

Testing the performance of a spatial consistency test for data quality control

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Meteorological network

Automatic stations;

Complex orography;

Hourly data;

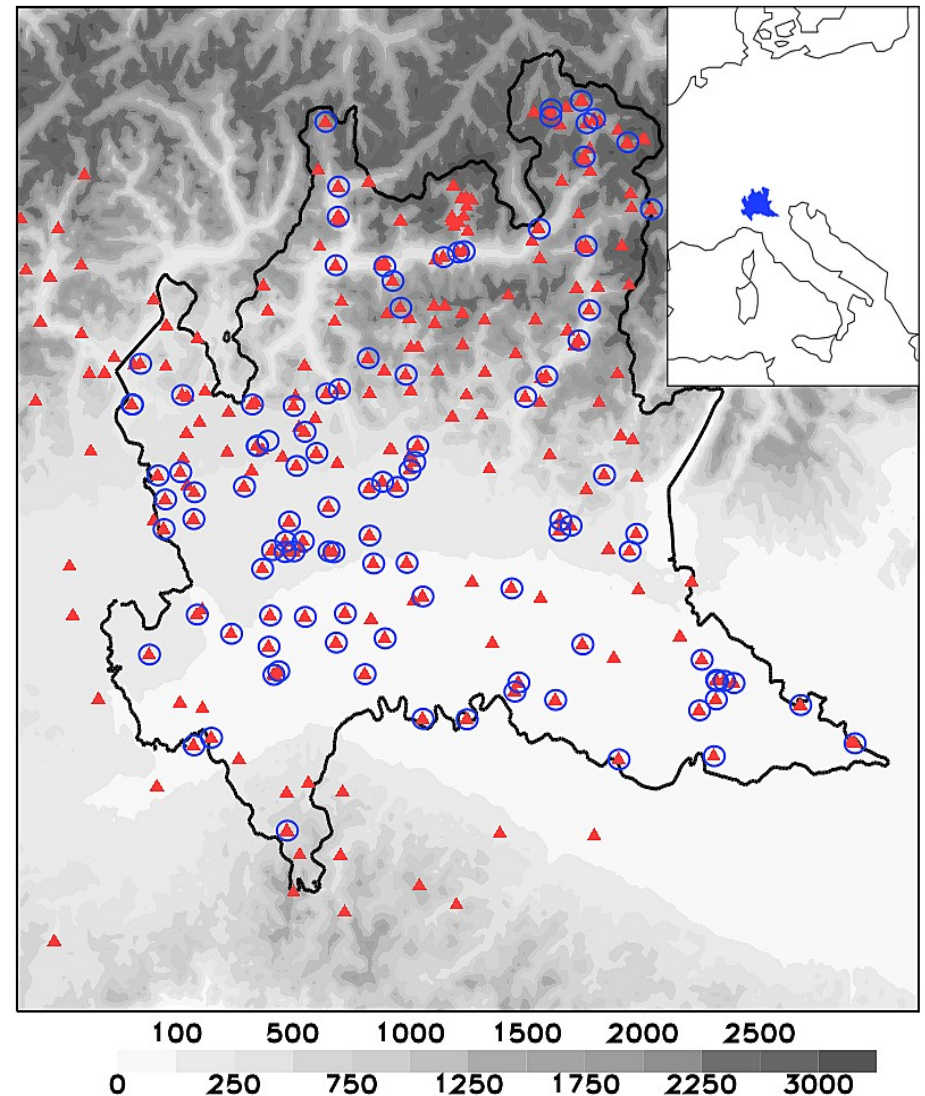
High station density;

Station altitudes: 10m - >3000m

AMSL;

Grid: 1.5 Km (177x174);

Grid orography from a high resolution DEM (250 m) without smoothing;



Meteorological network

Automatic stations;

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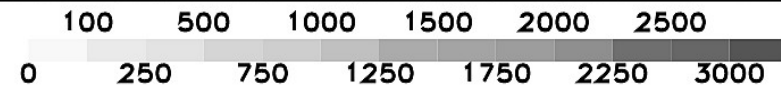
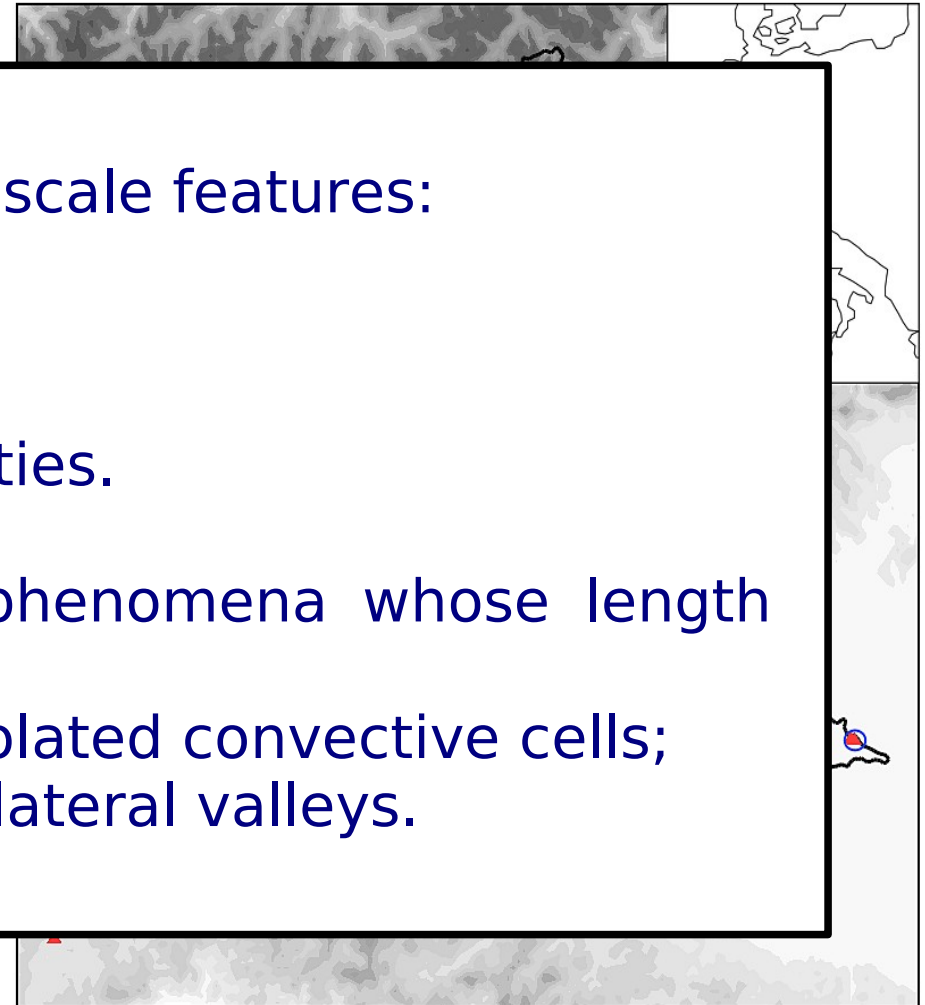
smoothing;

efficiently resolves most mesoscale features:

- major valley cold pools;
- thermal inversion;
- widespread foehn warming;
- heat-island effect of major cities.

unable to correctly resolve phenomena whose length scale < 10 Km:

- cold outflow associated to isolated convective cells;
- katabatic warmings in small lateral valleys.



Automated Quality Control

- Plausible value check
- Time consistency check 1: **step** (check on a maximum allowed variability of an hourly value)
- Time consistency check 2: **persistence** (check on a minimum required variability in a prescribed time interval)
- **Spatial Consistency Test (SCT)**



Decision Making Algorithms



Data disseminated to the users

Test, parameters and DMA implementation are variable dependent

Automated Quality Control *(details tomorrow in session AW6 !)*

- Plausible value check
- Time consistency check 1: **step** (check on a maximum allowed variability of an hourly value)
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Decision Making Algorithms

Data disseminated to the users

DMA-Temperature	1	2	3	4	5	6	7
Plausibility	P	P	P	P	F	-	-
SCT	P	P	W	W	-	F	-
Step	P	W	P	W	-	-	-
Persistence	P	P	P	P	-	-	F
Result	G	G	G	B	B	B	B

Test, parameters and DMA implementation are variable dependent

Spatial Consistency Test

The SCT is based on a spatial analysis scheme (*Uboldi et al., Meteorol. Appl., 2008*), an implementation of Optimal Interpolation

Main features of the analysis scheme:

- Background information derived from observations detrending;
- Background error covariance specified by means of 3D gaussian correlation functions;
- Efficient computation algorithm.

Background field
(first guess) (\mathbf{x}^b)

Observation
(measured data) (\mathbf{y}^o)

Background space = analysis space
(regular grid)

Observation space
(sparse data)

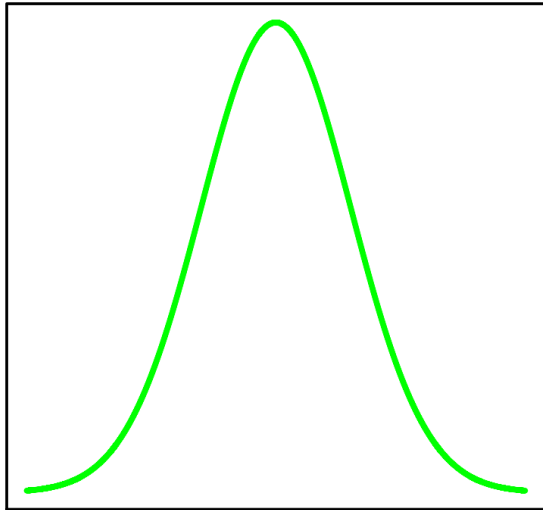
background error covariance
matrices \mathbf{G} , \mathbf{S}

observation error
covariance matrix \mathbf{R}

Analysis field

$$\mathbf{x}^a = \mathbf{x}^b + \mathbf{G} (\mathbf{S} + \mathbf{R})^{-1} (\mathbf{y}^o - \mathbf{y}^b)$$

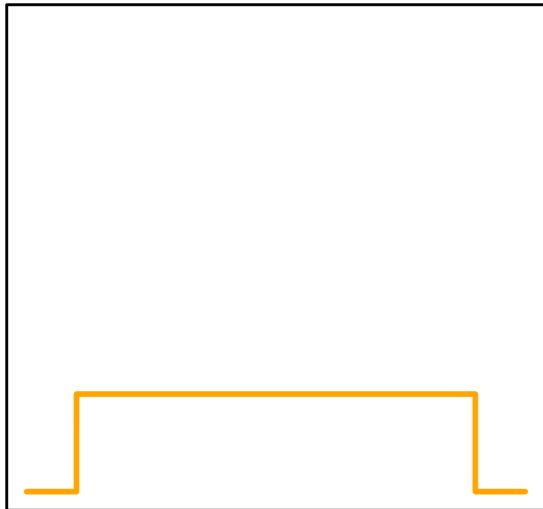
Observational Error Model (Lorenz and Hammond, *qjrms*, 1988)



Observation not affected by
Gross Error (GE)

$$P(O|\overline{GE})$$

Gaussian *pdf*

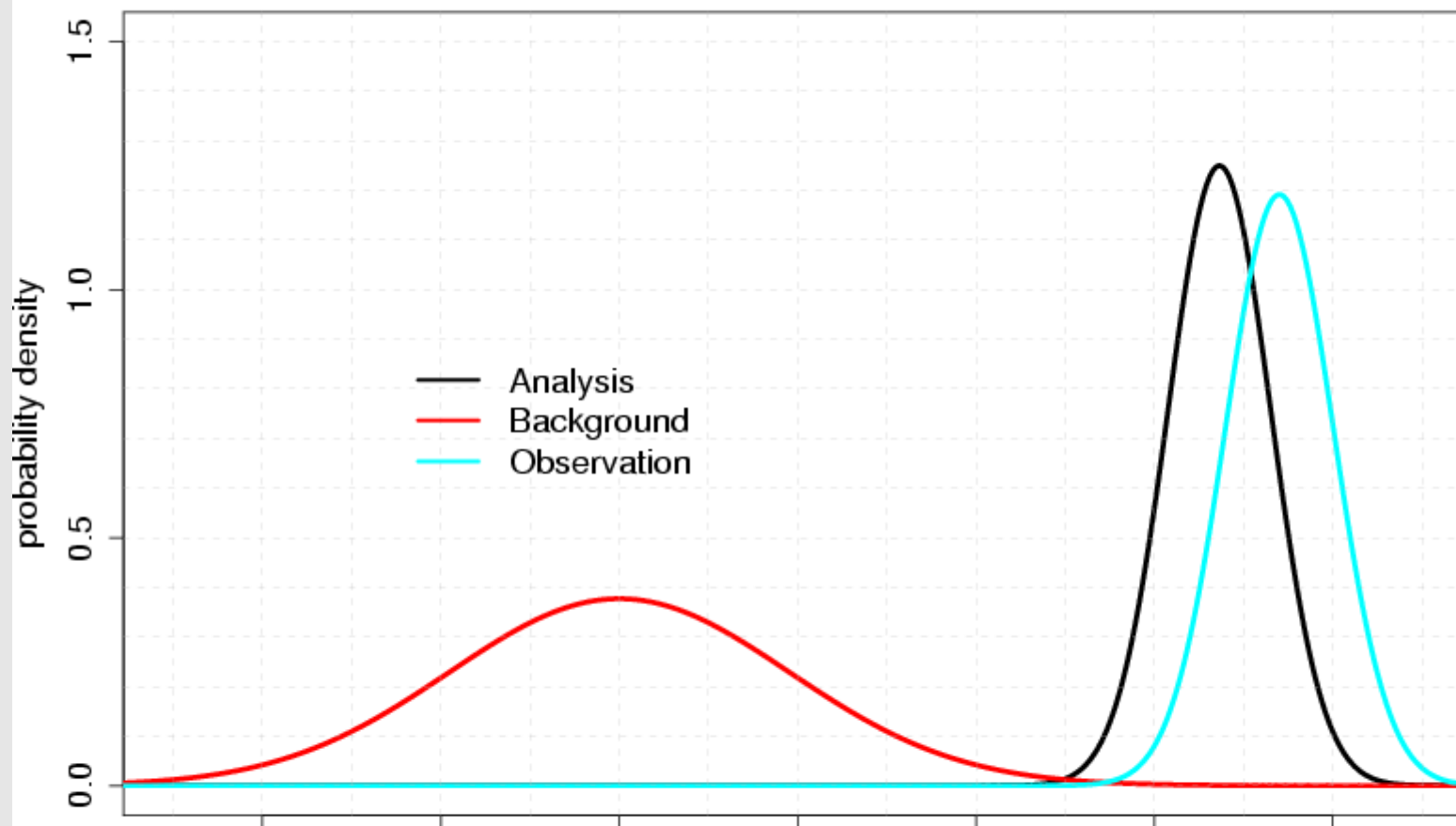


Observation affected by **GE**

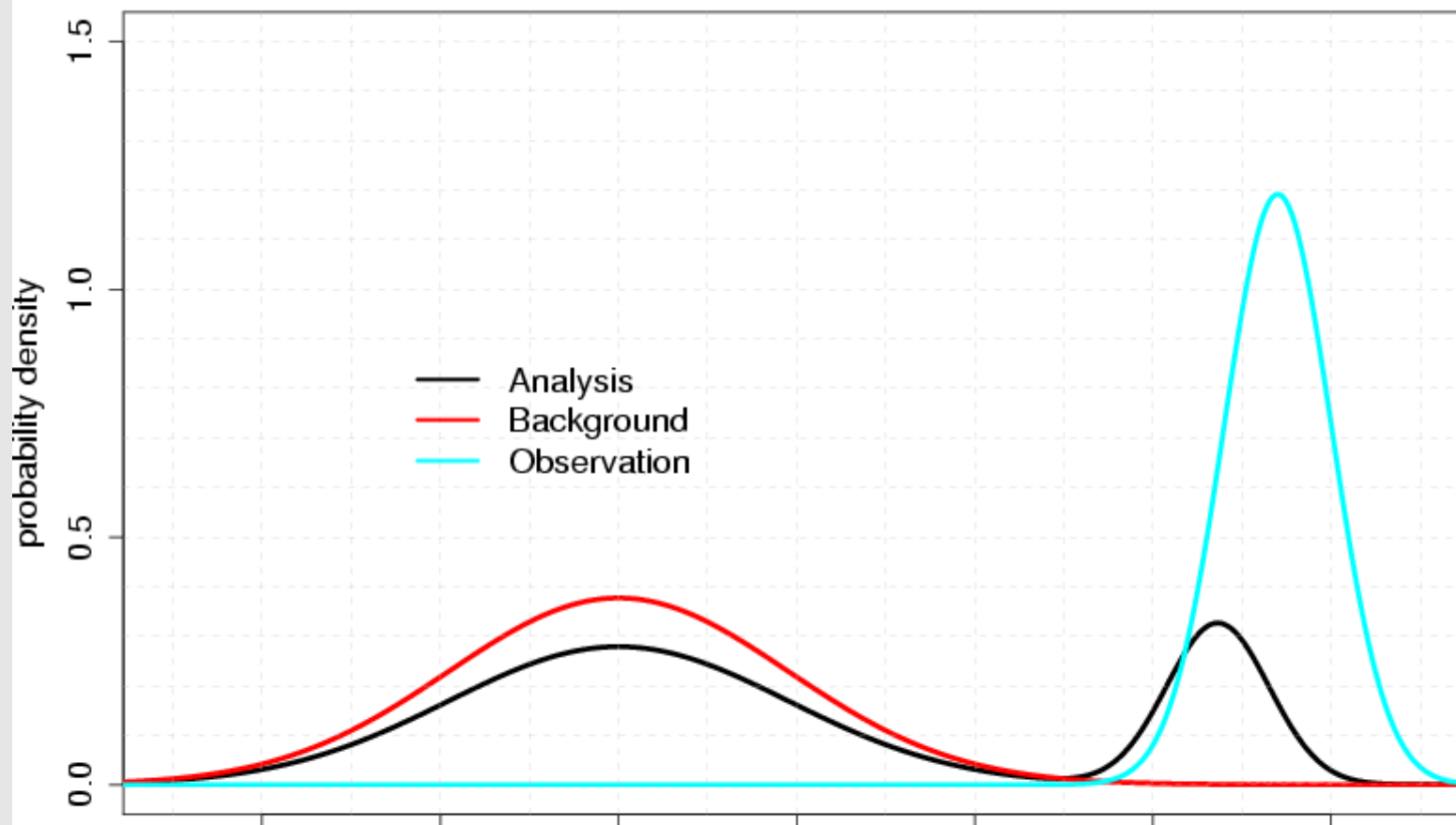
$$P(O|GE)$$

uniform *pdf*
over the climatological interval

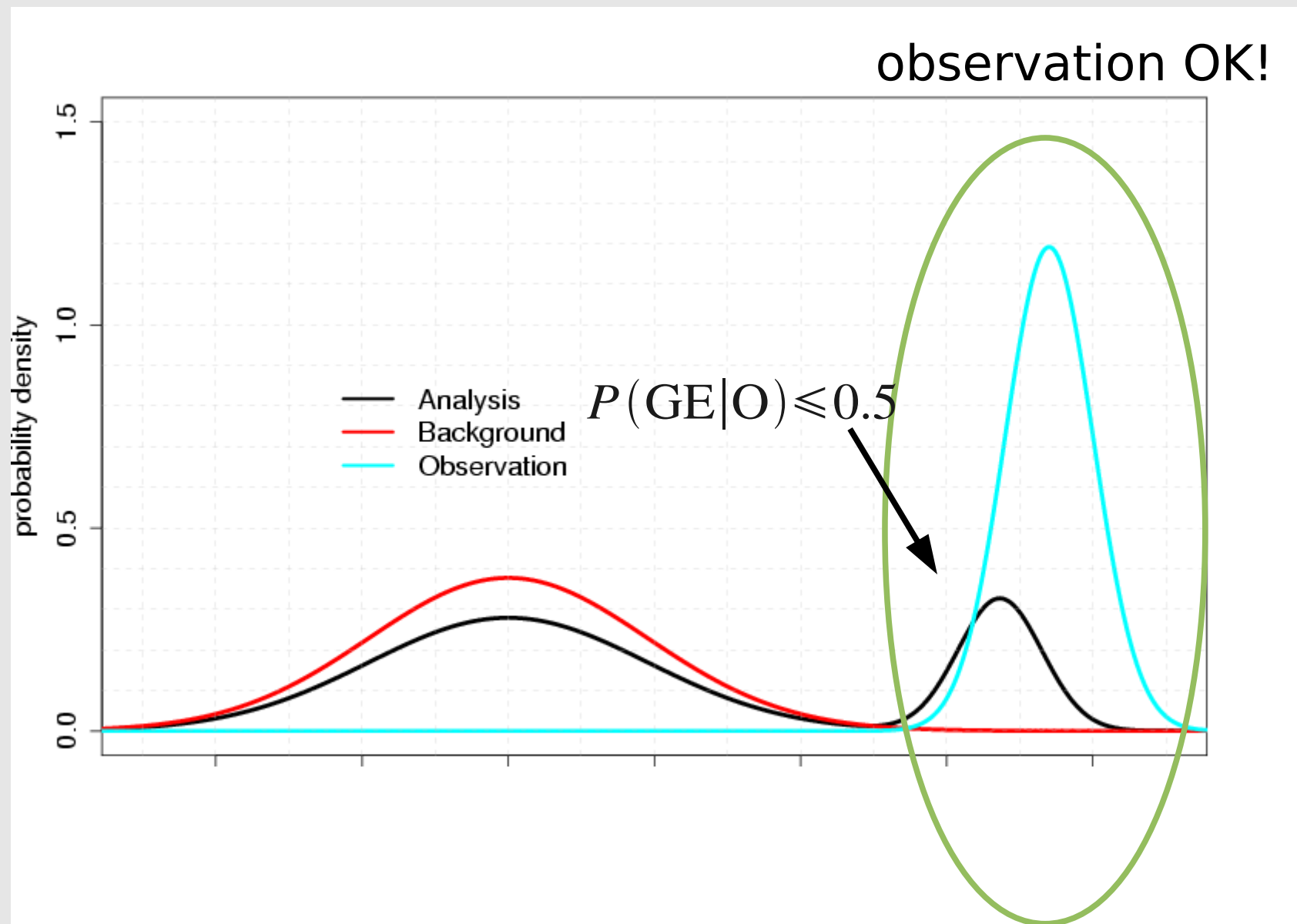
Observational Error Model: Gaussian only GE present but NOT accounted for



Observational Error Model: accounting for a GE

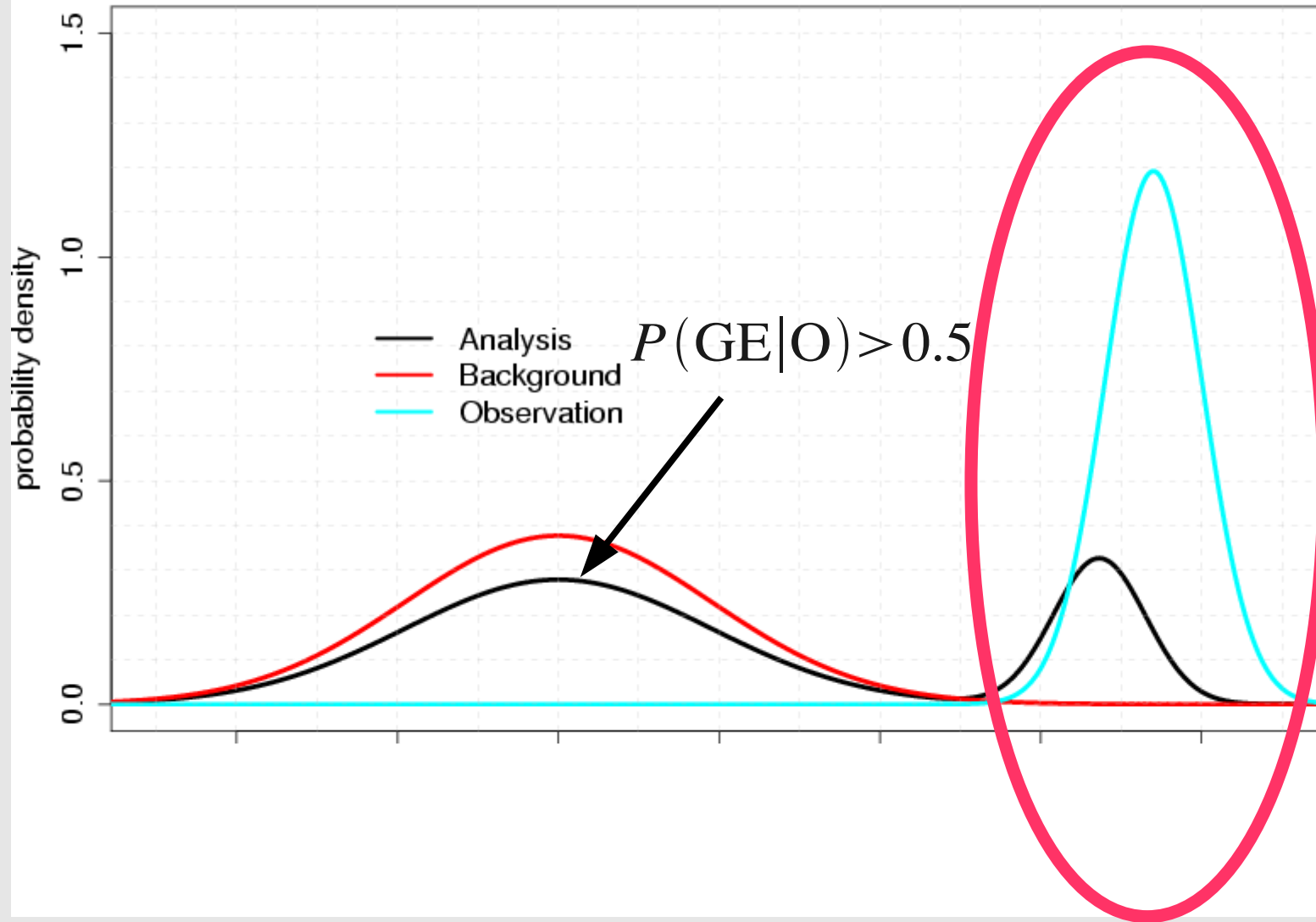


Observational Error Model



Observational Error Model

The observation is rejected



Cross Validation (CV) Analysis

In our case (no independent background field) the *a priori* estimate is the CV analysis:

$$y^{CVa}$$

The CV analysis is produced using ALL observations EXCEPT the observation undergoing the SCT.

As a consequence, observation error and CV analysis error are uncorrelated.

Spatial Consistency Test

$$(y^o - y^{CVa})^2 > T^2 (\sigma_o^2 + \sigma_{CVa}^2)$$

Only T^2 , σ_o^2 are needed: objectively estimated from the statistical hypothesis and 3-year statistics

The SCT automatically accounts for local data density:

- completely isolated stations (CV analysis = background): *permissive*

$$(y^o - y^b)^2 > T^2 (\sigma_o^2 + \sigma_b^2)$$

- totally redundant stations (CV analysis = analysis): *restrictive*

$$(y^o - y^{CVa})^2 = (y^o - y^a)^2 > T^2 \sigma_o^2$$

SCT Performance monitoring

			SCT	
Persistence	Step	Pass	Warning	Fail
Pass	Pass	1151771	2568	1252
Pass	Warning	426	177	1285
Fail	Pass	4815	1716	4323
Fail	Warning	0	0	0

Total observation tested = 1165339

January – June 2009 temperature observations

SCT Performance monitoring

SCT rejection frequency = 0.0045 (expected: from 0.0018 to 0.0031)

Estimate $f(\text{pass}|\text{GE}) = 0.29$ (expected: from 0.13 to 0.23)

- Network management differences
- Temporal GE correlations
- Systematic errors

Conclusions

SCT:

- particularly efficient for a high density network
 - based on clear statistical hypothesis (error model + OI)
 - objective scheme for estimating thresholds and parameters
 - *a priori* estimate of false and missed rejections
 - local station density automatically taken into account
- the threshold estimation procedure also provides a measure of network reliability;
- preliminary comparison of test performance in the first 6 months of operational use with the 3 years statistics used to estimate test parameters show agreement in the order of magnitude but larger values of P(GE) in the network.
- large amplitude representativity errors

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