# **Evaluating model skill: what is the half-life of a cloud-fraction forecast?**



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### 3. Desirable properties of a skill score

- *Equitable:* random forecasts have expected score of zero. This is essential!
- **Transpose symmetric:** no change if swap model and observations. Asymmetric scores tend to be *improper*; they can be *hedged* by over- or underestimating the frequency of occurrence.
- **Uses full range of cloud fraction**: better than assessing just when fraction exceeds a threshold.
- **Useful for rare events:** Most scores tend to meaningless limit as frequency of occurrence  $\rightarrow 0$ .
- *Linear:* To calculate a half-life, score must depend on the inputs in a reasonably linear fashion.

Score	Equitable as $n  ightarrow 0$	<b>Transpose</b> symmetric	Uses full range	Useful for rare events	Linear	Definition and notes
Hit rate, <i>H</i> False alarm rate, FAR	N	N	N	N	Y	H = a/(a+c) FAR = b/(a+b) These and other non-equitable scores used by Mace et al. (1998) for cloud evaluation
Heidke Skill Score, HSS	Y	Y	N	N	Y	Define number of correct forecasts $x = a+d$ , then define HSS to vary linearly between 0 for a random forecast and 1 for a perfect forecast using HSS = $(x-x_{random})/(x_{perfect}-x_{random})$
Log of Odds Ratio, $ln\theta$	Y	Y	N	~	~	<b>Inθ = In(<i>ad/bc</i>)</b> Analyzed by Stephenson (2000); property that a perfect forecast scores infinity
Yule's Q (also known as Odds Ratio Skill Score)	Y	Y	N	N	N	<b>Q = (<i>ad–bc</i>)/(<i>ad+bc</i>) = (θ–1)/(θ+1)</b> Equivalent to Inθ, but bounded to 0-1 at the expense of being very non-linear
Mean Absolute Error Skill Score, MAESS	Y	Y	Y	N	Y	As HSS but with $\mathbf{x} = \Sigma  f_{model} - f_{obs}  / n$
Extreme Dependency Score, EDS	N	N	N	Y	~	<b>EDS = 2ln[(<i>a</i>+<i>c</i>)/<i>n</i>]/ln(<i>a</i>/<i>n</i>) –1, where <i>n</i> = <i>a</i>+<i>b</i>+<i>c</i>+<i>d</i> Shown by Stephenson et al. (2008) to tend to a meaningful limit for rare events</b>
Symmetric Extreme Dependency Score, SEDS	Y	Y	N	Y	~	SEDS = {In[(a+b)/n]+In[(a+c)/n]}/In(a/n)-1 = In(a <sub>r</sub> /a)/In(a/n), where a <sub>r</sub> is expected a for random forecast. Desirable properties of EDS plus transpose symmetry & equitability

But this only tests model climatology, how can we test whether clouds were forecast in the right place?

Plot scores against HSS for particular frequencies of occurrence in observations, p, and model,  $p_m$ :



2. Joint probability distributions

Consider "DWD-EU" 7-km model over Murgtal in 2007:





Simplify information by defining a contingency table:

	Observed cloud ( $f_{obs} > f_{thresh}$ )	Observed clear-sky ( $f_{obs} \leq f_{thresh}$ )
Forecast cloud ( $f_{model} > f_{thresh}$ )	a: Number of cloud hits	<i>b:</i> Number of false alarms
Forecast clear-sky ( $f_{model} \leq f_{thresh}$ )	<i>c:</i> Number of misses	d: Number of clear-sky hits

#### Threshold cloud fraction 0.1 10000



## 5. Skill versus height

### 6. Estimating forecast "half life"

SEDS over first 1.5 days fitted by an inverse exponential

- DWD half-life **2.87 days** in 2004 and **3.15 days** in 2007
- Met Office half-life **2.91 days** in 2004 and **3.07** in 2007 DWD forecasts available out to 3 days
- DWD half-life **4.31 days** for 1.5-3 day forecasts
- Forecast skill at short range dominated by convective timescales, at long range by large-scale weather systems Half-life for ECMWF 500-mb geopotential height is **9 days**



### Mid-level clouds most skilfully forecast

- Surprising: physics of mixed-phase clouds not represented well
- Large-scale ascent has largest amplitude in mid-troposphere so cloud response most strong here?



- Met Office performs best: arguably the most sophisticated microphysics with separate liquid and ice Boundary layer clouds least skilfully forecast
- Not a surprise: well-known forecasting problem
- Occurrence a subtle function of subsidence, surface fluxes, entrainment, stability, drizzle formation...





### Lead time (hours)

#### Temporal averaging

- Absolute skill and half-life increase with temporal averaging
- Larger-scale features more predictable
- Full results presented by Hogan et al. (2009)

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