

# The effect of ensemble size on verification measures for binary-event forecasts

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- ▶ Brier score
- ▶ Effect of ensemble size
- ▶ RPS, Logarithmic and ROC scores

# Brier score definition

Times  $t = 1, \dots, n$

Binary observations  $I_t = 0$  or  $1$

Probabilistic forecasts  $P_t$

Brier score

$$B = \frac{1}{n} \sum_{t=1}^n (P_t - I_t)^2$$

Brier (1950, MWR)

# Brier score definition

Times  $t = 1, \dots, n$

Binary observations  $I_t = 0$  or  $1$

Probabilistic forecasts  $P_t =$  proportion of ensemble members that forecast the event

Brier score

$$B(m) = \frac{1}{n} \sum_{t=1}^n (P_t - I_t)^2$$

for ensemble size  $m$ .

Brier (1950, MWR)

# Ensemble size effect

- Assume
1. data are stationary
  2. ensemble members are exchangeable

Expected Brier score

$$E[B(m)] = E[B(\infty)] + \frac{A}{m}$$

Richardson (2001, QJRMS)

$A$  measures sharpness

$$A \propto \frac{1}{4} - E \left[ \left( P - \frac{1}{2} \right)^2 \right]$$

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## Estimate Brier score for other ensemble sizes

Given ensemble size  $m$ , estimate

$$\begin{aligned} E[B(M)] &= E[B(\infty)] + \frac{A}{M} \\ &= E[B(m)] - \frac{A}{m} + \frac{A}{M} \end{aligned}$$

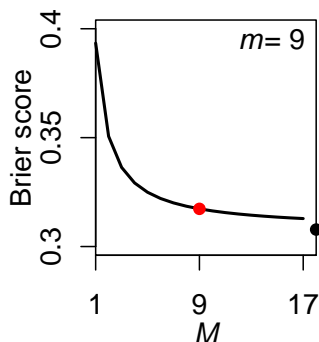
Unbiased estimator for  $E[B(M)]$  is

$$B(m) - \frac{M - m}{M(m - 1)n} \sum_{t=1}^n P_t(1 - P_t)$$

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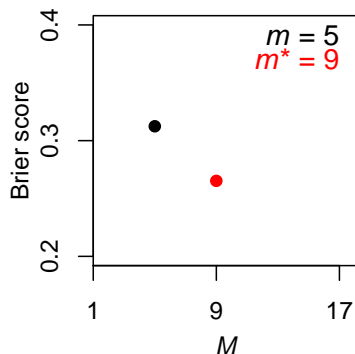
# Compare Brier scores

Compare two sets of forecasts:

$P_t$  with ensemble size  $m$

$P_t^*$  with ensemble size  $m^*$

Identical verifying observations



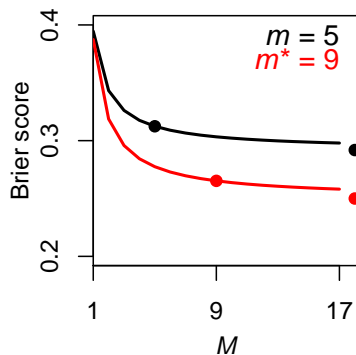
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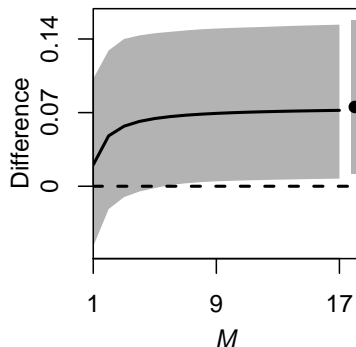
$P_t$  with ensemble size  $m$

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Identical verifying observations

Bootstrap confidence interval for

$$E[B(M)] - E[B^*(M)]$$



## Other scores

- ▶ Multi-category Brier / Ranked probability scores

$$\frac{1}{K} \sum_{k=1}^K B_k(m)$$

cf. Müller et al. (2005, JCLim)

- ▶ Logarithmic score

$$S(m) = - \sum_{t=1}^n [I_t \log P_t + (1 - I_t) \log(1 - P_t)]$$

- ▶ Area  $R(m)$  under ROC curve
- ▶ Unbiased estimators for  $E[S(M)]$  and  $E[R(M)]$  iff  $M \leq m$

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# Conclusion

- ▶ Easy to estimate effect of ensemble size on Brier scores
- ▶ Can use confidence intervals to compare scores
- ▶ Also possible for RPS, Logarithmic and ROC scores

Paper and R code available at

[www.met.rdg.ac.uk/~sws02caf](http://www.met.rdg.ac.uk/~sws02caf)

[c.a.t.ferro@reading.ac.uk](mailto:c.a.t.ferro@reading.ac.uk)