

Comparative Evaluation of Earthquake Forecasts: Application to Italy

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(My) background

- T. Gneiting's HITS group working on forecast evaluation
- Evaluation based on [scoring rules/functions](#) is common in meteorology and economics
- (How) can we use them in a point process/spatial setting?

Consistent scoring functions

- Assign score/loss $S(x, y)$, where x is the forecast and y is the outcome
- **Consistency** ensures that the true value minimizes the score/loss on average
- Example: S is consistent for the **mean** if

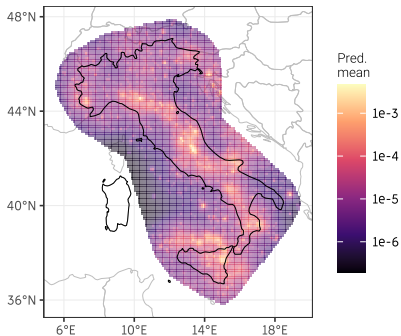
$$\mathbb{E}_{Y \sim F} S(\text{mean}(F), Y) \leq \mathbb{E}_{Y \sim F} S(x, Y) \quad \text{for all } x$$

- $S(x, y) = (x - y)^2$ is consistent for the mean
- $S(x, y) = |x - y|$ is not consistent for the mean (but for the median)

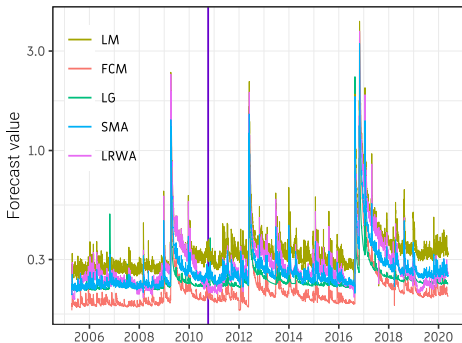
M4+ earthquakes in Italy

We get $\left\{ \begin{array}{l} \text{predicted means of model LM, FCM, LG, SMA, LRWA} \\ \text{observed counts (0, 1, 2, \dots) of M4+ earthquakes} \end{array} \right.$

Forecasts $m = \text{LM}$, $t = 2010-10-06$



Combined Forecasts for all of Italy



→ Which model issues the best predictions?

→ How well do the predictions represent reality?

Model Scores

Quadratic score

$$S(x, y) = (x - y)^2$$

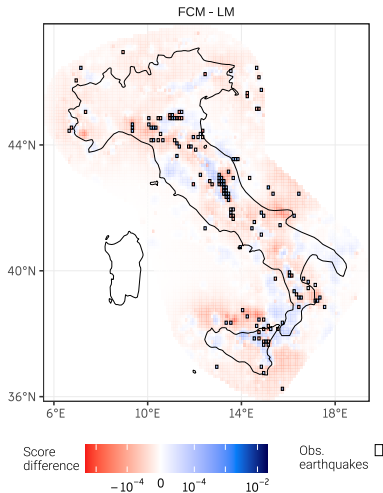
Poisson score (preferred)

$$S(x, y) = x - y \log x$$

Overall mean scores \bar{S}

Model	Poisson	Quadratic
LM	2.71	0.841
FCM	2.80	0.846
LG	3.02	0.847
SMA	2.73	0.844
LRWA	2.69	0.842

Poisson Score Difference



The Poisson Score $S(x, y) = x - y \log x$

- Commonly used in CSEP methodology as *Poisson log-likelihood*
- Consistent for the mean. Connection to Poisson distribution is **purely formal**
 - Forecasts x don't need to come from a Poisson model
 - Observations y don't have to be Poisson
- Evolve T-test (Rhoades et al. 2011) into **Diebold-Mariano test** of equal predictive performance (measured in terms of Poisson score)

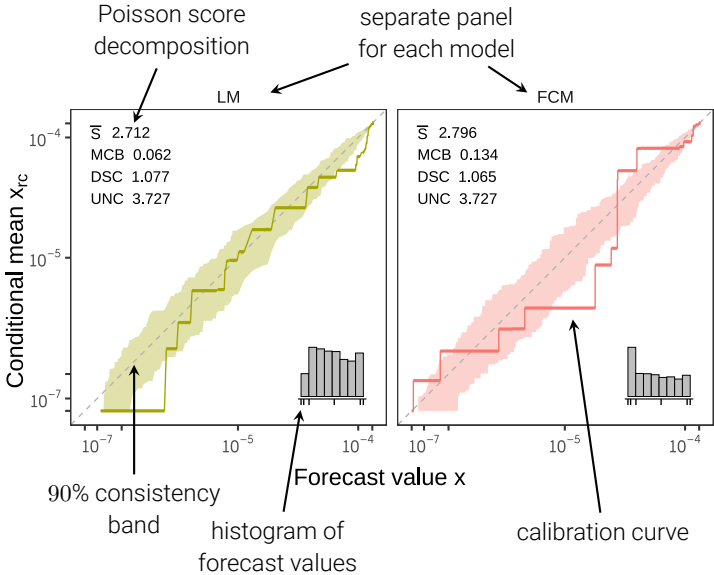
Are forecasts and outcomes compatible?

- Reliability quantifies how good forecasts represent outcomes
- The forecast X is **mean-calibrated** for the outcome Y if conditional mean of Y given X equals X

$$\mathbb{E}[Y|X] = X$$

- Estimate conditional mean X_{rc} and plot X_{rc} vs. X in a **reliability diagram**

Reliability Diagrams



High level overview

	CSEP Framework	vs.	New Proposal
Comparative	Poisson metric + heuristic statistical test	\approx	consistent scoring function + theory-based statistical test
Absolute	statistical tests	\neq	diagnostic tools reliability diagram
	often many assumptions		non-parametric weak assumptions

Summary

- Evaluate earthquake forecasts in the form of **predicted mean counts**
- The Poisson scoring function is convenient, **no distributional assumptions** are needed
- For forecasts in **other formats** (e.g. quantile forecasts), 'simply' change the scoring function

References

In review. Brehmer, J. R., Kraus K., Gneiting, T., Herrmann, M., and Marzocchi, W. (2024). Comparative evaluation of earthquake forecasting models: An application to Italy. Preprint, <https://arxiv.org/abs/2405.10712>.

Mathematical background (1). Brehmer, J. R., Gneiting, T., Herrmann, M., Marzocchi, W., Schlather, M., and Strokorb, K. (2024). Using scoring functions to evaluate point process forecasts. *Annals of the Institute of Statistical Mathematics*, 76, 47–71, <https://doi.org/10.1007/s10463-023-00875-5>.

Mathematical background (2). Gneiting, T. and Resin, J. (2023). Regression diagnostics meets forecast evaluation: Conditional calibration, reliability diagrams, and coefficient of determination. *Electronic Journal of Statistics*, 17, 3226–3286, <https://doi.org/10.1214/23-EJS2180>.