

Prequential Approaches for the Assessment of Stochastic Spatio-temporal Models for Infectious Disease Data



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Introduction

- Despite their theoretical advantages, prequential methods (Dawid, 1984) for model selection and comparison have not been extensively explored.
- Only requirement: models under consideration are able to generate predictions for the next observation.
- We investigate the advantages of one-step-ahead predictions for assessing model fit in a Bayesian framework.
- The methodology is applied to (sequential in nature) sheep pox and foot and mouth disease epidemic data.

Prequential methodology

- Examines each model's prediction for the next data point x_{n+1} after fitting the model to data x_1, x_2, \dots, x_n .
- Criterion for model selection: distances, known as scoring rules, measuring the discrepancy between the forecasts and the data.
- The best model is the one with the smallest average score among the entertained models.
- We utilize four scoring rules for the evaluations:

Single-valued predictions:

$$AES = |x_t - \mu_{p_t}|$$

$$SES = (x_t - \mu_{p_t})^2$$

Probabilistic predictions:

$$\log S = S(x_t, F_t) = -\log q(x_t)$$

$$RPS = \sum_{k=0}^{\infty} (P(X \leq k) - 1(x \leq k))^2$$

Spatio-temporal models

Let y_i be the number of occurrences of infected farms at time t_i . Poisson, NB and ZIP/ZINB models are special cases of:

$$y_i \sim g(y_i | \theta_i, p_i)$$

$$g(y_i | \theta_i, p_i) = p_i I_{\{y_i=0\}} + (1 - p_i) f(y_i | \theta_i)$$

$$\theta_i = h(\mu_i) = \exp(\mu_i)$$

$$d\lambda_t = \phi(\lambda_t - \mu_t) dt + dB_t$$

$$\mu_t = \mathbf{X}_{(t)} \cdot \Theta_{\beta} + \theta_{\tau} \cdot y_{(t-1)} + K(d_t, \Theta_K)$$

$$\Theta = (\Theta_{\beta}, \theta_{\tau}, \Theta_K)^t$$

with model parameters including environmental factors, transmission kernel functions and the mean reversion rate of the O-U process.

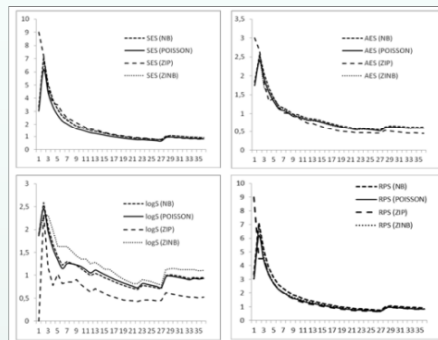
Implementation details

- Inference and sampling from the predictive distributions was performed in WinBUGS.
- Prequential analysis was done in MatBUGS, a tool developed for combining WinBUGS and Matlab.

Results

FMD data:

Model	SES	AES	log S	RPS	\bar{D}	DIC
Poisson	0.824	0.603	0.933	0.824	142	167.6
Negative binomial	0.914	0.633	0.956	0.914	146.2	174.3
ZIP _p	0.798	0.451	0.491	0.700	113.4	---
ZINB	0.924	0.597	1.109	0.851	116.4	---

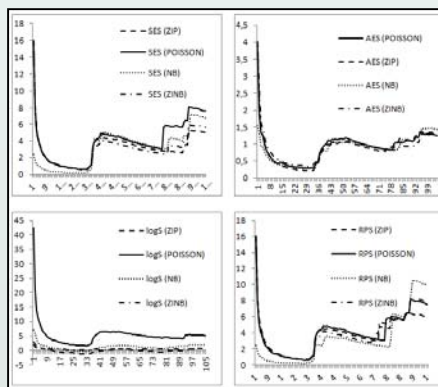


Series of one-step-ahead average prediction error estimates per mean prediction score for each data point

The analysis did not confirmed theoretical claims, such as high sensitivity of RPS to distance (Czado et al., 2009), or sensitivity of logS to extreme cases (Gneiting and Raftery, 2007)

Sheep pox data:

Model	SES	AES	log S	RPS	\bar{D}	DIC
Poisson	7.549	1.242	5.115	7.569	288.9	324.7
NB	6.746	1.422	8.636	10.338	276.48	329.6
ZIP _p	5.008	1.295	0.568	5.972	228	---
ZINB	5.818	1.275	0.591	6.049	277	---



Discussion

- Assessment based upon the scoring rules indicates the superiority of the ZIP distributional assumption.
- Main difference with the deviance-based criteria: scoring rules tend to select simpler models.
- A principled advantage of the prequential approach is that it respects the time ordering of the data.
- Allows for the temporal assessment of each model's performance.
- We prefer the probabilistic one-step-ahead predictions based on logS and RPS since they properly account for uncertainty.
- Efficient (sequential) calculation of predictions represents an important area for future research.