

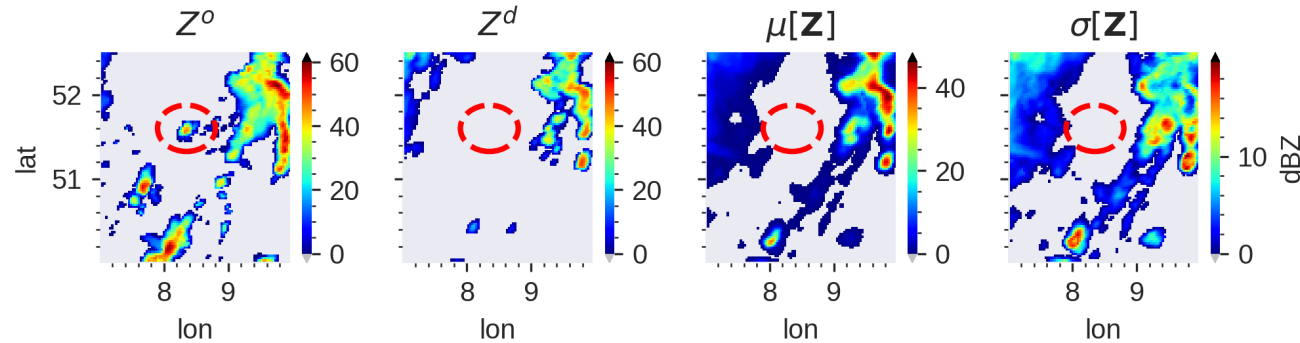
Improvements on the Assimilation of radar reflectivities (P3)

R. Potthast, K. Vobig



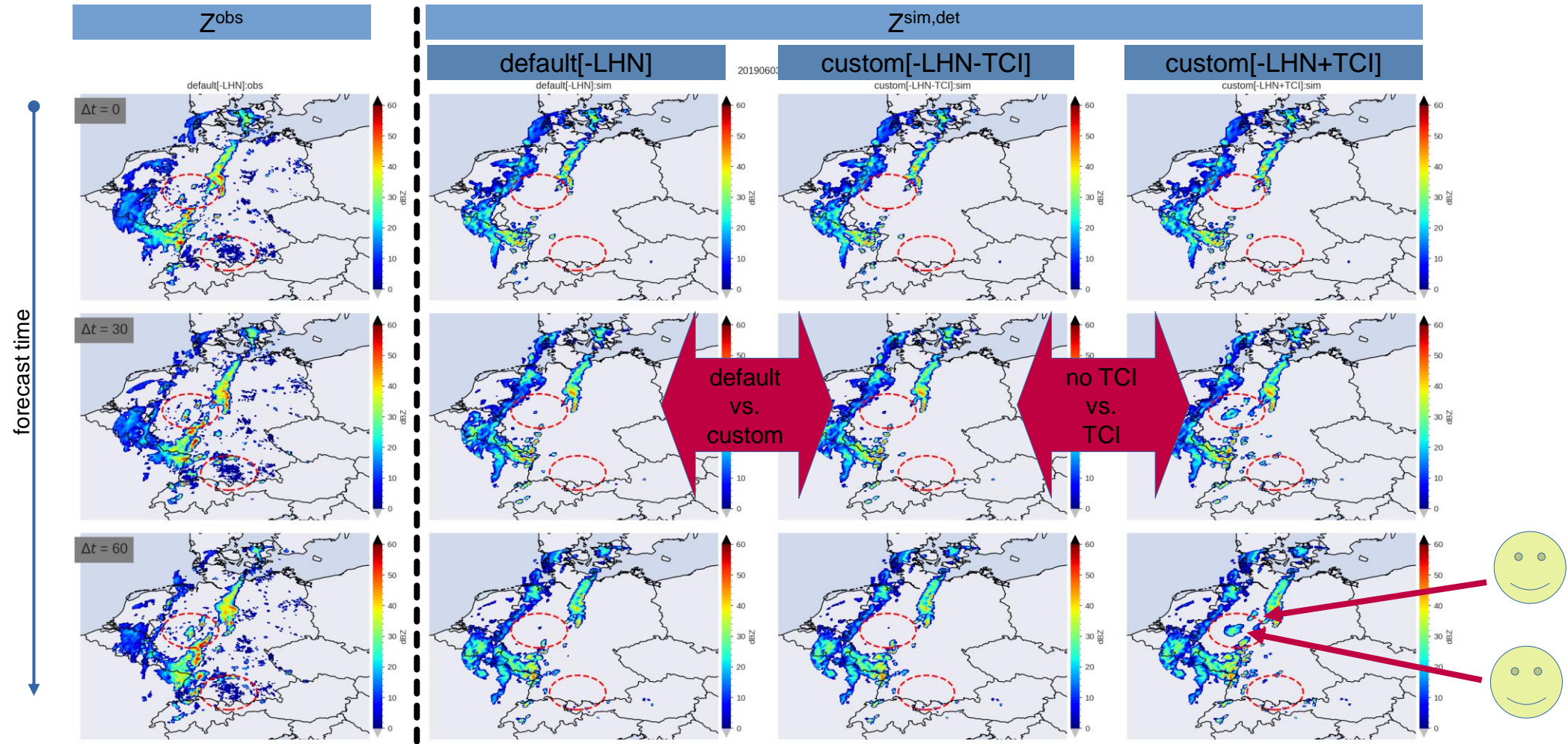
Deutscher Wetterdienst
Wetter und Klima aus einer Hand



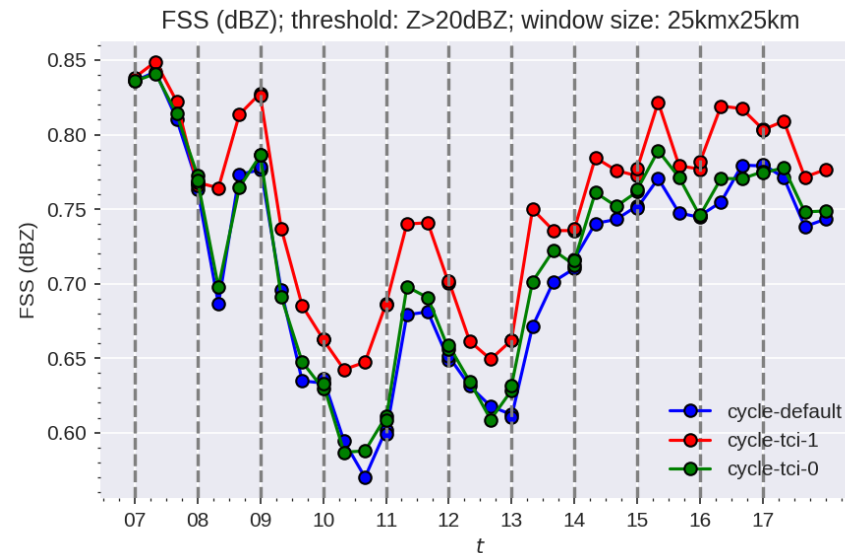
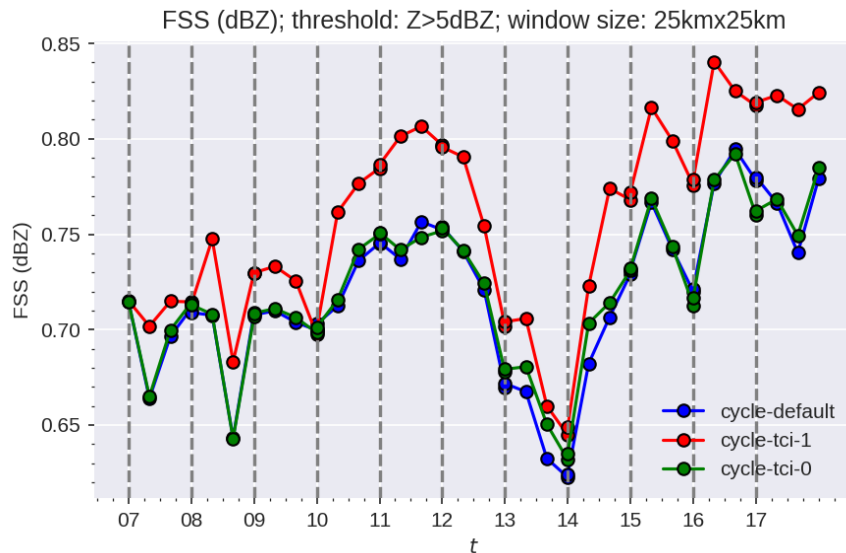


- even for large discrepancies between obs./sim. REFL LETKF might give small increments due to very **small ensemble spread** $\sigma[Z] \ll 1$
- approach: **increase spread** via (additive) targeted covariance inflation (TCl) based on **correlations** between Z and QV
- overall, **TCl results are promising**
 - ◆ production of “new” REFL cells (consistent with observations)
 - ◆ positive impact on fractional skill score (w.r.t. REFL)

Evolution of REFL



Verification: Fractional Skill Score



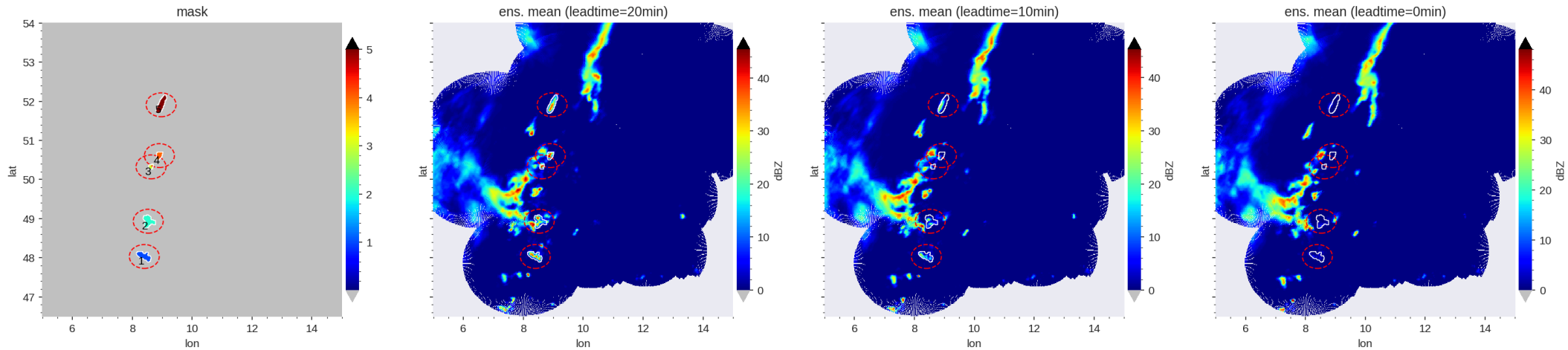
default[+LHN]
custom[+LHN+TCI]
custom[+LHN]

- performing cycle starting at 7 UTC
- TCI applied at each assimilation (hourly)
- FSS for two dBZ thresholds shown

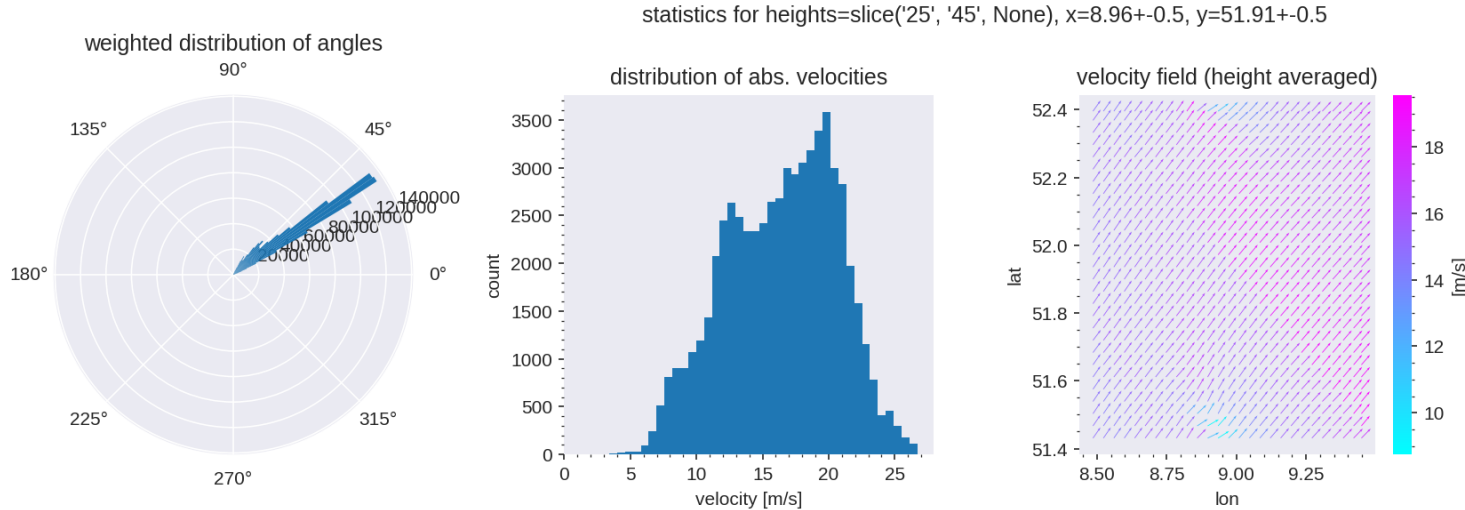
positive impact
of TCI on FSS

- AIREP/TEMP **observation error statistics for humidity**:
 - ◆ negative impact of TCI
 - ◆ contribution (positive/negative) to statistics highly time/location dependent
 - ◆ time/spatial/process dependence of correlations
- optimize the **capturing and use of correlations** of TCI method
 - ◆ better **data filtering/pre-processing** necessary
 - ◆ towards more “**process-aware**” TCI
 - ◆ first step here: only include data associated with **new emerging cells** for correlation analysis

adate:20190603130000, ensemble_slice:slice(1, None, None)



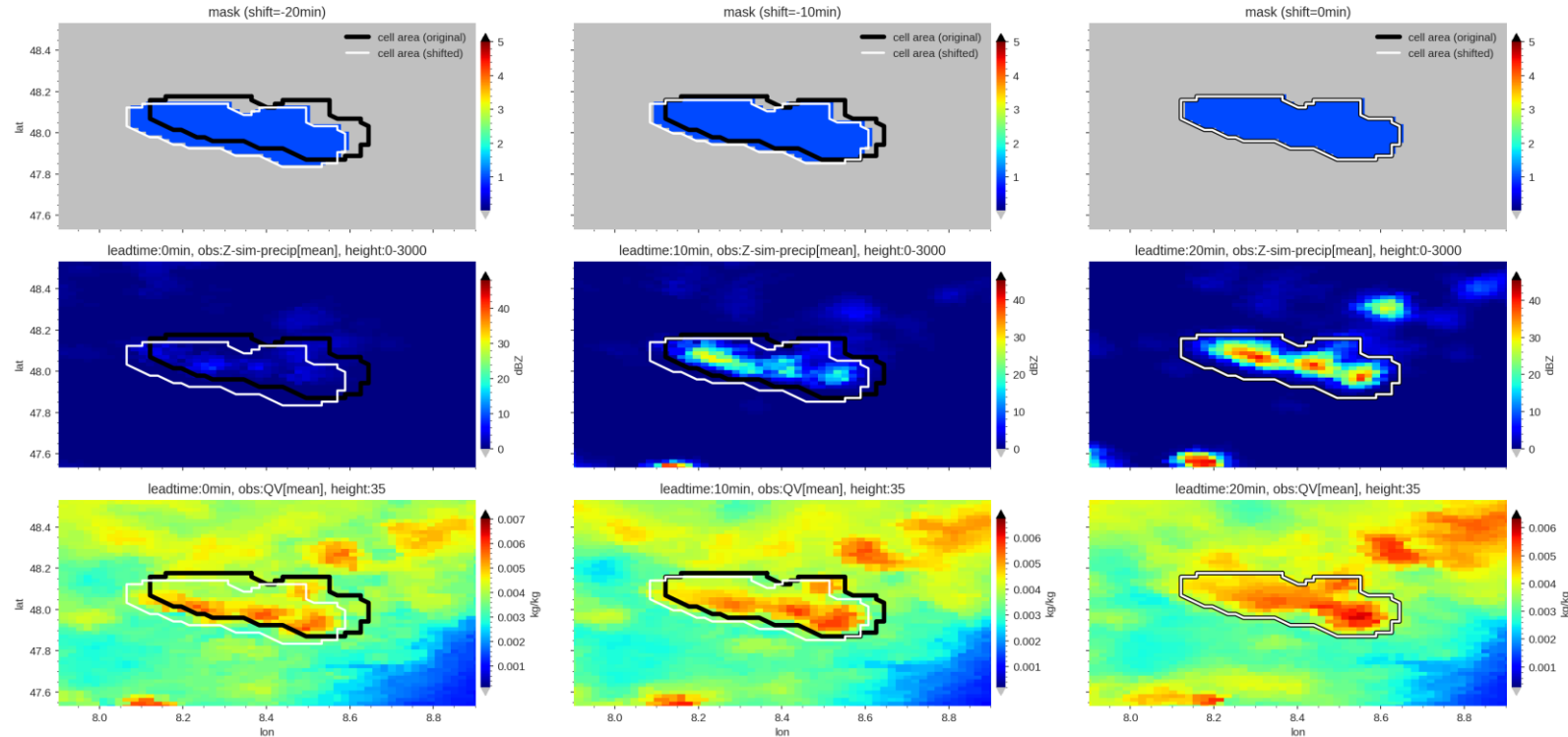
- Implemented simple algorithm for the **detection of new cells**
- employs time series of (binned) Radar data at 0-3000m
- gives **area of new cells** at certain leadtime
- here: 'mask' shows area of **5 new cells** detected for **20min leadtime**



- using **statistical analysis** of wind fields for obtaining dominant **angle and velocity of horizontal wind** in region of detected cells
- next: **shift area** associated with each cell using the corresponding wind field information → “**backward propagation in time**” / obtain environment cell eventually originated from

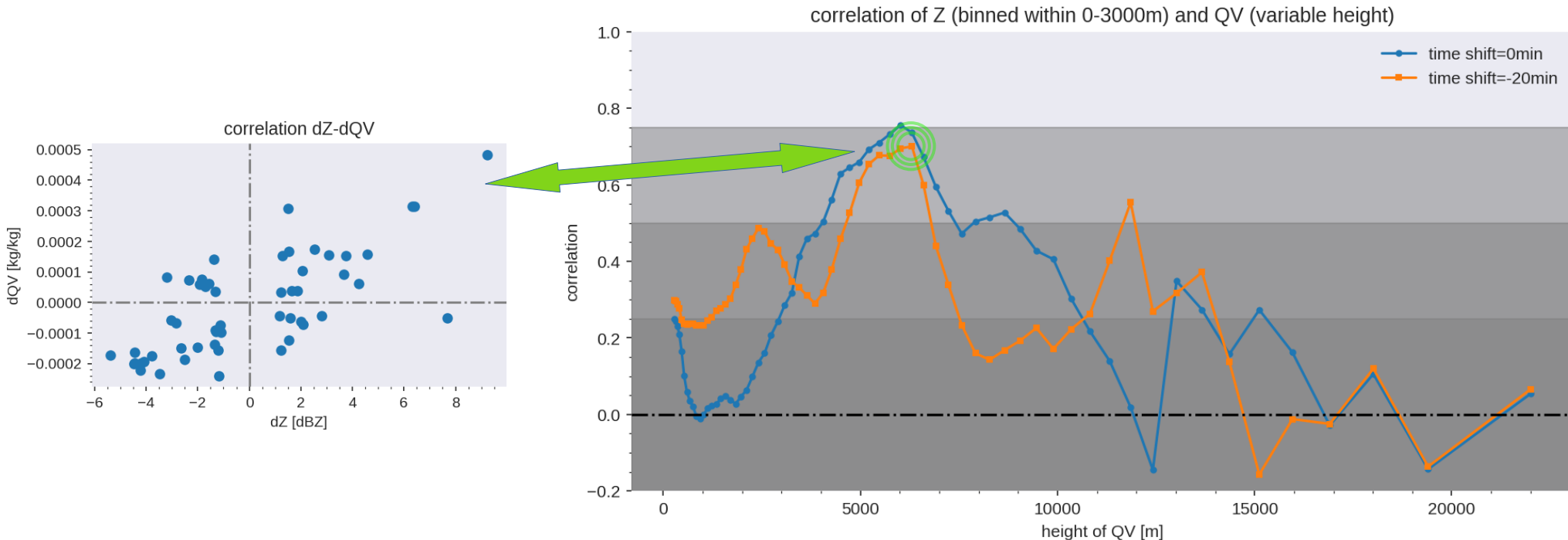
Cell Drifting: Application

cell-label:1, adate:20190603130000, leadtimeBase:20min, ensemble_slice:slice(1, None, None)



- zoom into the area of one of the detected cells
- assumed drift seems to match “real” drift of structure in QV field

- previous plot already hints at relationship between Z “now” and QV at an earlier time (contained within shifted area)
- use the previously shown cell areas (for each time t and cell c) $A_{c,t}$ for calculating **spatial mean of ensemble perturbations** $dQV_{e,h}$ and dZ_e (for each ICON level h and ensemble member e) yielding $dQV_{e,h,c,t}$ and $dZ_{e,c,t'}$
- given t, t', h the correlation $\text{corr}_{t,t',h}$ is then based on the dataset $\{(dQV_{e,h,c,t}, dZ_{e,c,t'}) \mid \text{all members } e \text{ and all cells } c\}$
- perform this correlation calculation for several t, t' and all ICON levels h

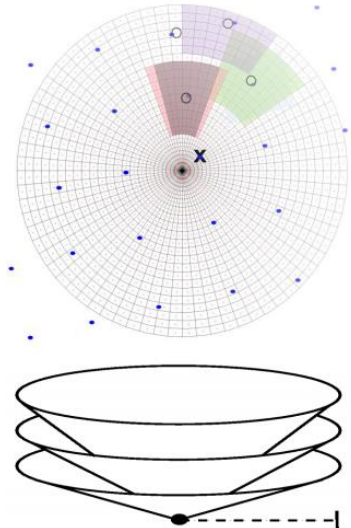
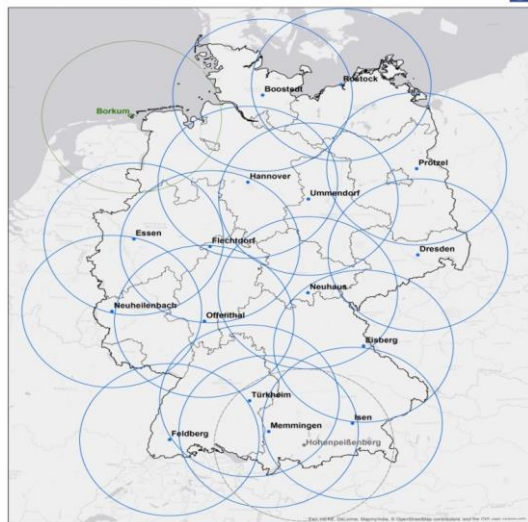


- depiction of **correlation between Z and QV** (at variable height)
- **QV shifted** against Z by 0min or -20min
- clear maximum at around 6000-7000m height for QV

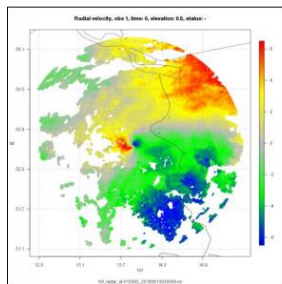
- further **optimize the process/data-filtering**
- use these “**process-aware**” **correlations** as a **basis for the TCl** approach
- for the application of these correlations (within the TCl approach) the **procedure for their extraction is basically inverted**:
 - ◆ check if **discrepancies** exist between obs./sim. REFL
 - ◆ check if **spread** for sim. REFL is vanishing
 - ◆ check if (obs.) **cell has just emerged**
 - ◆ **use tailored Z-QV correlations** for these regions

OLD

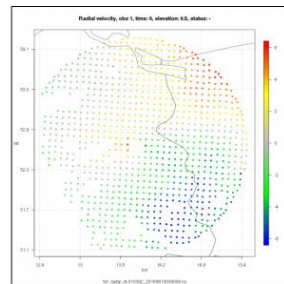
Radarverbund des Deutschen Wetterdienstes Deutscher Wetterdienst
Wetter und Klima aus einer Hand



Radar network of DWD (left); generation of superobservations (upper right); volume scan modus (lower-right)



Original Data



Superobing (~ 10 km)

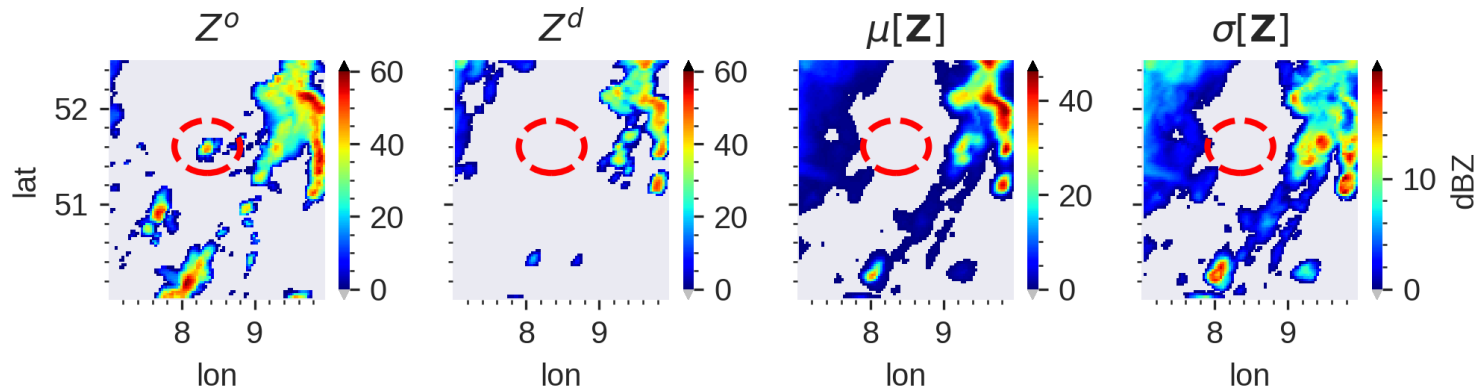
- 16 **Dualpolarization Radars** with 3D-Volume scans every 5 minutes
 - ◆ radial winds (**RW**)
 - ◆ reflectivities (**REFL**)
 - ◆ dual polarization moments (**DP**)
- generation of **superobservations**
 - ◆ average over specific volume
 - ◆ makes handling of large data sets feasible

- assimilate **2D REFL** based on **latent heat nudging** ✓
- assimilation of **3D-Volume** Radar data via LETKF and EMVORADO (by Blahak and Zheng)
 - ◆ assimilation of Radar **RW** ✓
 - ◆ assimilation of **REFL** ✓
- assimilation of Radar-derived **objects** and seamless integration of Radar objects into nowcasting and short-range NWP
- **Jana Mendrok** works on extending EMVORADO to simulate **DP** (✓)
 - ◆ enables **direct** assimilation of DP
 - ◆ alternatively: “indirect” assimilation of DP via derived hydrometeor mixing ratios (→Lucas Reimann)

- assimilation of **nowcasted information**⁽¹⁾
 - ◆ tested assim. of nowcasted information via LETKF (based on oscillator model / Lorenz 63 model system)
 - ◆ positive impact of assimilating nowcasted information demonstrated
 - ◆ first tests assimilating nowcasted states (REFL) with KENDA
- overall topic here: improve assimilation of REFL via **targeted covariance inflation**⁽²⁾ (TCI)

- ⁽¹⁾: R. Potthast et al., MWR, (2022), accepted for publication
- ⁽²⁾: K. Vobig et al., <https://doi.org/10.1002/qj.4157>, (2021)

TCI – Motivation & Basics



- even for large discrepancies between observed/simulated REFL LETKF might still produce small increments
- problem: very **small ensemble spread** $\sigma[\mathbf{Z}] \ll 1$
- approach: **increase spread** via (additive) targeted covariance inflation (TCl)

- assume **correlation** of Z with model variable Ψ

$$Z'_i(r) = Z_i(r) + \alpha_{\text{TCl}} (\Psi_i(r) - \mu[\Psi(r)])$$

- α_{TCl} serves as scaling factor for “strength” of TCl

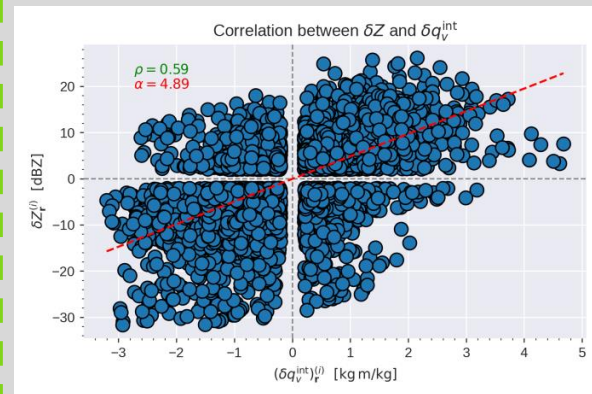
- use $q_v^{\text{int}}(\lambda, \mu, l_0, l_1, \beta) \equiv \int_{\mathcal{A}} d\lambda' d\mu' f_{\beta}(\lambda' - \lambda, \mu' - \mu) \int_{h(l_0)}^{h(l_1)} q_v(\lambda', \mu', h) dh$

- overall idea:

- ◆ spread of q_v “imprinted” onto **spread of Z**
- ◆ assim. “**favors**” members with more humidity: additional q_v (q_r, q_s, \dots) increments via corr.

integral details

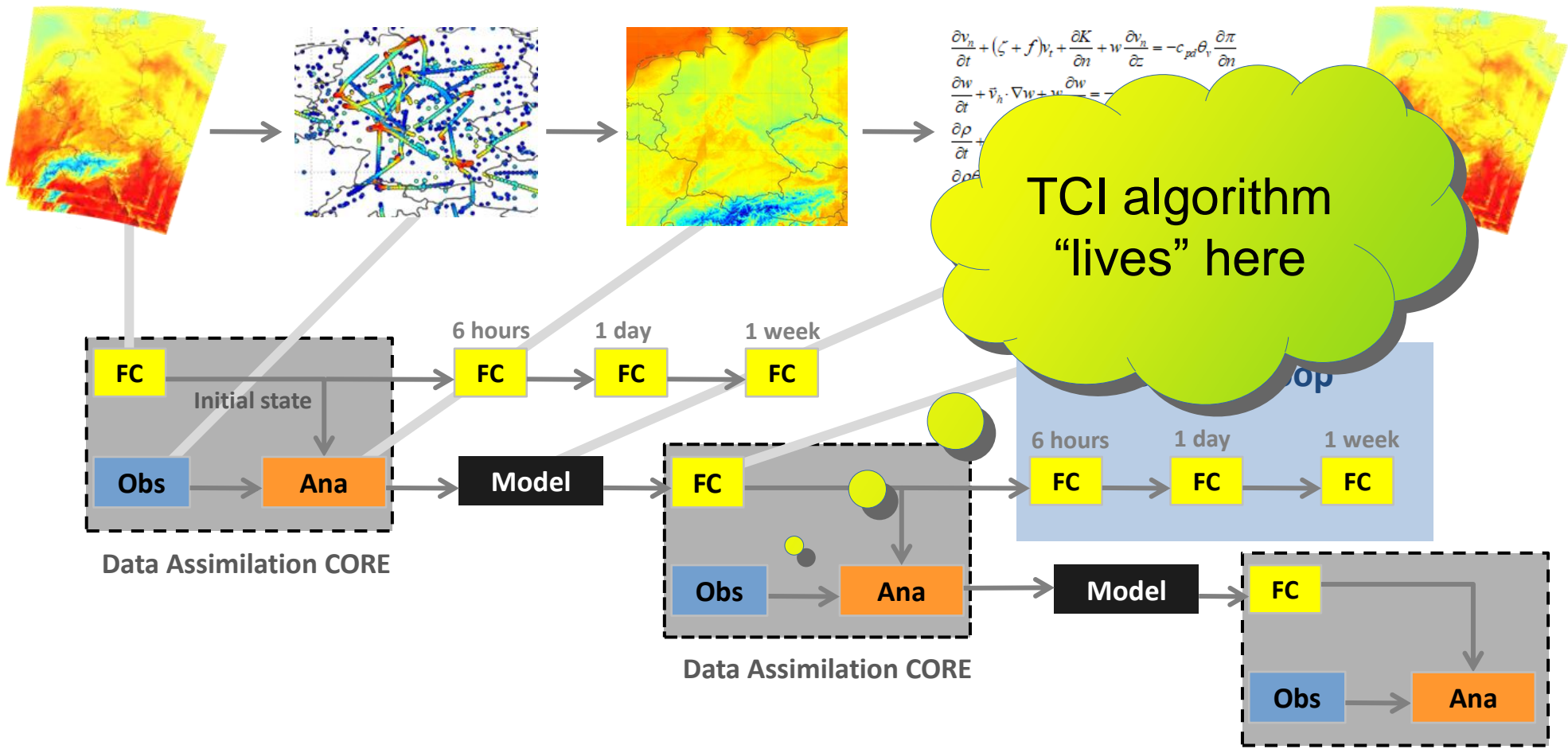
- β : strength of running mean factoring in time uncertainty
- l_0, l_1, β determined via optimization of corr. coefficient



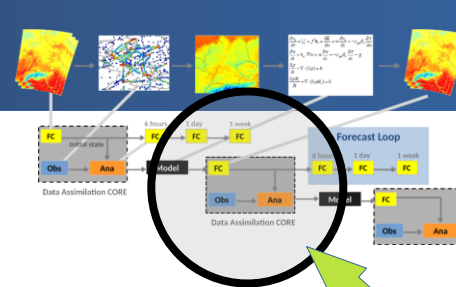
corr. coefficient

- several thresholds for data filtering and process determination
- $\alpha_{\text{TCl}} \rightsquigarrow$ “slope” of correlation

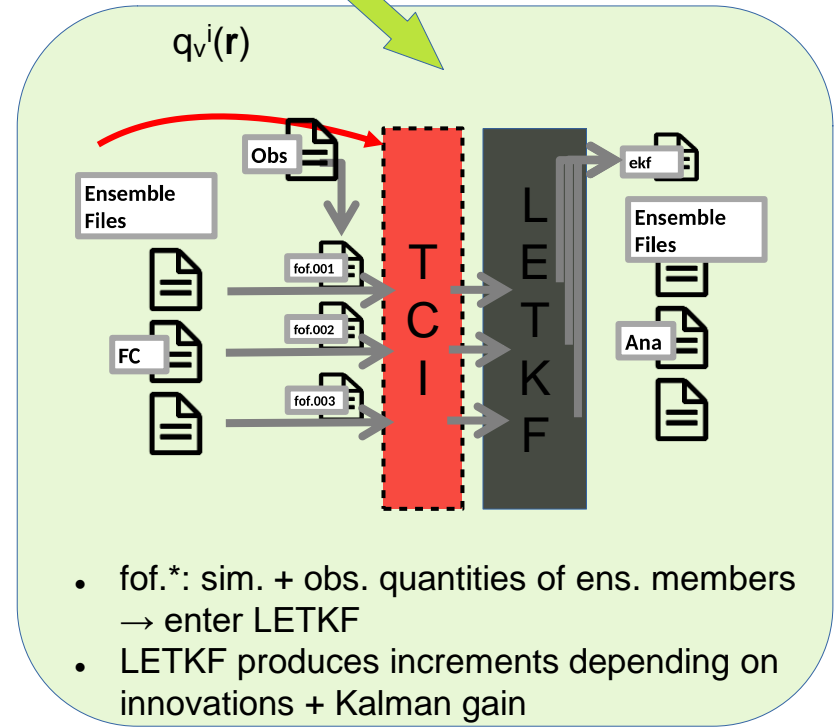
NWP: Assimilation Cycle



TCl: Technical Steps

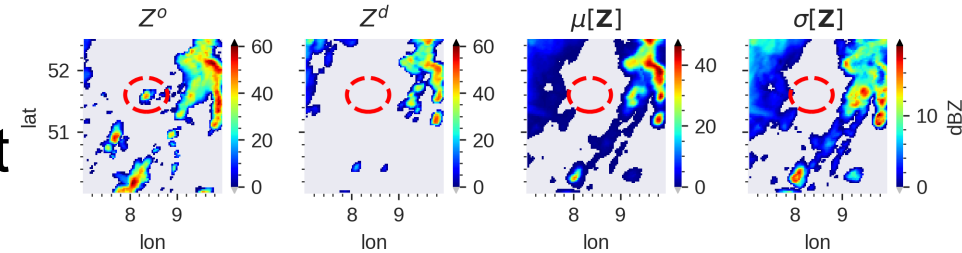


- implemented via **pre-processing feedback (fof) files** before entering the **LETKF**
- apply TCl algorithm and **alter simulated Z** in feedback files
- each member processed separately
- use altered feedback files as input for LETKF

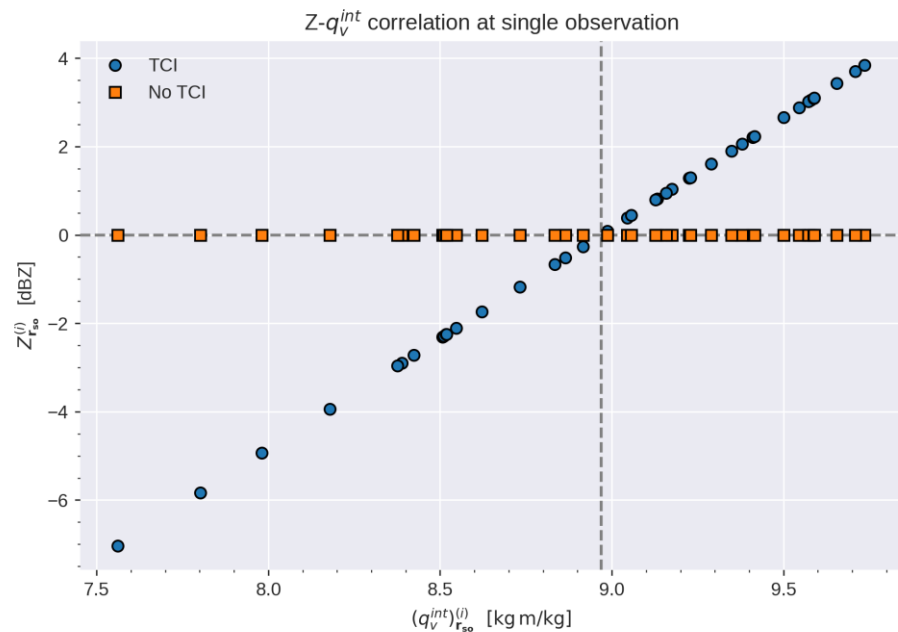


TCI – Single-Observation Experiments

- study effects of TCI in **single-observation** (SO) experiment



- ◆ **assimilating only single reflectivity** at $(51.60^\circ, 8.35^\circ, 1035\text{m})$ for 2019-06-03 at 12 UTC
- ◆ data from Radar station Flechtdorf at **elevation angle 0.5°**
- ◆ **other observation set to ‘passive’** within feedback files
- relevant changes to “default” BACY settings
 - ◆ **obs. error** reduced to 2 dBZ
 - ◆ **vertical localization** increased to $v_{\text{loc}}=10.3$
 - ◆ no multiplicative cov. inflation / no relaxation to prior perturbation



- **without TCI: no spread** in Z at single observation at all
- **with TCI: spread** in Z is produced
 - ◆ result: analysis produces **increments** for Z (linearized) and q_v

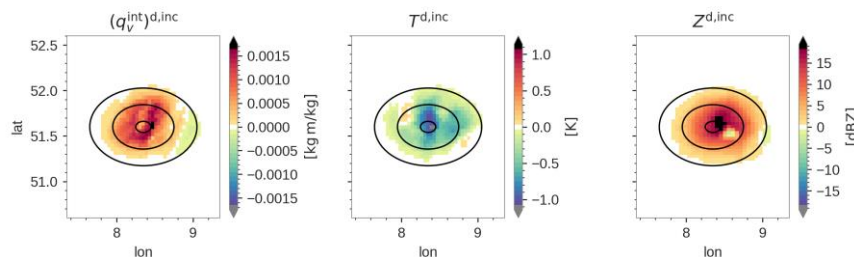
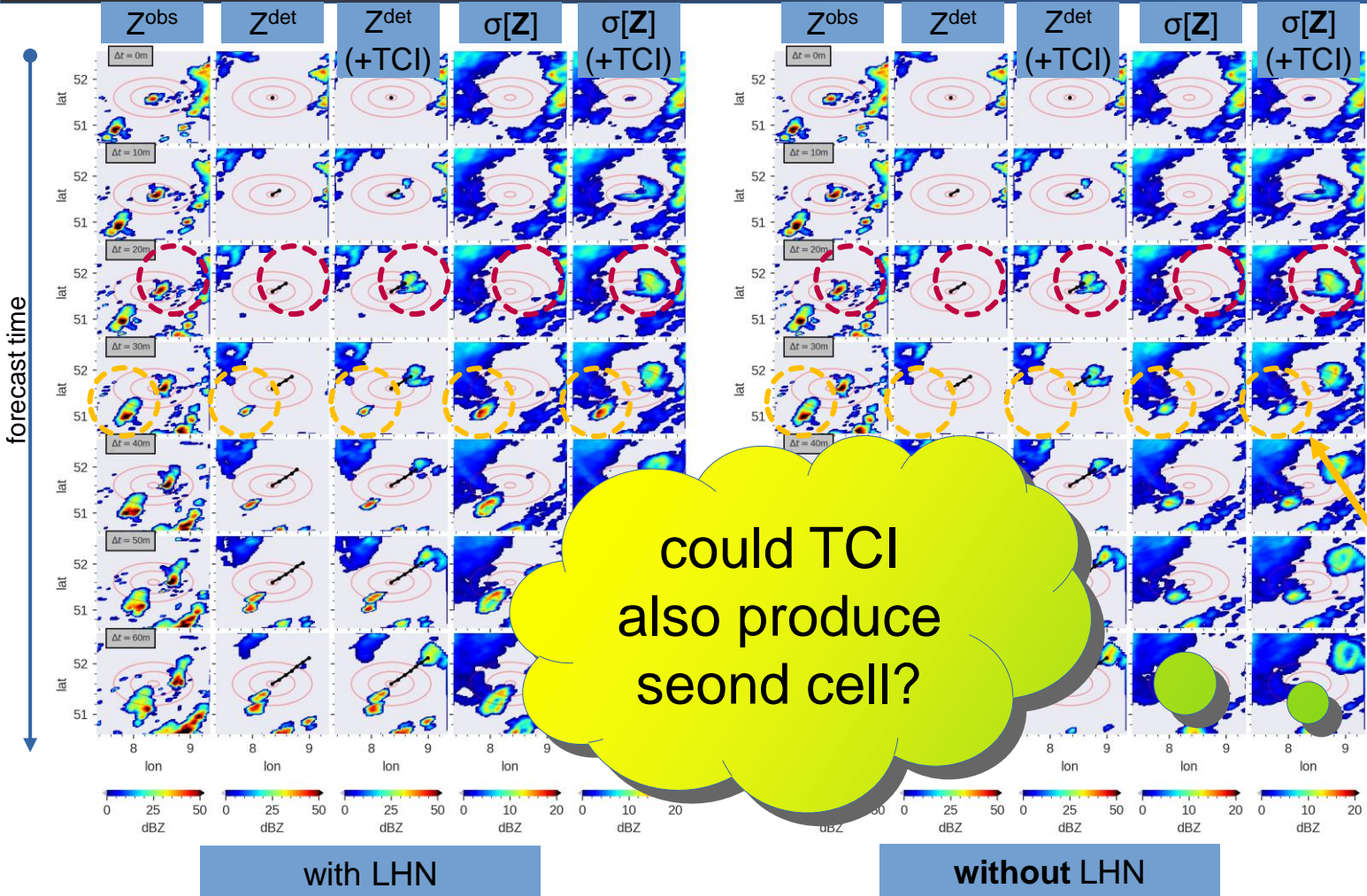


FIG: Increments at ICON-D2 layer 35, concentric circles indicate location of single observation

SO Exp.: Evolution of REFL

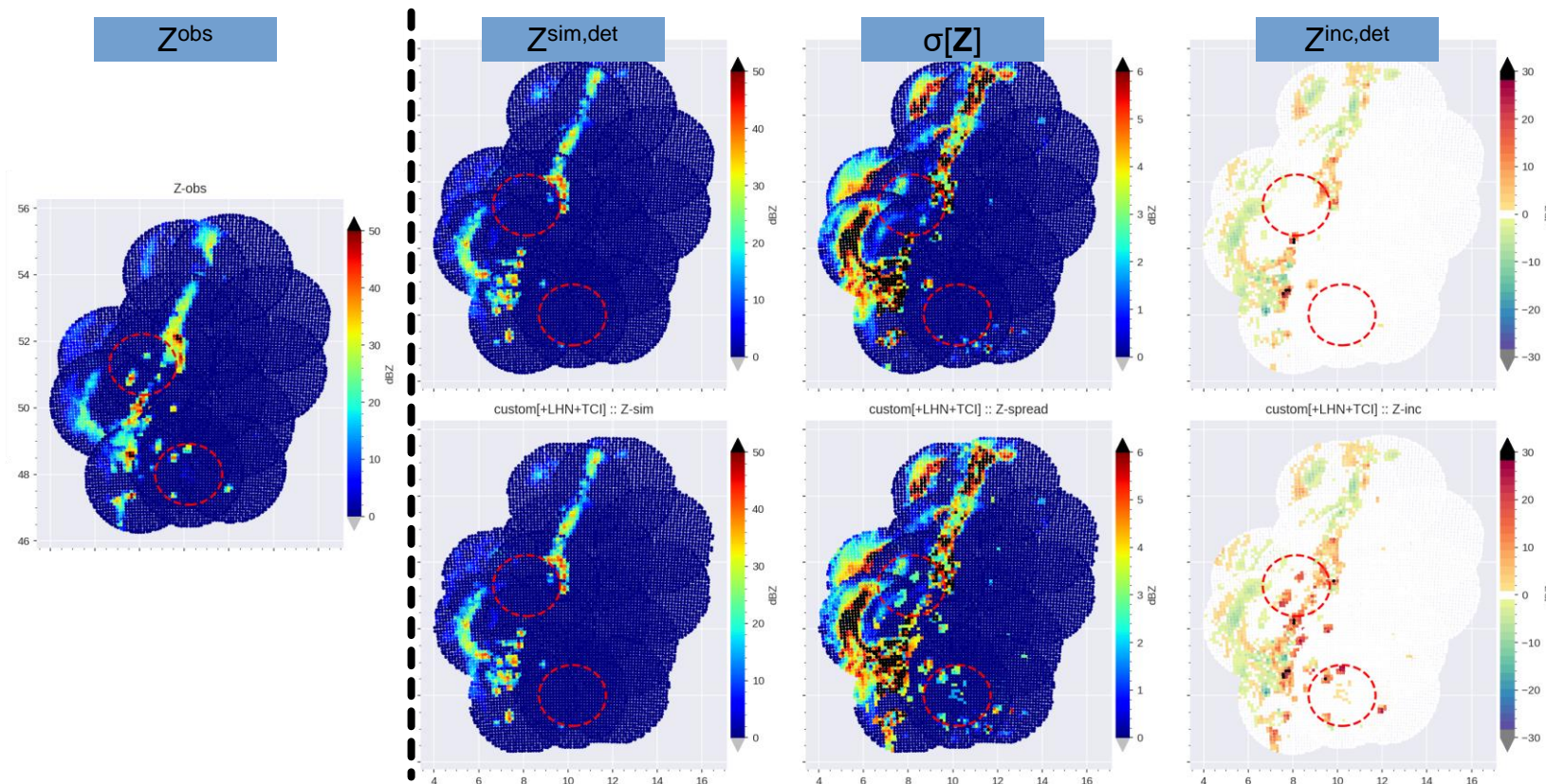


- ass. at 12 UTC followed by 1h free forecast
- “new” cell emerges consistent with observed cell
- second cell only with LHN (out of spatial reach of TCI)

TCI – Beyond Single-Obs Exps.

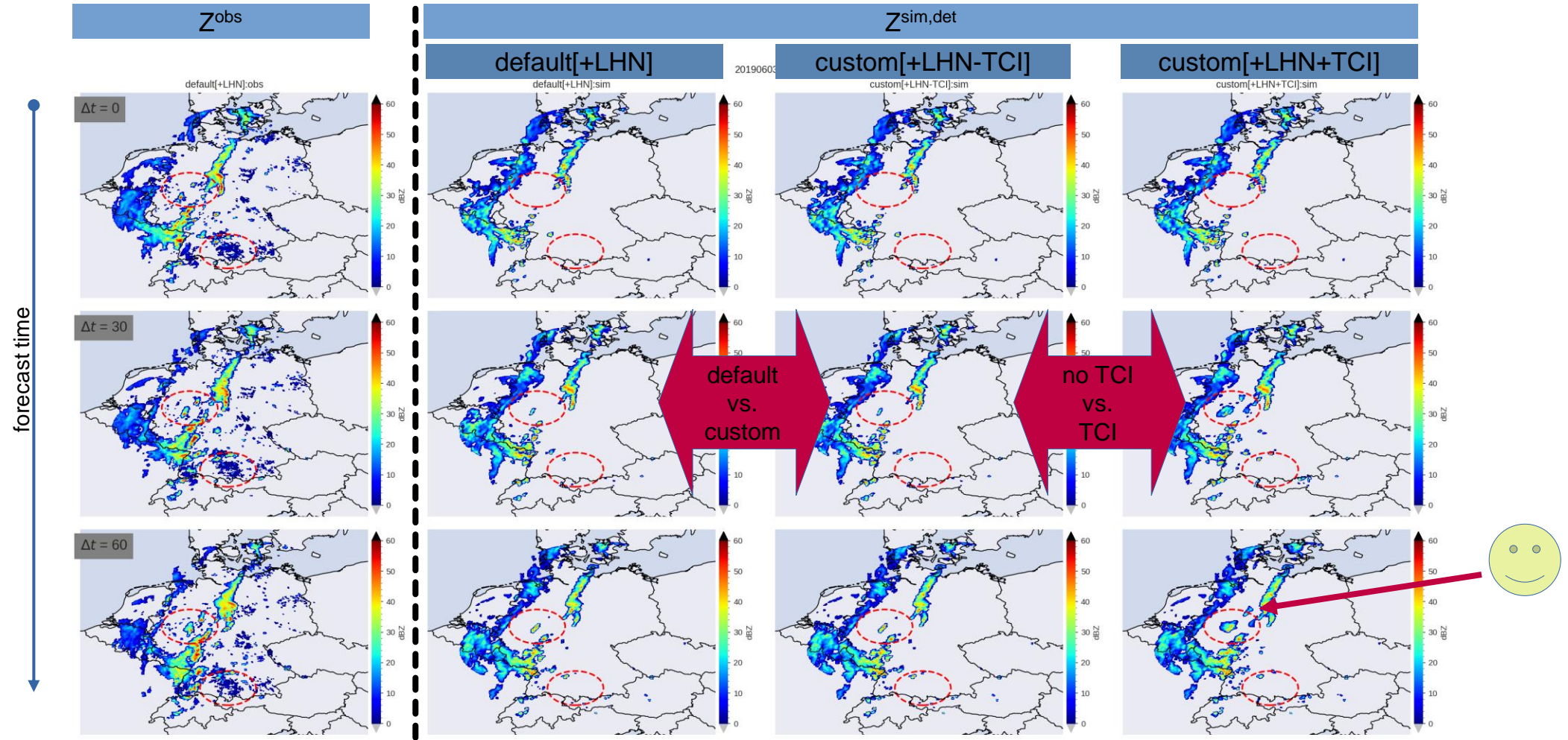
- **previously:** studied effects of TCI in **SO experiments**
- **now:** study effects of **TCI applied to all radar data**
- as before: TCI is applied via **modification of feedback files** before entering LETKF machinery
- prerequisites and effects of TCI application at **r**:
 - ◆ **discrepancy** between observed/simulated REFL
 - ◆ **small ensemble spread**
 - ◆ **modify Z** for all ensemble members via integrated q_v correlation

- “default”: (mostly) default BACY configuration
 - ◆ assimilation of conv. data and REFL (at several elevations)
 - ◆ LHN may be turned on/off [\pm LHN]
 - ◆ assimilation takes place on 2019-06-03 at $t_0 = 12$ UTC
- “custom”: minor changes w.r.t. default configuration
 - ◆ serves as reference for assessing direct impact of TCI
 - ◆ increased first-guess check for REFL
 - ◆ TCI may be turned on/off [\pm TCI]



- TCI produces spread → additional increments for REFL

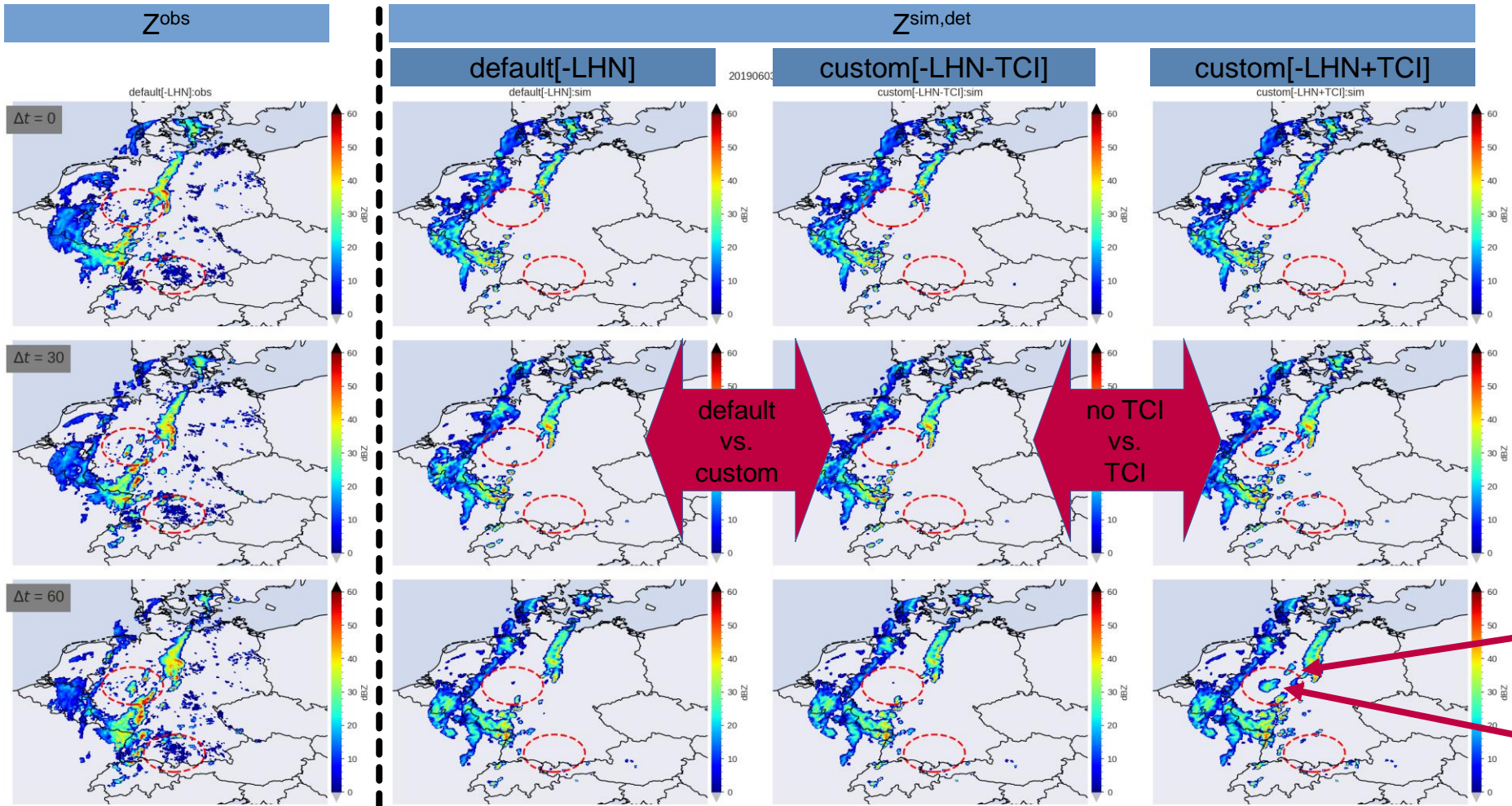
Evolution of REFL (with LHN)



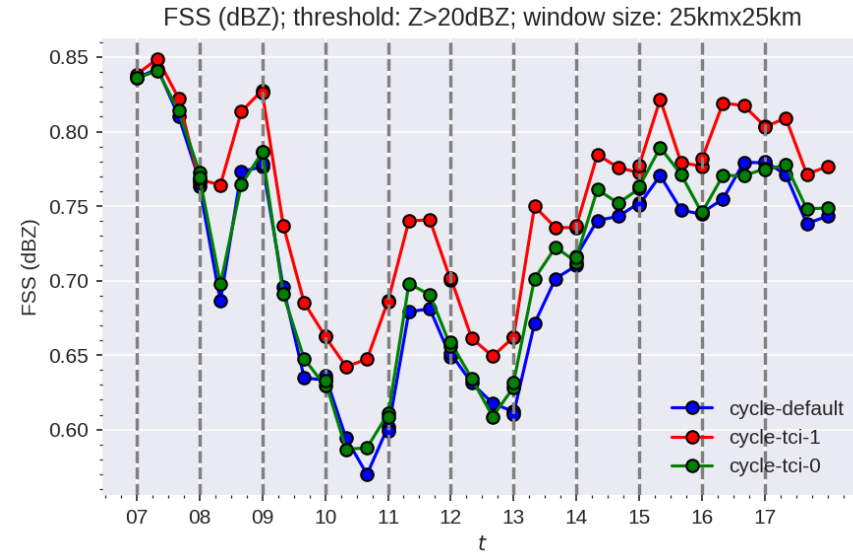
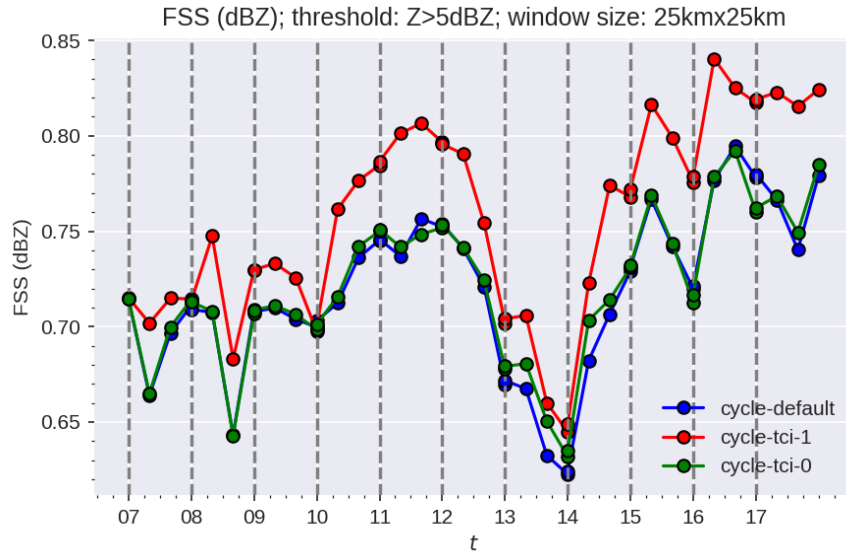
Evolution of REFL (without LHN)



forecast time



Verification: Fractional Skill Score

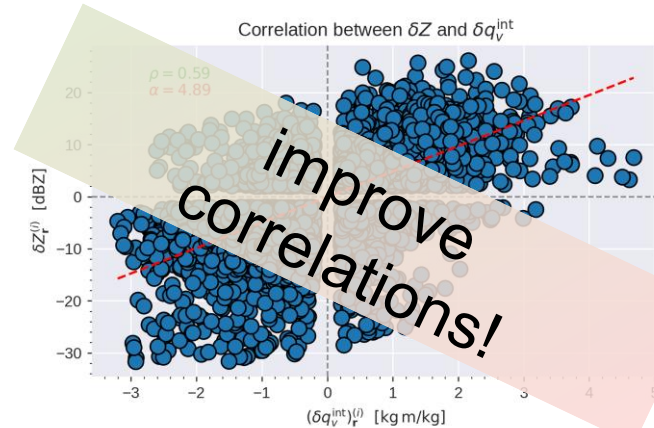


default[+LHN]
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- performing cycle starting at 7 UTC
- TCI applied at each assimilation (hourly)
- FSS for two dBZ thresholds shown

positive impact
of TCI on FSS

- overall, TCI results are promising
 - ◆ production of “new” cells (consistent)
 - ◆ positive impact on FSS
- AIREP/TEMP observation error
 - ◆ negative impact of TCI
 - ◆ contribution (positive/negative) to standard deviation
 - ◆ time/spatial/process dependence of correlations
- optimize the capturing and use of correlations of TCI method
 - ◆ better data filtering/pre-processing necessary
 - ◆ towards more “process-aware” TCI



- assim. of information on **convective initiation**
 - ◆ employ total column water vapor obtained from satellite data
 - ◆ also apply TCI-like approach (?)
- assim. of data from **Commercial Microwave links (CMLs)**
- assim. of **nowcasted states**
 - ◆ employ advanced nowcasting for assim. nowcasted states (REFL) with KENDA

Thank you for your attention!