

Application of neighborhood-based contingency scores to AROME verification

Fabien Stoop, Joël Stein Météo-France - DirOP/COMPAS

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Motivation

Météo-France main meteorological deterministic forecasting systems:

- **ARPEGE** : global low resolution (~5km over France),
- **AROME** : regional high resolution (\approx 1.3km over France)

Goal: Define a new headline score for AROME:

- for precipipation events,
- without double penalty effects,
- user-friendly (easy to explain).

AROME verification - Case study

QPF between 18 and 21H UTC, August 6th 2019 (convective period).





ARPECE leadtime 21H



Forecasters/modelers prefer AROME (better location, patterns, maxima)

Reference (RADAR+obs)

AROME verification - Case study

Currently, AROME headline score derives from FSS (Amodei et al (2015)).





Which model is better according to FSS ?

2mm : **AROME** better.

10mm : **ARPEGE** better for large neighborhoods.

Idea: include neighborhood strategy in contingency tables

Neighborhood-based contingency tables - Definition







Neighborhood-based contingency tables - Definition



Final contingency table: summation over all neighborhoods

AROME verification - Case study





Which model is better according to PSS ?

2mm : AROME better.

10mm : AROME better.

Bonus : Frequency bias, POD, FAR,... Easier to explain.





Application of neighborhood-based contingency scores to AROME verification- 6

AROME verification - Results over 2019



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Application of neighborhood-based contingency scores to AROME verification-7

AROME verification - Conclusion

Use of neighborhood in contingency tables (error compensation). Limit double penalty effects, like in the FSS.

Several benefits:

- contingency skill scores closer to forecasters/modelers insights than FSS,
- user-friendly measures (POD, FAR, ...)
- \implies replacement of FSS by PSS as headline score.

Perspectives:

- application to probabilistic forecasts of events,
- application to other events (gust, fog, ...) and extreme events (SEDI).

Merci

Additional content

Algorithm

Algorithm from Faggian et al (2015) (summed area tables) adapted to deal with:

- neighborhood contingency tables,
- masked data (useful for station data).



k classes

Classical			Error compensation
$\int t_{1,1}$	•••	$t_{1,k}$	$\begin{pmatrix} t'_{1,1} & \cdots & t'_{1,k} \end{pmatrix}$
	·	:	
$\left\{t_{k,1}\right\}$	• • •	$t_{k,k}$	$\left(\begin{array}{ccc} t_{k,1}' & \cdots & t_{k,k}' \end{array} \right)$

with:

$$\begin{aligned} t'_{i,i} &= t_{i,i} + \sum_{j=1..k} \min(t_{i,j}, t_{j,i}) \text{ for } i = 1..k \\ t'_{i,j} &= t_{i,j} - \min(t_{i,j}, t_{j,i}) \text{ for } i \neq j \end{aligned}$$

Links with probabilistic scores - In progress

Each grid point : event if $p \ge p_t$

Contingency tables evaluated for each threshold p_t (+error compensation)



Amodei, M., I. Sanchez and J. Stein, 2015: Verification of the French operational high-resolution model AROME with the regional Brier probability score. *Meteor. Appl.*, **22**, 731-745

Faggian, N., B. Roux, P. Steinle and B. Ebert, 2015: Fast calculation of the Fractions Skill Score. *Mausam*, **66**, 457-466

Schwartz, C. S., 2017: A comparison of methods used to populate neighborhood-based contingency tables for high-resolution forecast verification. *Wea. Forecasting*, **32**, 733-741

Stein, J., F. Stoop, 2019: Neighborhood-Based contingency tables including error compensation. *Mon. Wea. Rev.*, **147**, 329-344