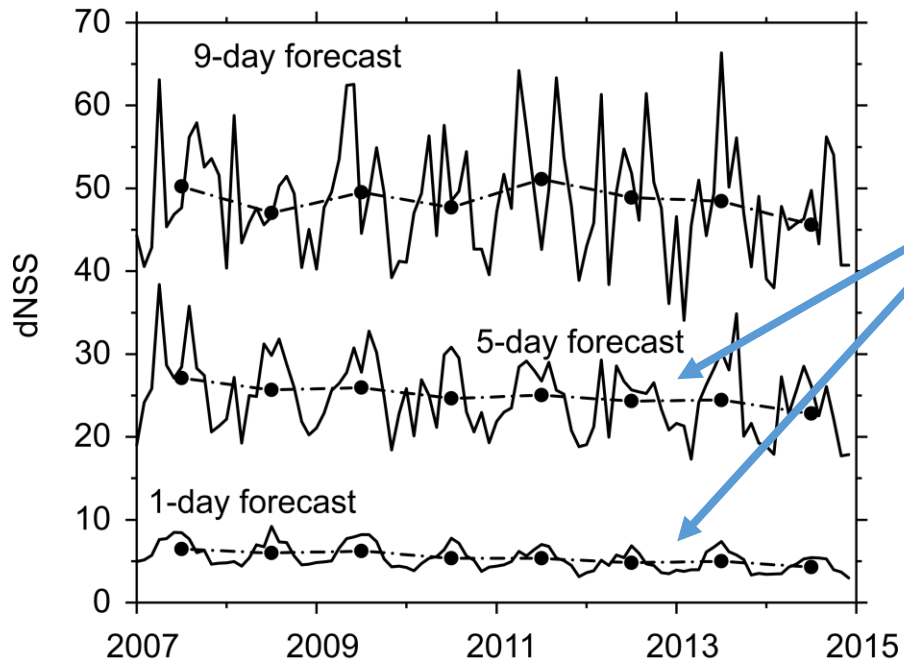




Real examples

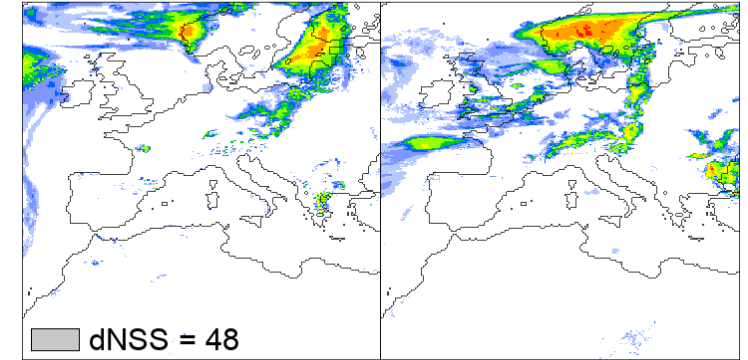
ECMWF forecast 6-hourly precipitation

(a) dNSS of ECMWF 6-hourly precipitation

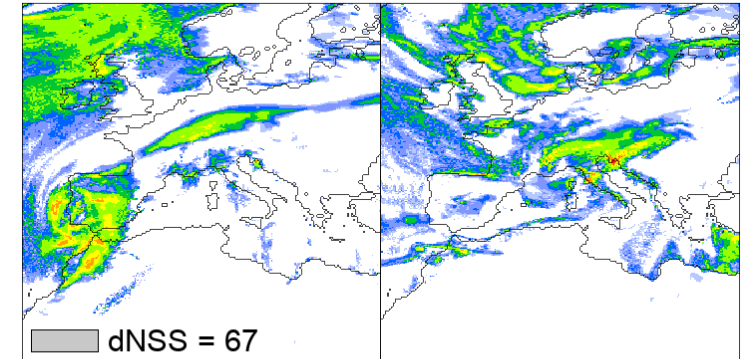


Statistically significant
decreasing trends of dNSS:
-0.49 grd.points/year, $p > 0.0026$
-0.29 grd.points/year, $p > 0.016$

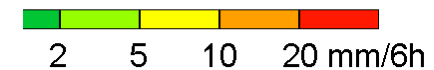
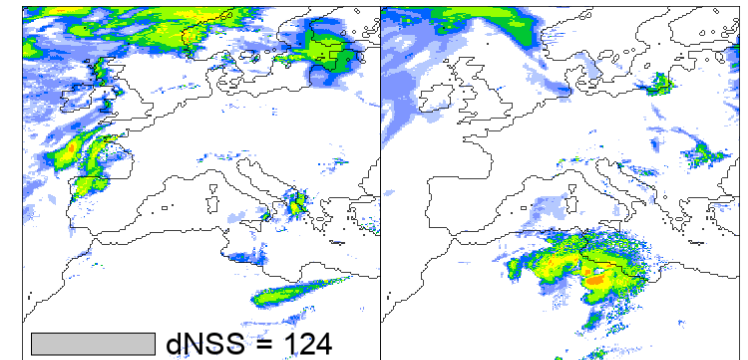
(b) 2014-06-12 9 day forecast $R = 0.62$, $Q = 0.15$



(d) 2014-12-14 9 day forecast $R = 1.33$, $Q = 0.28$



(f) 2010-04-29 9 day forecast $R = 1.27$, $Q = 0.17$





Conclusions 1/2

- A new score was devised to **avoid the problems** associated with **thresholding**
- It behaves similarly to dFSS
 - can be used to determine **spatial displacement** in a **meaningful** way
 - is not sensitive to **noise**
 - results are directly related to the **actual displacements** of events
 - events with **higher magnitude** have a larger influence on the resulting value
 - cannot be used in case of **empty fields**, is somewhat sensitive to **orientation** and closeness of the events to the **border**.



Conclusions 2/2

- dNSS has some important **advantages** compared to dFSS
 - it is not limited by **bias**
 - a more **proportional response** to the **magnitude** of events
 - can be used for direct analysis of **continuous** or **multi-level** fields
 - avoids the potential **problems** associated with **thresholding**
- Submitted to MWR
- The code will be made available in SpatialVx R-package



Thank you !



The dNSS measure

- **Smooth** the whole field using a square neighbourhood (n)
- **Define** a new score: neighbourhood skill score (NSS)

$$NSS(n) = 1 - \frac{\sum_{i,j} |f_A(n)(i,j) - f_B(n)(i,j)|}{\sum_{i,j} f_A(n)(i,j) + \sum_{i,j} f_B(n)(i,j)}$$

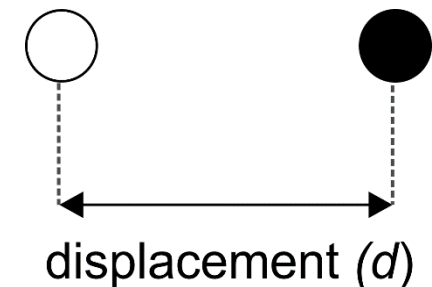
values between **0 (worst)**
and **1 (best)**

- A simple idealized setup with a **single displaced event**

$$NSS = 1 - d/n$$



$$d = (1 - NSS) \cdot n$$





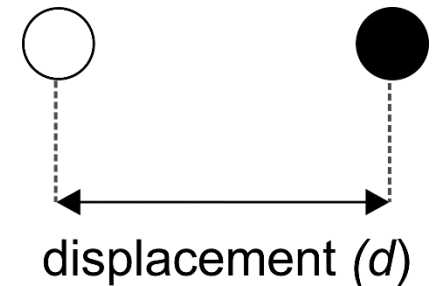
The dNSS measure

$$d = (1 - NSS) \cdot n$$

- Determine the **neighbourhood size where $NSS=0.5$** (n')
- The displacement can then be estimated as

$$dNSS = \frac{n'}{2}$$

- Although this result is based on a **very simple idealized setup** it works quite well **also in other situations**
- A **special adjustment** needs to be made if events are **overlapping**. The fields need to be **unbiased** prior to analysis.





Idealized examples

New idealized cases

