



On the value of reforecasts for the TIGGE database

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Motivation

- One goal of TIGGE is to investigate whether multi-model predictions are an improvement to single model forecasts
- The goal of using reforecasts to calibrate single model forecasts is to provide improved predictions
- Questions:
 - What are the relative benefits (costs) of both approaches?
 - What is the mechanism behind the improvements?
 - Which is the “better” approach?

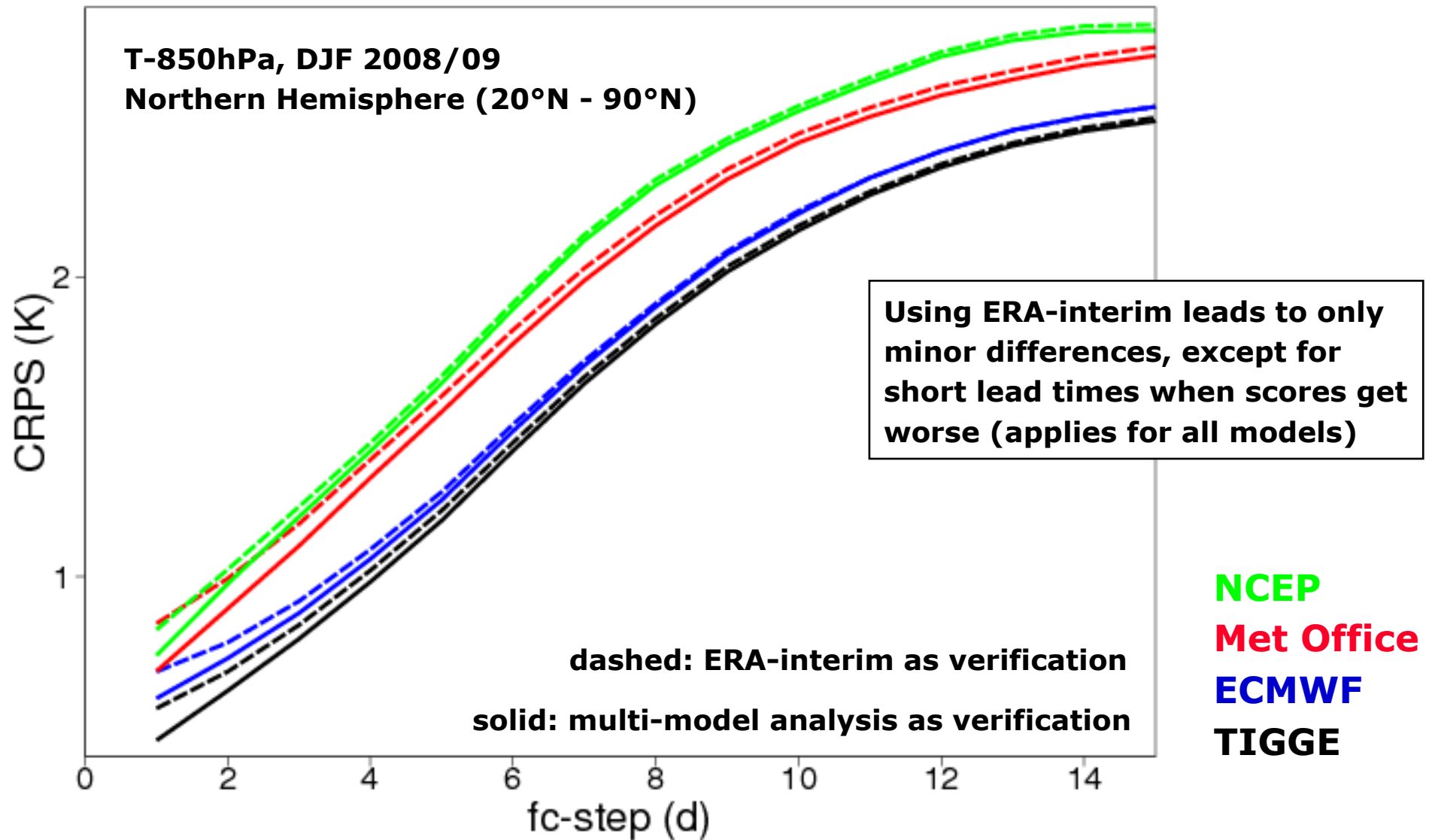


Possible verification datasets

- If we don't verify against model independent observations we need to agree on a '**fair**' but also '**most useful**' verification dataset
- Use each model's own analysis as verification
 - Multi-model has no "own analysis"
 - Intercomparison of skill scores "difficult" because reference forecast scores differently for different analysis
- Use a multi-model analysis as verification
 - Incorporating less accurate analyses might not necessarily lead to an analysis which is closest to reality
 - Calibration needs a consistent verification dataset used in both training and application phase, MM-analysis not available for reforecast training period
- Use "semi-independent" analysis: ERA-interim
 - Assumed to be as close as possible to reality
 - Available for long period in the past and near real-time
 - For upper air fields in Extra-Tropics close to analyses of best models / MM-analysis
 - For Tropics and near-surface fields use bias-corrected forecasts for 'fair' assessment

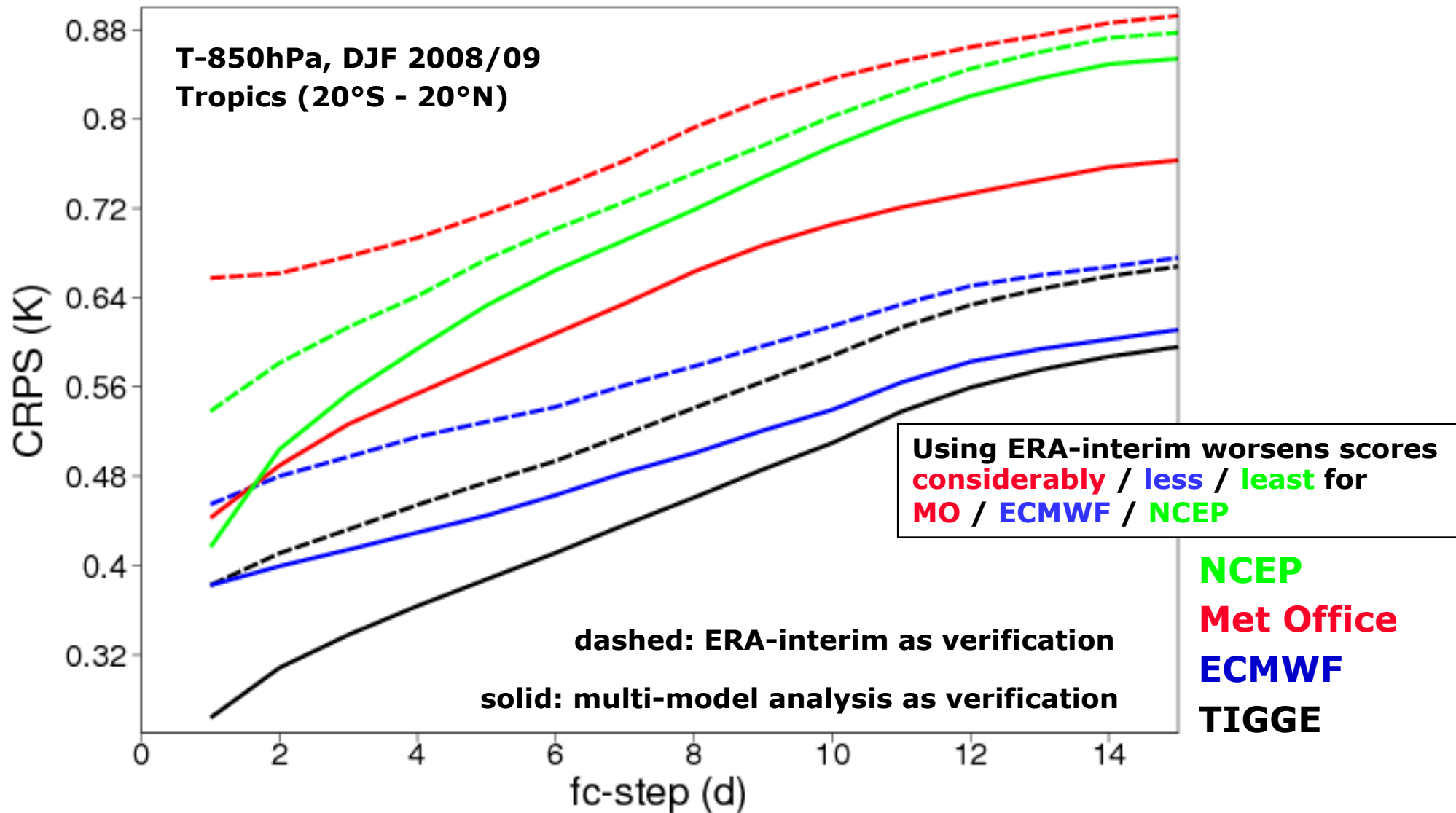


Choice of analysis: upper air, extra-tropics



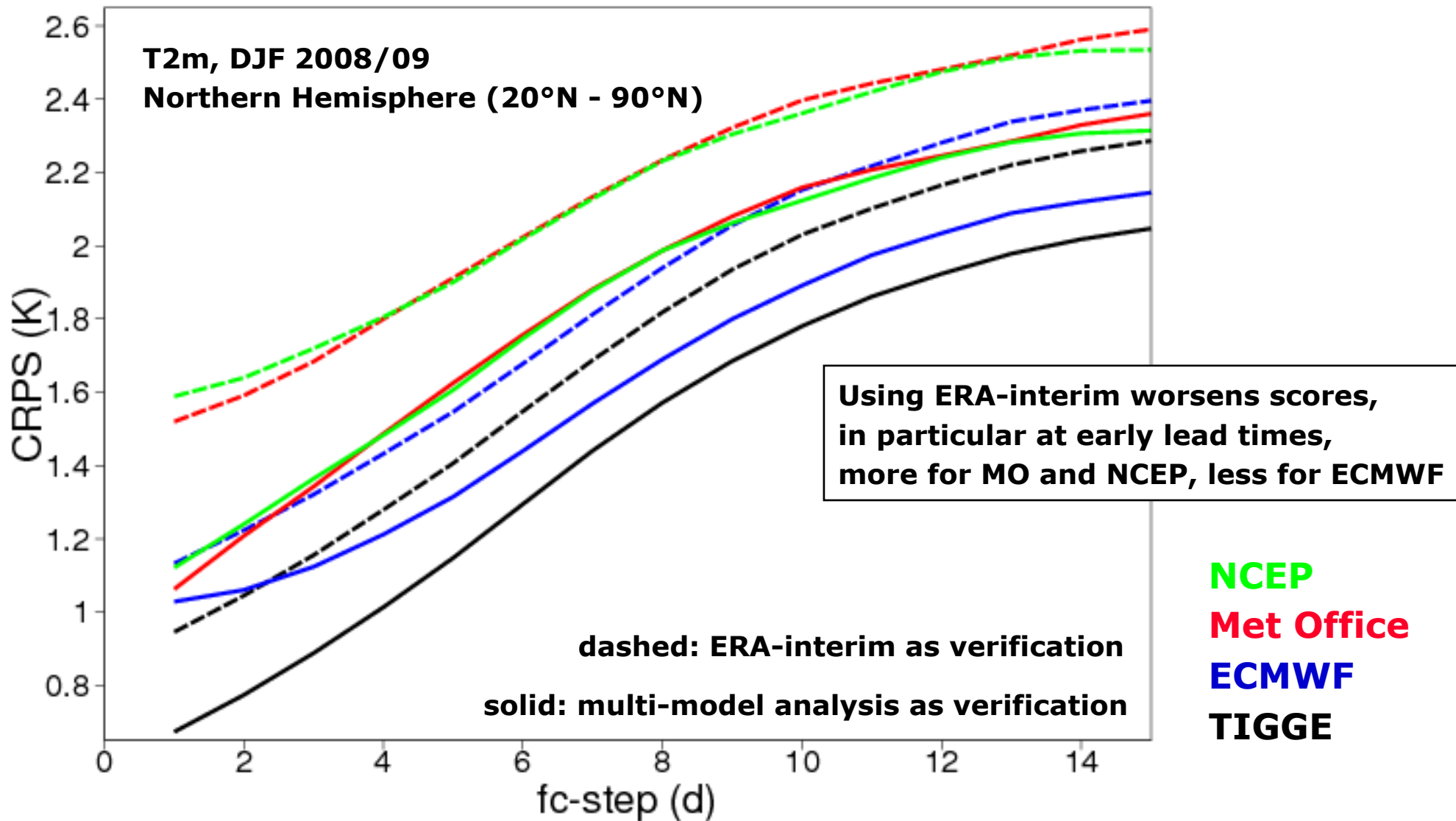


Choice of analysis: upper air, tropics



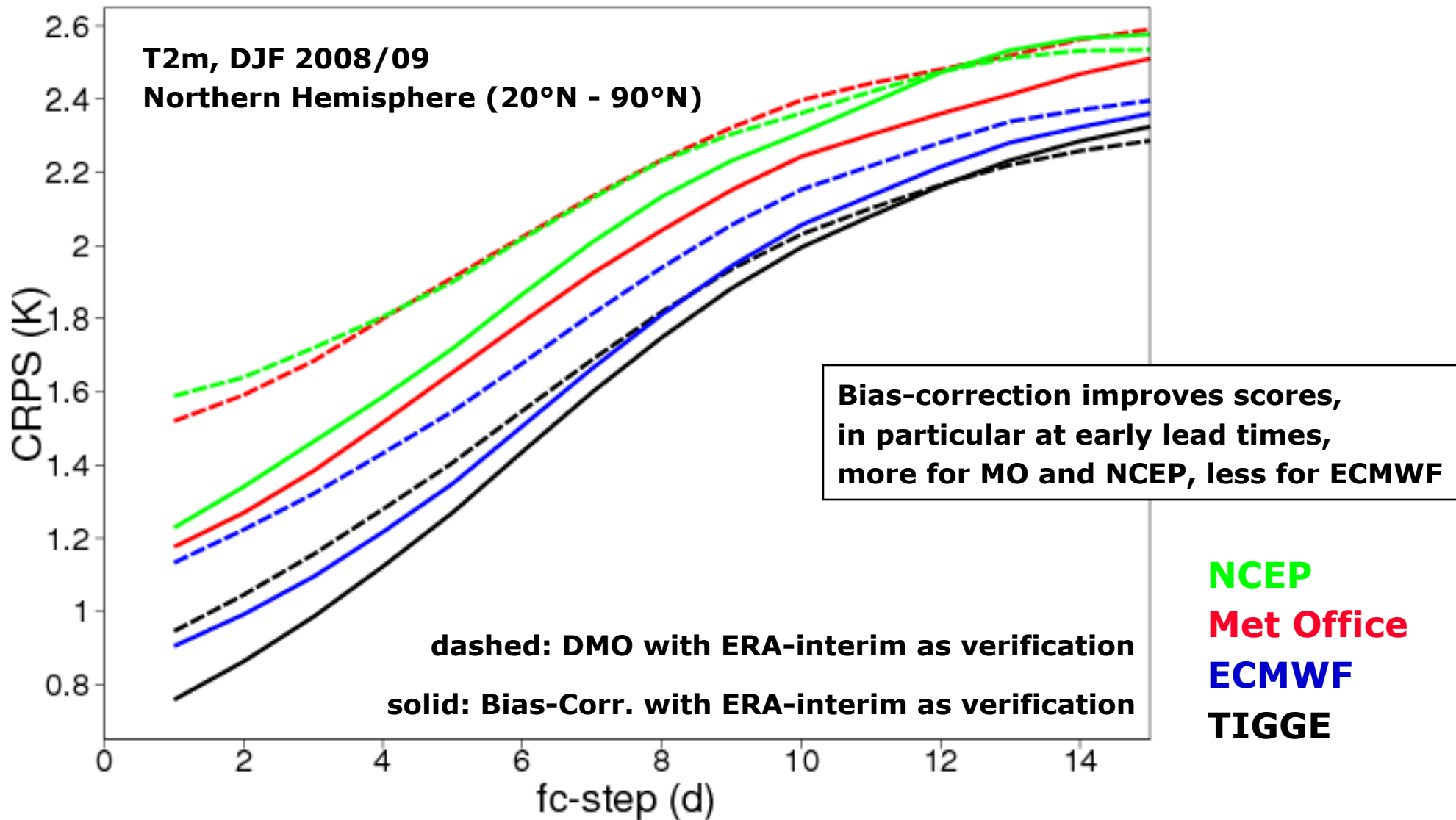


Choice of analysis: surface



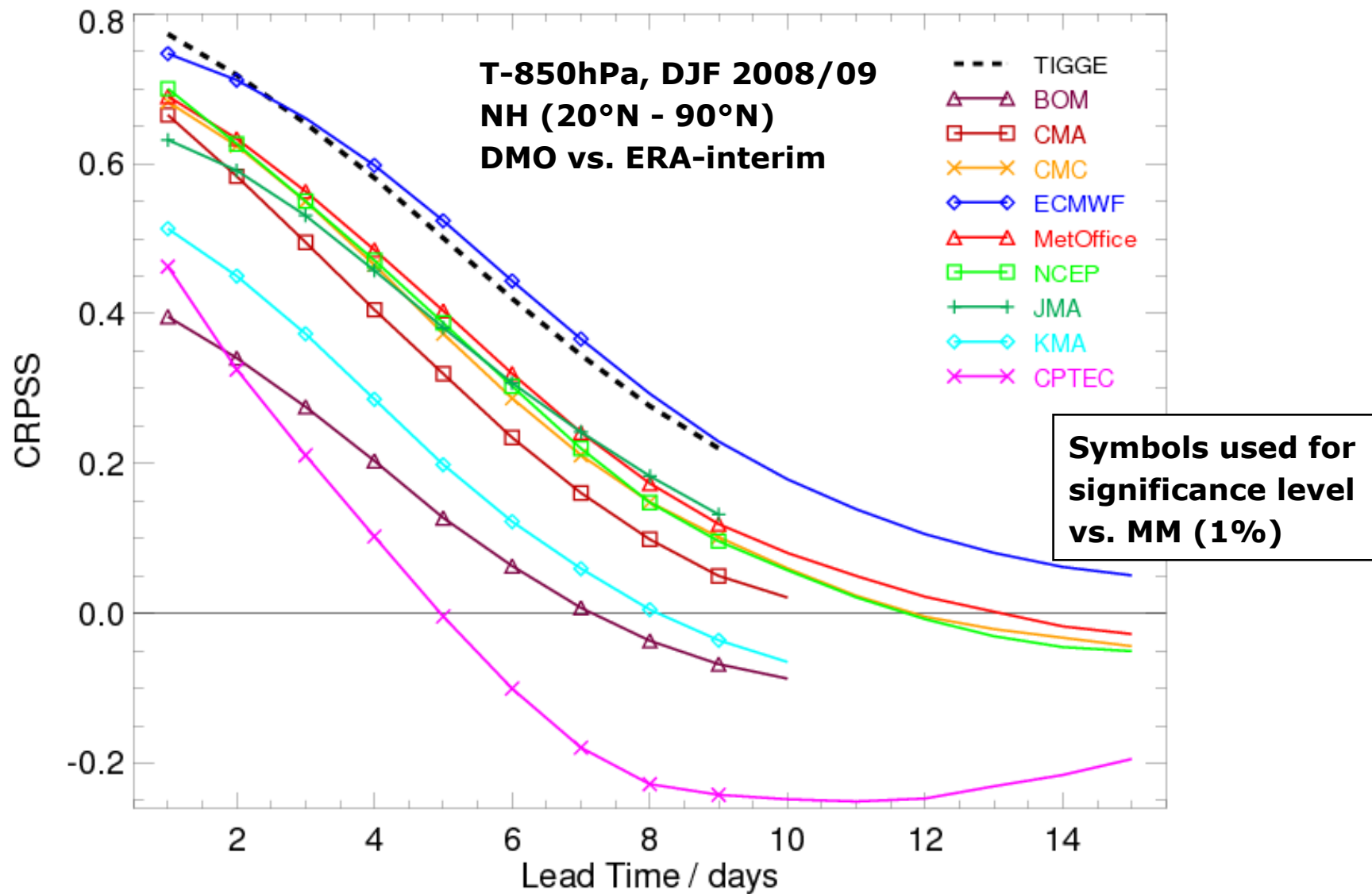


Choice of analysis: surface, bias-corrected



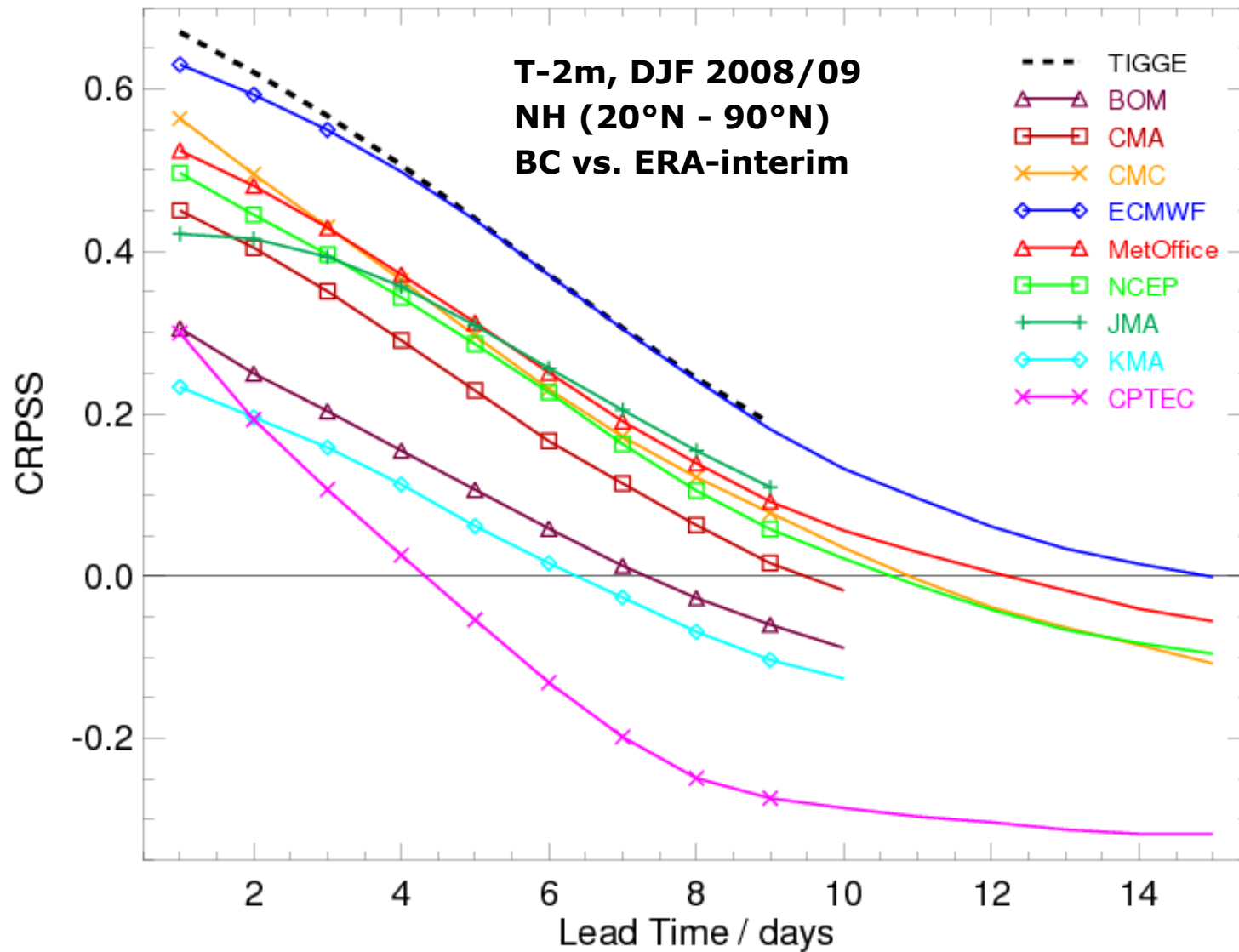


Comparing 9 TIGGE models & the MM



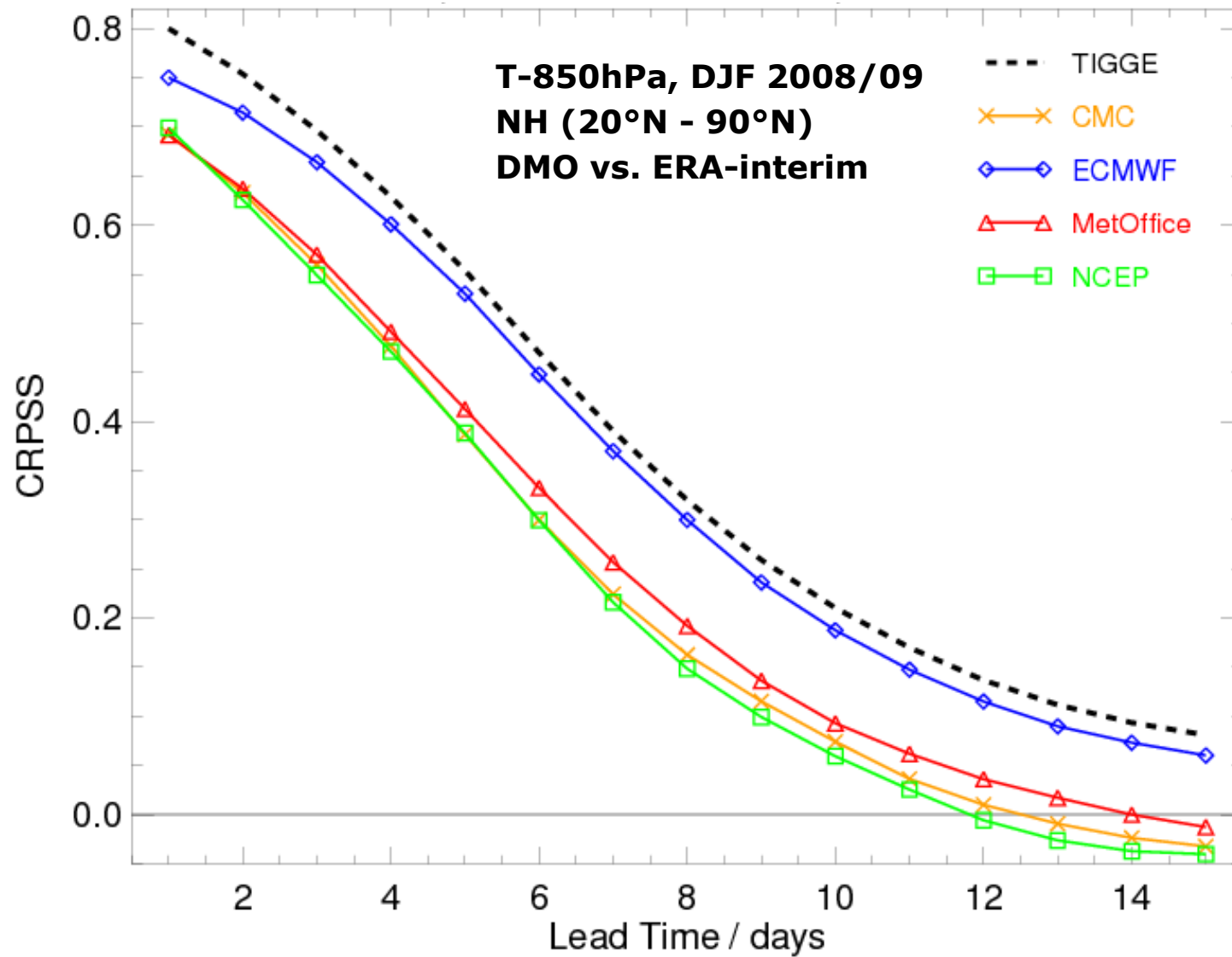


Comparing 9 TIGGE models & the MM



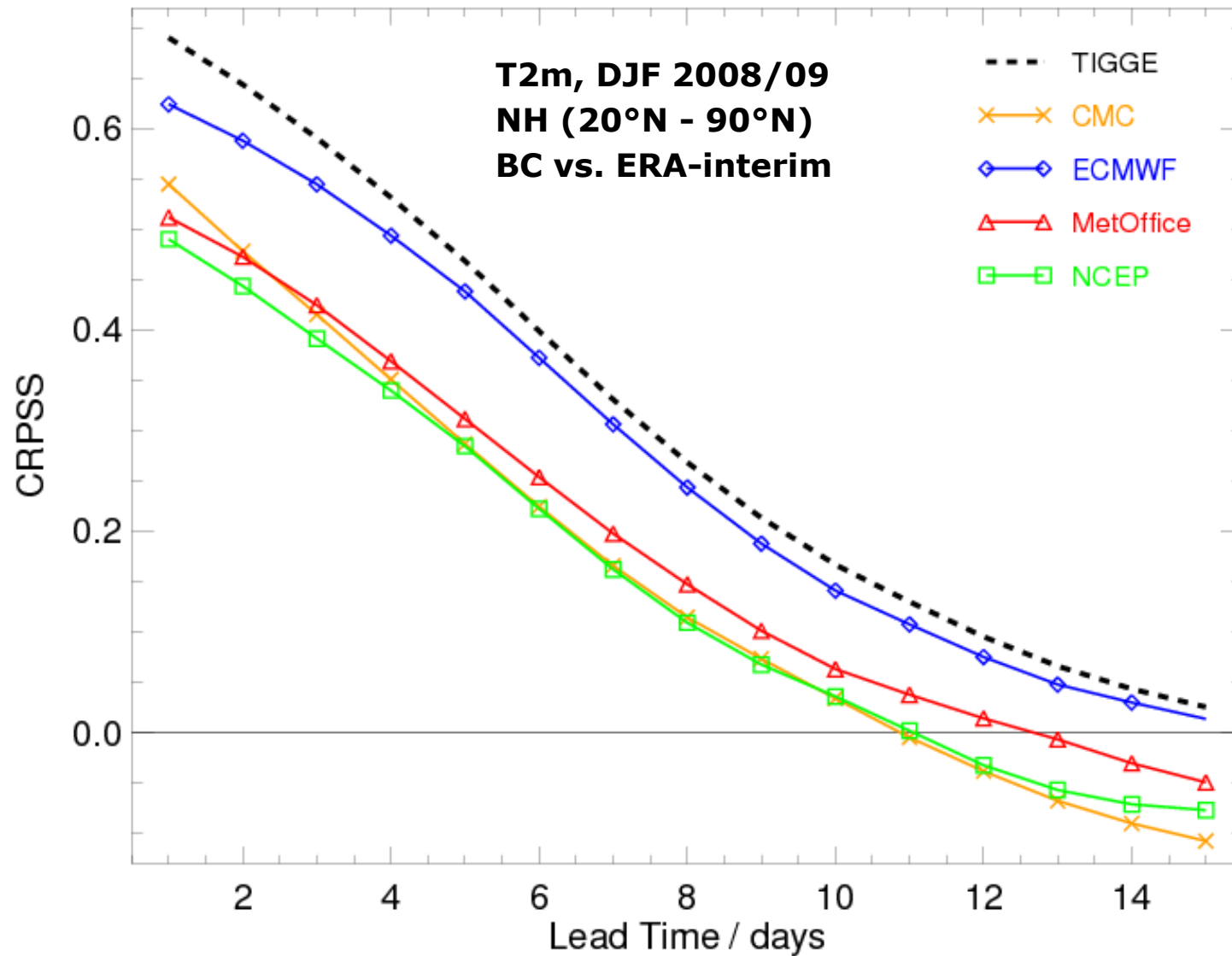


Comparing 4 TIGGE models & the MM





Comparing 4 TIGGE models & the MM





Calibration using reforecasts

- All calibration methods need a training dataset, containing a number of forecast-observation pairs from the past
- Non-homogeneous Gaussian Regression (NGR) provides a Gaussian PDF based on the ensemble mean and variance of the raw forecast distribution

$$P(v \leq q) = \Phi \left[\frac{q - (a + b\bar{x}_{ens})}{\sqrt{c + ds_{ens}^2}} \right] \quad \text{with: } \Phi = \text{CDF of standard Gaussian distribution}$$

- Calibration process:
 - Determine optimal **calibration coefficients** by minimizing CRPS for training dataset
 - Apply calibration coefficients to determine calibrated PDF from ensemble mean and variance of actual forecast to be calibrated
 - Create calibrated NGR-ensemble with 51 *synthetic* members
 - Combine NGR-ensemble with '30-day bias corrected' forecast ensemble



The reforecast dataset

	2008																															2009			
	Nov		Dec																													Jan			
	29	30	01	02	03	04	05	06	07	08	09	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	01	02
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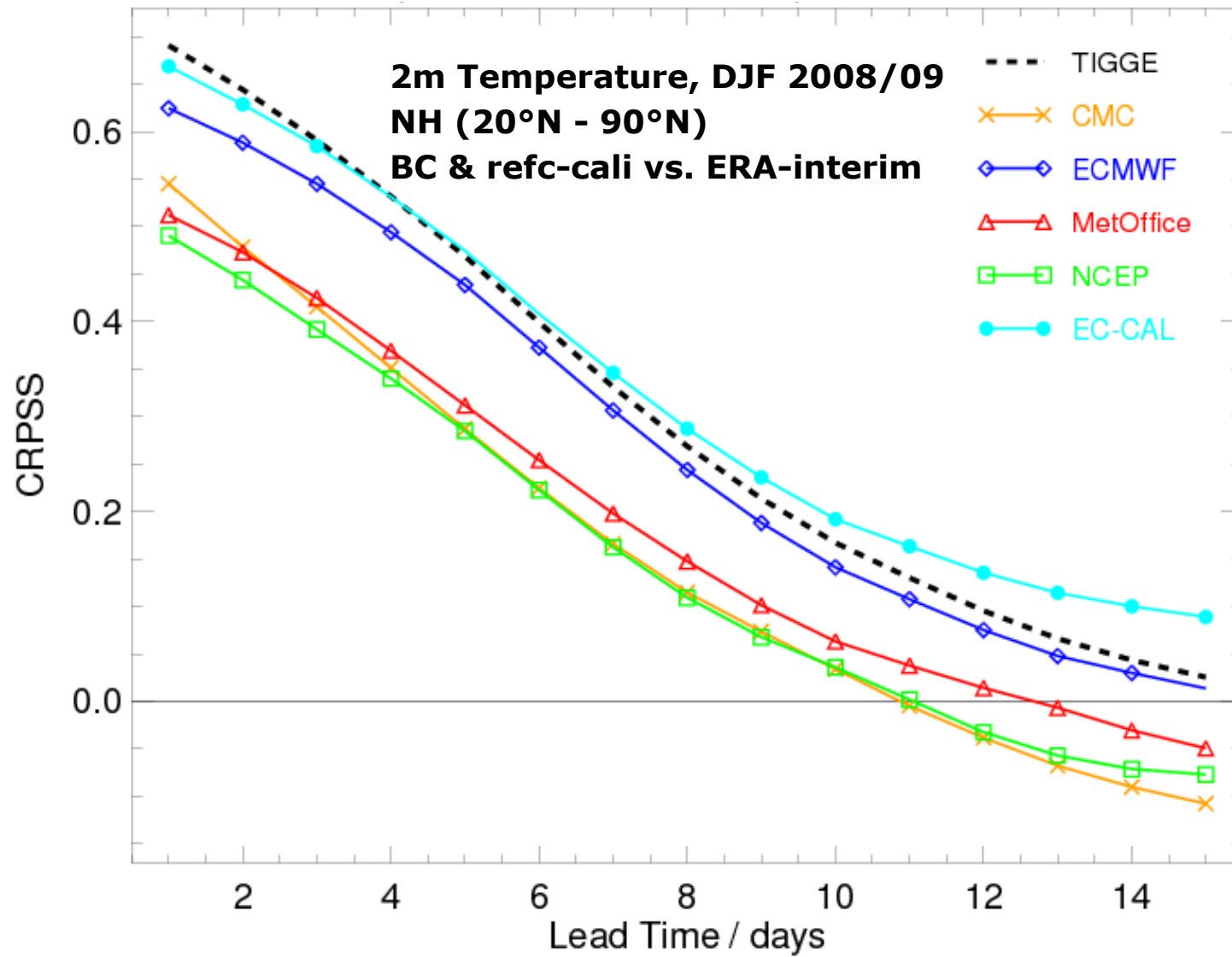


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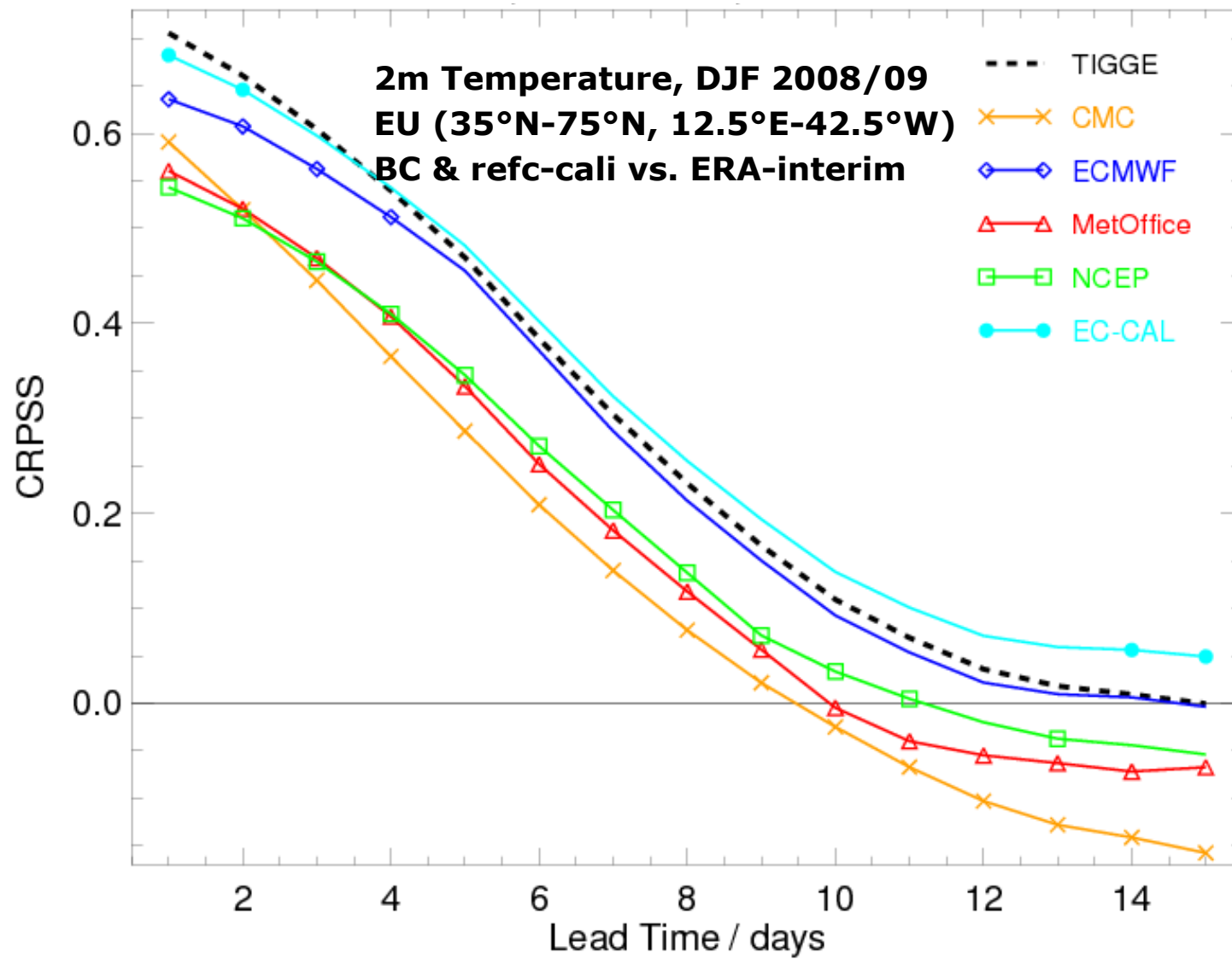


Comparing 4 TIGGE models, MM, EC-CAL



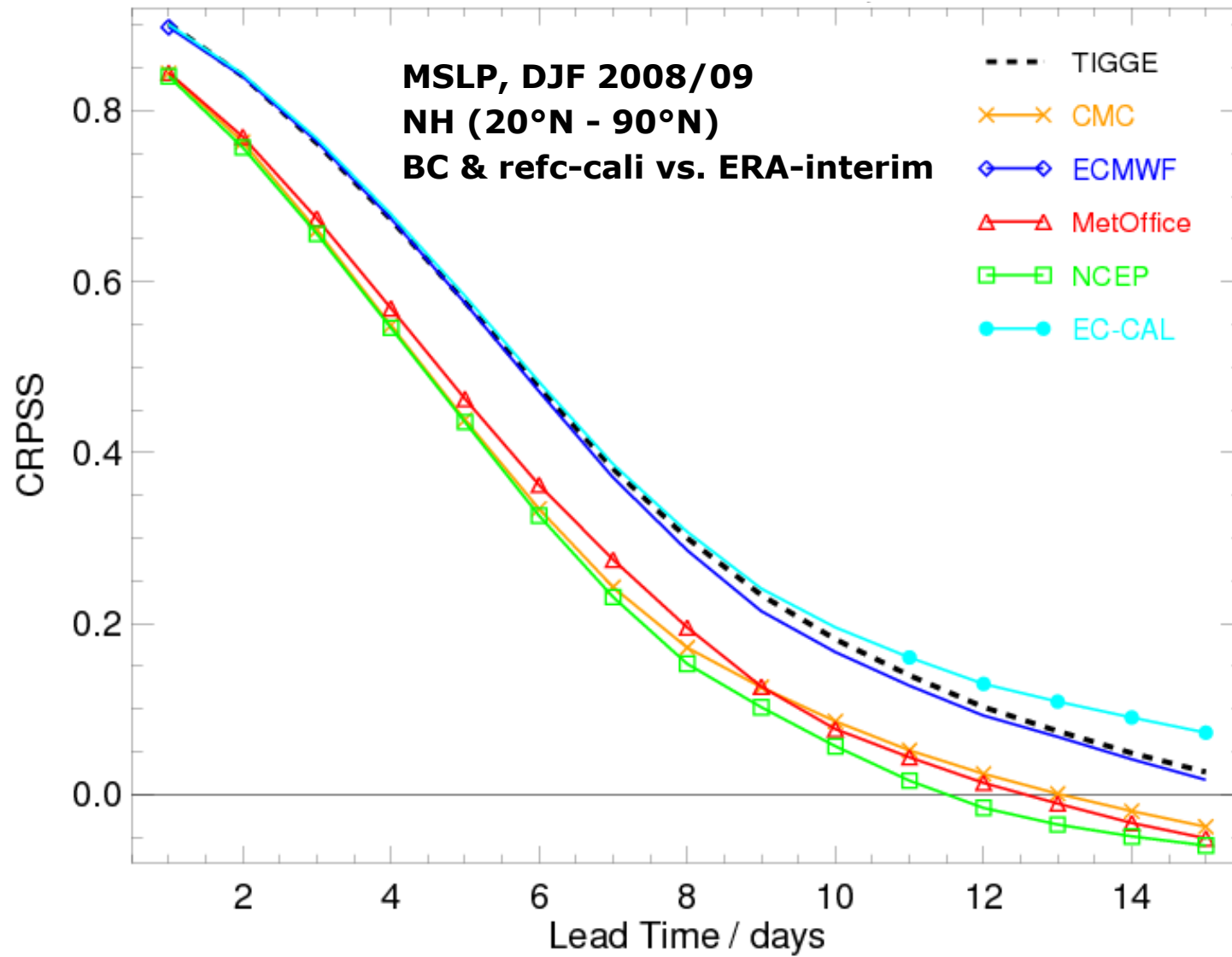


Comparing 4 TIGGE models, MM, EC-CAL



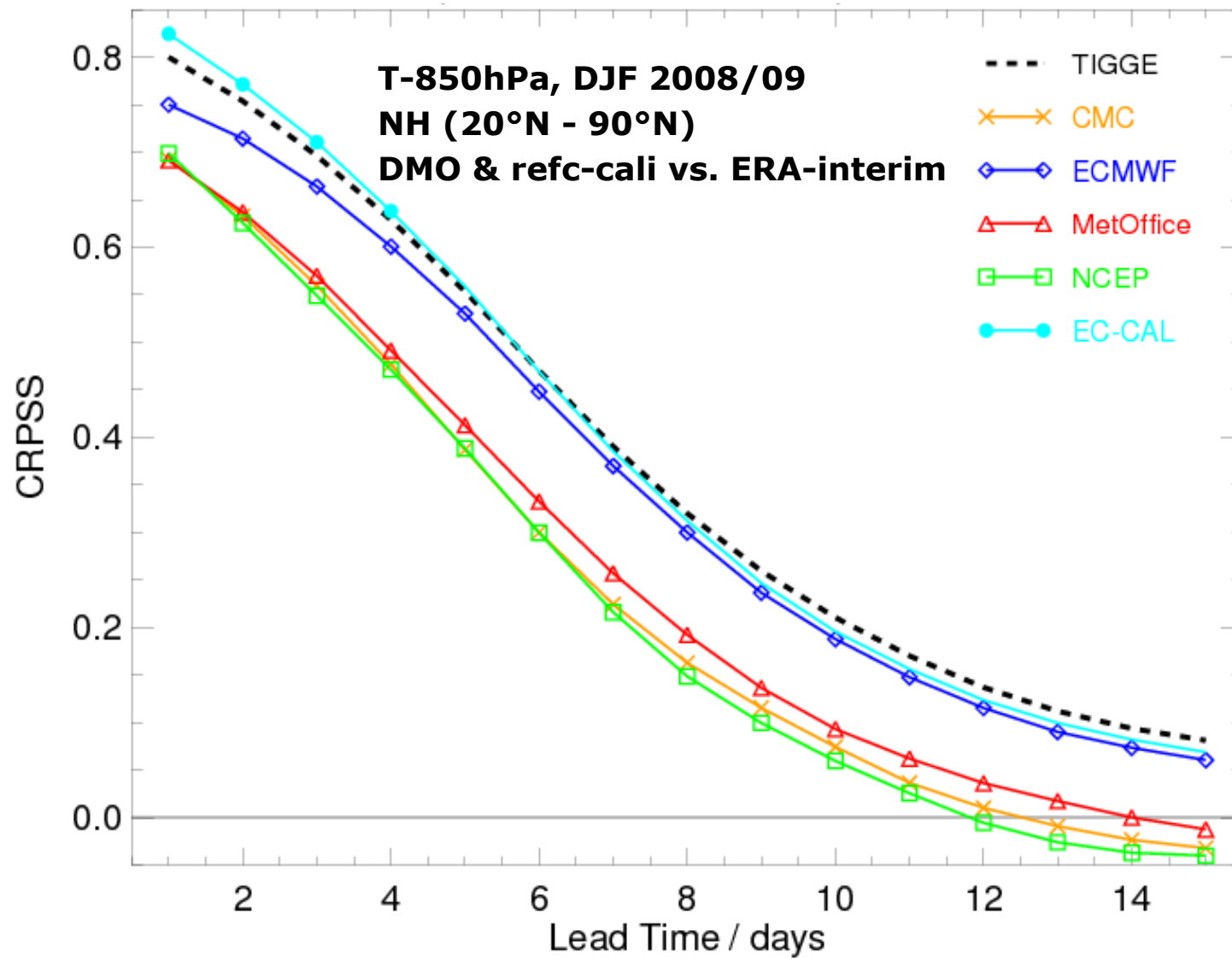


Comparing 4 TIGGE models, MM, EC-CAL

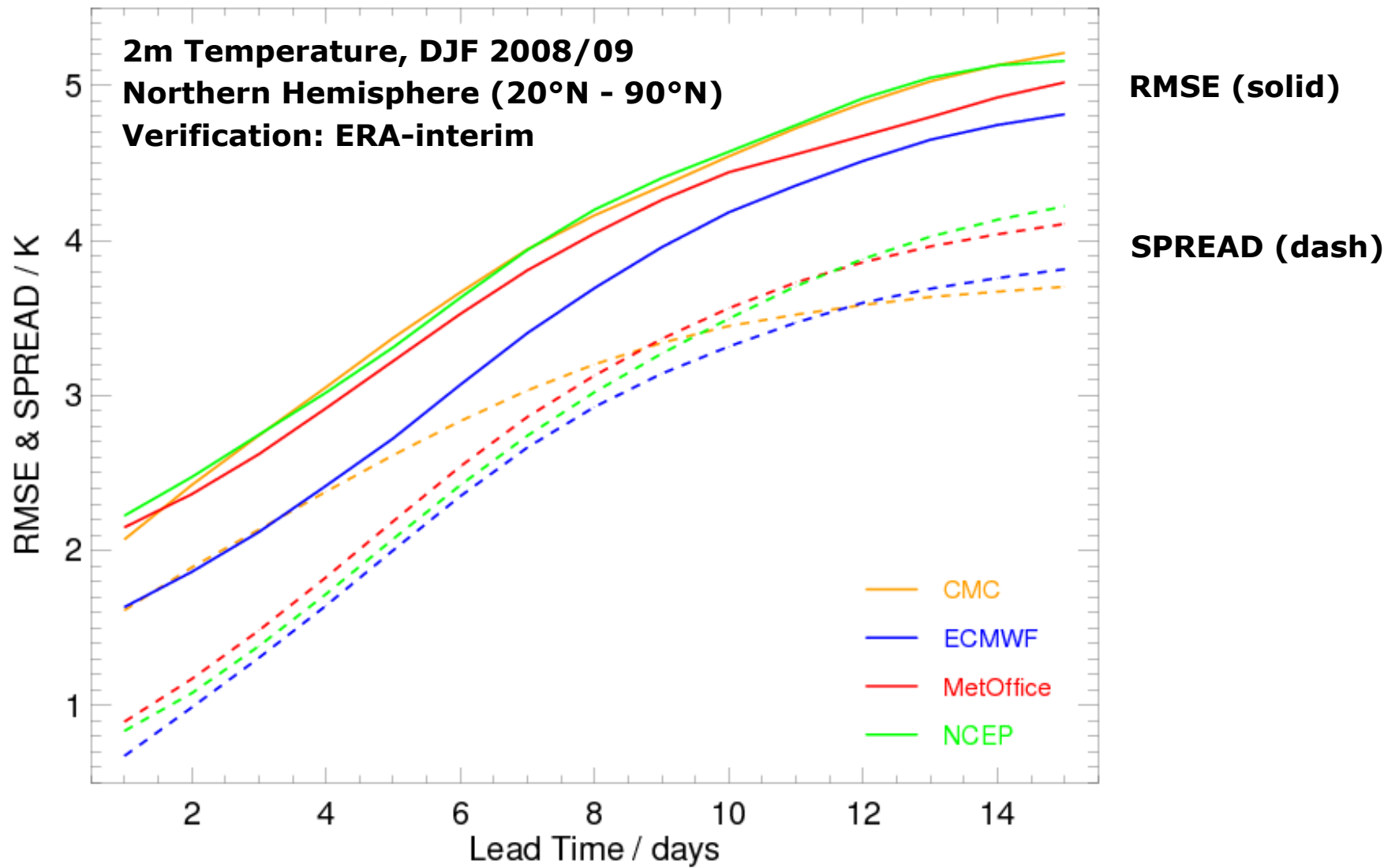




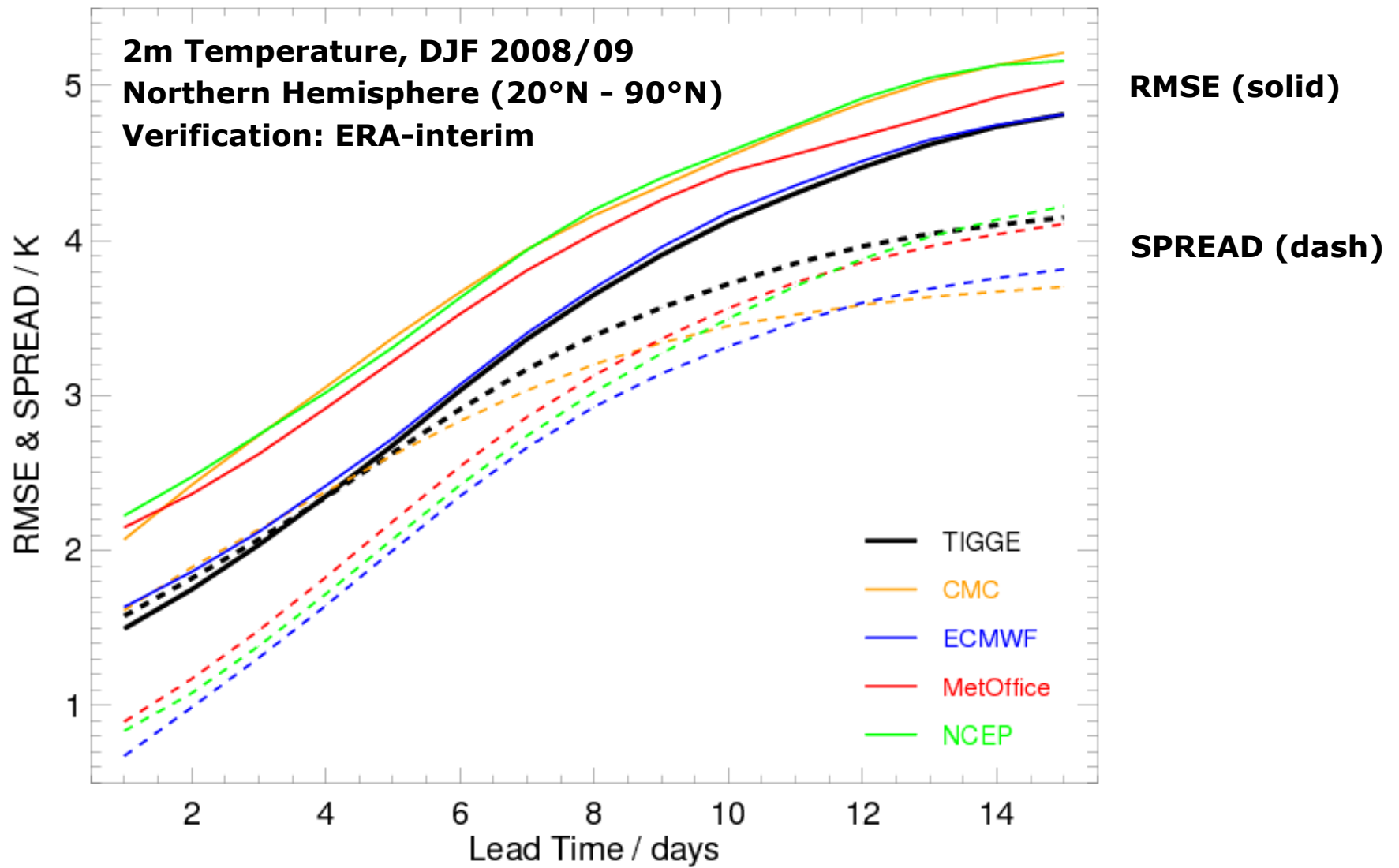
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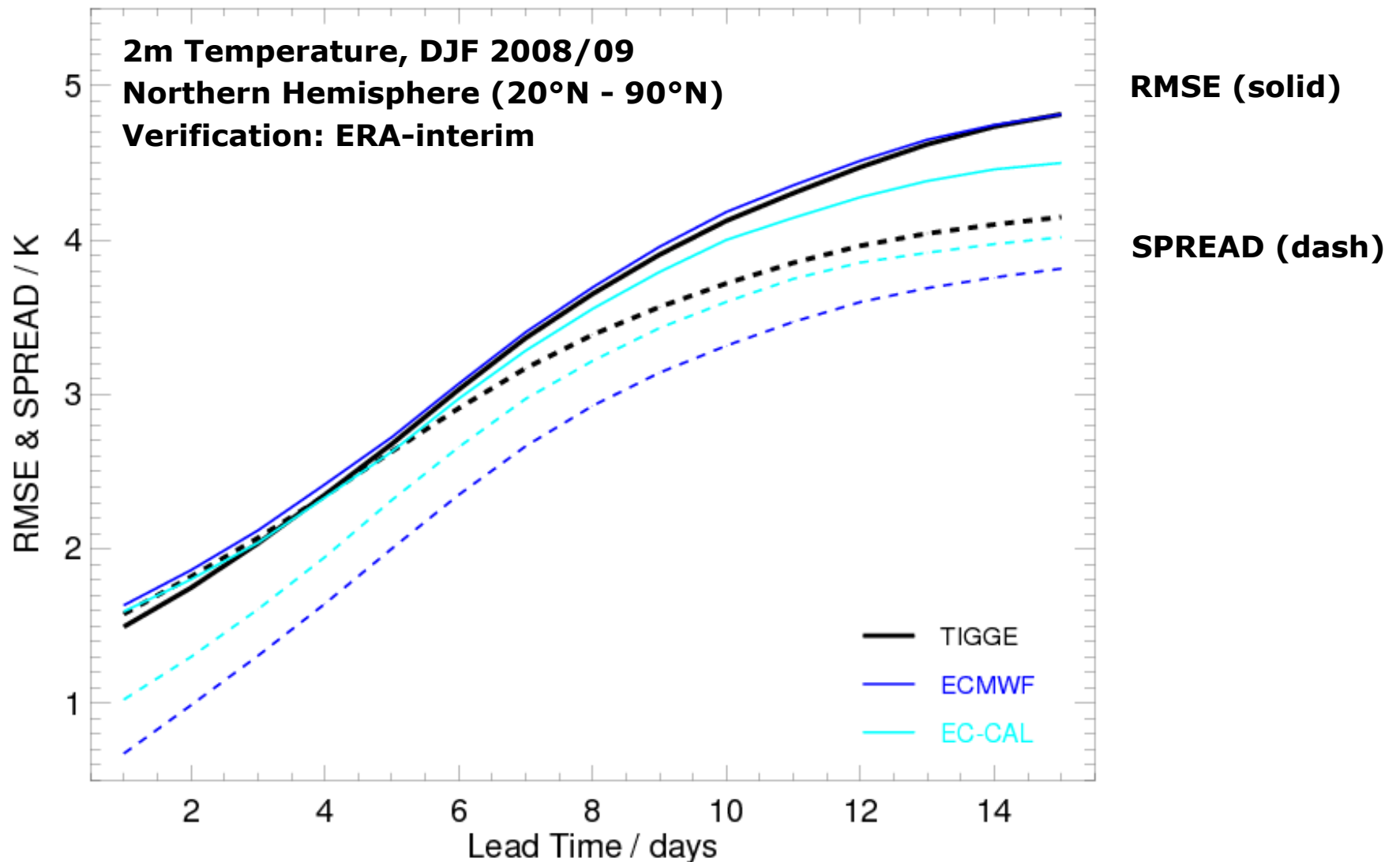
Mechanism behind improvements



Mechanism behind improvements

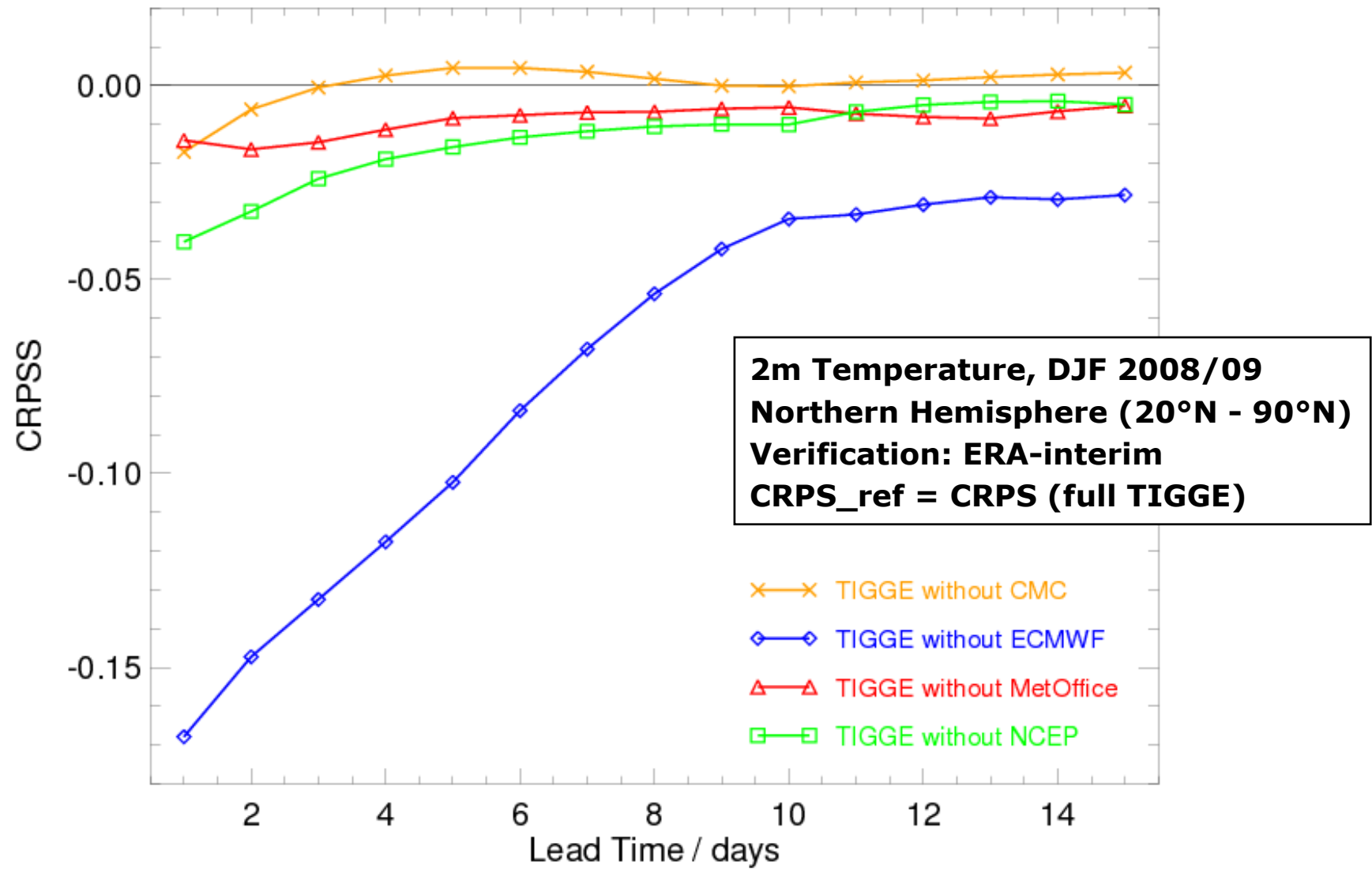


Mechanism behind improvements



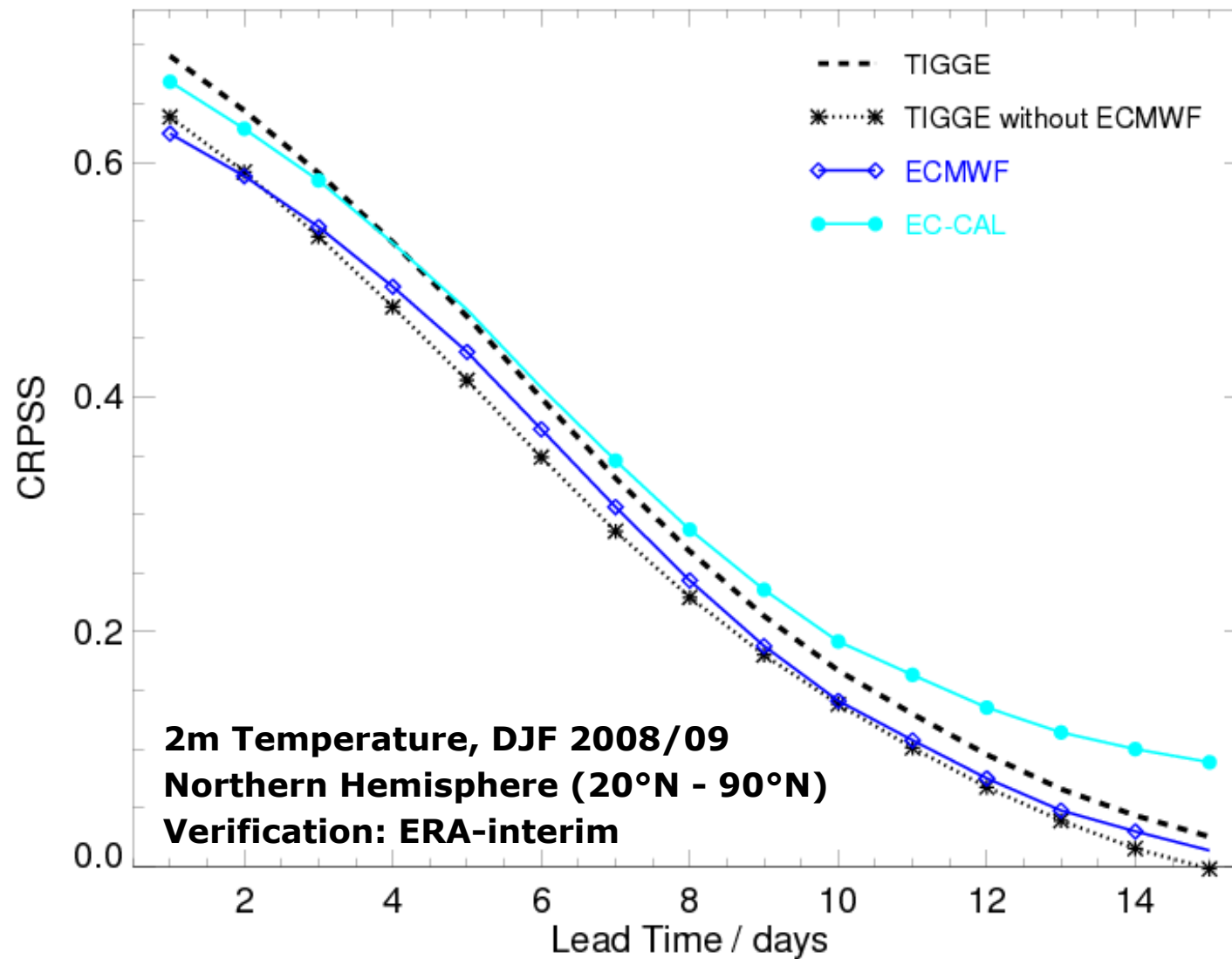


Reduced TIGGE multi-model



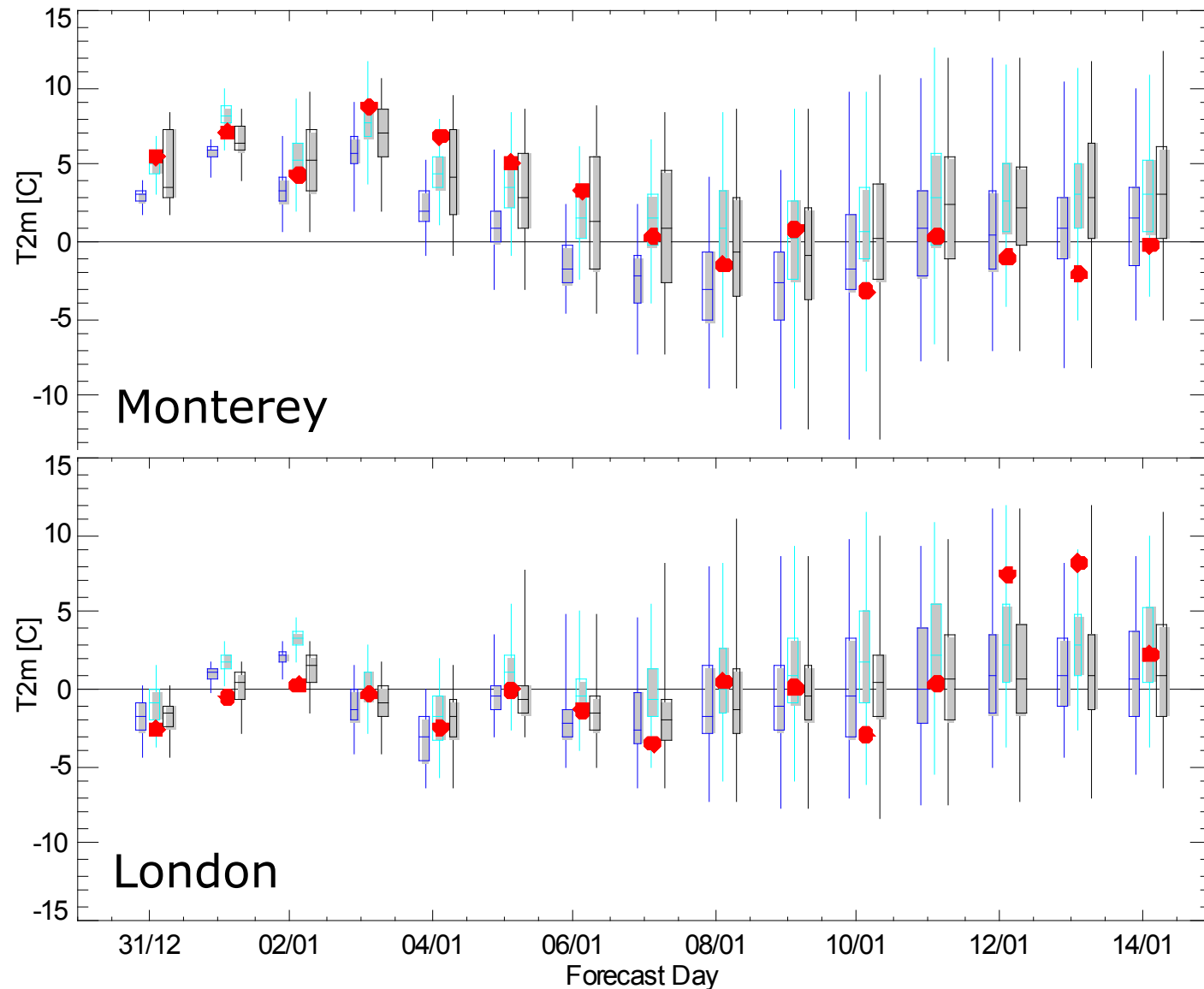


TIGGE vs. ECMWF vs. EC-CAL





Impact of calibration & MM in EPSgrams

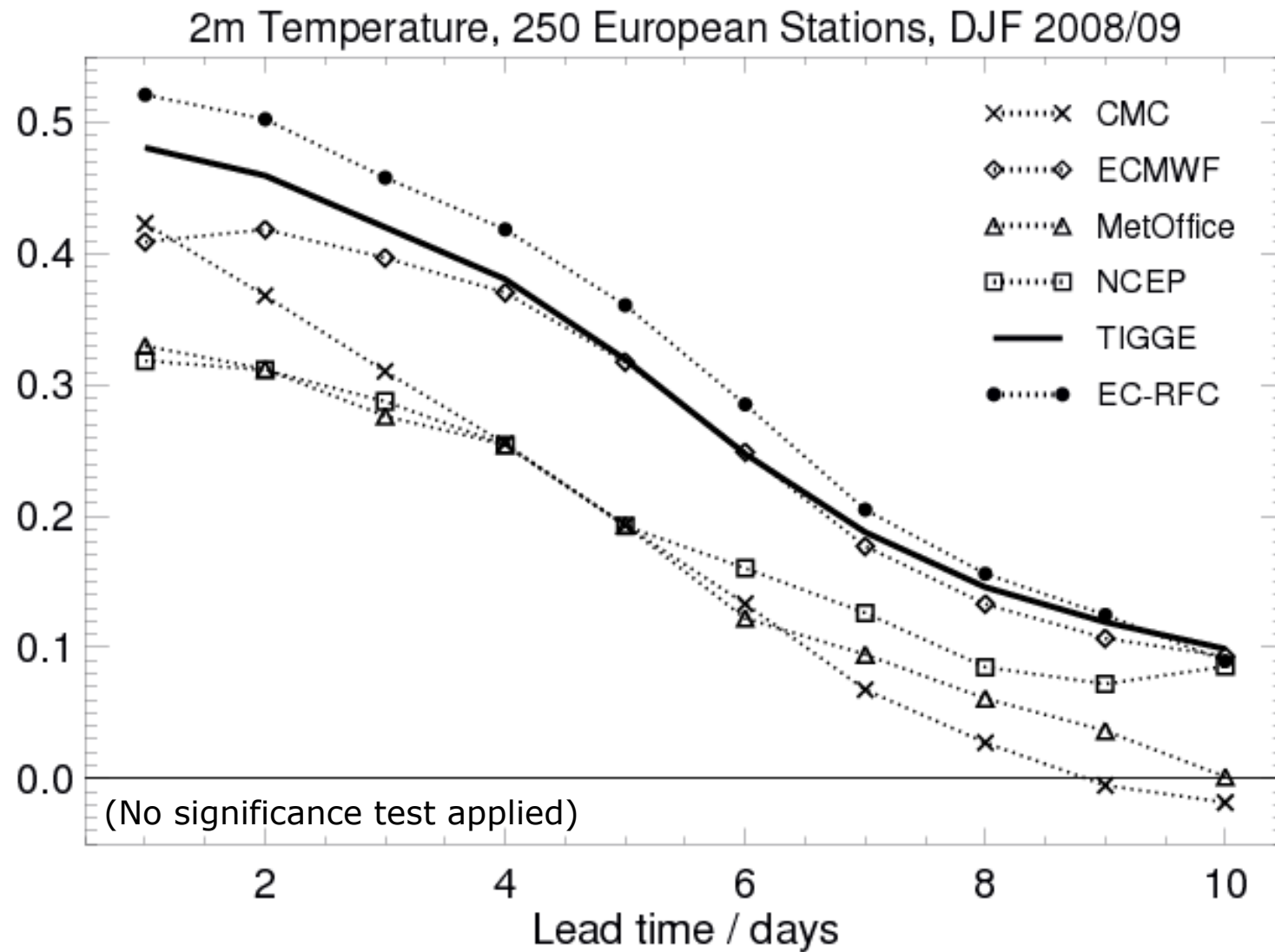


2m Temperature
FC: 30/12/2008

ECMWF
ECMWF-NGR
TIGGE
Analysis



What about station data?





Relative benefits and costs

	TIGGE multi-model	NGR Calibration using reforecasts
Benefits: upper air fields	limited	limited
Benefits: surface fields	Improved scores through reduced systematic error and increased spread	Improved scores through reduced systematic error and more appropriate spread
Costs: Computational aspects	No extra computer time but data transfer costs	Moderate increase in computing time (~10%), “for free” if reforecasts are produced for other purposes
Costs: Logistic aspects	Significantly increased complexity could make system more prone to failures, and timing issues could arise	Slight increase in complexity, e.g. when changing model cycles



Summary

- What are the relative benefits (costs) of both approaches?
 - Both multi-model and reforecast calibration approach can improve predictions, in particular for (biased and under-dispersive) near-surface parameters
- What is the mechanism behind the improvements?
 - Both approaches correct similar deficiencies to a similar extent
- Which is the “better” approach?
 - On balance, reforecast calibration seems to be the easier option for a reliable provision of forecasts in an operational environment
 - Both approaches can be useful in achieving the ultimate goal of an optimized, well tuned forecast system