## II. Frequentist probabilities

## II. 2 Verification of probability forecasts

# II.2.1 What is a good probability forecast? 

Probability The average evolution of probability values


## Observed rainfall (radar) Probability forecast 1 Probability forecast 2



A scientist at a meeting showed these images: The radar observed rain fall and the probability forecast from system 1

And then he showed the forecasts from system 2 Are they worse??

## Answer: -We cannot say

## 1.It is only one forecast

2.If rain only fell when the probabilities were > 40\% and not when they were below, something is wrong

## The reliability diagram




Good sharpness $=$ forecasts draw towards $\mathbf{0 \%}$ and $\mathbf{1 0 0 \%}$

## Good reliability, but poor sharpness




## Do not confuse "sharpness" and "resolution"

## The resolution



## II.2.2 The Brier Score (BS)

## The Brier score



## BS and RMSE have identical mathematical structures



# The notation of the Brier score can be simplified as with RMSE 

$$
\begin{aligned}
& B S=\overline{(p-o)^{2}} \\
& E^{2}=\overline{(f-a)^{2}}
\end{aligned}
$$

## II.2.3 Decomposition of the Brier score

Two alternatives, Murphy (1983) which is very quoted but rarely used, or one similar to the RMSE decomposition

## The Non-Murphy decomposition is identical to the RMSE one:

$B S=\overline{(p-o)^{2}}=$

$$
\overline{(p-\bar{o})^{2}}+\overline{(o-\bar{o})^{2}}-\overline{2(p-\bar{o})(o-\bar{o})}
$$



## Atmospheric variability



## "Sharpness" Model variability

## $(p-\bar{o})^{2}$



## "Reliability" or "skill"

$$
\overline{2(p-\bar{o})(o-\bar{o})}
$$

|  | $0-\bar{o}$ | $1-\bar{o}$ |
| :---: | :---: | :---: |
| $p_{1}-\bar{o}$ | 45 | 5 |
| $p_{2}-\bar{o}$ | 7 | 3 |
| $p_{3}-\bar{o}$ | 5 | 5 |
| $p_{4}-\bar{o}$ | 3 | 7 |
| $p_{5}-\bar{o}$ | 2 | 18 |

# ... 01 <br> <br> graphically <br> <br> graphically <br> $\overline{2(p-\bar{o})(o-\bar{o})}$ <br> <div class="inline-tabular"><table id="tabular" data-type="subtable">
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</tr>
</tbody>
</table>
<table-markdown style="display: none">| Probability |
| :---: |
| forecasts not |
| followed by rain |</table-markdown></div> <br>  

## II.2.4 Alan Murphy's decomposition

## Brier Score decomposition

$$
B S=\frac{1}{N} \sum_{k=0}^{M} N_{k}\left(f_{k}-\bar{o}_{k}\right)^{2}-\frac{1}{N} \sum_{k=0}^{M} N_{k}\left(\bar{o}_{k}-\bar{o}\right)^{2}+\bar{o}(1-\bar{o})
$$

The first term is a reliability measure:

For perfectly reliable forecasts, the subsample relative frequency is exactly equal to the forecast probability in each sub-sample.

$$
B S=\frac{1}{N} \sum_{k=0}^{M} N_{k}\left(f_{k}-\bar{o}_{k}\right)^{2}-\frac{1}{N} \sum_{k=0}^{M} N_{k}\left(\bar{o}_{k}-\bar{o}\right)^{2}+\bar{o}(1-\bar{o})
$$



The second term is a resolution measure:

$$
B S=\frac{1}{N} \sum_{k=0}^{M} N_{k}\left(f_{k}-\bar{o}_{k}\right)^{2}-\frac{1}{N} \sum_{k=0}^{M} N_{k}\left(\bar{o}_{k}-\bar{o}\right)^{2}+\bar{o}(1-\bar{o})
$$

The uncertainty term ranges from 0 to 0.25. If the event either always occurs or never occurs, then there is high certainty. With a 50-50 probability it is most uncertain

$\bar{o}=0.3$ yields less "uncertainty" ( 0.21 ) than $\bar{o}=0.5(0.25)$
Compare with a sack of balls with two colours with proportions $\bar{o}$ and $1-\bar{o}$ where "certainty" $=\bar{o}^{2}+(1-\bar{o})^{2}$

## II.2.5 Pitfalls with the Brier Score

The BS will appear to improve if the sharpness gets worse.
A contest between a real and fake doctor trying to forecast the sex of not yet born children.

The fake doctor will score BS = 0.5 just by guessing.

## If the real doctor is $65 \%$ correct in his forecasts he will score BS = 0.35.

By saying "fifty-fifty" in 60\% of the cases the fake doctor can "improve" his score to exactly the same $\mathrm{BS}=0.35$.
$B S=1$ in $50 \%$ and
$B S=0$ in $50 \%$
$B S=1$ in $35 \%$ and $B S=0$ in $65 \%$
$B S=1$ in 20\%
BS = 0 in 20\%
$B S=0.25$ in $60 \%$
$0.2+0.15=0.35$


# The guessing hoaxer's reliability diagram. Brier score $=\mathbf{0 . 5 0}$ 

The rather skilful ( $60 \%$ hit rate) scientist's reliability diagram. Brier score $=\mathbf{0 . 3 5}$

The hoaxer's "improved" reliability diagram. He is still guessing but the Brier score has decreased to $\mathbf{0 . 3 5}$ since he increased a "useless" reliability

## I.2.6 Extensions of the Brier Score

## Brier skill score

A new "Brier Skill Score" with climate as reference would be

$$
B S S_{n e w}=\frac{\overline{(p-\bar{o})(o-\bar{o})}}{\overline{(o-\bar{o})^{2}}}
$$

In analogy with the Anomaly Correlation Coefficient

$$
A C C=\frac{\overline{(f-c)(a-c)}}{\overline{(a-c)^{2}}}
$$

Instead the following definition has been agreed

## Brier skill score

## $B S S=\left(B S_{\text {ref }}-B S\right) / B S_{\text {ref }}$

## Rank probability score (RPS)

## It is just the Brier score <br> $$
B S=\overline{(p-o)^{2}}
$$

applied for different thresholds, defining new probabilities, and then integrating or summing up

## END

