

Two extra components in the Brier Score Decomposition

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$$\begin{aligned} BS &= \frac{1}{n} \sum_{k=1}^m \sum_{j=1}^{n_k} (f_{kj} - o_{kj})^2 \\ &= \bar{o}(1 - \bar{o}) + \frac{1}{n} \sum_{k=1}^m n_k (f_k - \bar{o}_k)^2 - \frac{1}{n} \sum_{k=1}^m n_k (\bar{o}_k - \bar{o})^2 \\ &= \text{Uncertainty} + \text{Reliability} - \text{Resolution} \end{aligned}$$

Brier score components

To calculate the components (e.g. $E(o|f)$):

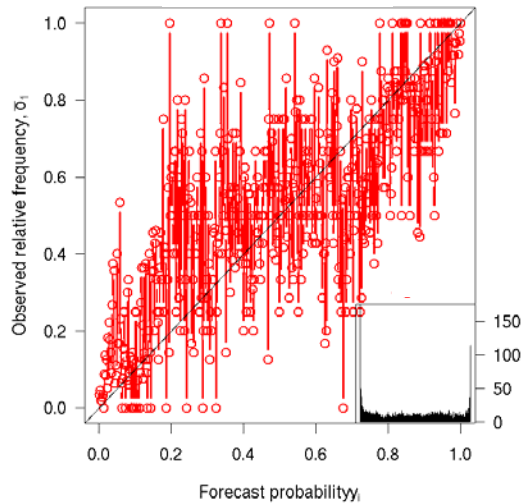
Stratify on ALL issued probability values $\{f\}$

OR

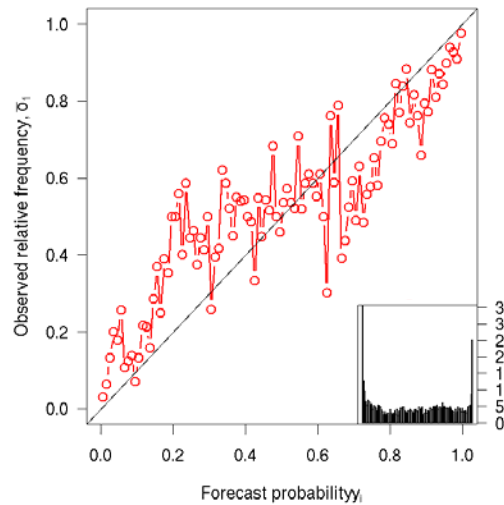
Stratify into m distinct probability bins:

- More reliable estimates (smoothing);
- Can avoid sparseness issues;
- Comparison of different forecasting systems.

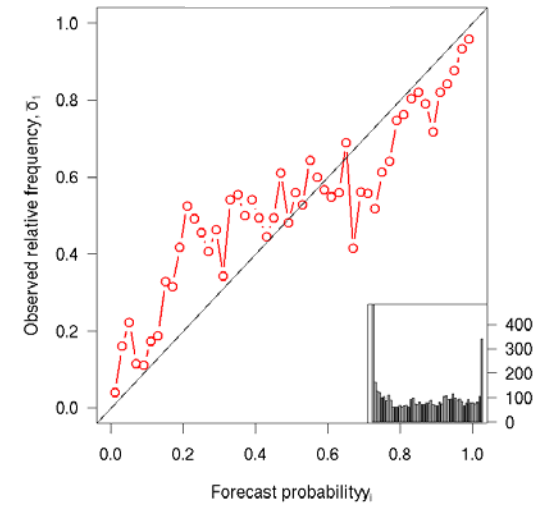
500 bins



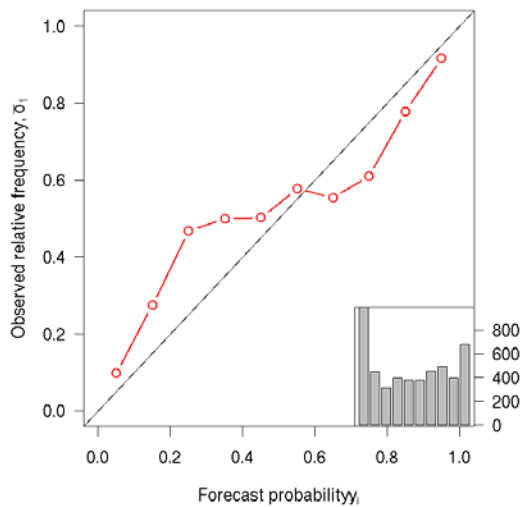
100 bins



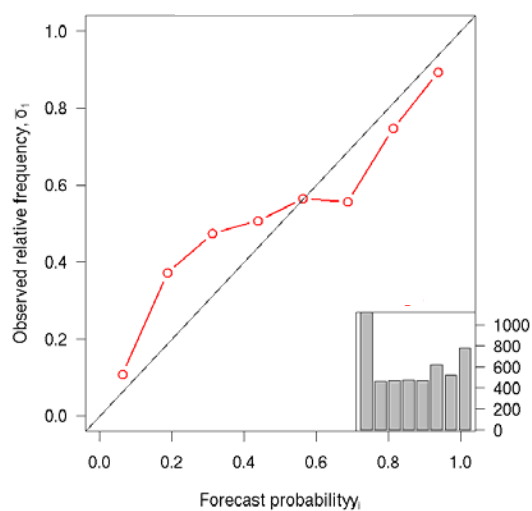
50 bins



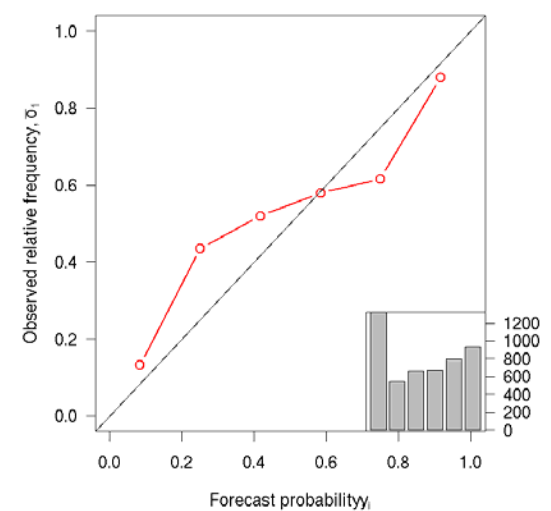
10 bins



8 bins



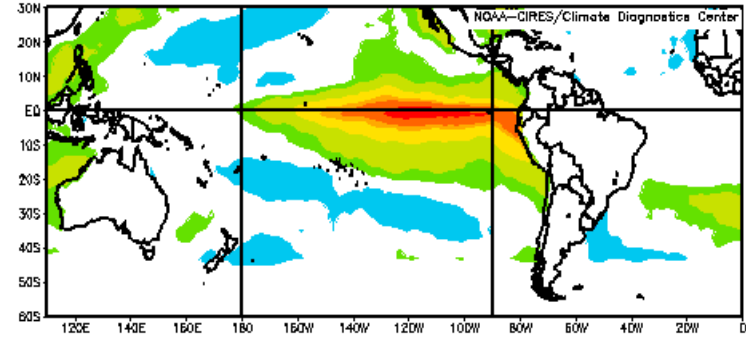
6 bins



→ Forecast system is over-confident

Example: Equatorial Pacific SST

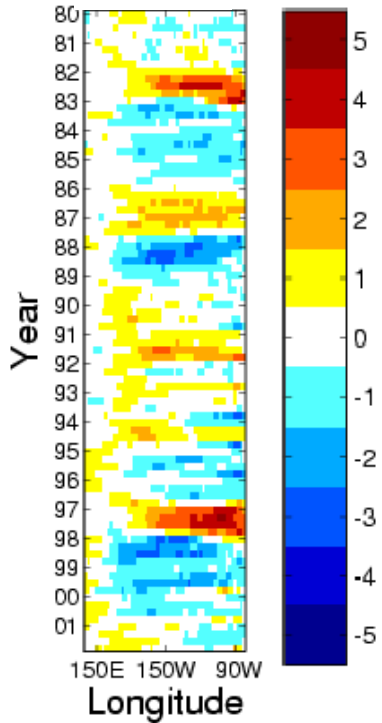
88 seasonal probability forecasts of binary SST anomalies at 56 grid points along the equatorial Pacific. Total of 4928 forecasts.



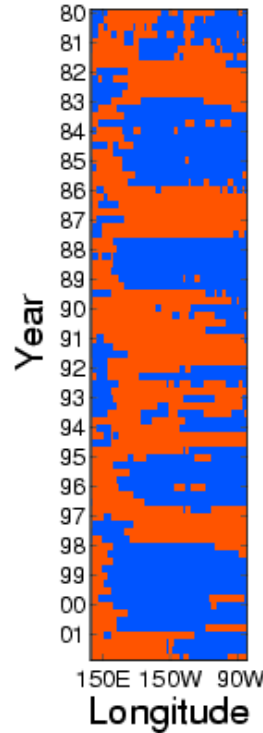
SST

$$o = (SST > 0) \quad f = \Pr(\hat{o})$$

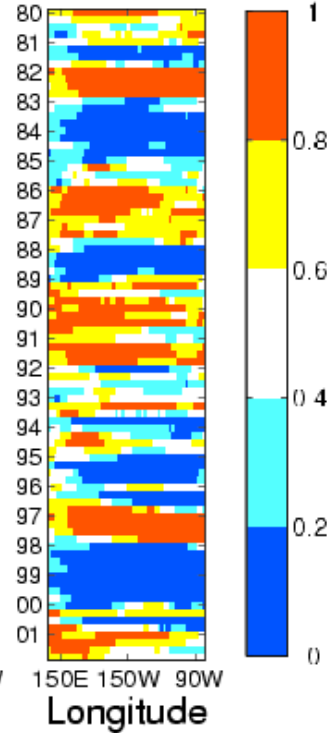
OBS



OBS



ENS

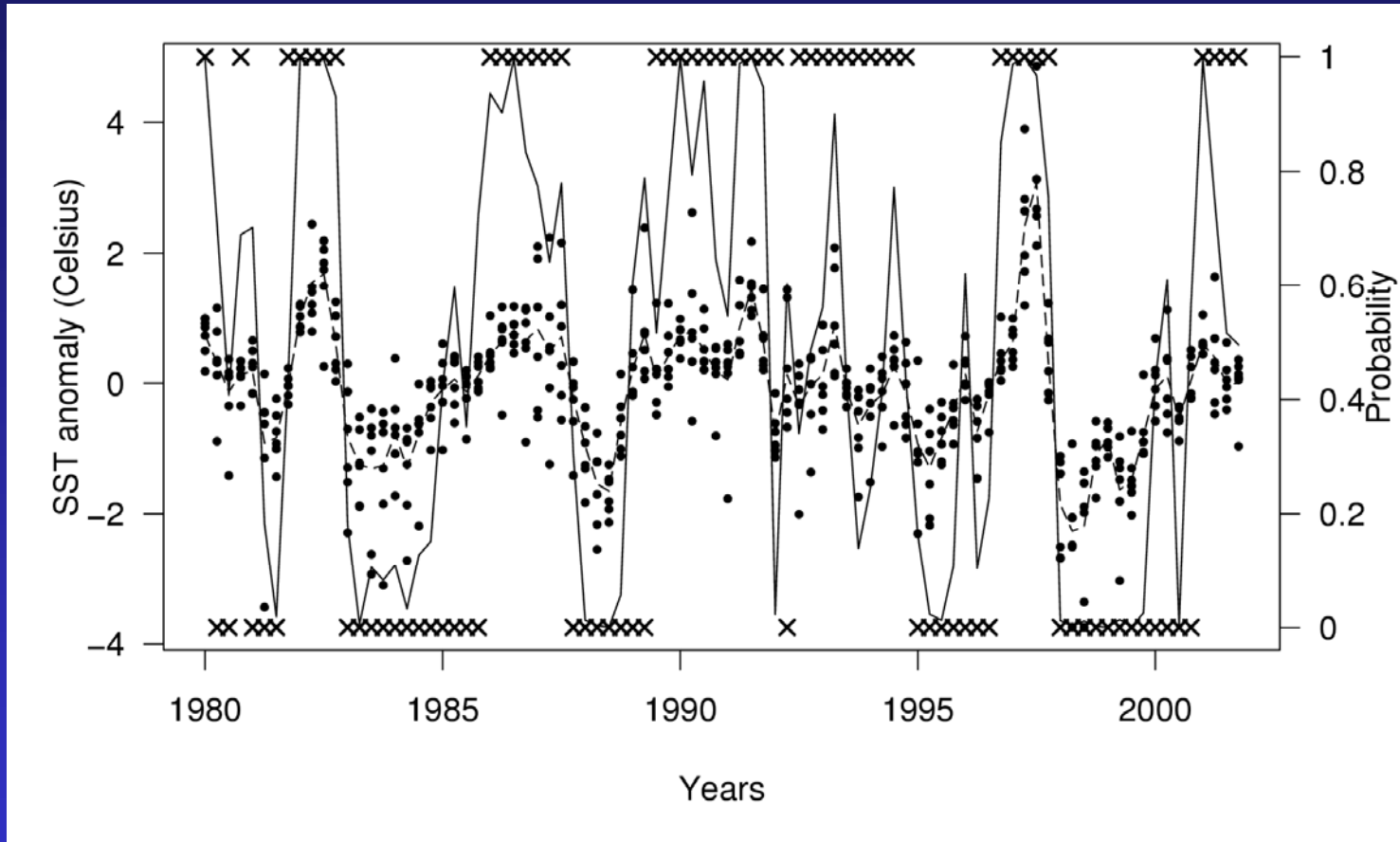


The probability forecasts were constructed by fitting Normal distributions to the ensemble mean forecasts from the 7 DEMETER coupled models, and then calculating the area under the normal density for SST anomalies greater than zero.

SST anomalies ($^{\circ}\text{C}$)

Forecast probabilities: f

Forecasts and observations at 150W

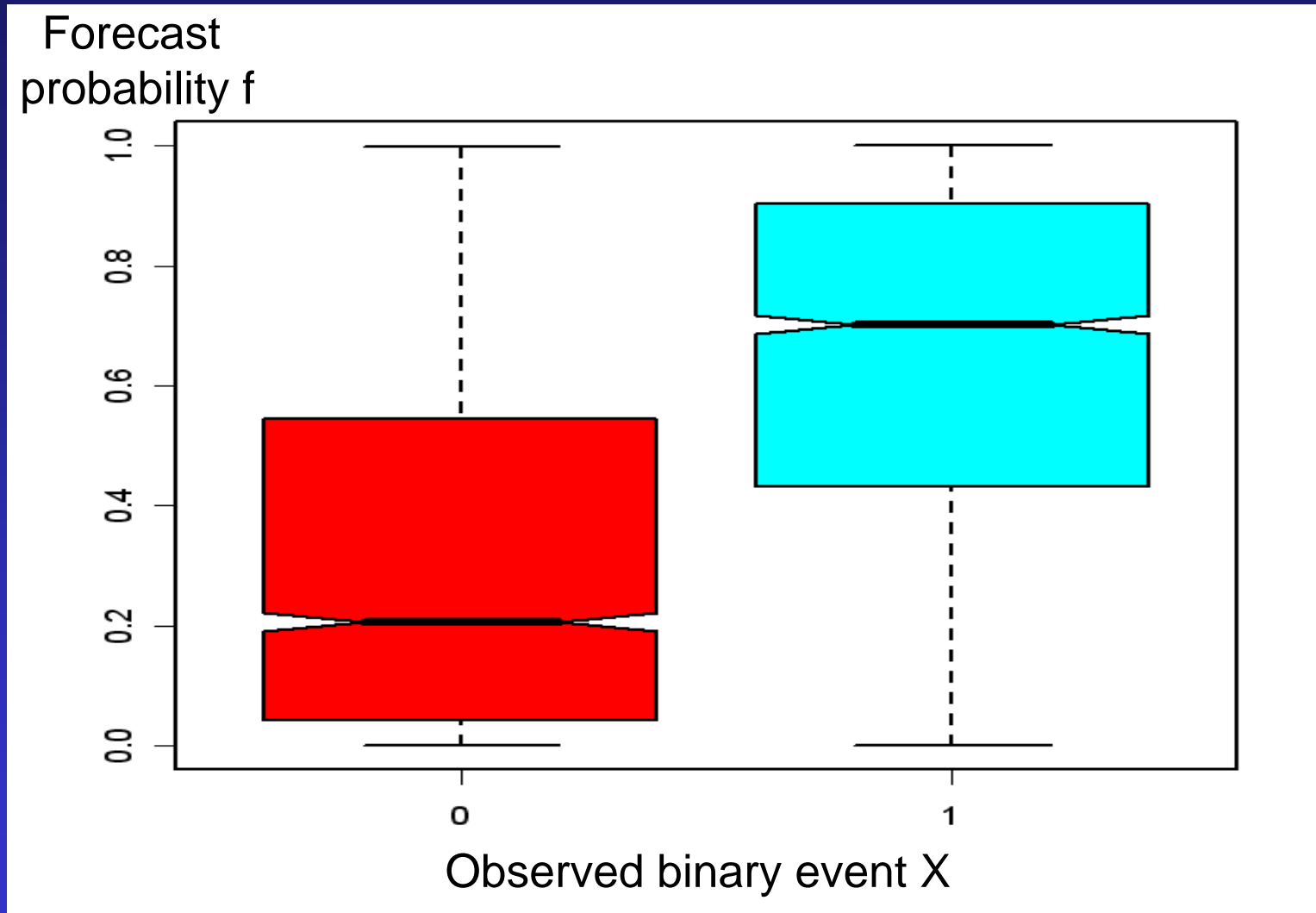


X = observed binary event: =1 for above average SST

Dots = ensemble mean forecasts of SST

Solid line = probability forecast estimated from ensemble means

Prob. forecasts stratified on observations



→ Forecast system has discrimination

Brier score for probabilities in m bins

$$\begin{aligned} BS &= \frac{1}{n} \sum_{k=1}^m \sum_{j=1}^{n_k} (f_{kj} - o_{kj})^2 \\ &= \bar{o}(1 - \bar{o}) + \frac{1}{n} \sum_{k=1}^m n_k (f_k - \bar{o}_k)^2 - \frac{1}{n} \sum_{k=1}^m n_k (\bar{o}_k - \bar{o})^2 \\ &\quad + \frac{1}{n} \sum_{k=1}^m \sum_{j=1}^{n_k} (f_{kj} - \bar{f}_k)^2 - \frac{2}{n} \sum_{k=1}^m \sum_{j=1}^{n_k} (o_{kj} - \bar{o}_k)(f_{kj} - \bar{f}_k) \end{aligned}$$

←NEW!!

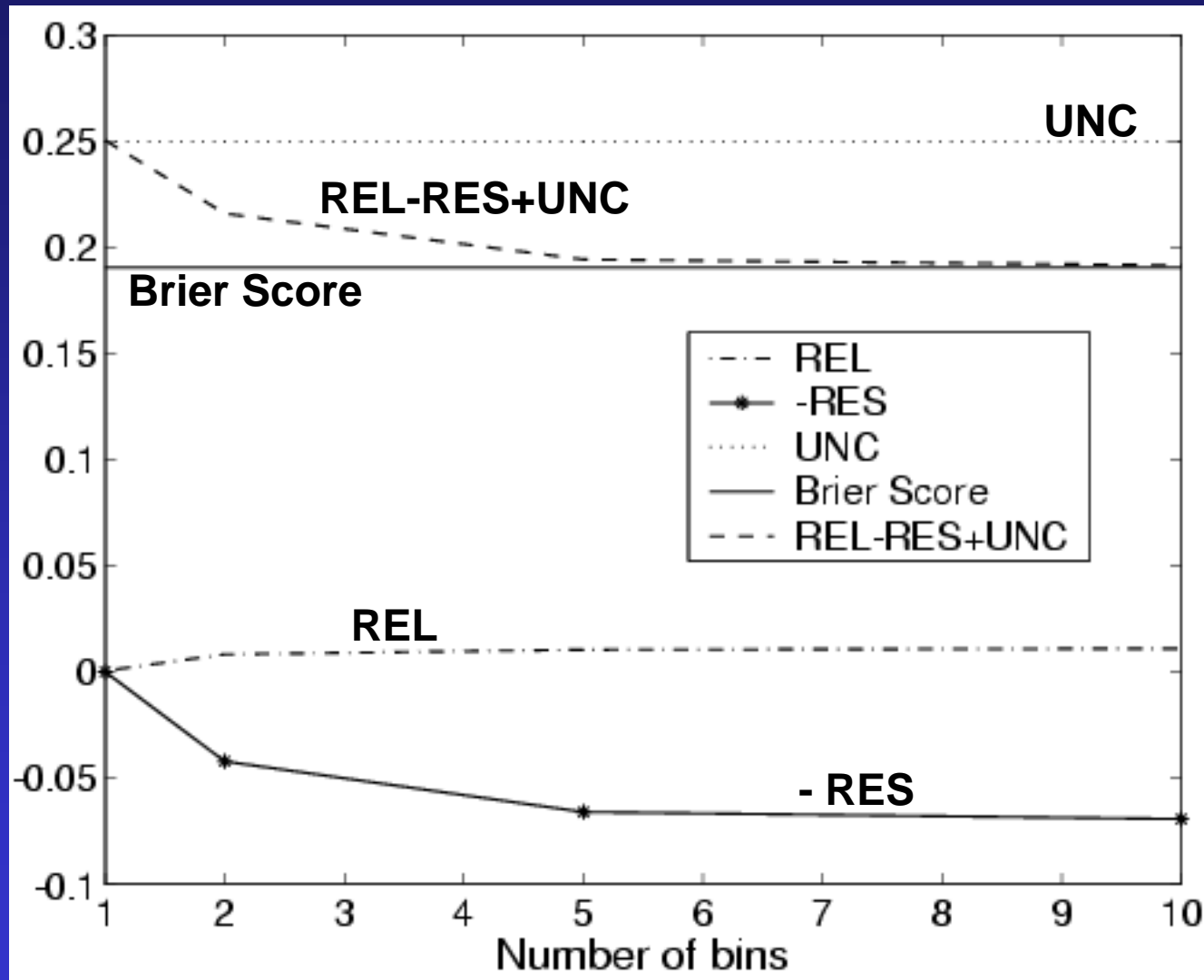
= Uncertainty + Reliability - Resolution

+ Within-Bin Variance - Within-Bin Covariance

For mathematical derivation please refer to:

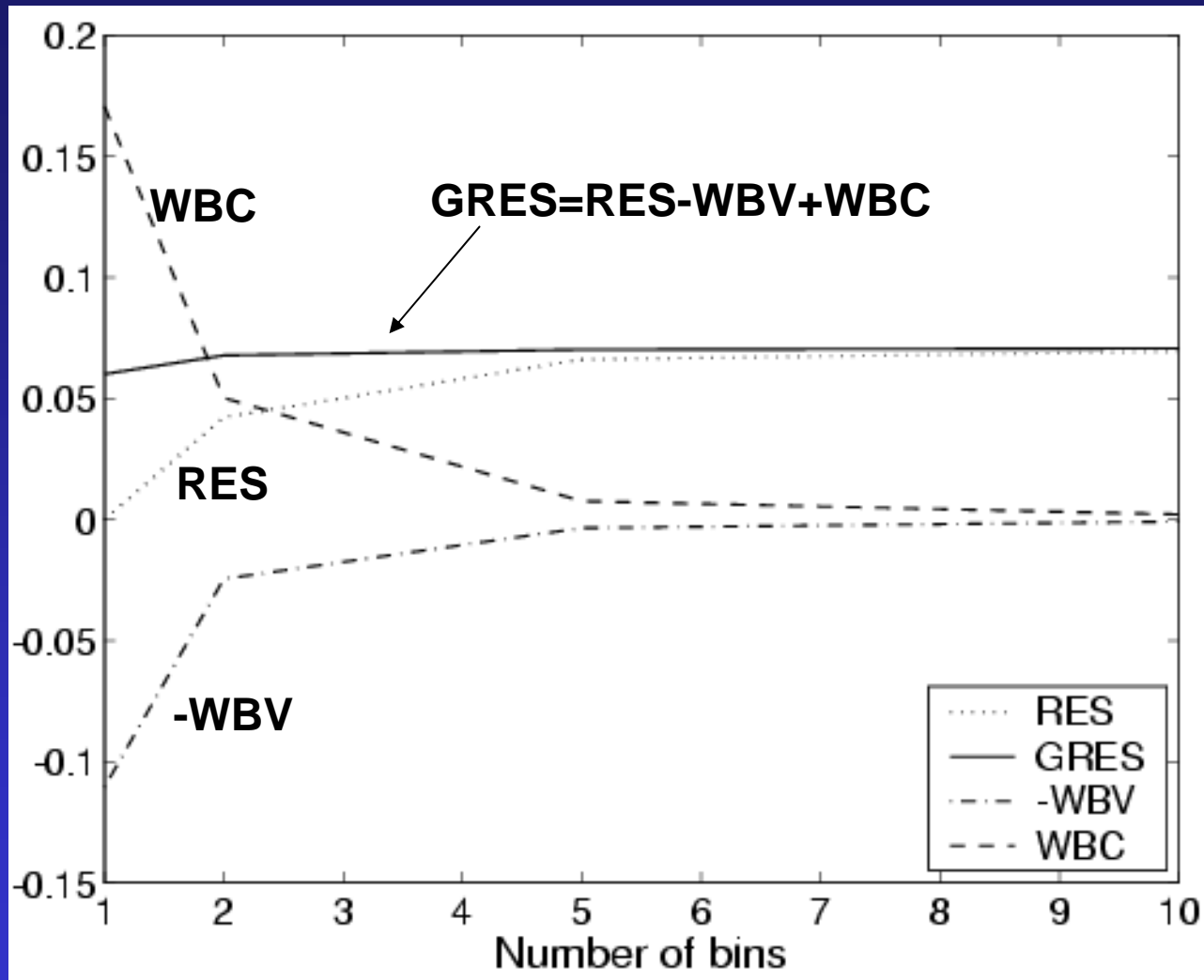
Stephenson, D.B., Coelho, C.A.S., and Jolliffe, I.T., 2007:
Two extra components in the Brier Score decomposition,
Weather and Forecasting (submitted).

Brier score components vs. num. of bins



→ Brier score is less than REL-RES+UNC!

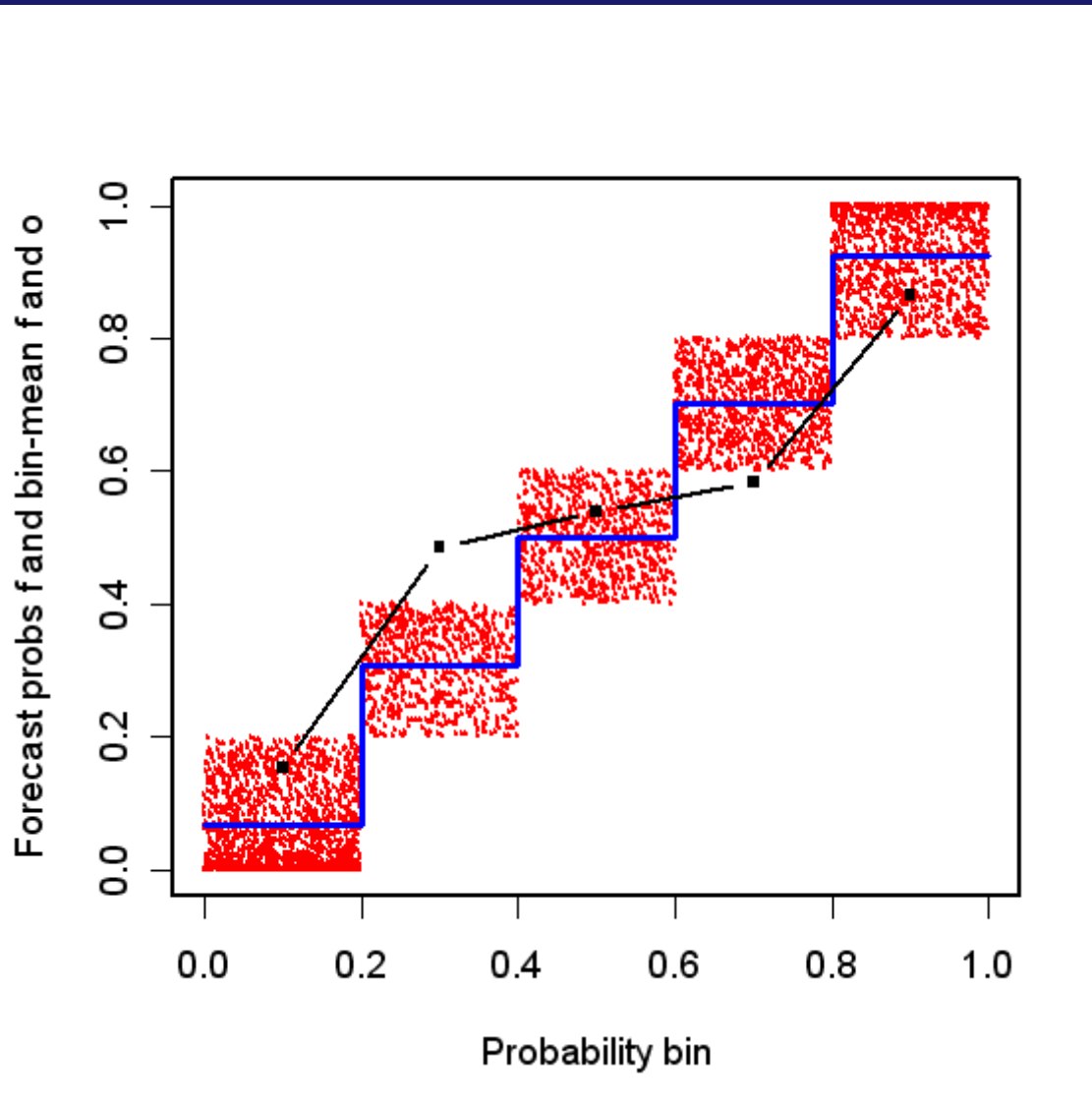
Within-bin terms and Generalised RESolution



→ $GRES = RES - WBV + WBC$ is more constant than RES

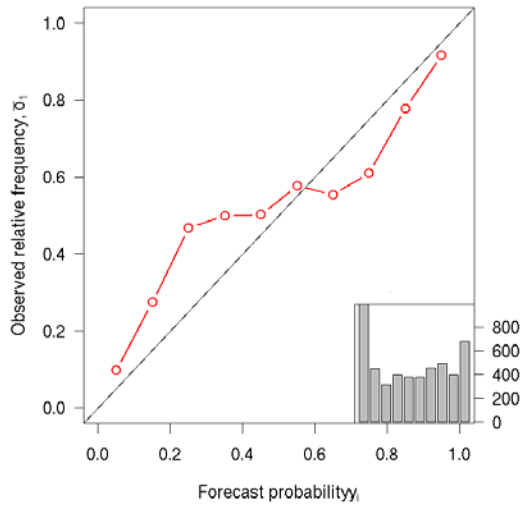
The End

Within-bin Variance of Probabilities

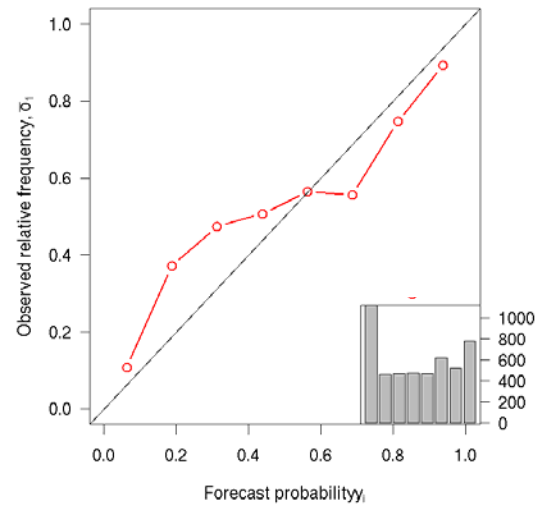


Red dots = probabilities f
Blue line = bin-average f
Black line = bin-average o
(reliability diagram)

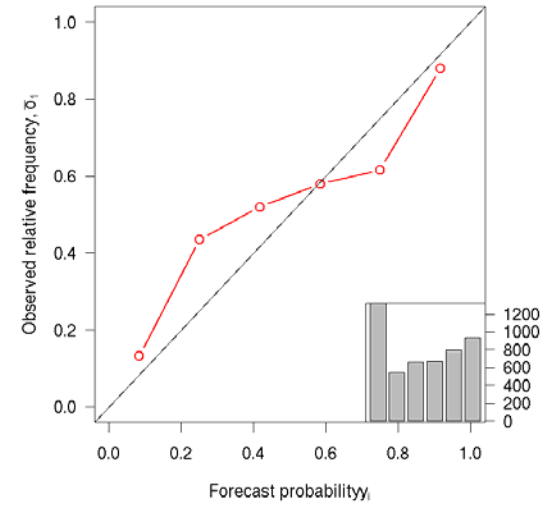
10 bins



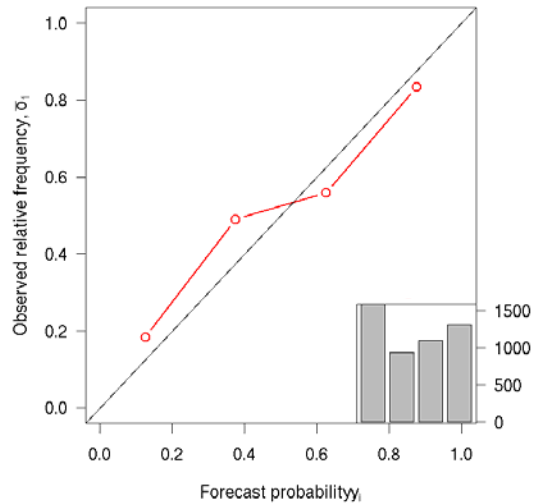
8 bins



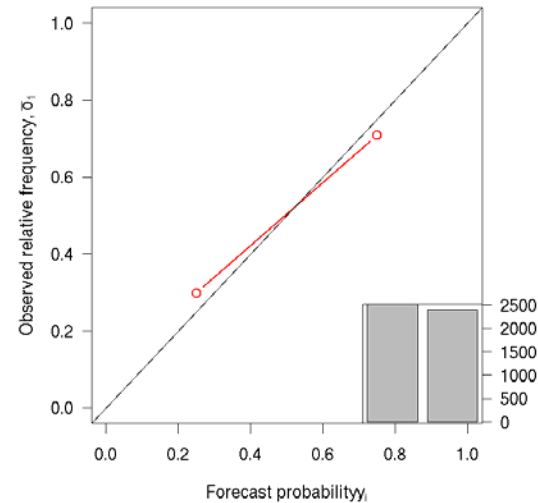
6 bins



4 bins

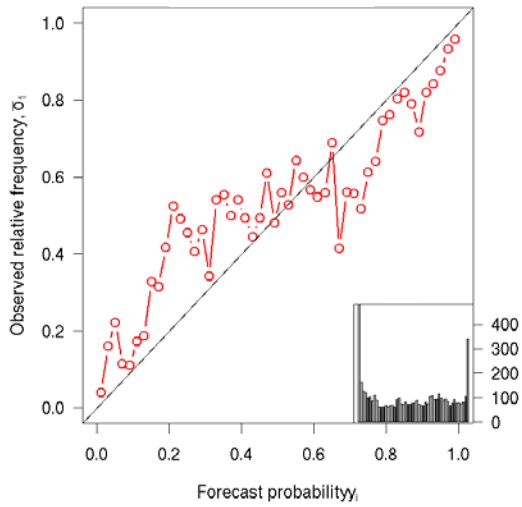


2 bins

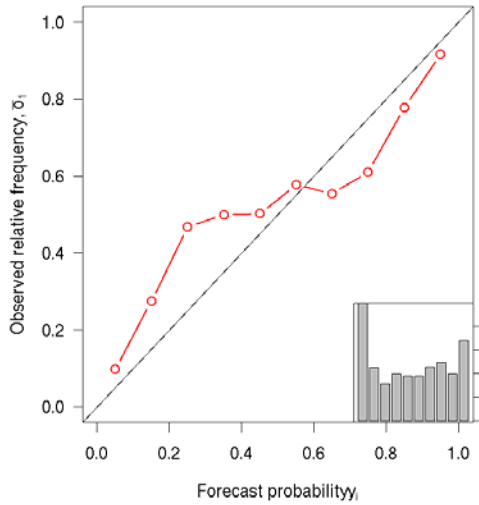


→ Forecast system is over-confident

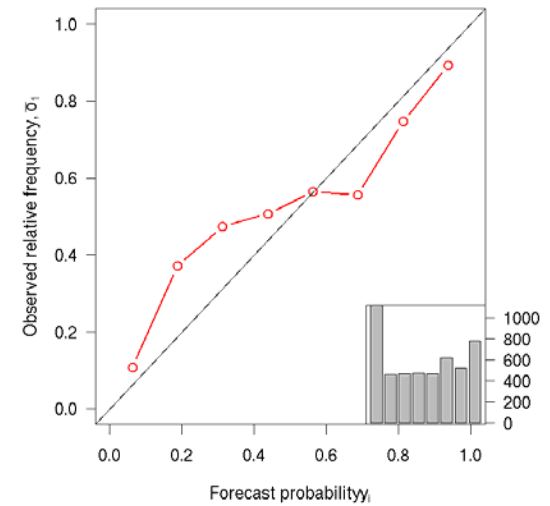
50 bins



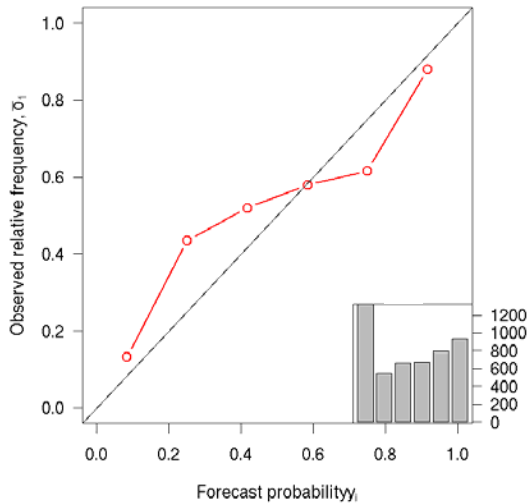
10 bins



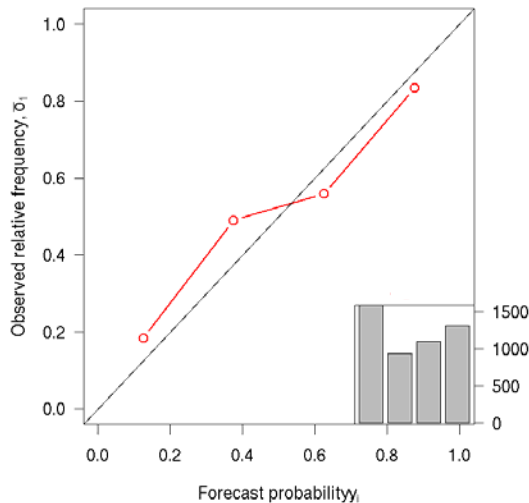
8 bins



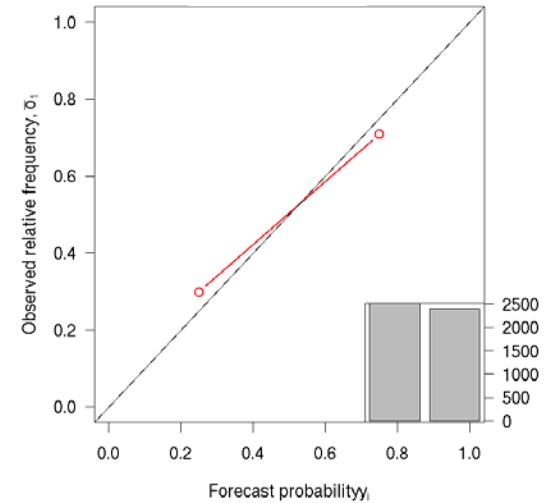
6 bins



4 bins



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