

Achieving seamless verification across sub-seasonal time scales from weather to climate

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How we (still) do forecasting

- Traditional weather forecasts – day-by-day, deterministic or probabilistic, out to about 7 days with current models

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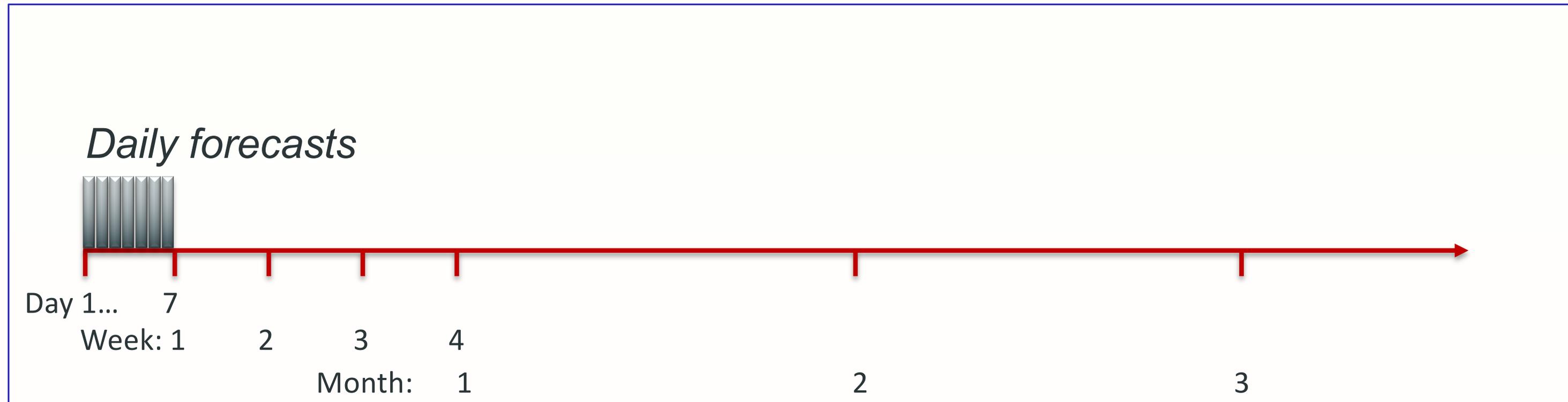
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 - Usually step by weeks, months, etc.

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- Arbitrarily we jump to time-averages after some lead time
 - Usually step by weeks, months, etc.
- These divisions dictated by the construct of our calendar, not the nature of phenomena – **Is this the best? Can we be more flexible?**

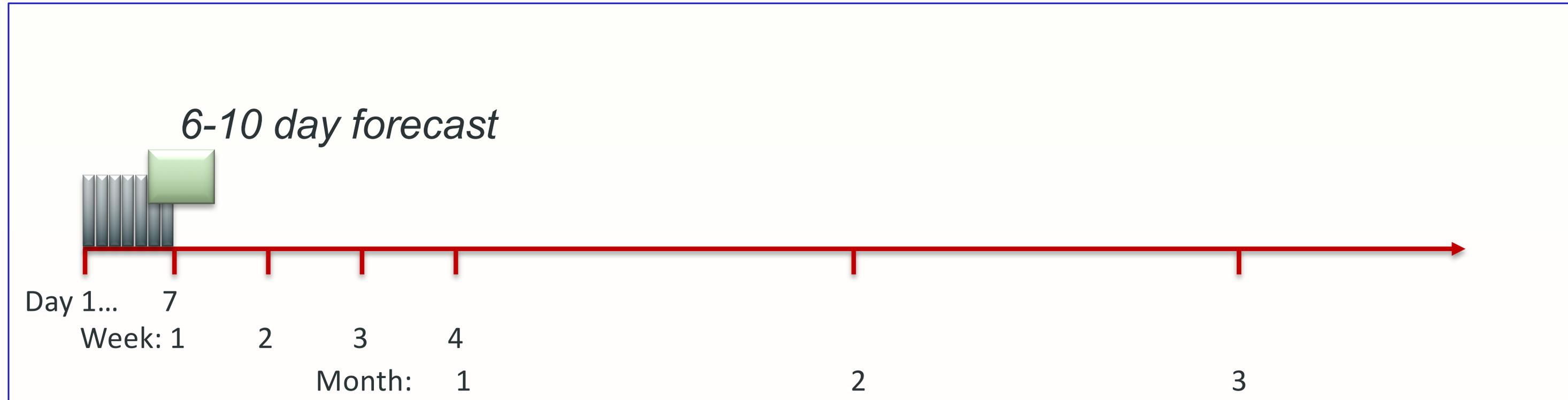
For example: NCEP

- The longer the forecast lead, the wider the “window”.



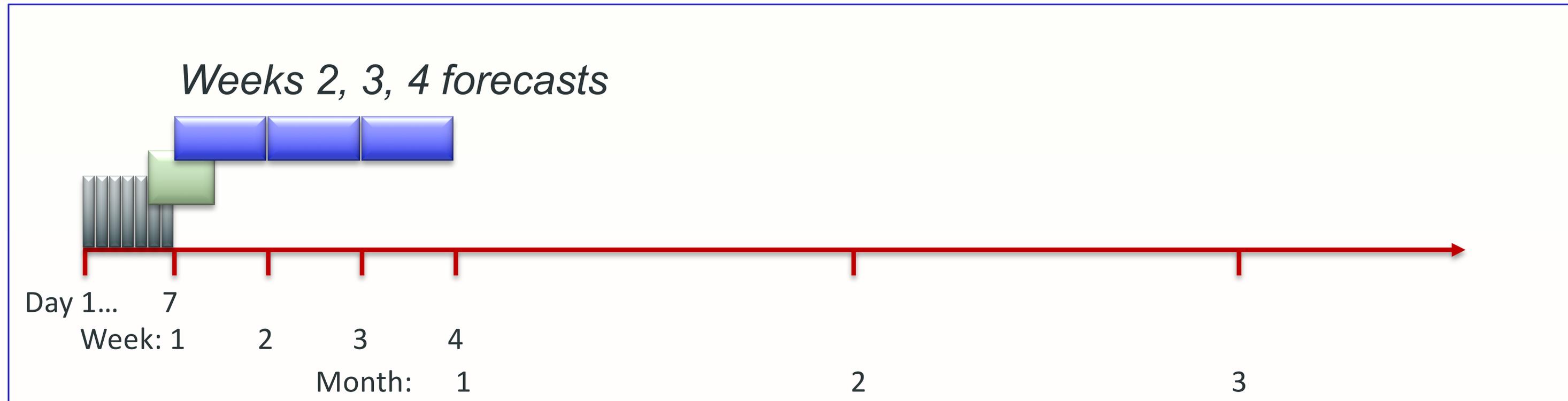
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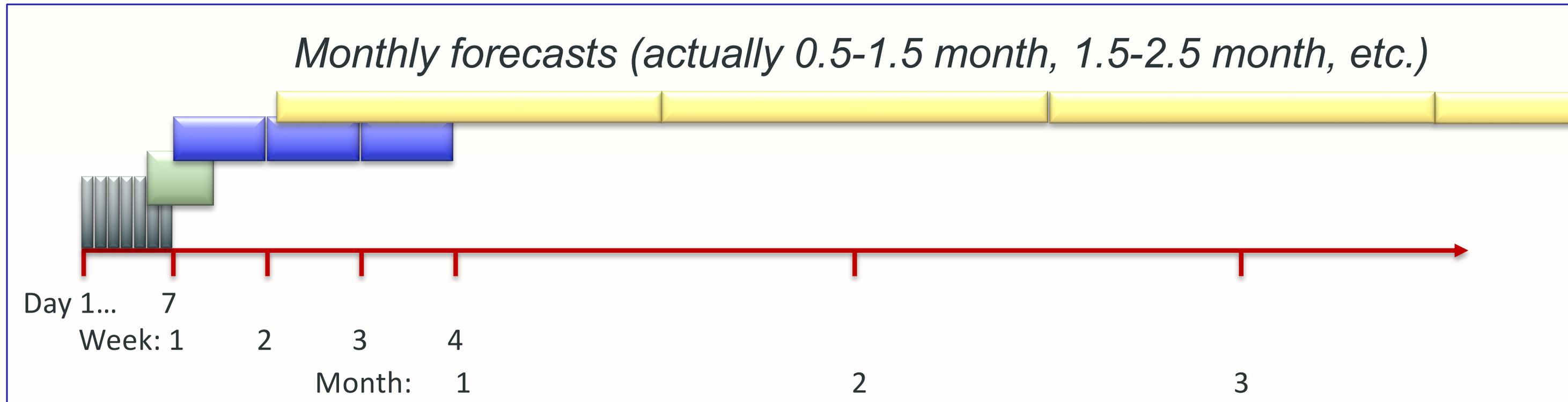
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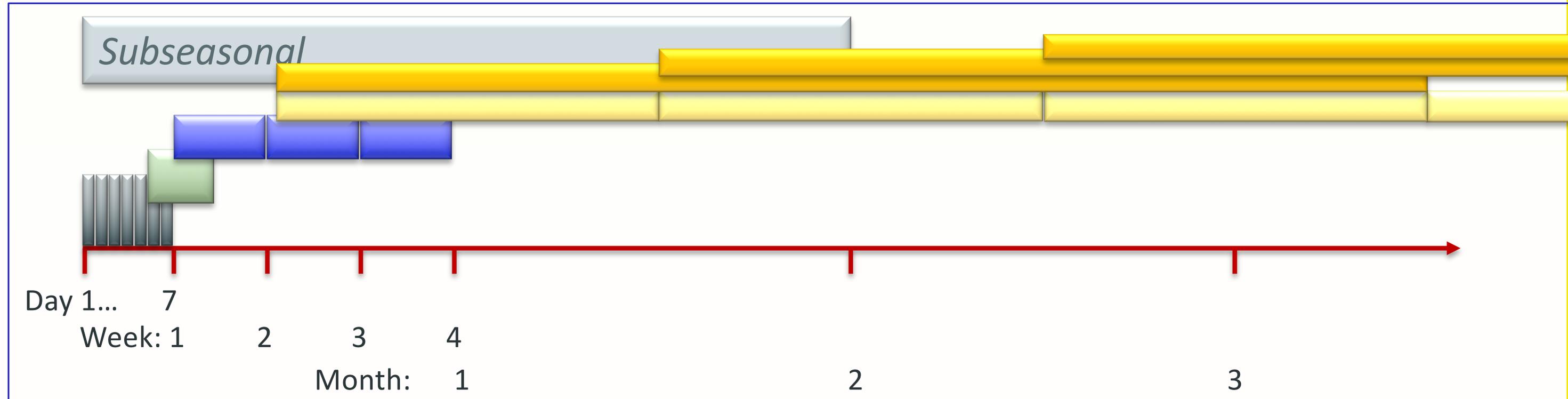
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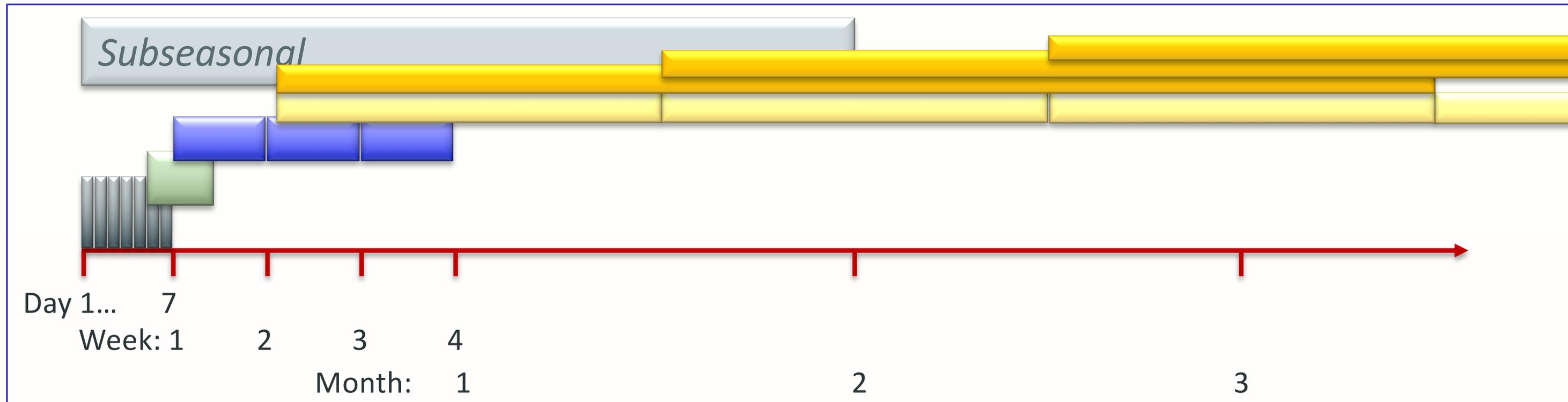
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- This is intuitive – we measure error relative to the distance to the target.
- Begs the question: what constitutes a successful forecast?

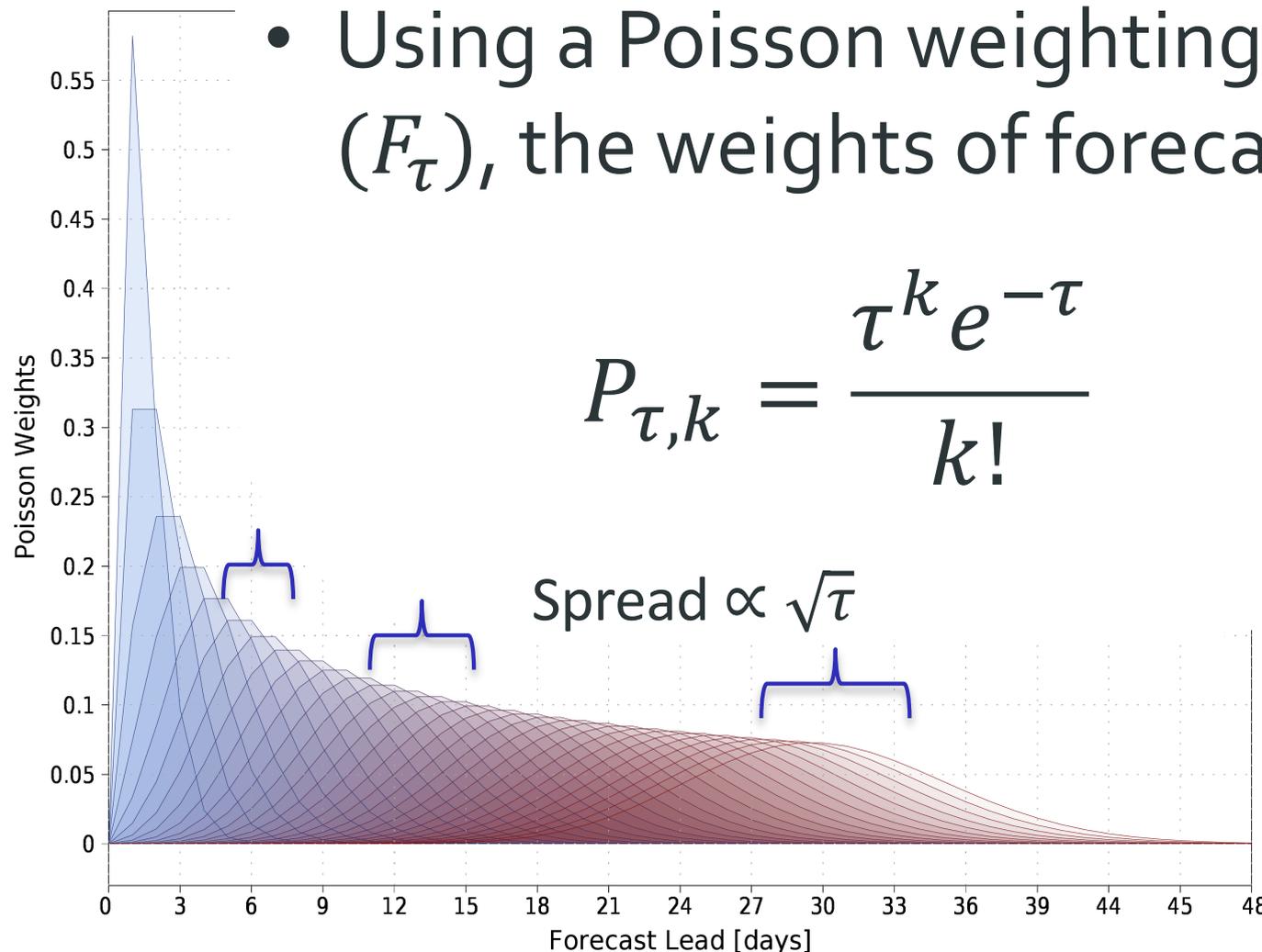
Seamless Validation (Poisson Weighting)

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- Using a Poisson weighting in time for a forecast at lead τ : (F_τ), the weights of forecast daily means at each lead k (A_k):



$$P_{\tau,k} = \frac{\tau^k e^{-\tau}}{k!}$$

$$F_\tau = \frac{\sum_{k=1}^N A_k P_{\tau,k}}{\sum_{k=1}^N P_{\tau,k}}$$

Ford, T. W., P. A. Dirmeyer and D. O. Benson, 2018: *npj Climate Atmos. Sci.*, doi: 10.1038/s41612-018-0027-7.

Weakness of Poisson weighting approach

- The Poisson function has an advantageous shape and behavior for this approach to seamless forecasting, but it could be better.
 - Not flexible – its shape is what it is.
 - Even at 1-day lead, it is a blend of forecasts from several lead times, not a deterministic forecast only for day 1.

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 - Not flexible – its shape is what it is.
 - Even at 1-day lead, it is a blend of forecasts from several lead times, not a deterministic forecast only for day 1.
- Would like to have an approach that **transitions smoothly from day-to-day forecasts to a time-average** (Poisson → Normal distribution).

Kronecker delta

- The area under the curve of the Poisson function always equals 1, making it ideal as a weighting function.
- The Kronecker delta:

$$\delta_{\tau,k} = \begin{cases} 0 & \tau \neq k \\ 1 & \tau = k \end{cases}$$

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- Thus, a **linear combination of Poisson and Kronecker weighting functions**, with weights for each function that sum to 1, is well behaved as a versatile, compound weighting function.

Blending $P_{\tau,k}$ and $\delta_{\tau,k}$

- We want the deterministic forecast represented by $\delta_{\tau,k}$ to last for some number of days before transitioning to the $P_{\tau,k}$ weighting.
- The **2-parameter Hill equation** is ideal for this:

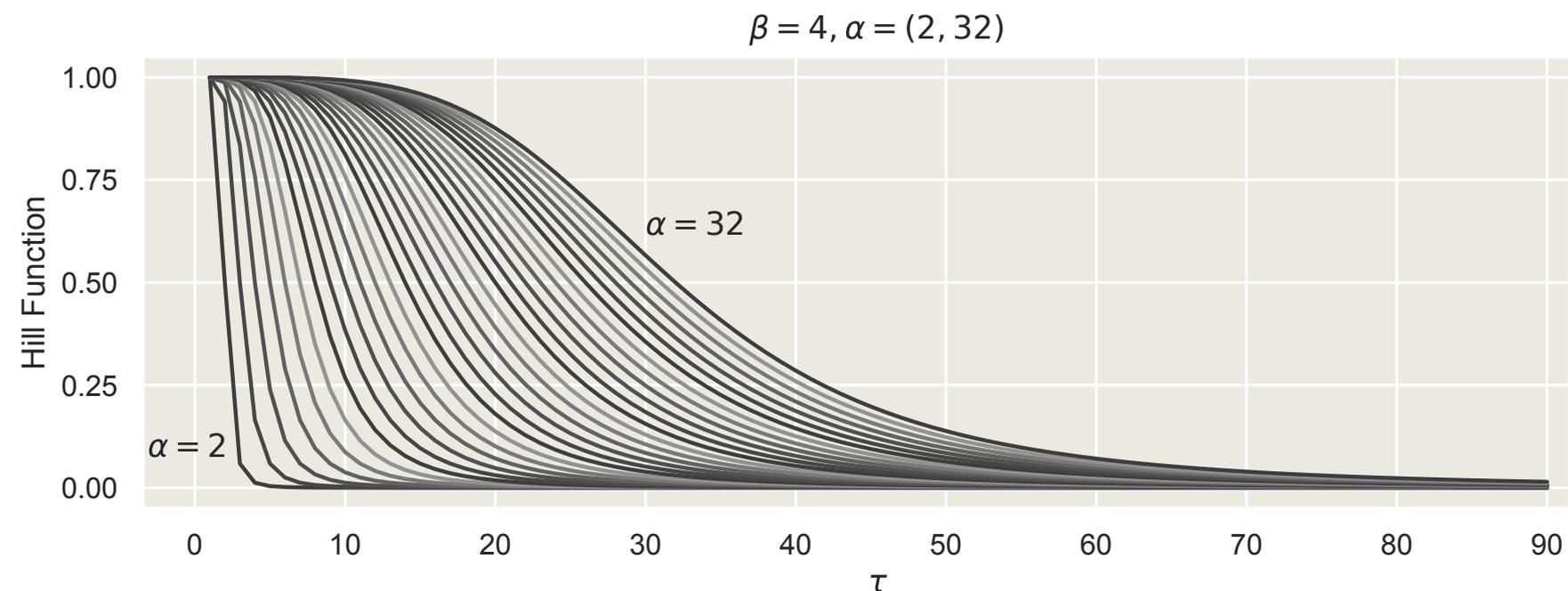
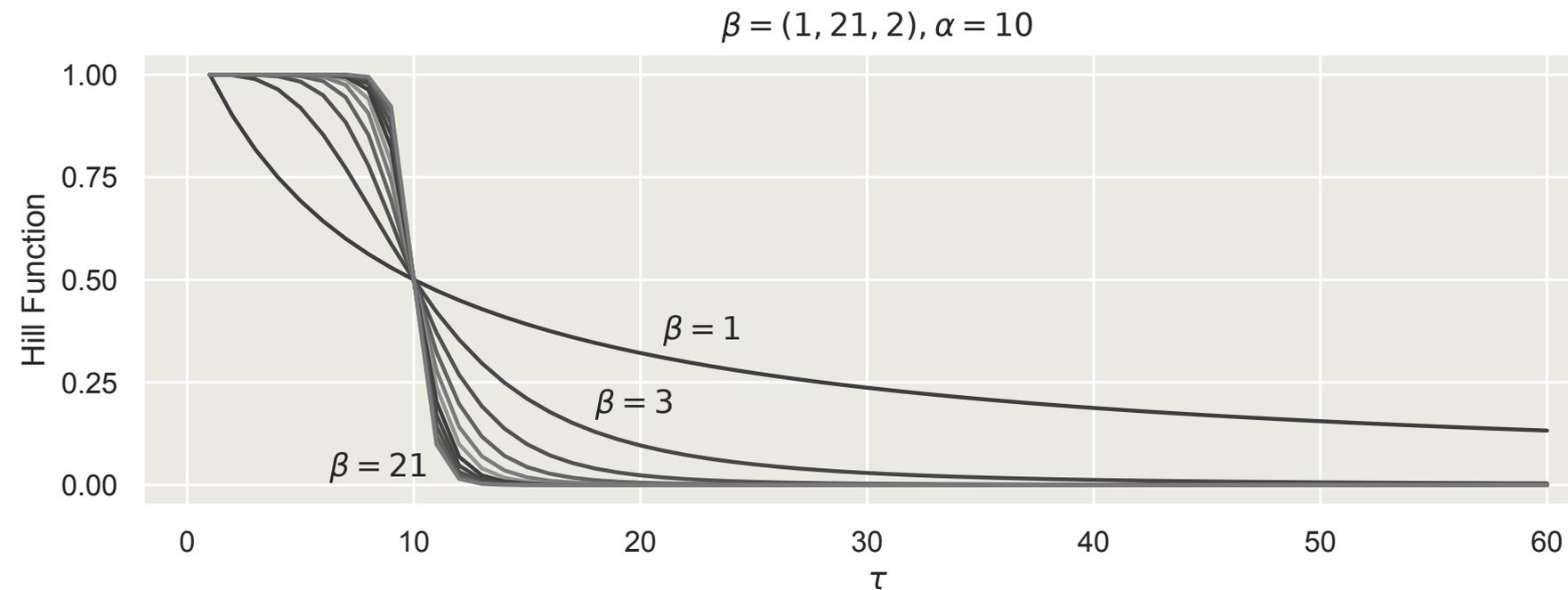
$$H_{\tau} = \frac{1}{\left(\frac{\tau - 1}{\alpha - 1}\right)^{\beta} + 1}$$

- α determines the transition point (50/50 weighting between $P_{\tau,k}$ and $\delta_{\tau,k}$)
- β gives the abruptness of the transition.
- The compound weight is: $W_{\tau,k} = H_{\tau}\delta_{\tau,k} + (1 - H_{\tau})P_{\tau,k}$

Hill, A. V., 1910: *J. Physiol.*,
doi:10.1113/jphysiol.1910.sp001386

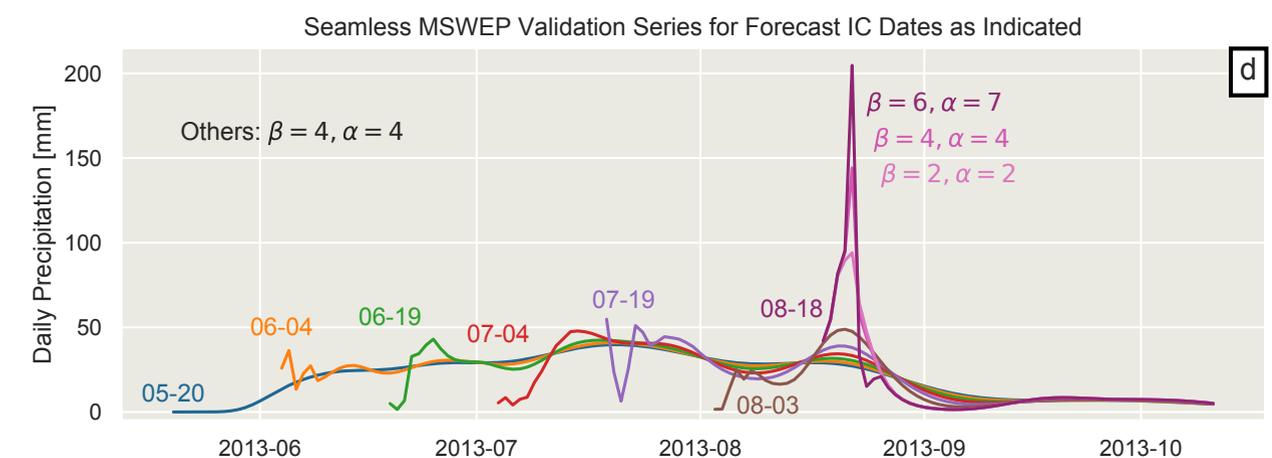
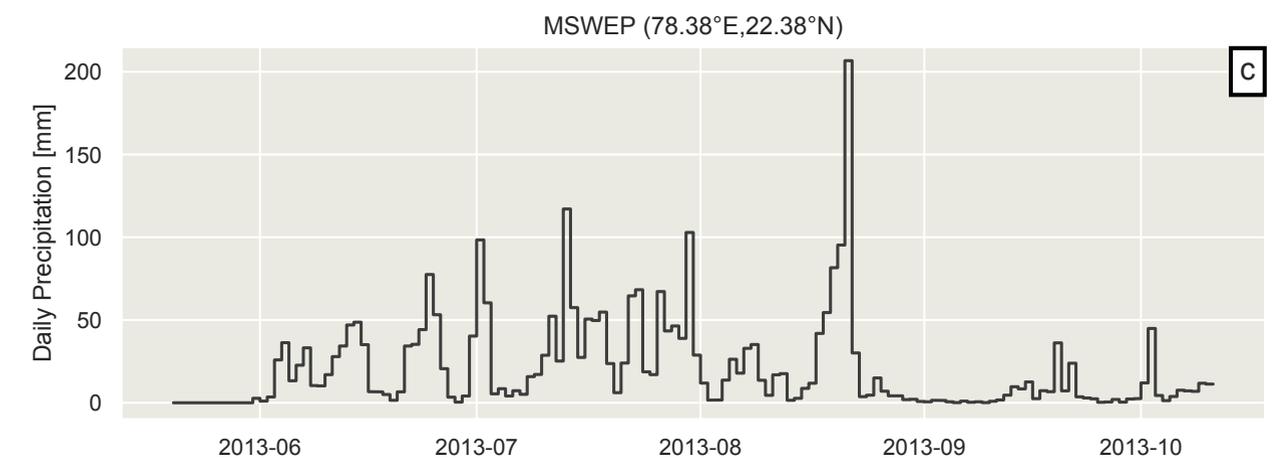
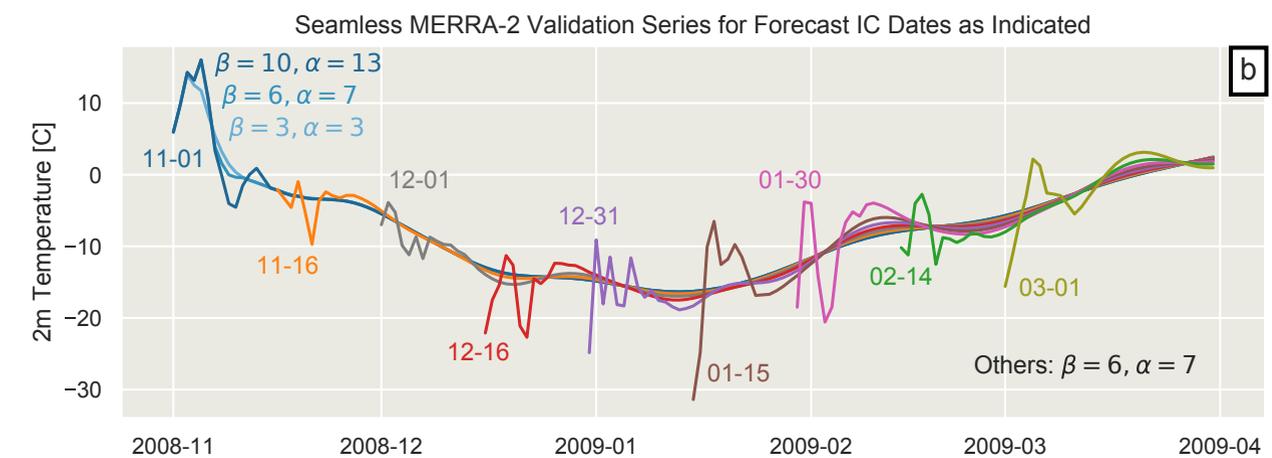
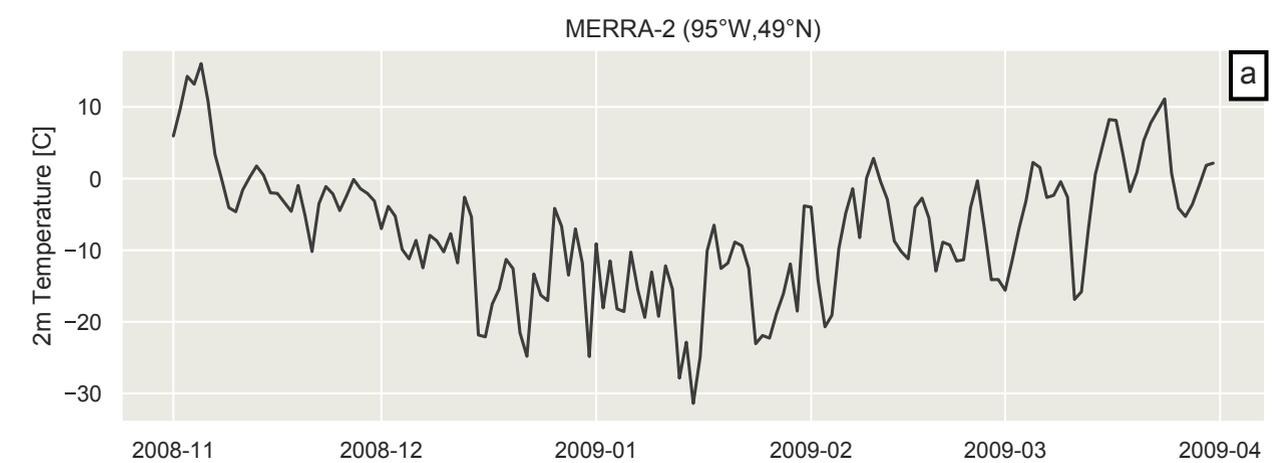
Hill equation

- Low β gives smooth transition with lead time, large β yields a sharp transition.
- α is the lead time at which the Hill equation has a value of 0.5. For any given β , larger α also produces a smoother transition.
- α and β can be chosen to give desired effect.



Seamless validation

- Validation becomes more complicated, as there are two time dimensions: validation time and lead (or initialization) time.
 - This should be done anyhow for forecasts with any models that drift (i.e., all dynamical models).
- MERRA-2 temperature (top) and MSWEP precipitation (bottom) examples show how validation time series smooth out with lead time.

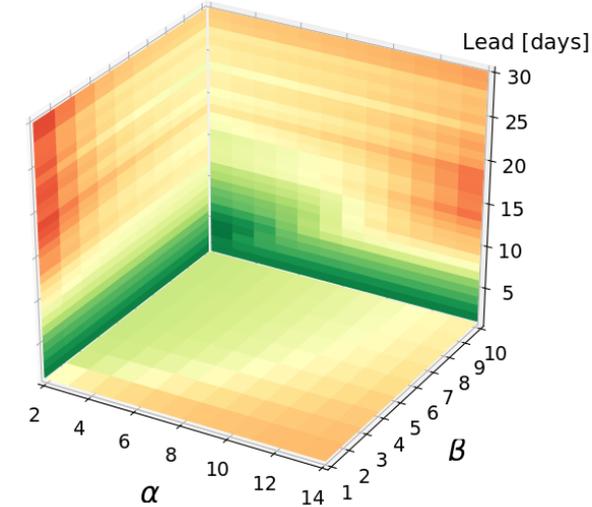


Skill(α, β, τ)

- Obviously skill tends to decrease with lead time, but the transition to time scales means that is not always the case.
- The choice of α and β that gives the best skill scores is not usually the most useful – the goal is not to make models look good.

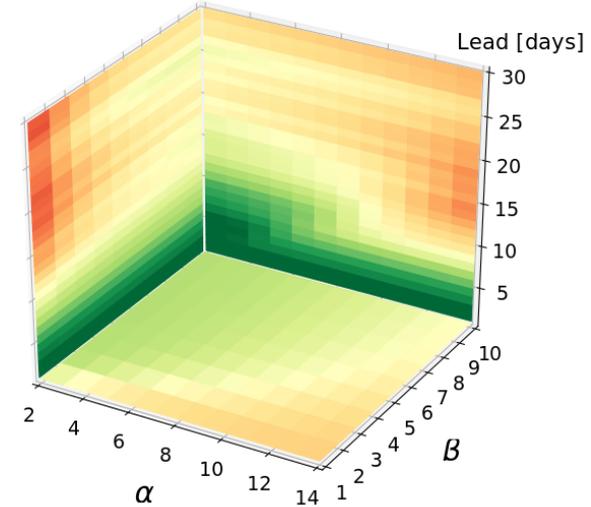
Anomaly Correlation Coefficient (DJF)

Daily Maximum Temperature

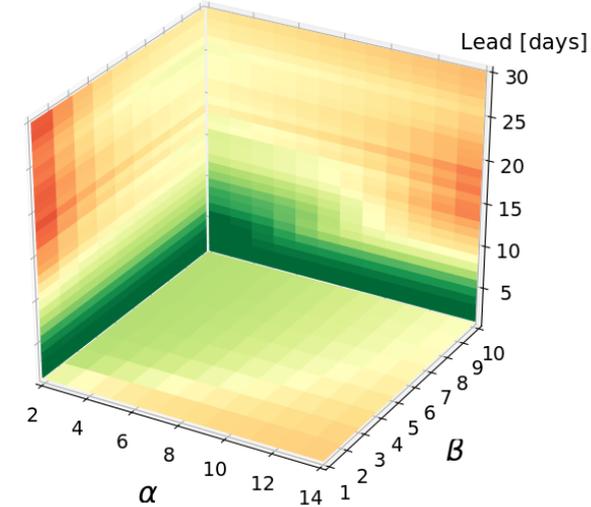


Minneapolis [45.0°N,93.0°W]

Daily Minimum Temperature

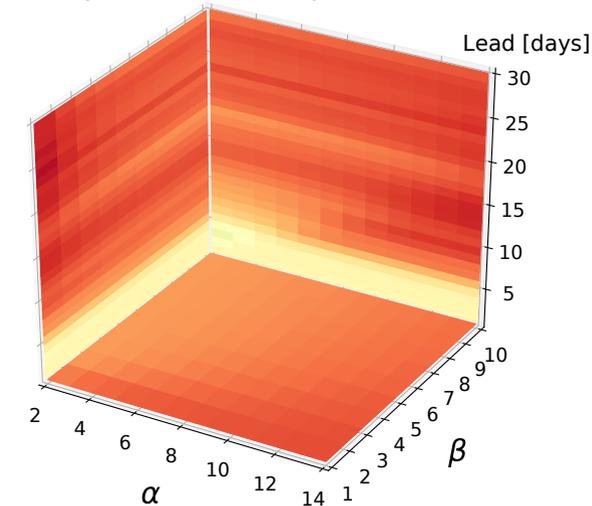


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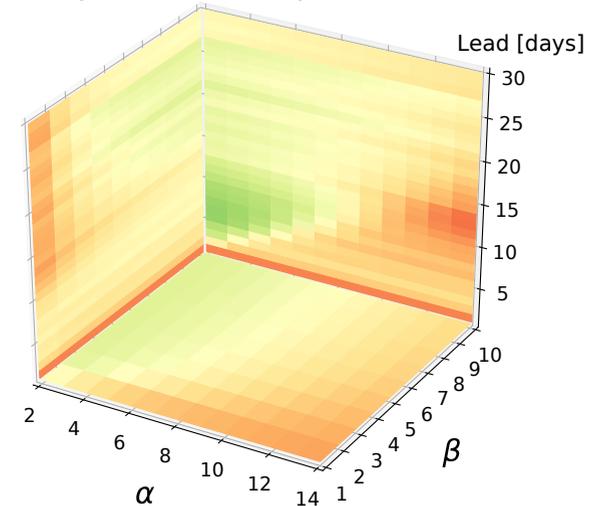
Anomaly Correlation Coefficient (JJA)

Daily Maximum Temperature

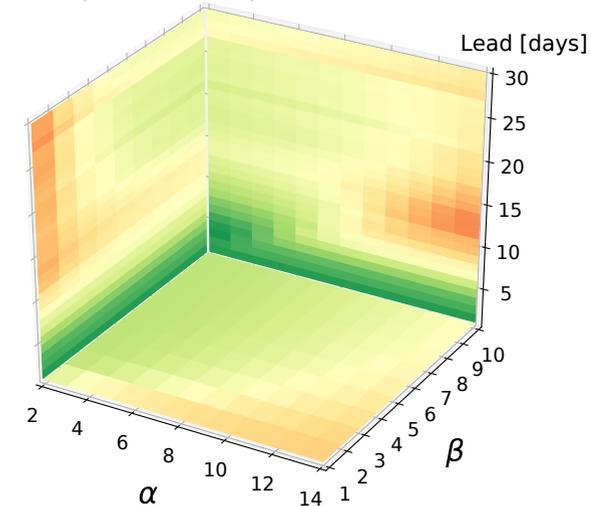


Atlanta [33.0°N,84.0°W]

Daily Minimum Temperature



Daily Mean Temperature

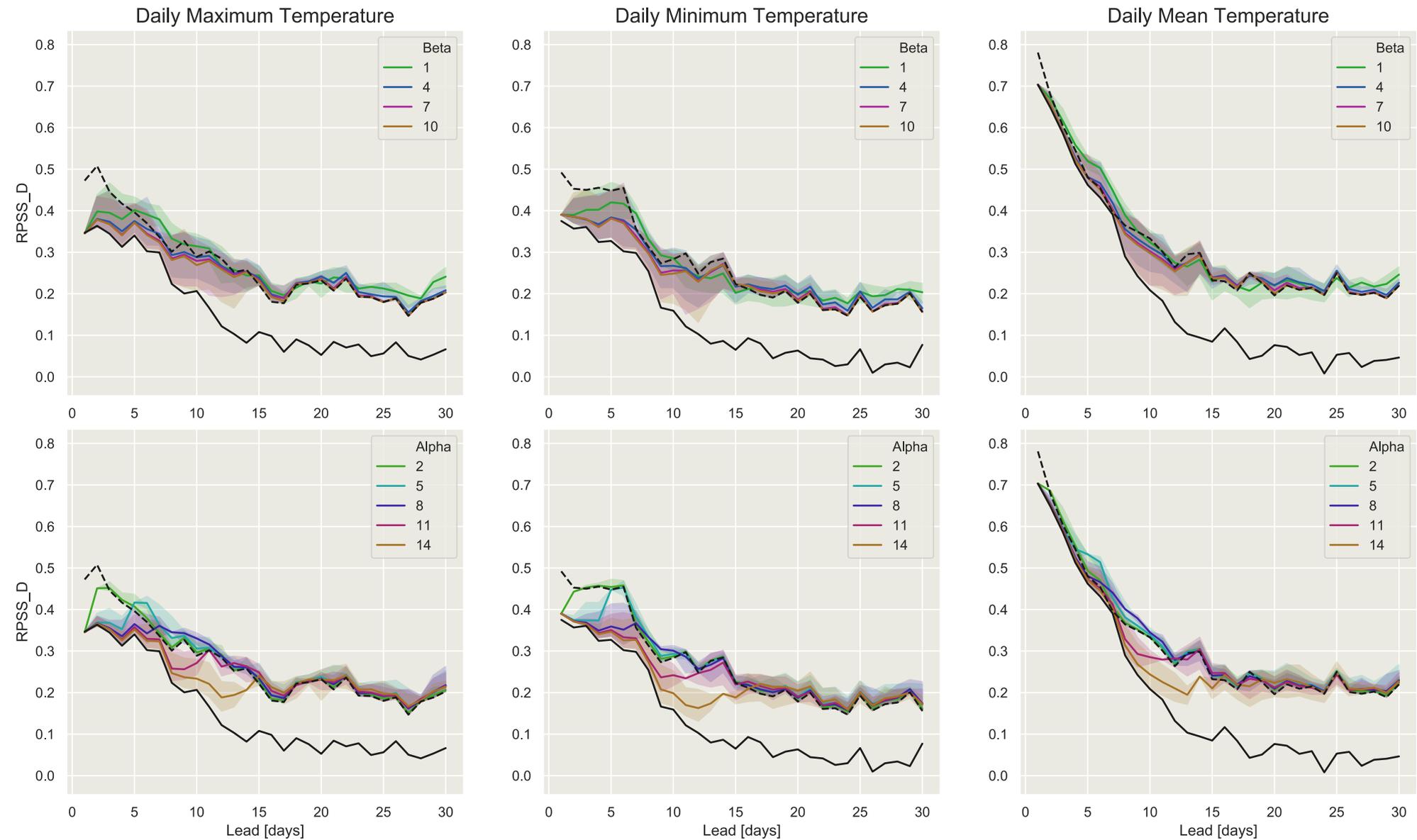


CMA forecast skill (ACC) [1999-2010], color shows arithmetic mean along the perpendicular dimension.

Ensembles

- Approach can be used for ensemble-based skill metrics as well.
- See most of the sensitivity to choices of the Hill equation parameters is in the 1-3 week range.

Minneapolis [45.0°N,93.0°W]

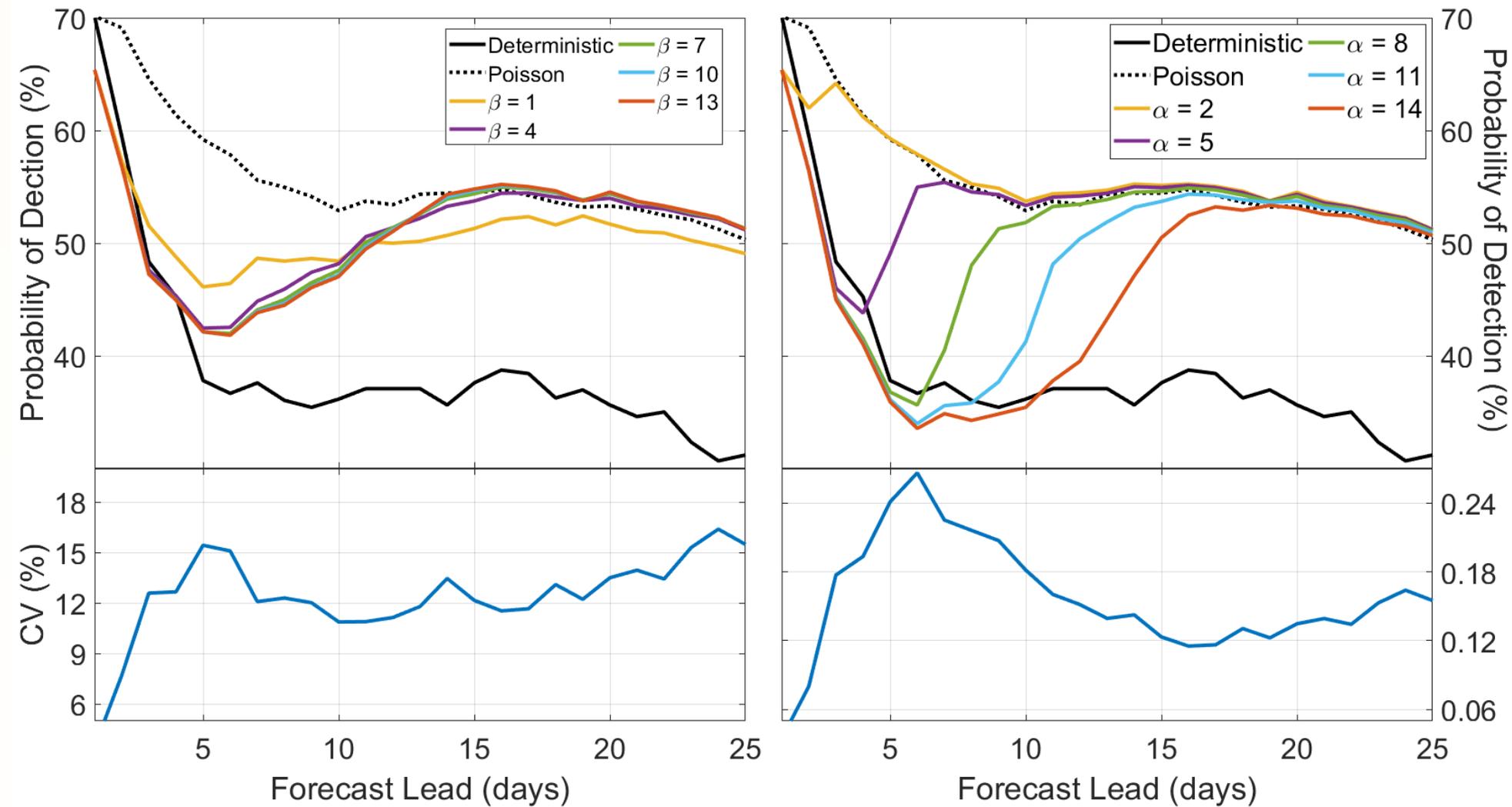


NCEP CFS $RPSS_D$ [1999-2010]; Black line = purely deterministic forecasts, dashed line = purely Poisson-weighted forecasts. Colored lines and spreads for different α and β are across all values of β and α respectively.

Heat waves

- Event-based statistics require a flexible means of definition for events that can vary with window.
- A forecast 12 days in advance for an event that occurs on day 11 or 13 should not be penalized – it is a useful forecast.

Minneapolis [45°N, 93°W]



NCEP GEFS probability of detection (POD) [1999-2016]; Black line = purely deterministic forecasts, dashed line = purely Poisson-weighted forecasts.

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- Highly tailorable for different applications, user interests.
- Linear combination of Kronecker and Poisson-weighted forecasts can be done a posteriori with various choices of α and β .
- Requires a more complicated observed climatology to calculate anomalies: two time dimensions.
- Open question whether there is an objective approach to optimize the choice of the parameters.
- Large values of β reintroduce the “seam”.

Dirmeyer & Ford, 2019: *Mon. Wea. Rev.*, submitted.