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Nowcasting of thunderstorm severity with Machine Learning in the Alpine Region

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Hering A., Nerini D., Nisi L., Sassi M., Germann U.



COALITION-3 – Faces



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COALITION
project lead



Joel Zeder
Intern in 2018/2019
COALITION-3



Luca Nisi
COALITION-1
forecaster



Alessandro Hering
TRT developer



Elena Leonarduzzi
Intern in 2016
COALITION-2



Urs Germann,
head of MDR



Lorenzo Clementi
head of MDRD



Loris Foresti
precip. Attractor
pySTEPS



Daniele Nerini
pySTEPS



Marco Sassi
NWC-SAF,
deployment

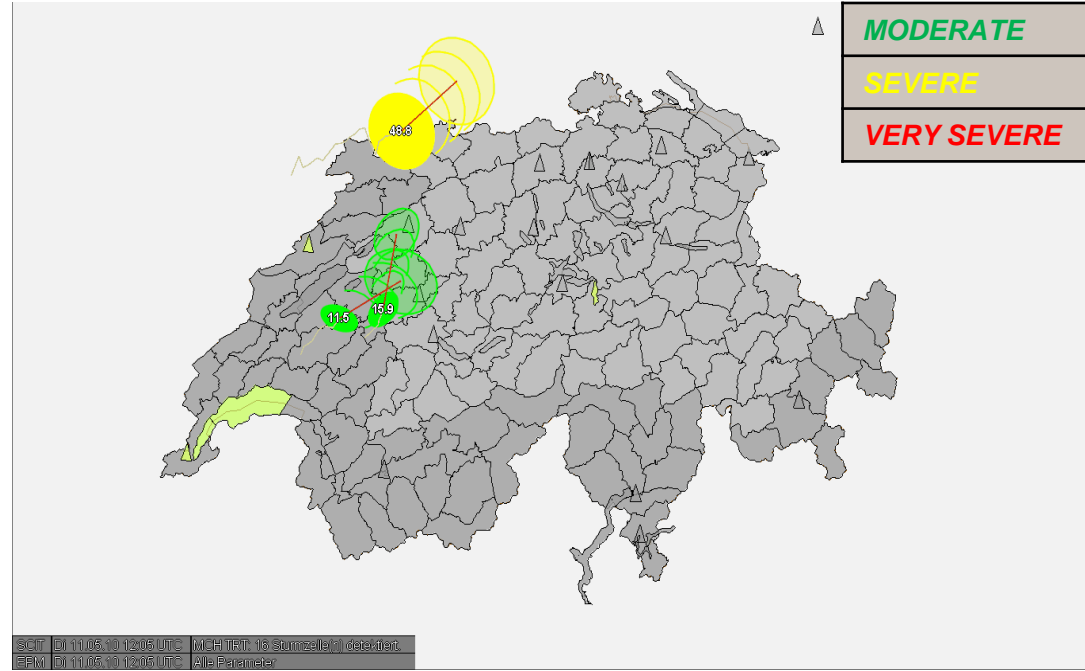


Lea Beusch
Intern in 2017
Satellite Rainfall

Starting point before our study

- Thunderstorm cells are identified by radar with adaptive reflectivity thresholds.
- Thunderstorm intensity is expressed as heuristic TRT rank (in colours).
- Future position is extrapolated with the current motion.
- TRT rank is kept constant.
- Multi-sensor cell parameters are monitored.
- Automatic warning suggestions are generated for warning regions.
- Warning suggestions are modified by forecasters.

Example for Thunderstorm Radar Tracking (Hering et al) 11 May 2010 for Switzerland. TRT rank expressed in colours.

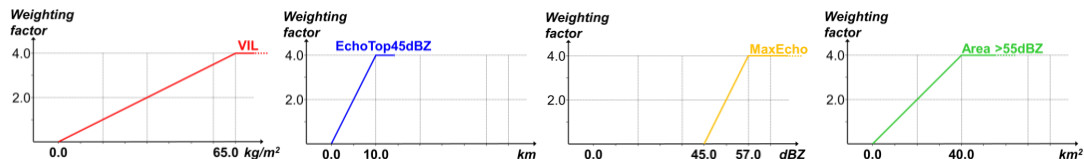




TRT rank (heuristic thunderstorm intensity)

Cell severity ranking: Single numerical score [0.0, 4.0] based on cell attributes integrated with a weighting scheme (fuzzy logic like approach):

- Vertical integrated liquid **VIL**
- Median of 45 dBZ Echo Top altitude **ET45**
- Maximum cell reflectivity **MaxEcho**
- Area of cell reflectivity ≥ 57 dBZ **area57dBZ**

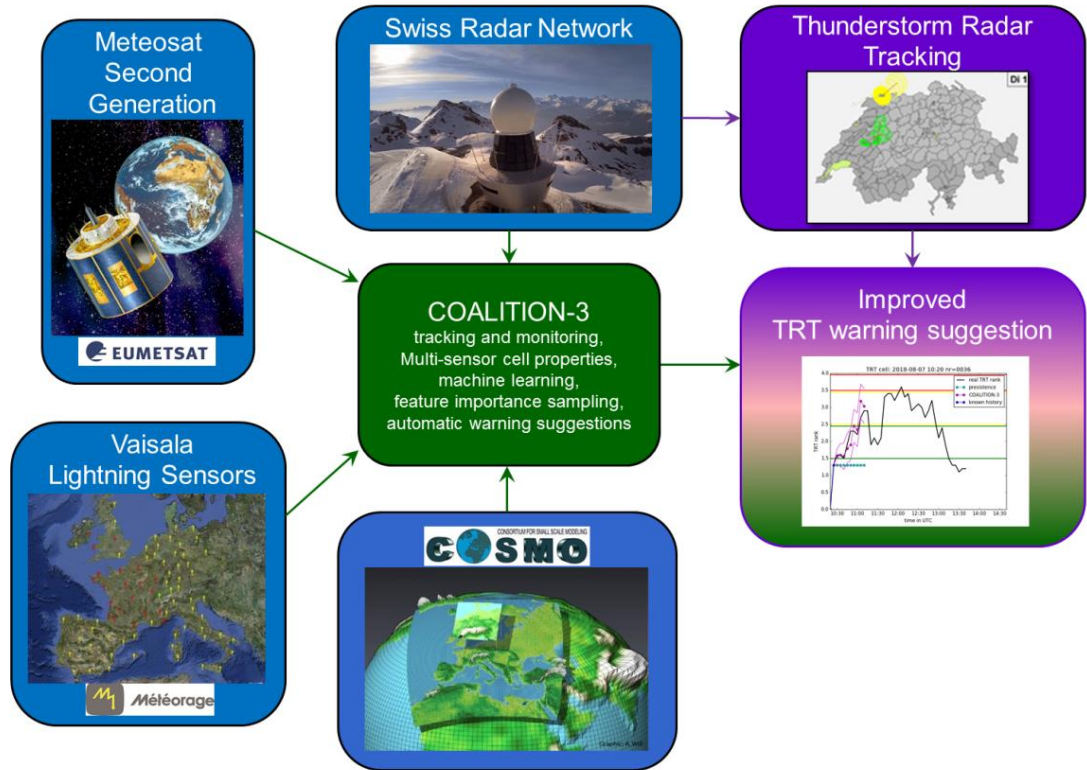


Severity	RANK
<i>DEVELOPING</i>	$RANK = [1.2 - 1.5[$
<i>MODERATE</i>	$RANK = [1.5 - 2.5[$
<i>SEVERE</i>	$RANK = [2.5 - 3.5[$
<i>VERY SEVERE</i>	$RANK = [3.5 - 4.0]$

$$RANK = \frac{2.0 * g(VIL) + 2.0 * g(ET45med) + 1.0 * g(dBZmax) + 2.0 * g(area57dBZ)}{7}$$

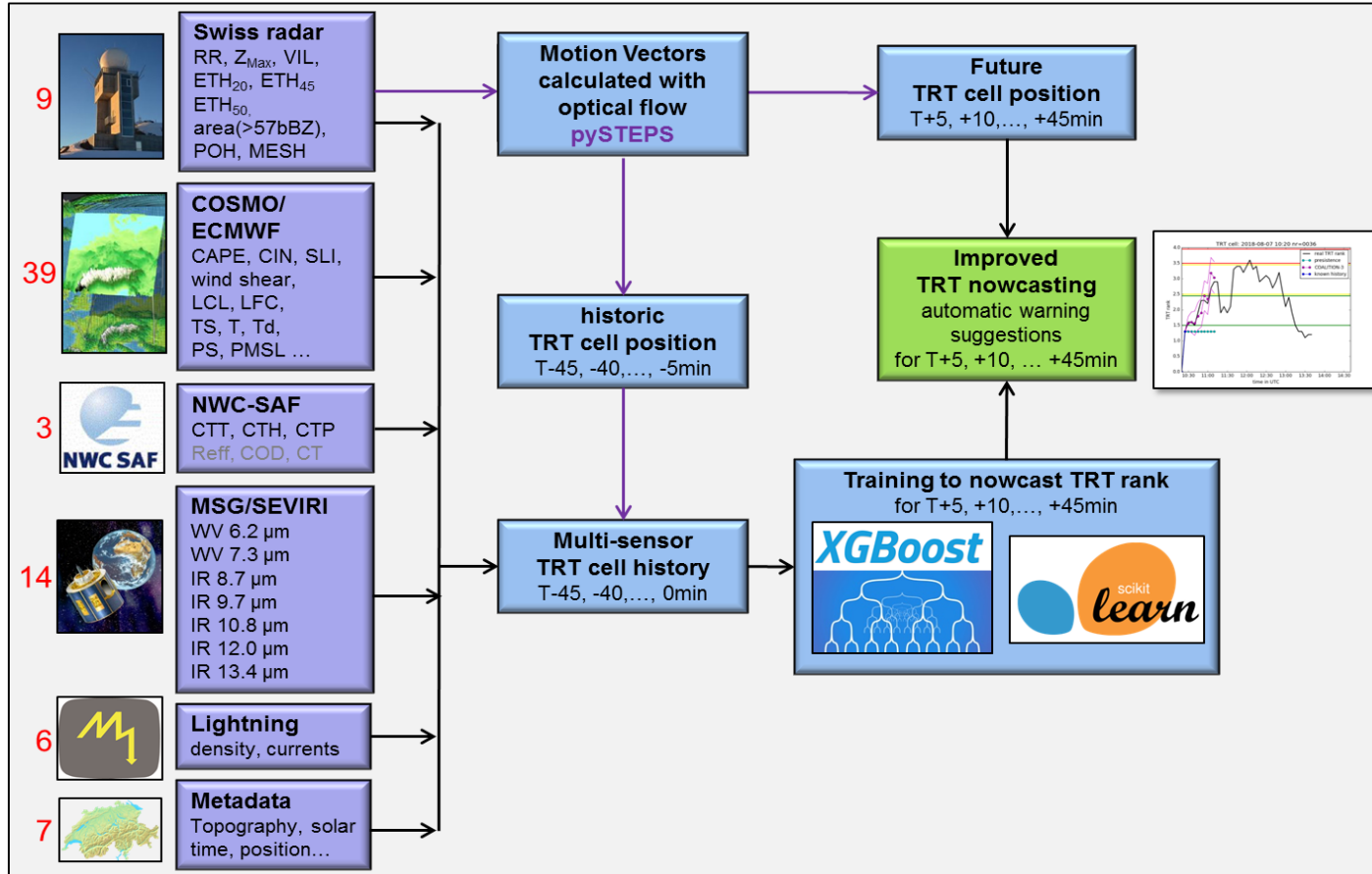
Goals for this study (COALITION-3)

- Improved automatic TRT warning suggestions
 - a) thunderstorm position
 - b) thunderstorm intensity (expressed as TRT rank)
- Multi-sensor retrieval (satellite, radar, lightning, COSMO, meta-data)
- Update cycle 5 min
- Nowcast up to 45 min
- Long warning lead times
- Quantitative, customer oriented output





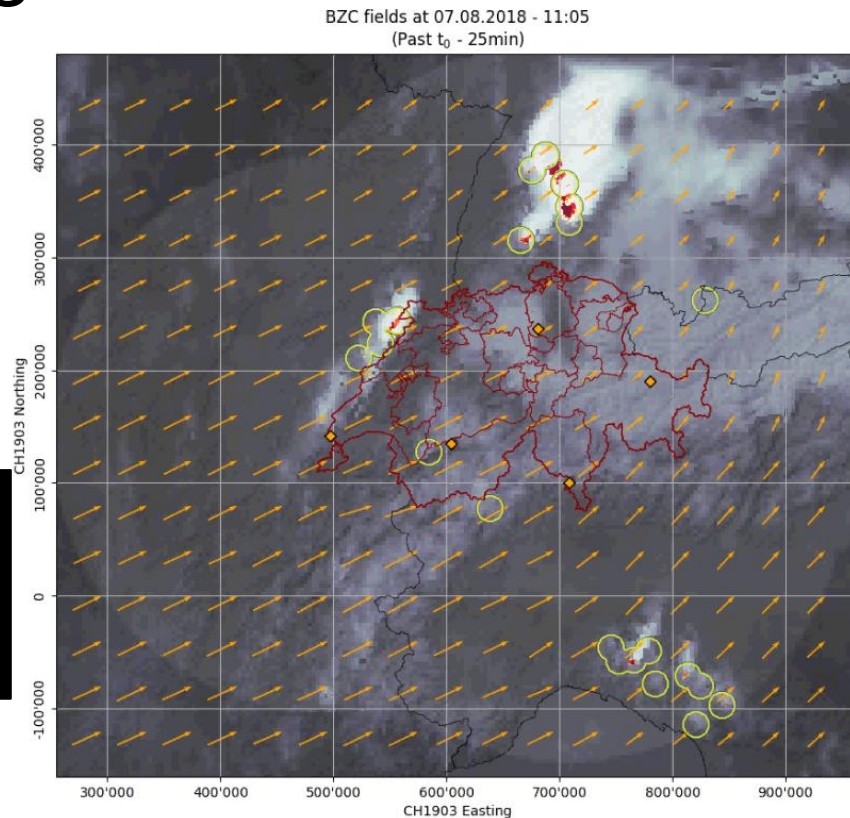
Flowchart of COALITION-3





Retrieving cell histories

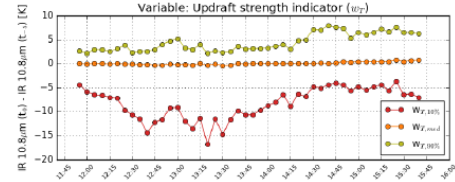
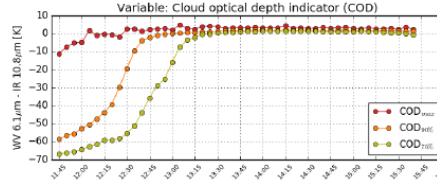
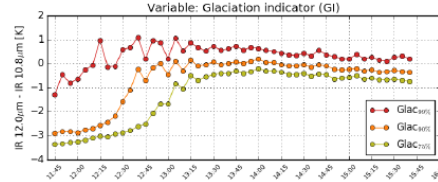
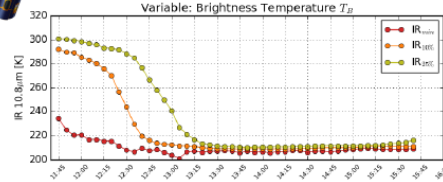
- Calculating motion vectors with pySTEPS.
Feature selection with Shi and Tomasi (1994).
Tracking with pyramidal implementation of Lucas-Kanade (1981) feature tracker algorithm
- Initial cell identification with TRT algorithm.
- Track position 45 min backwards in time to retrieve cell history (predictors).
- For each predictor following statistics are calculated in a 23 km diameter circle:
 - Mean, Sum, Standard Deviation
 - Minimum and maximum value
 - 1%, 5%, 25% 50%, 75%, 95%, 99% Quantiles
 - Number of pixels with certain properties, e.g. precipitation > 0mm/h
- Monitor the thunderstorm intensity (TRT rank) up to 45 min into the future (truth for training)



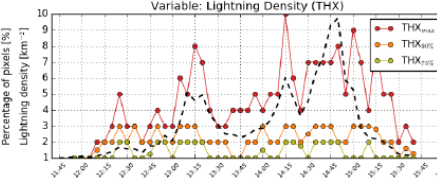
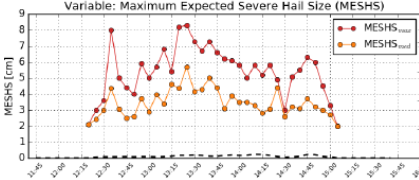
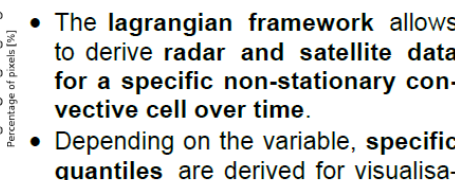
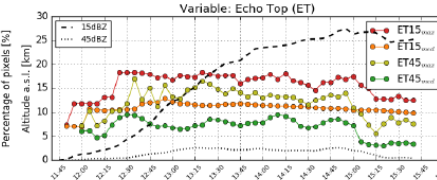
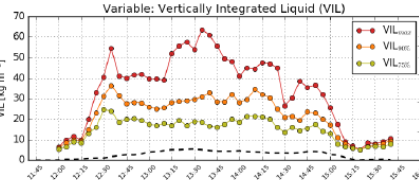
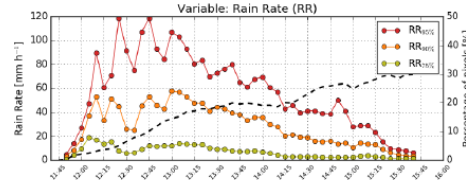
An example of typical cell histories

4. SEVIRI, Radar, and Lightning Data within Lagrangian Framework

SEVIRI Data



Radar and Lightning Data



- The **lagrangian framework** allows to derive **radar and satellite data** for a **specific non-stationary convective cell** over time.
- Depending on the variable, **specific quantiles** are derived for visualisation. Furthermore, the **fraction of covered pixels** is shown.
- This time series data can be further **exploited for automated nowcasting applications**, for example using machine learning algorithms.



Input data - Cardinality

Overall, the final count of features is composed of:

- ➞ • **68** input variables (RZC, IR 10.8 μm , PV at 500 hPa, Slope ...)
- ➞ • **10** time steps ($t_0, t_{-5\text{min}}, \dots, t_{-40\text{min}}, t_{-45\text{min}}$)
- ➞ • **12** Statistics (Sum, Mean, Quantiles, Pixel counts ...) plus some conditional statistics of Radar variables
- $68 \times 10 \times 12 = \mathbf{8'160}$ possible input parameters
- Training period summer 2018, about **10'000** observed cell histories



DATE_TRT_ID	RZC_stat -15 PERC75	RZC_stat -15 PERC95	RZC_stat -15 PERC99	RZC_stat -15 MAX	RZC_stat -10 SUM	...	TOPO_ALTITUDE_stat -20 PERC01
201804291235_2018042910550020	-22.248047	0.000000	0.000000	0.000000	-3424.883789	...	-11.875000
201804291235_2018042912000024	-1.099609	44.785156	94.321289	87.427734	1311.467773	...	-88.125000
201804291235_2018042912000025	-0.245117	-1.833008	-31.992966	-32.000977	-314.942383	...	-5.800003
201804291235_2018042912050022	-0.861328	3.979492	14.365040	23.396484	103.207031	...	-16.274994
201804291235_2018042912100033	-8.855469	-68.509766	-41.690041	0.000000	-1397.562500	...	-0.375000
201804291235_2018042912200024	0.636719	0.706055	1.180468	0.000000	254.544922	...	0.000000
201804291235_2018042912250029	0.804688	0.711914	-1.006836	1.375000	180.148438	...	-17.625000
...
201809181900_2018091819000008	-0.435547	-10.399414	-14.435349	-13.000000	-731.588867	...	8.199997
201809181900_2018091819000033	0.289062	2.375000	6.605469	-2.296875	208.524414	...	57.875000

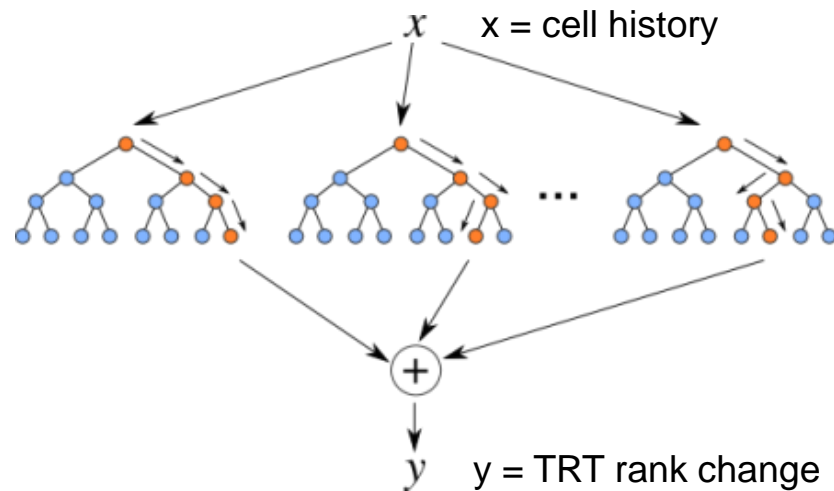
[34 rows x 10128 columns]

TRT Cells identified by (unique) ID and Date/Time

Machine learning nowcasting

XGBoost model (gradient boosted trees)

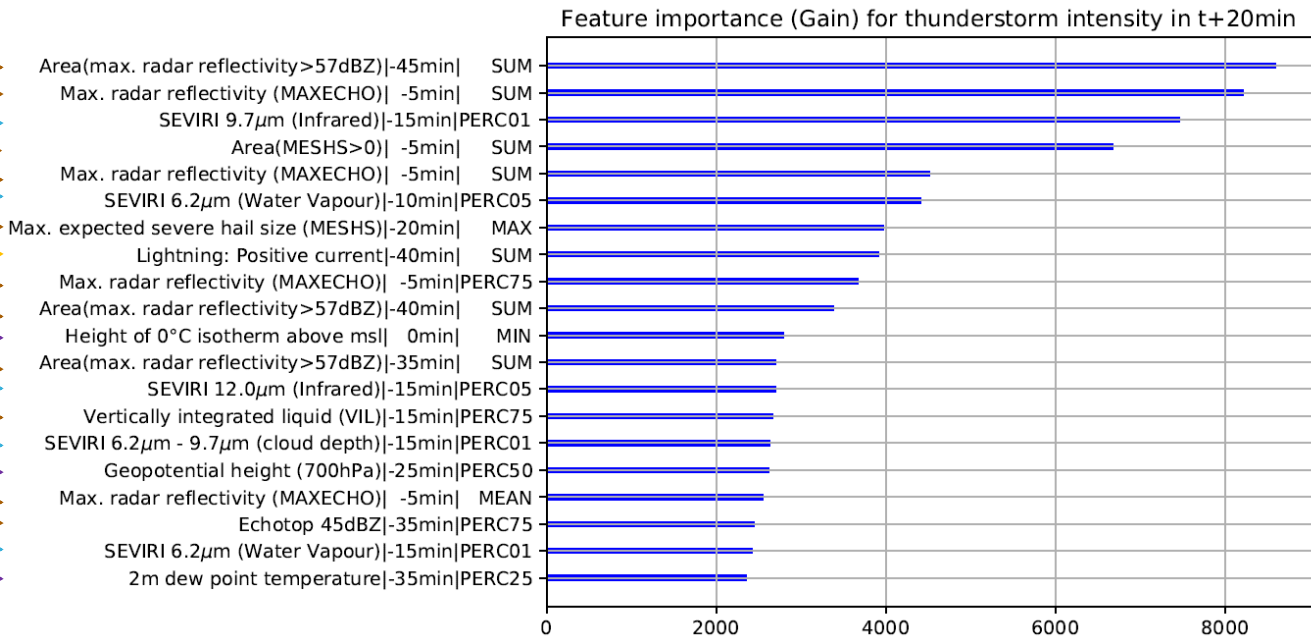
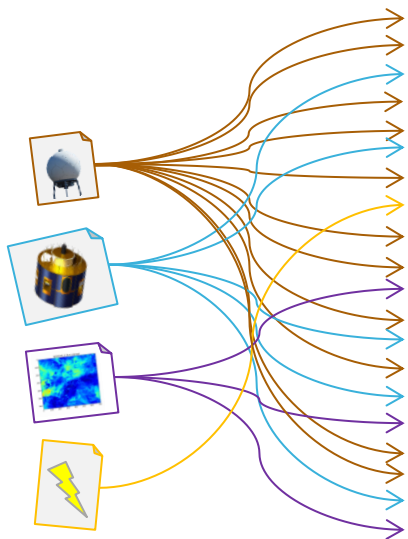
- roughly **equivalent to random forest** with an ensemble of weak-learners (trees) as model
- but where during **training**, trees are added until the objective function converges
- where the mean squared error **MSE** is the **loss function**, and the **regularisation** term is the **sum of scores** at the leaves.
- XGBoost also produce an estimate how important an input variable is.
- Number of input parameter could be reduced to 750.





Statistical learning – Feature selection

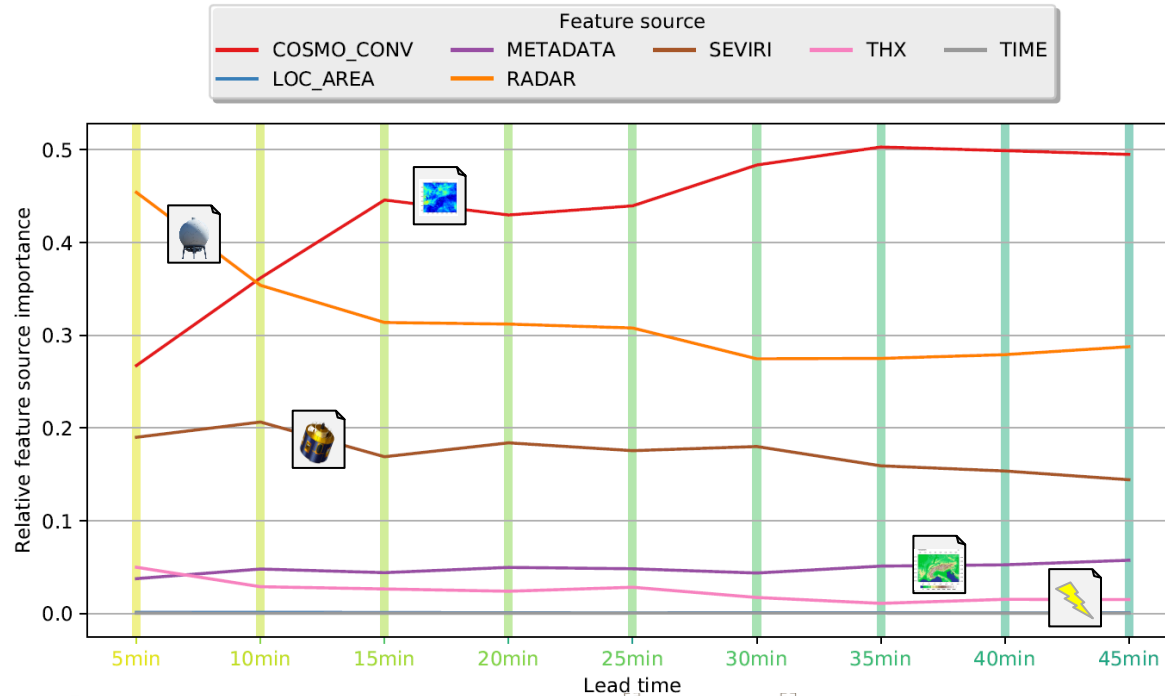
Top 20 features according to XGB model for t_{+20min} lead time:





Statistical learning – Feature selection

Relative importance of feature source:



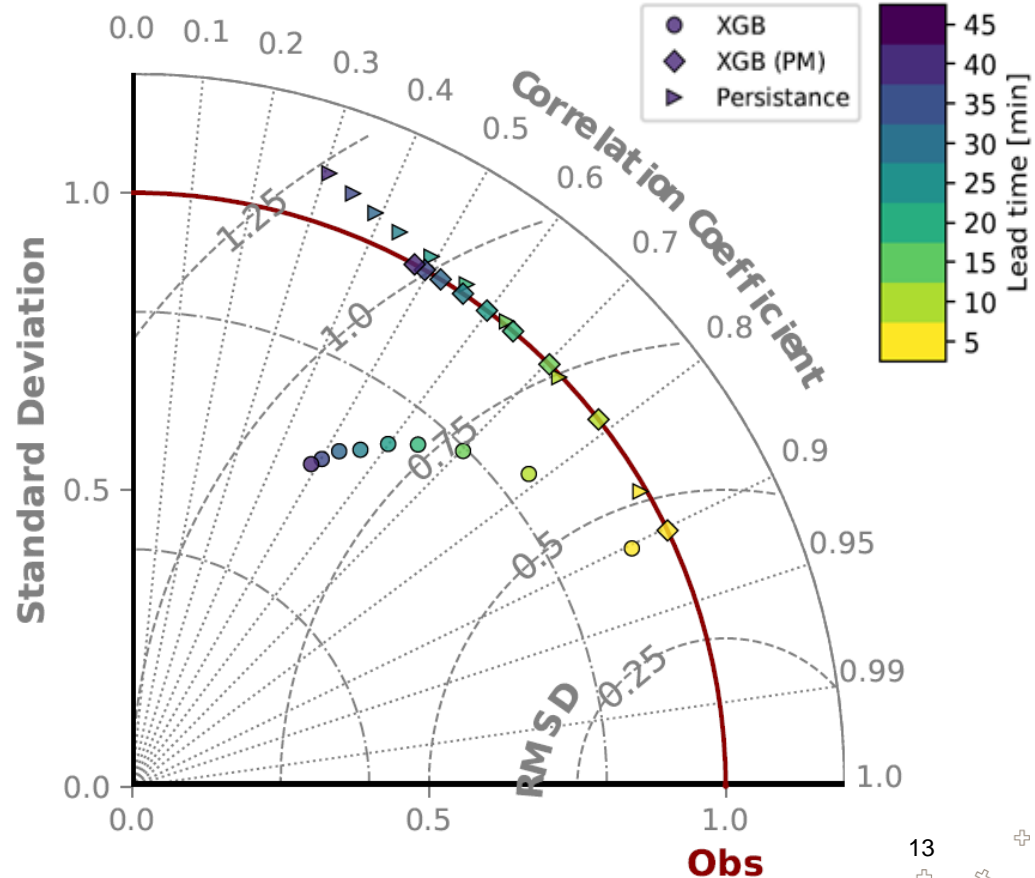


Model evaluation

- XGBoost nowcasts (XGB, circles) always have the smallest RMSD (skilful up to t_{+45min})
- Probability matching (diamonds) is used to correct for the standard deviation (skilful up until t_{+35min})
- Probability matched results have a smaller RMSD as persistence (triangles) for all lead times.

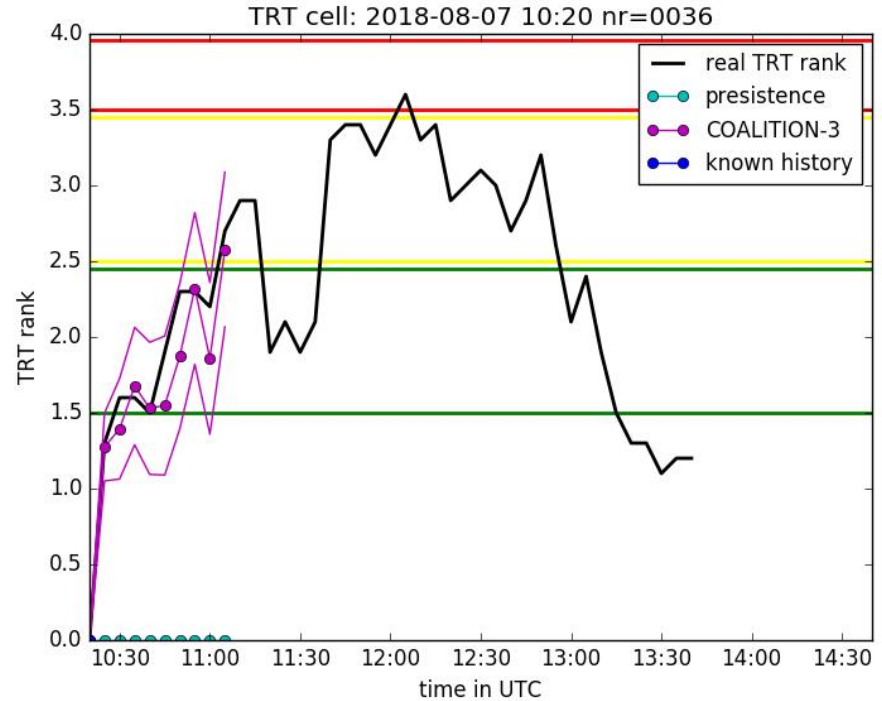
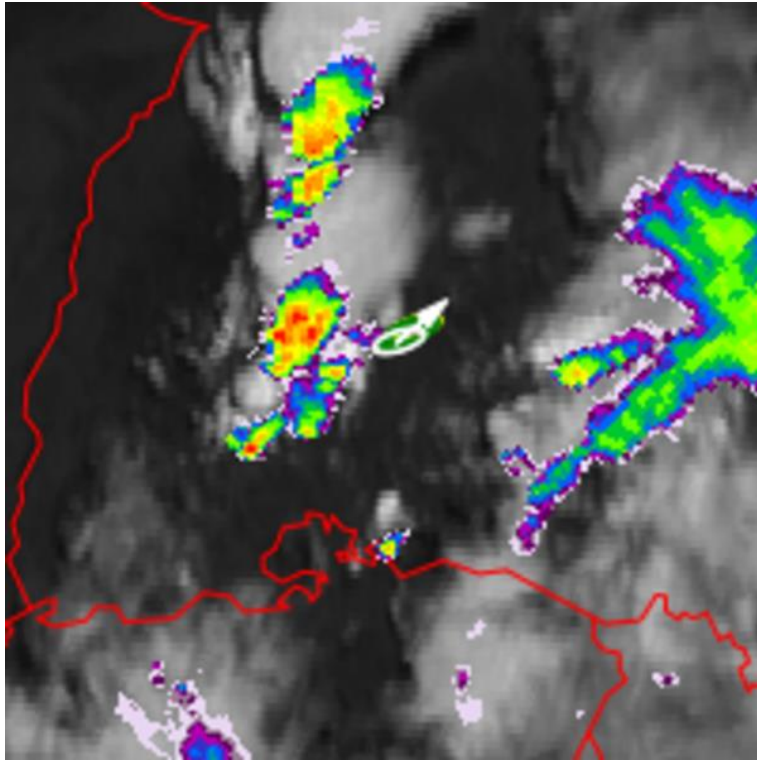
Forecast times with same RMSD

Persistence	5	10	15	20	25
XGBoost, PM	6	14	22	30	42
XGBoost	6	18	40	>45	>45



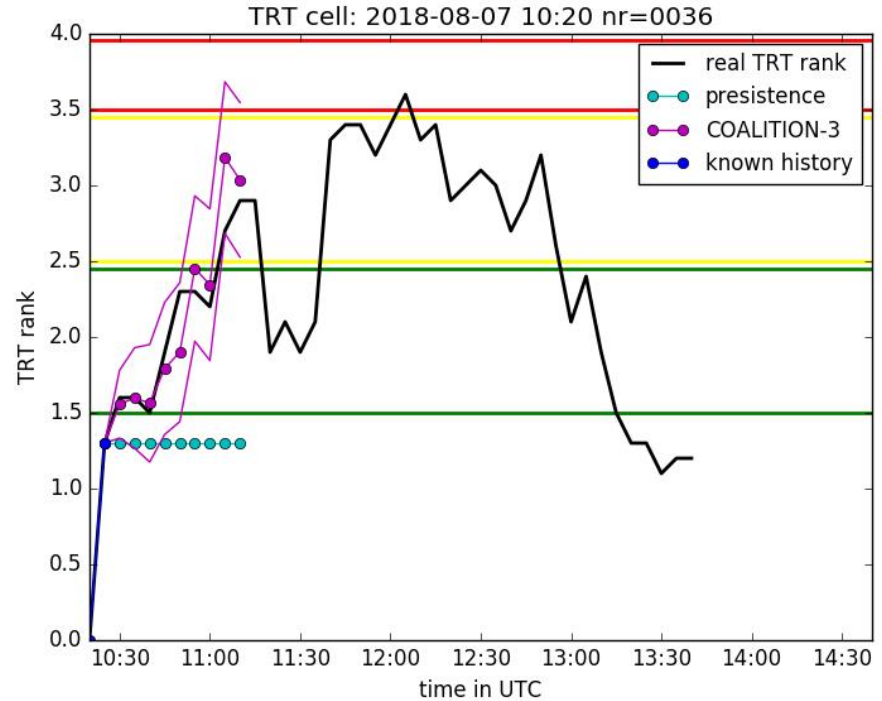
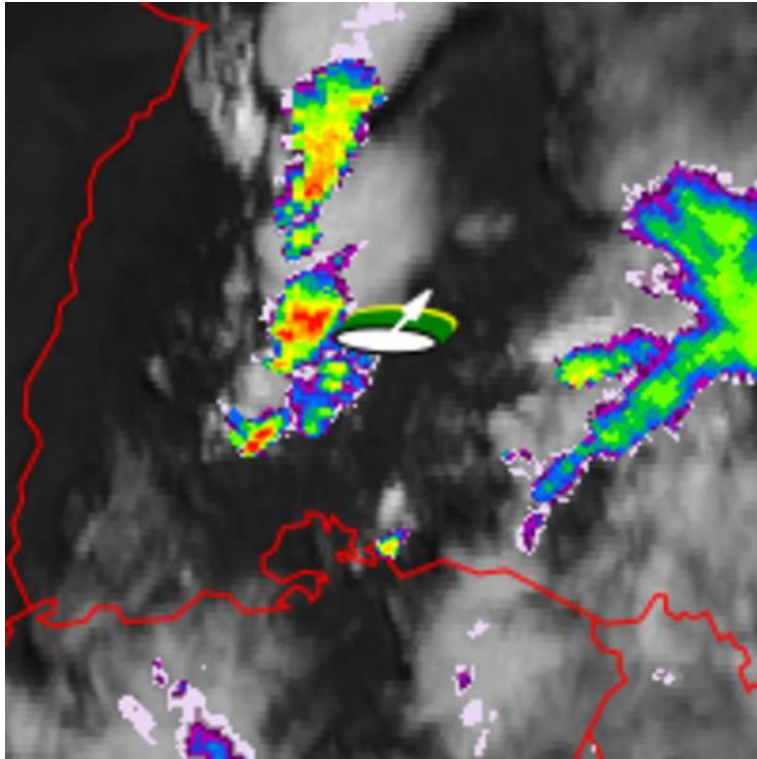


Visualization of the nowcasting results



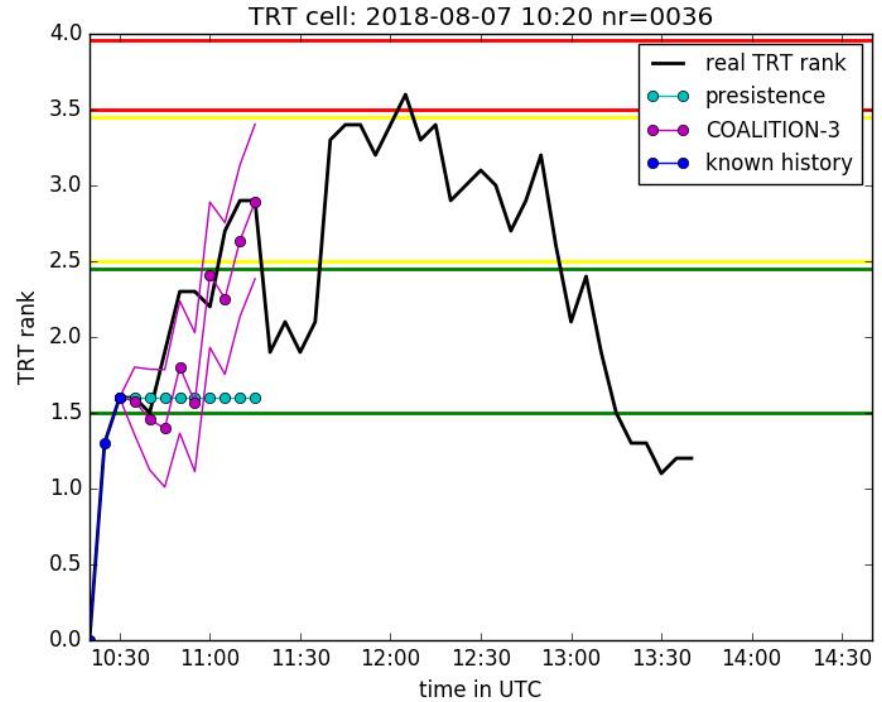
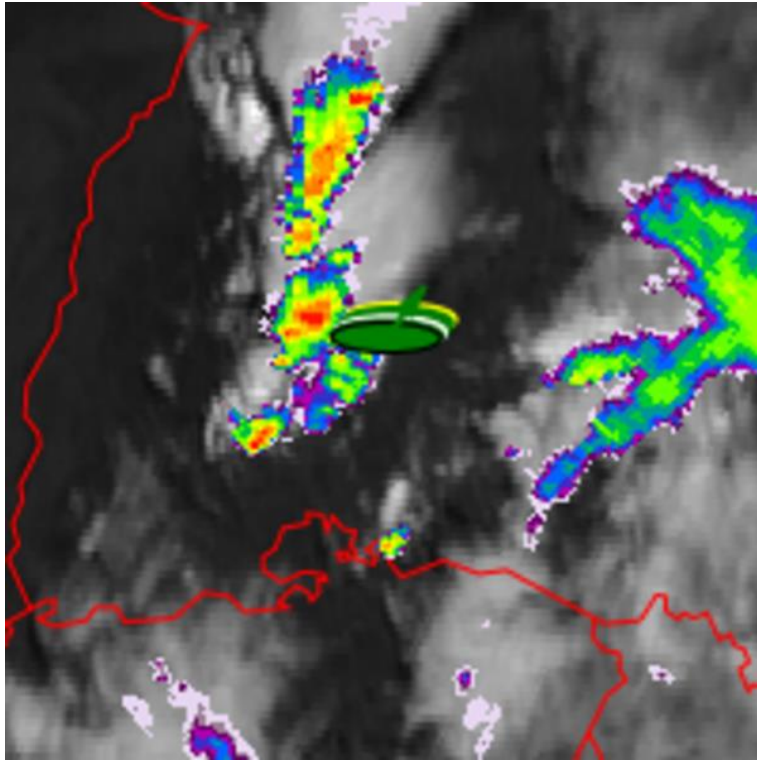


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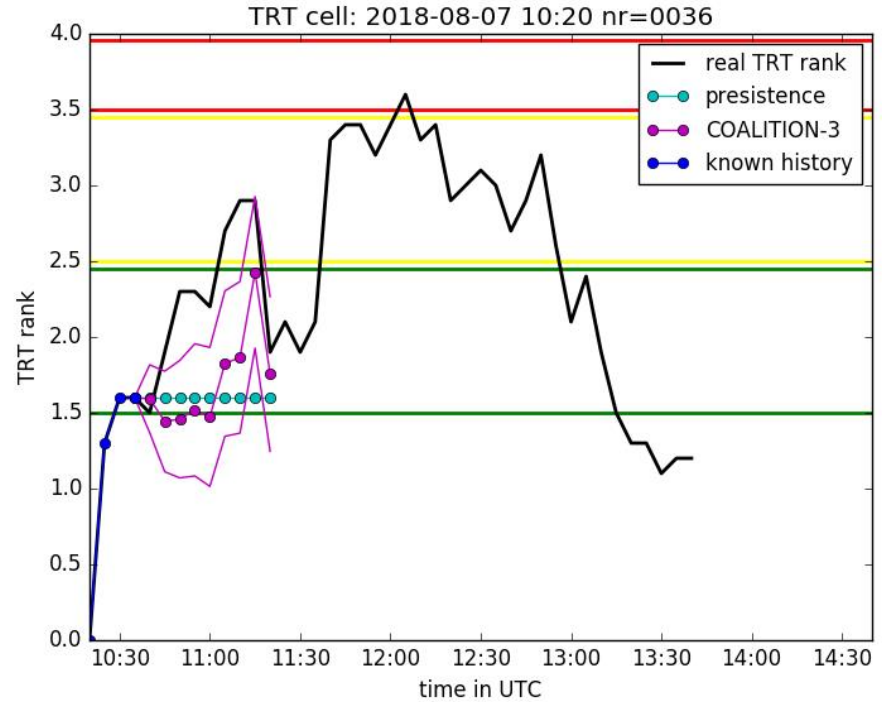
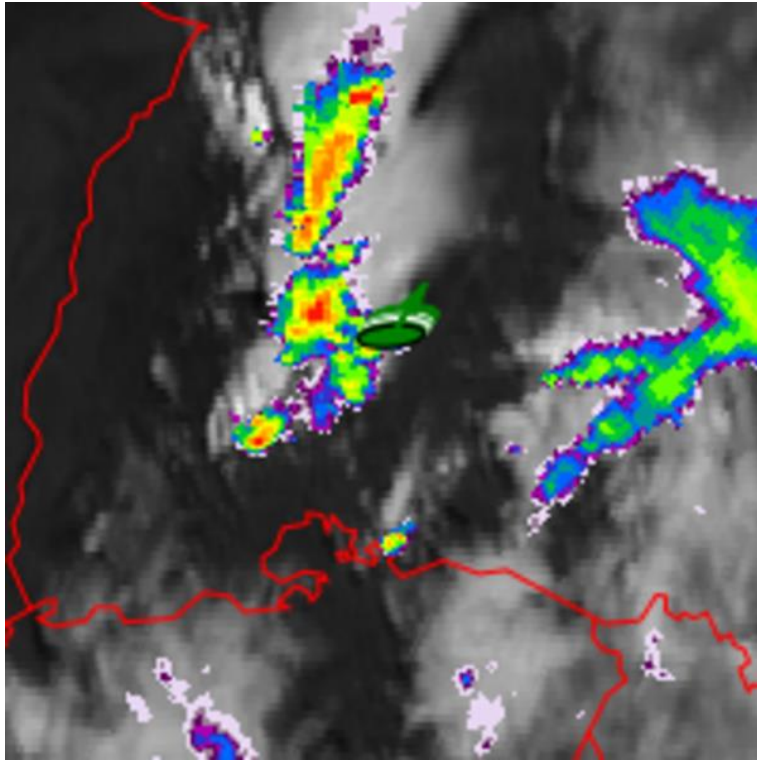


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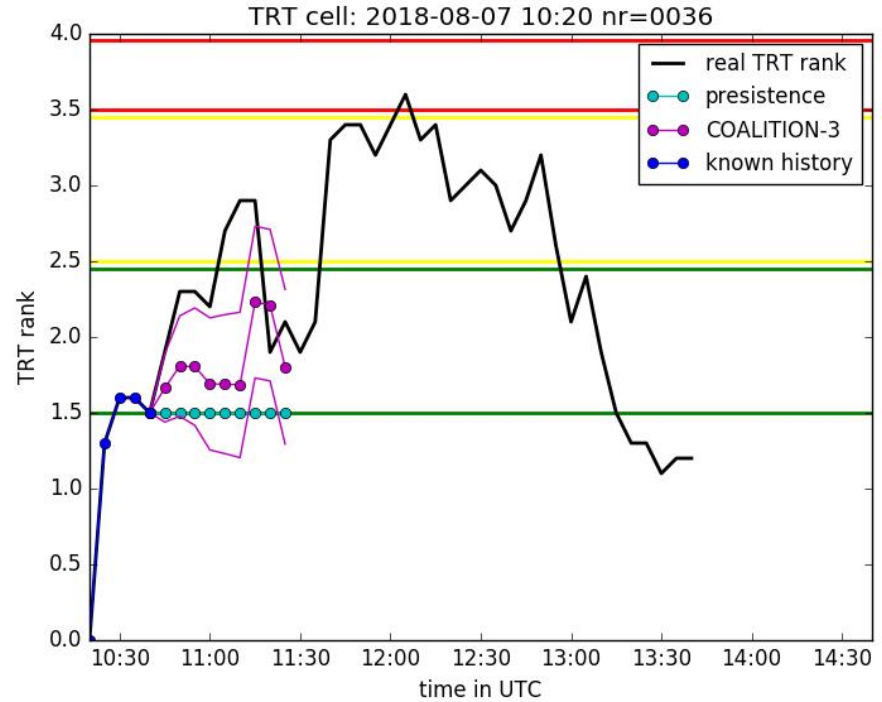
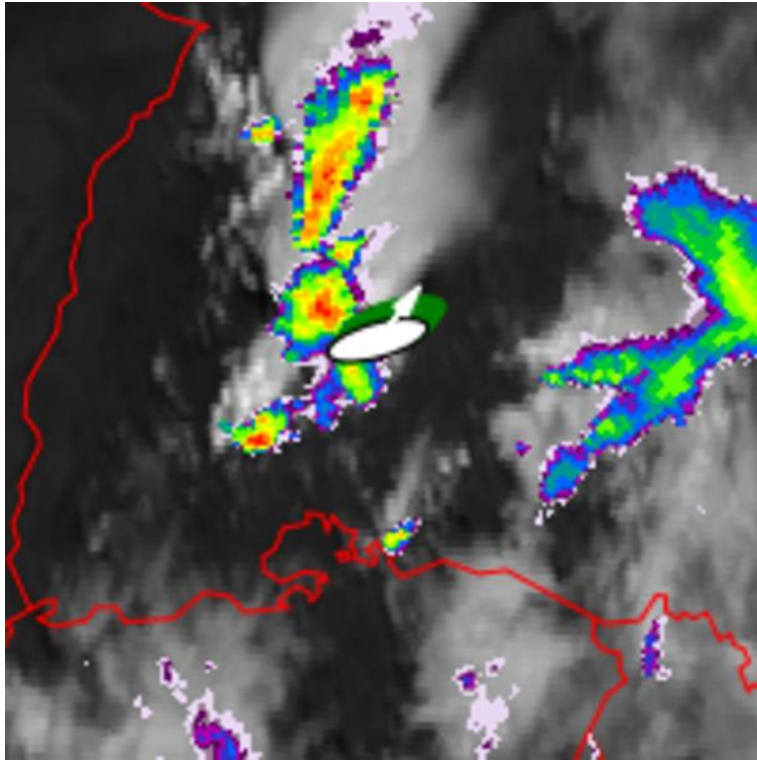


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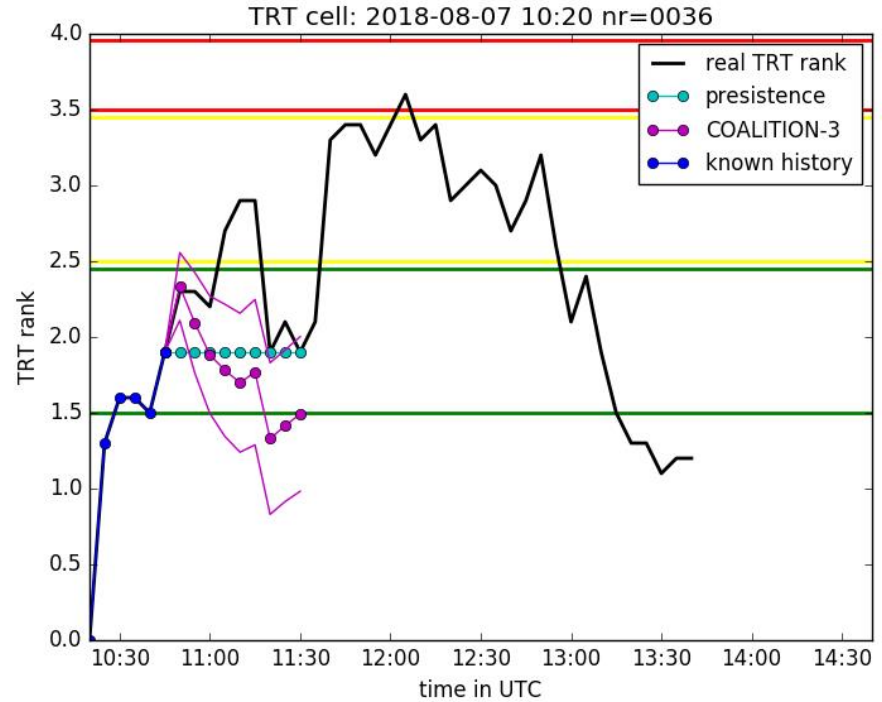
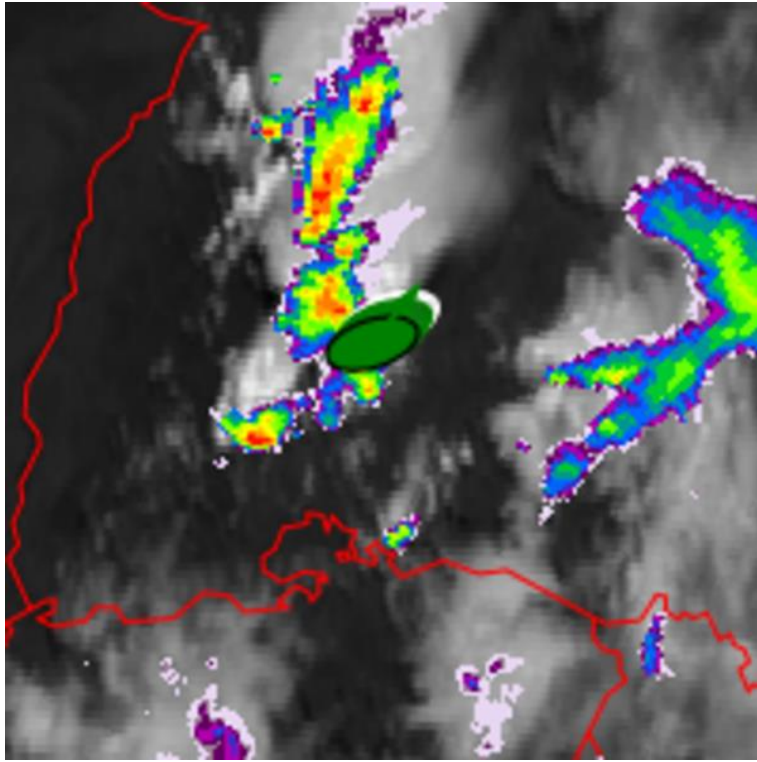


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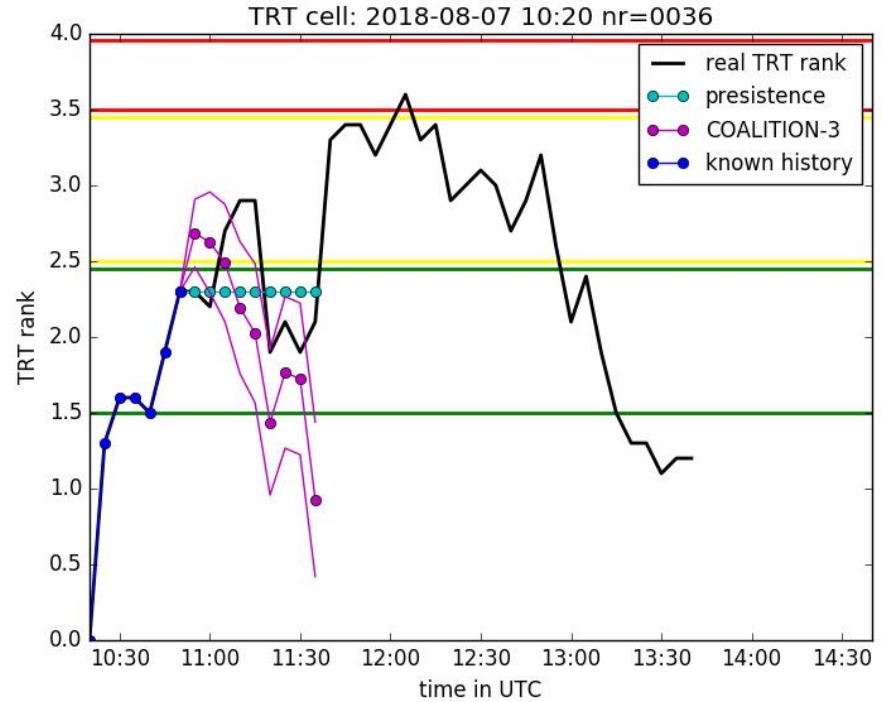
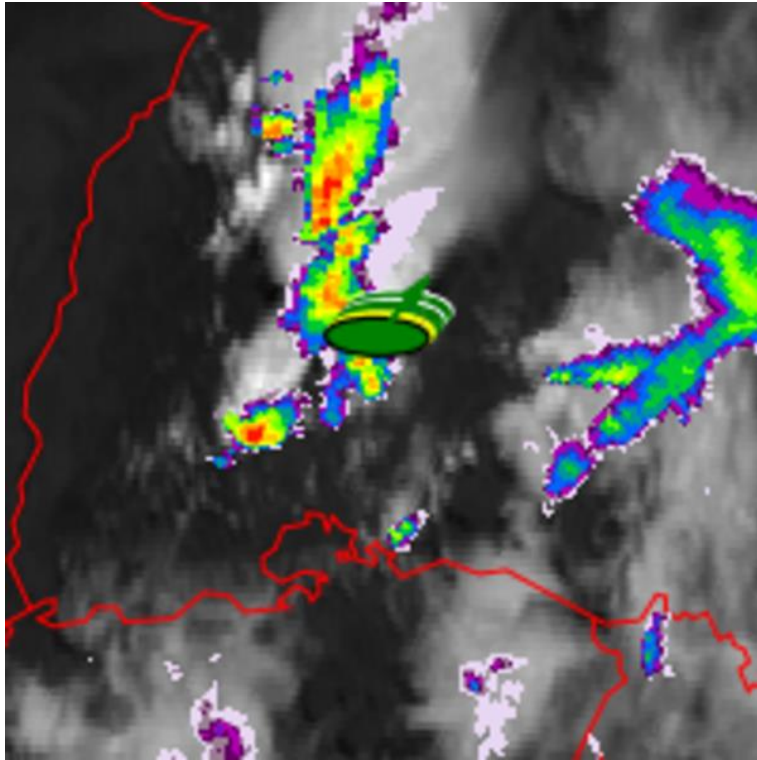


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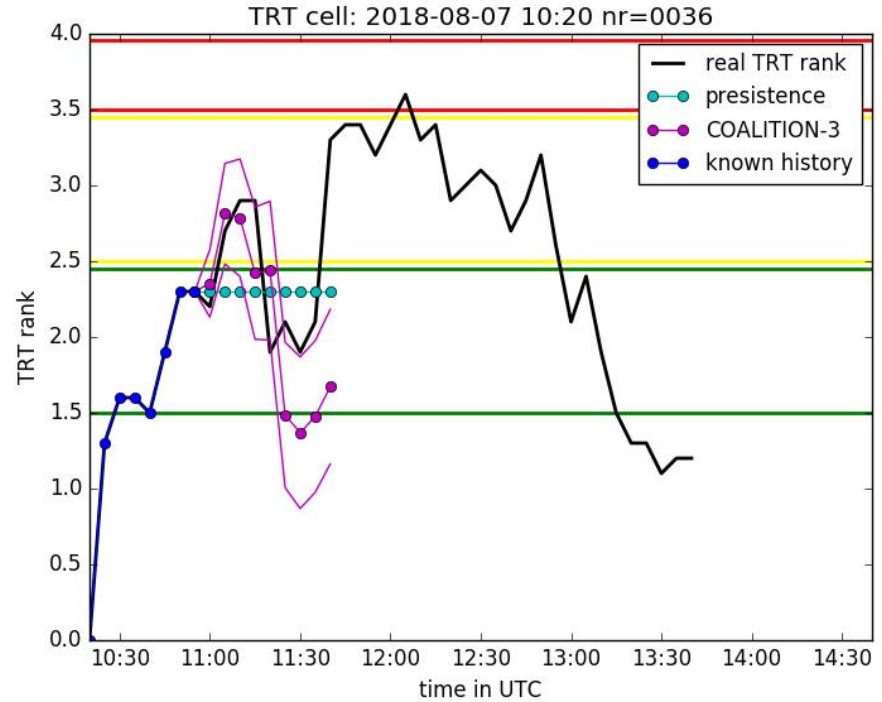
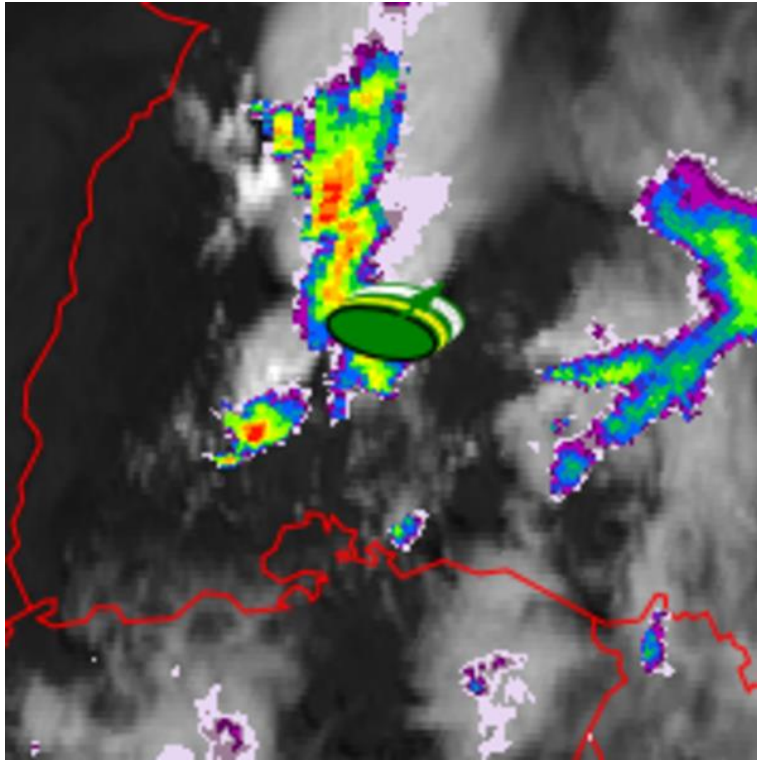


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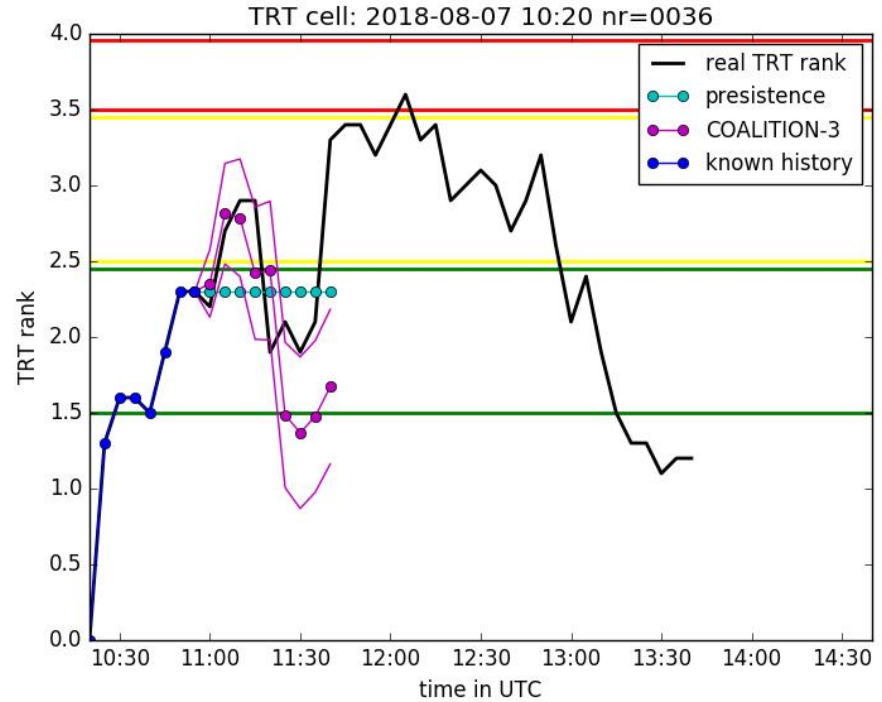
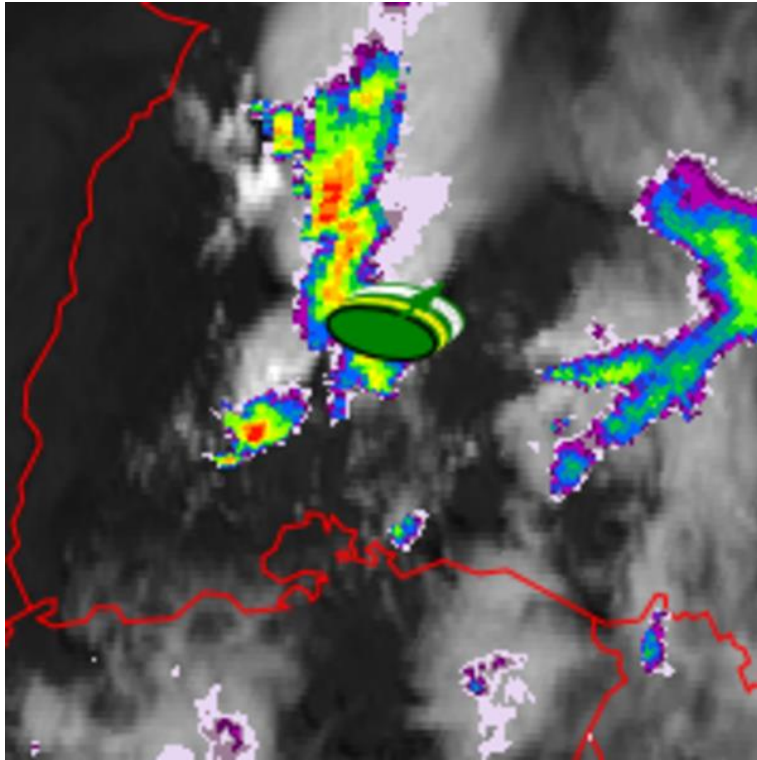


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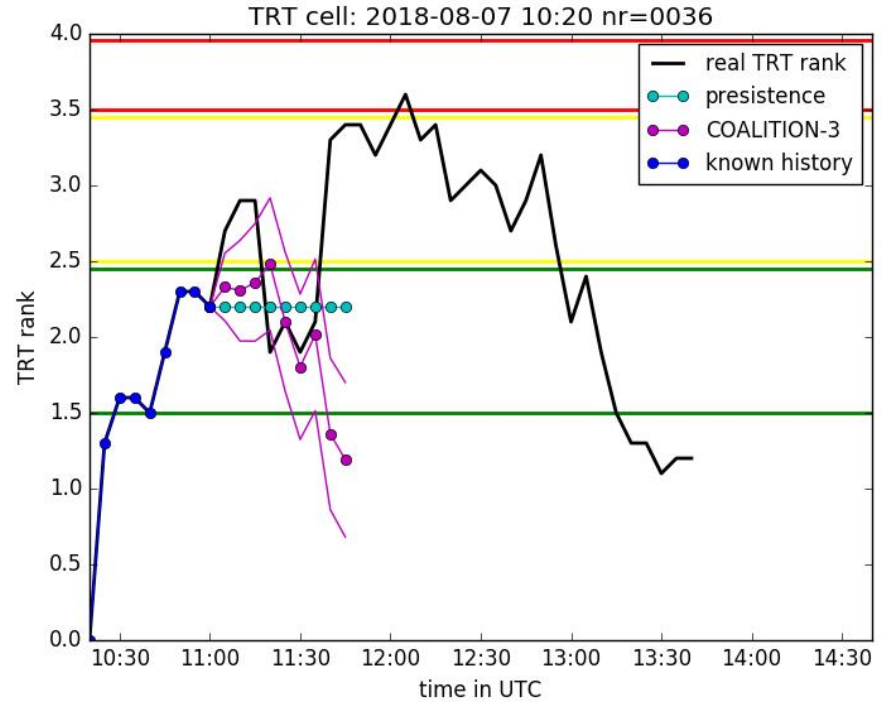
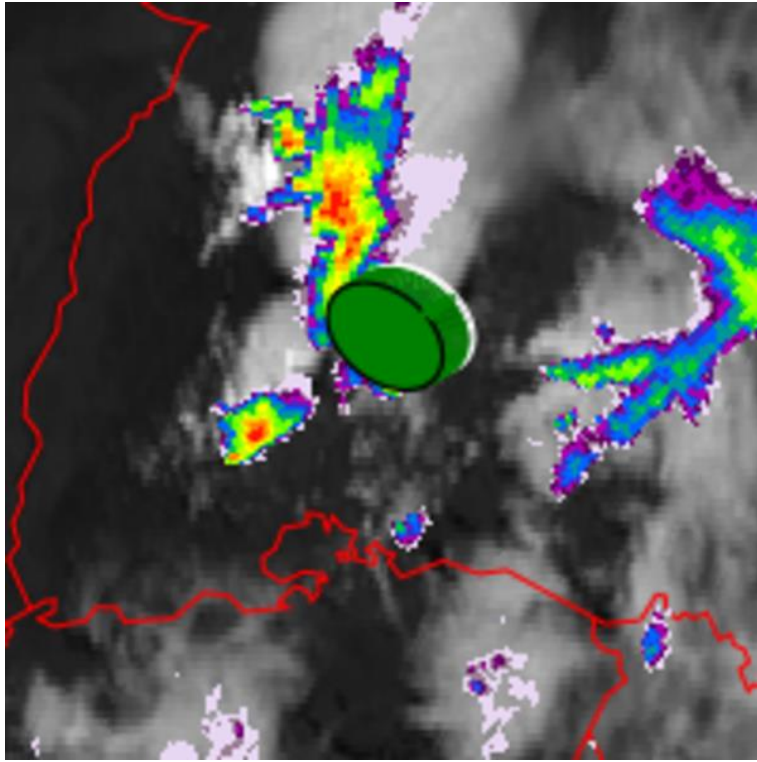


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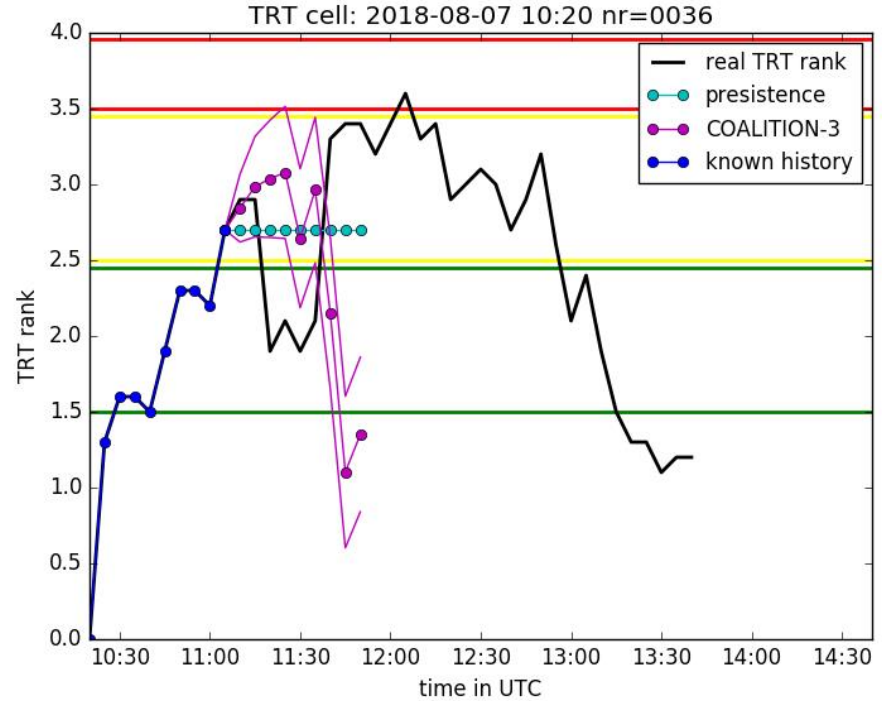
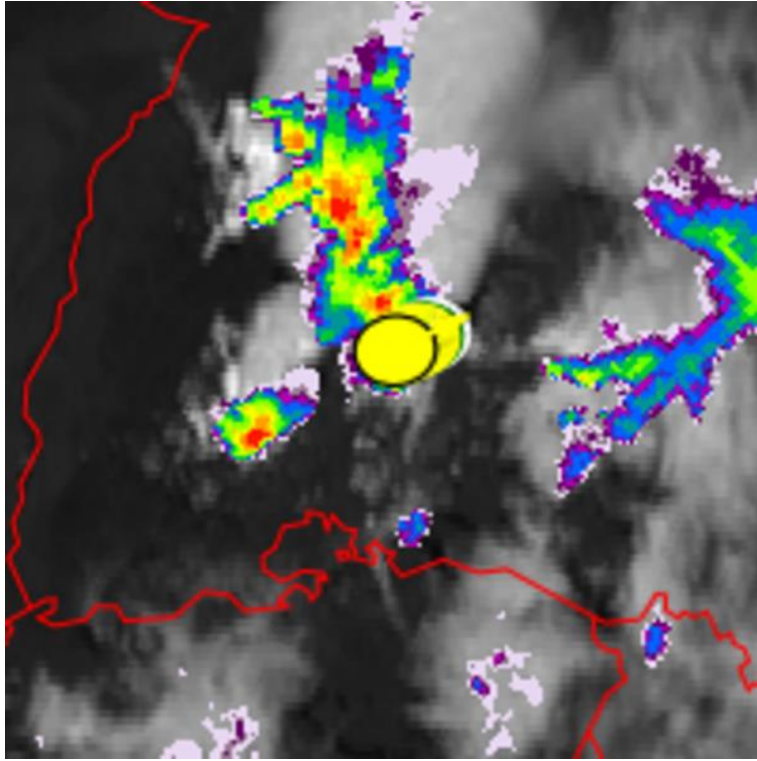


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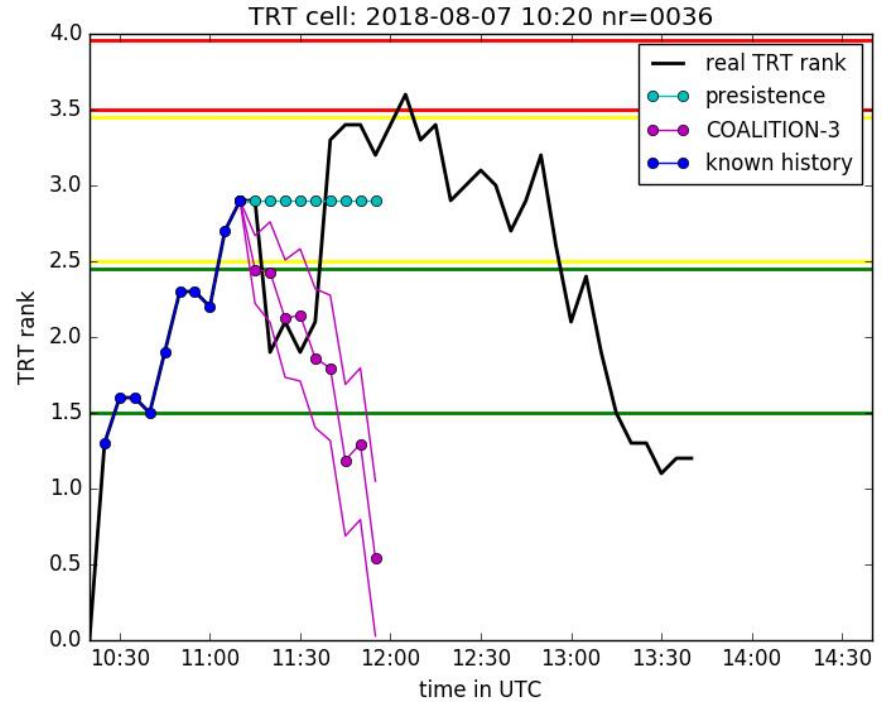
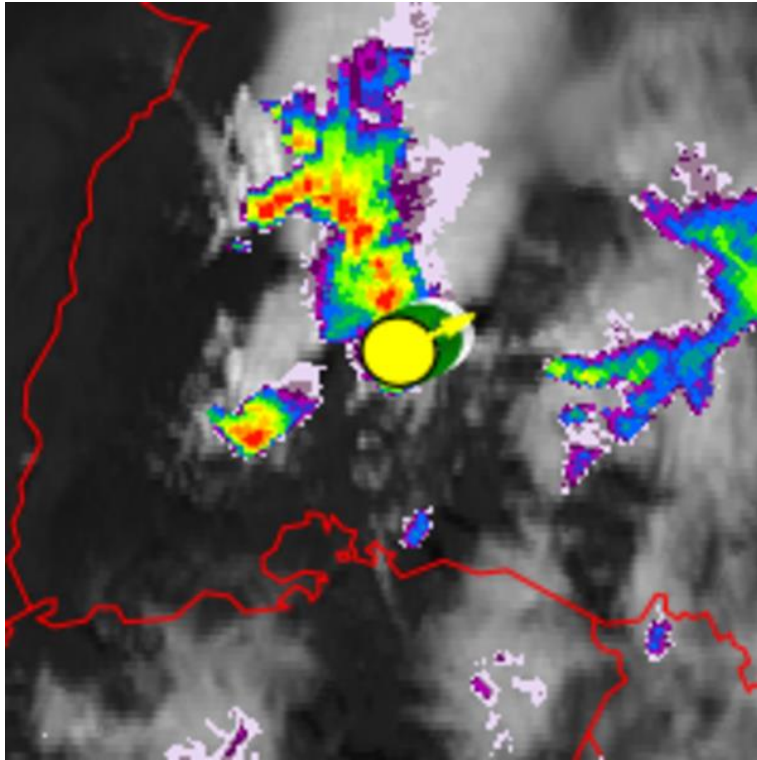


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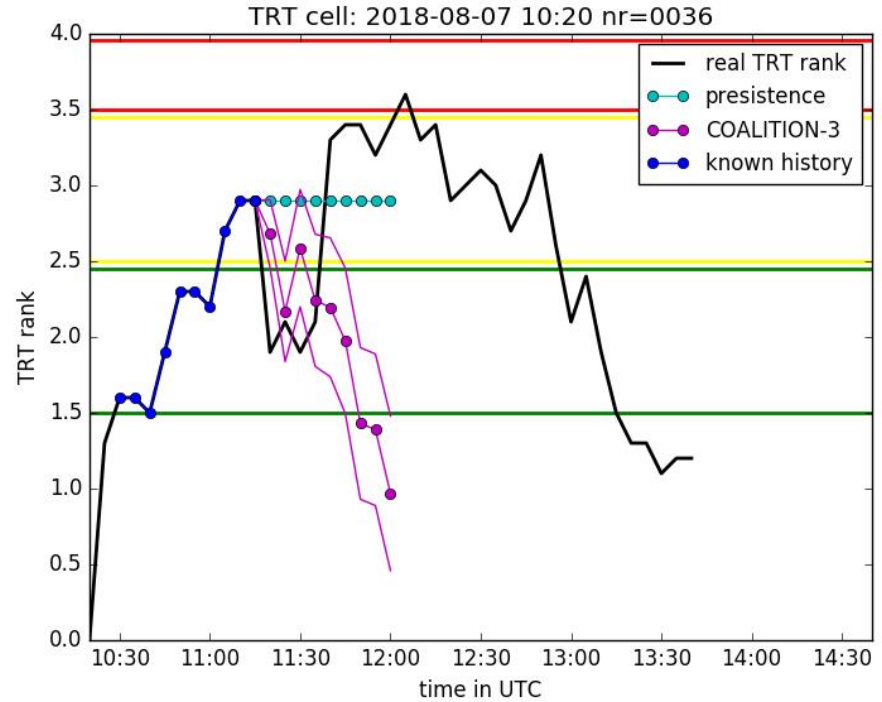
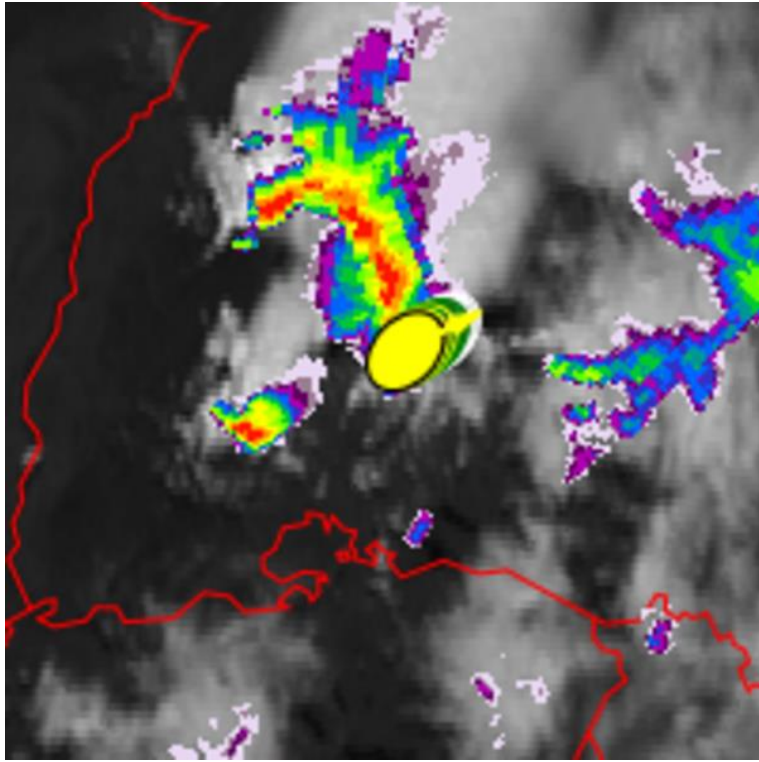


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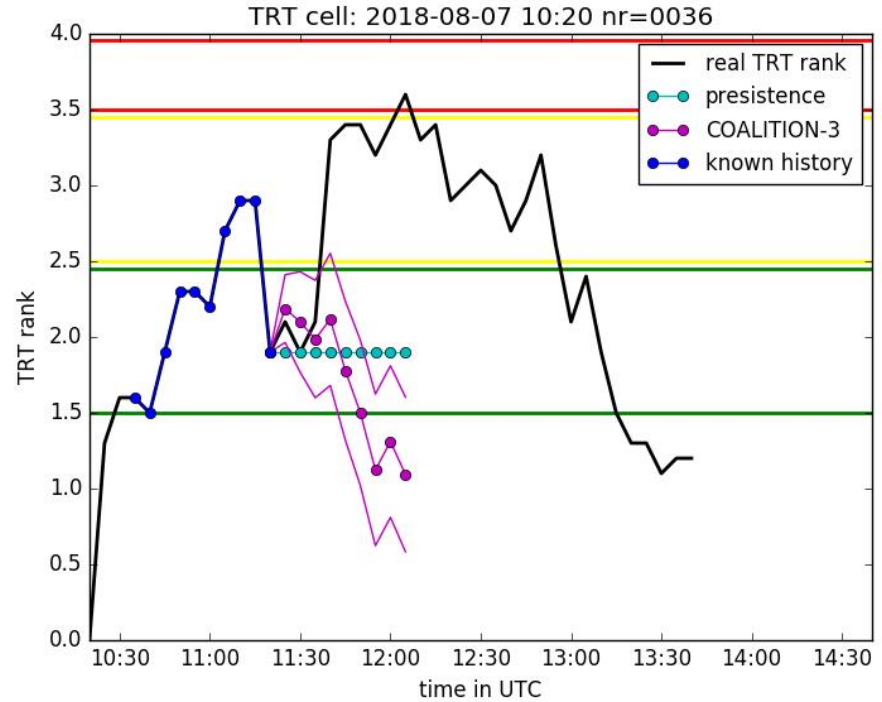
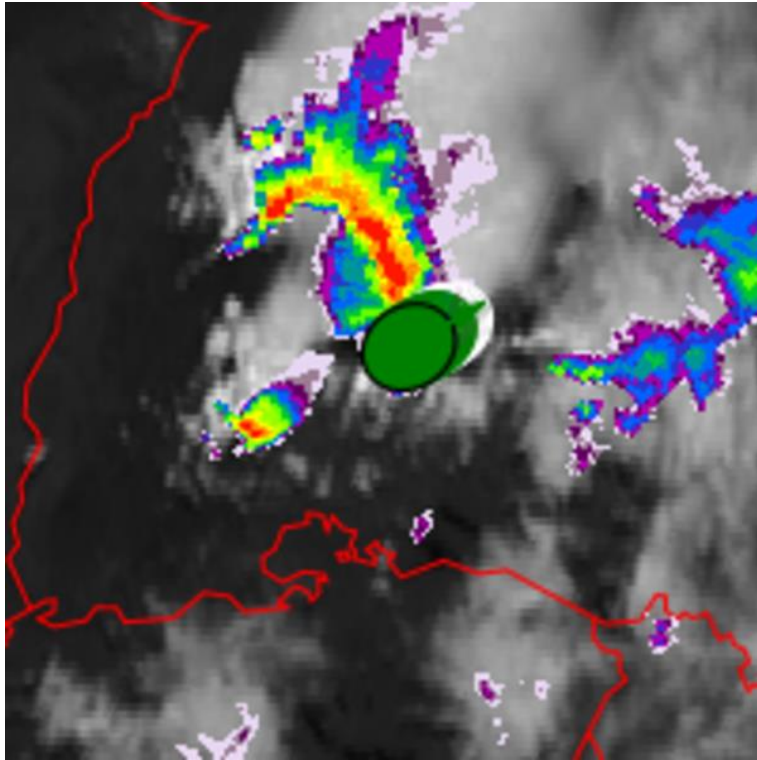


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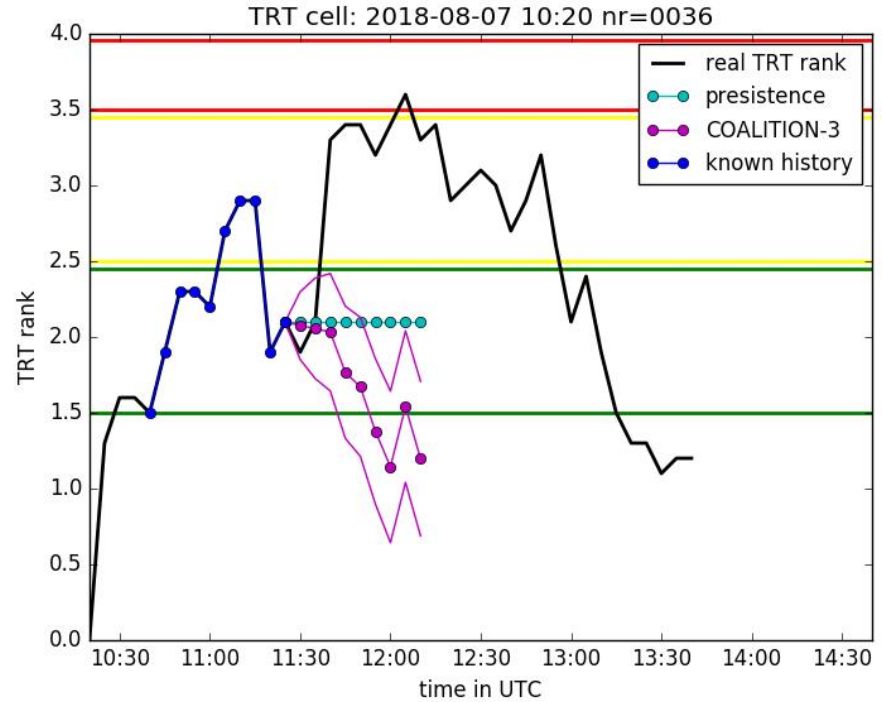
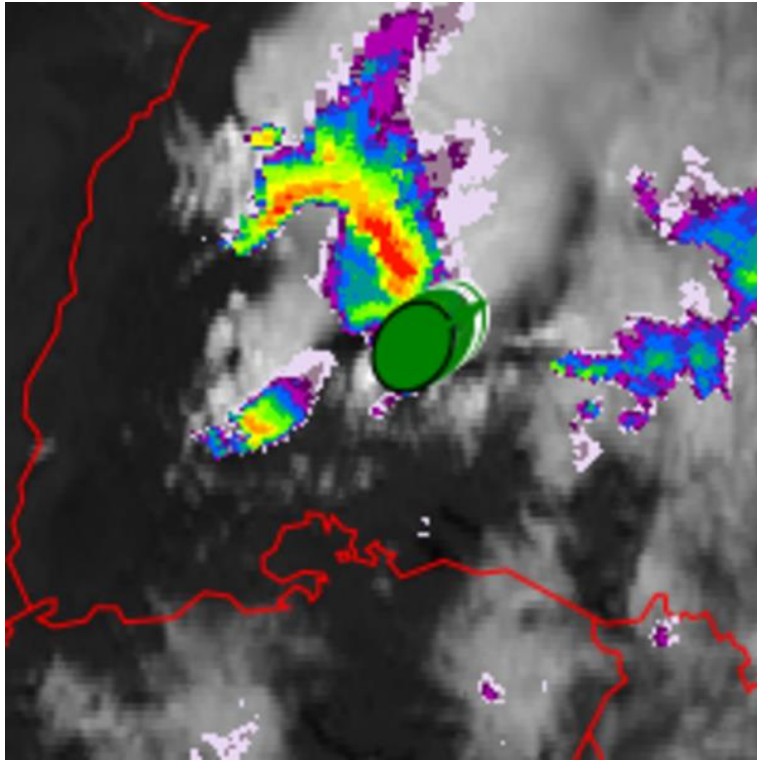


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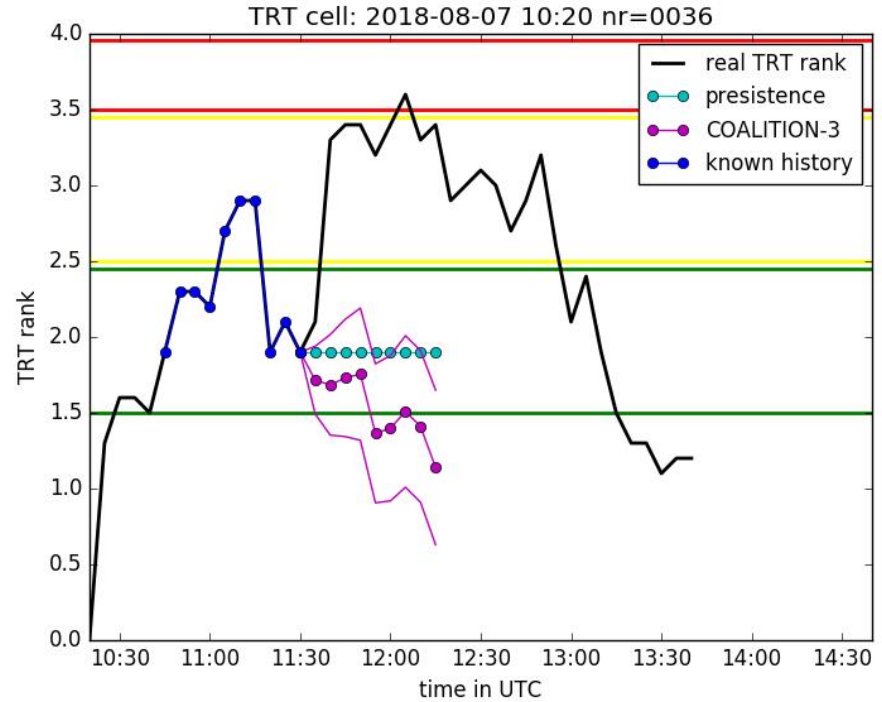
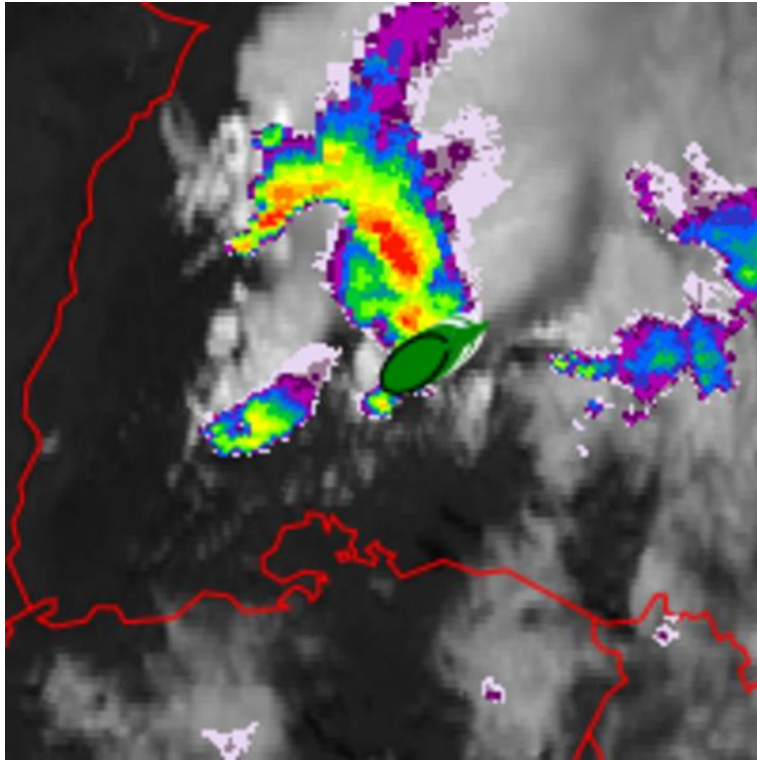


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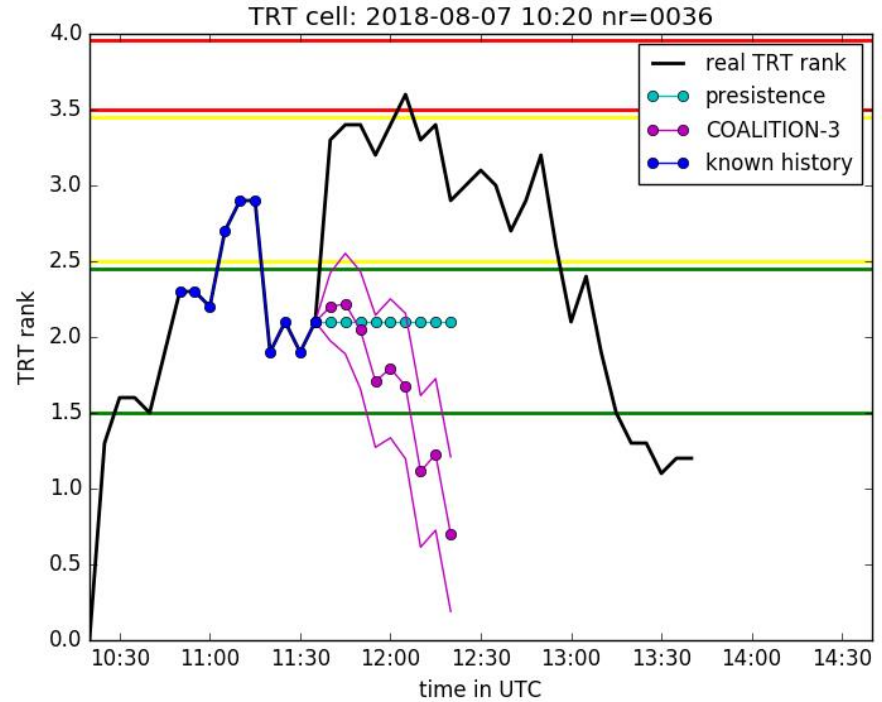
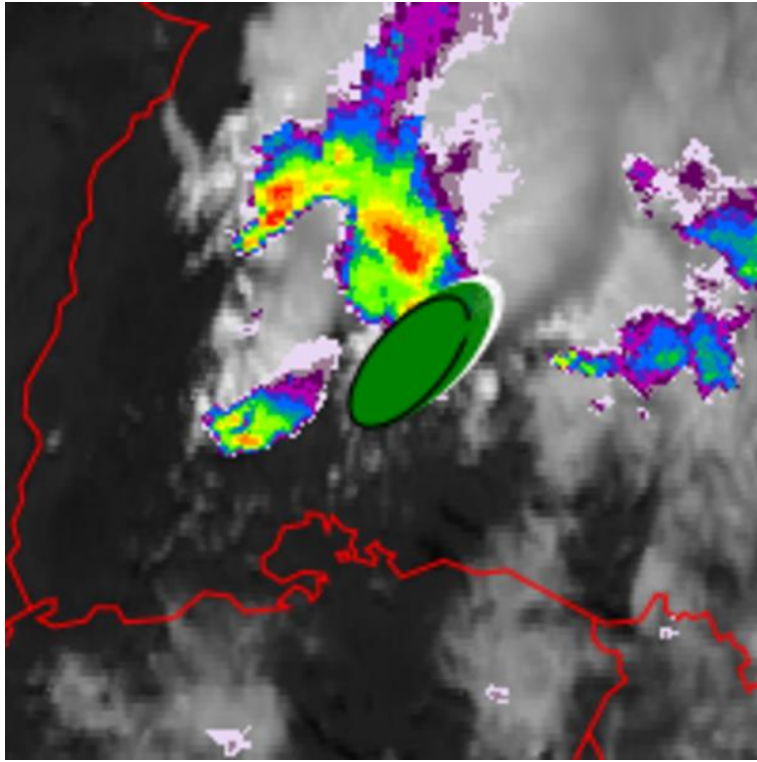


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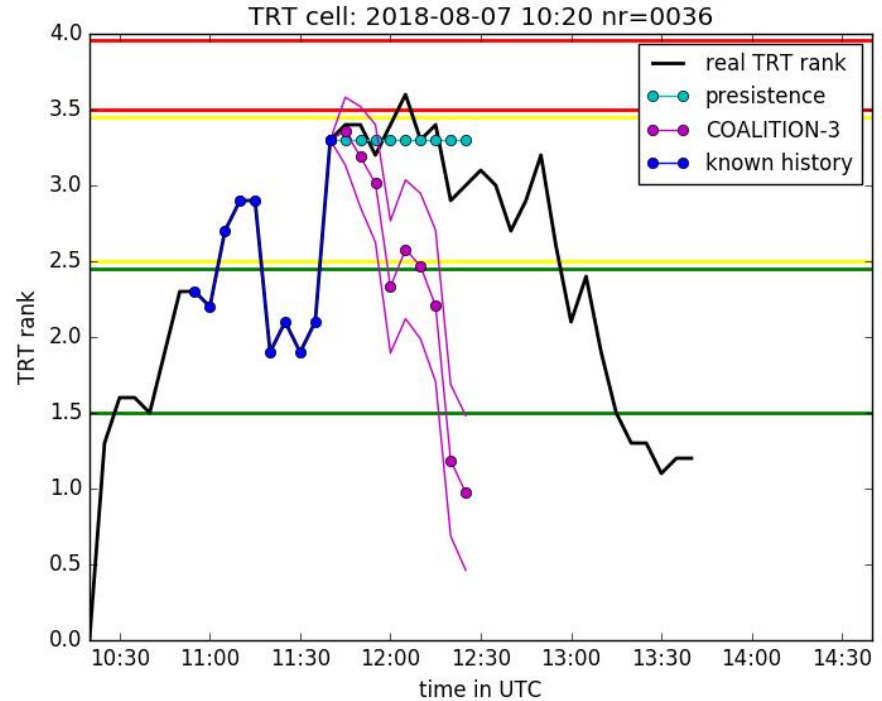
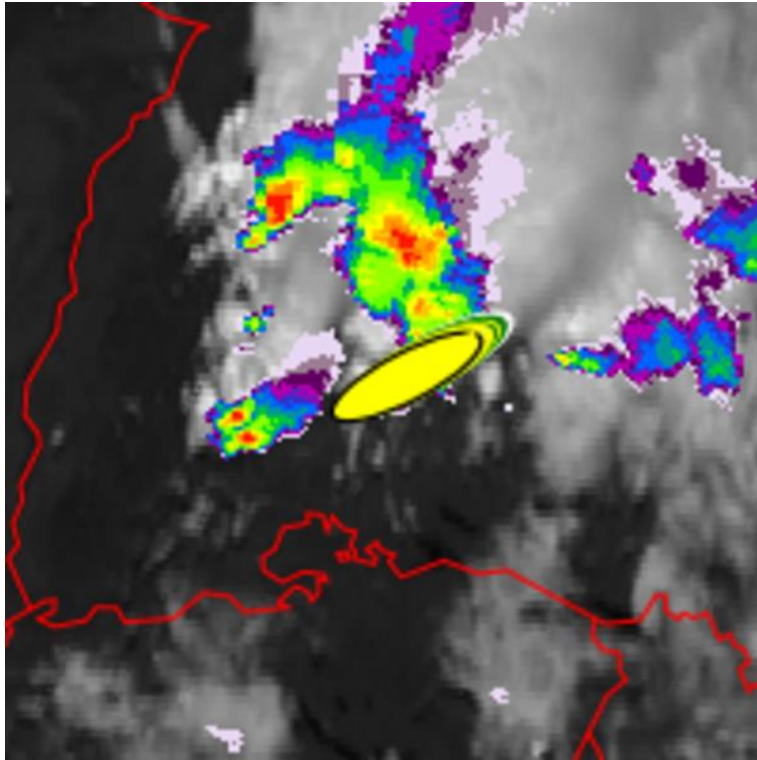


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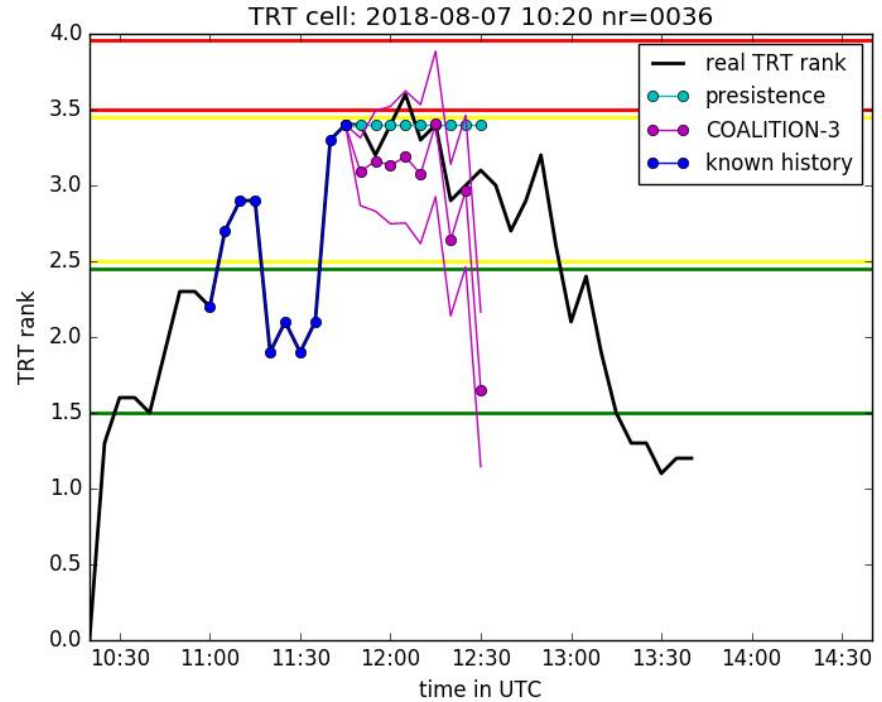
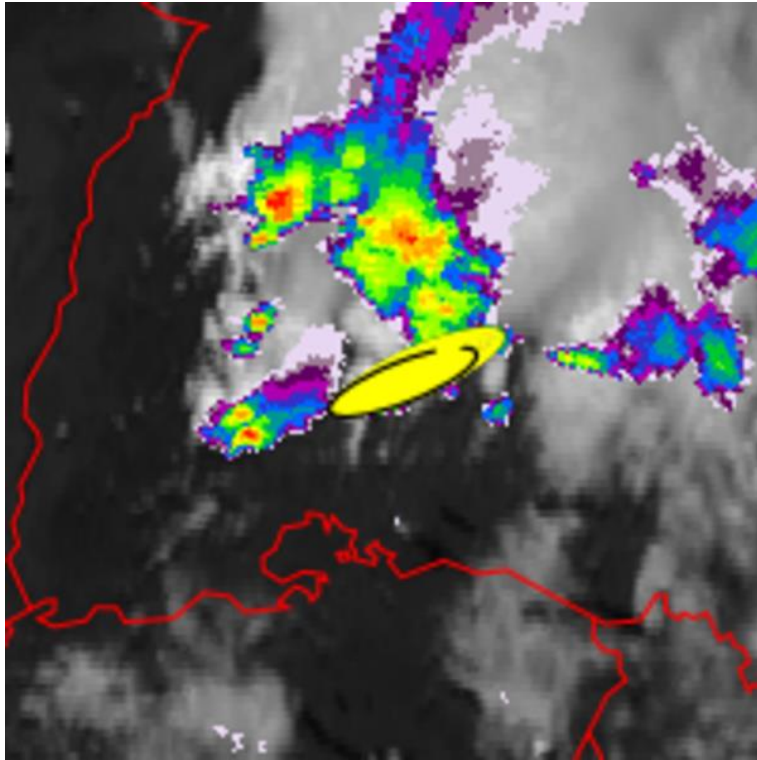


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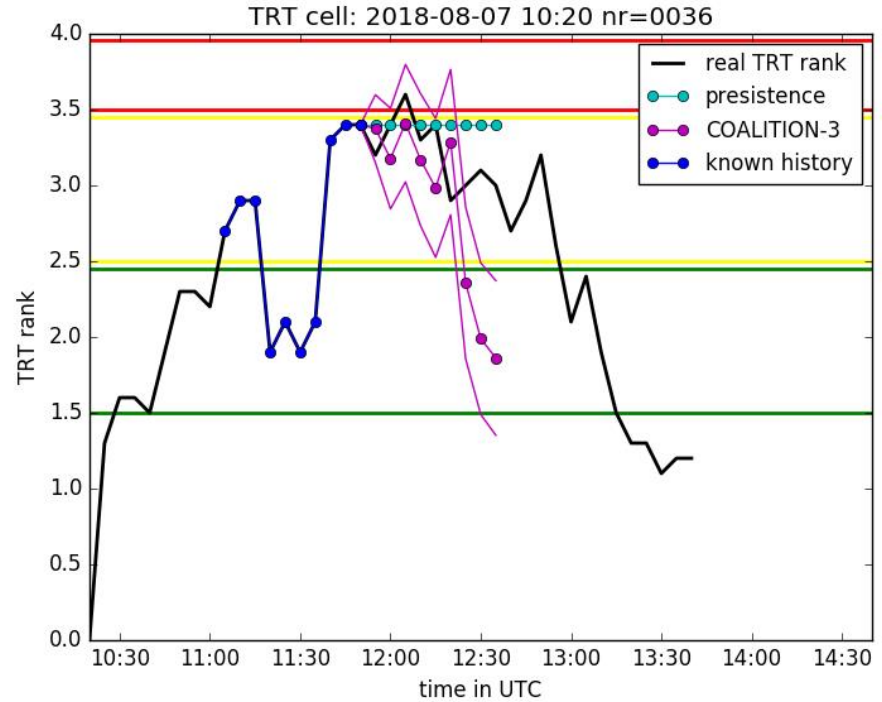
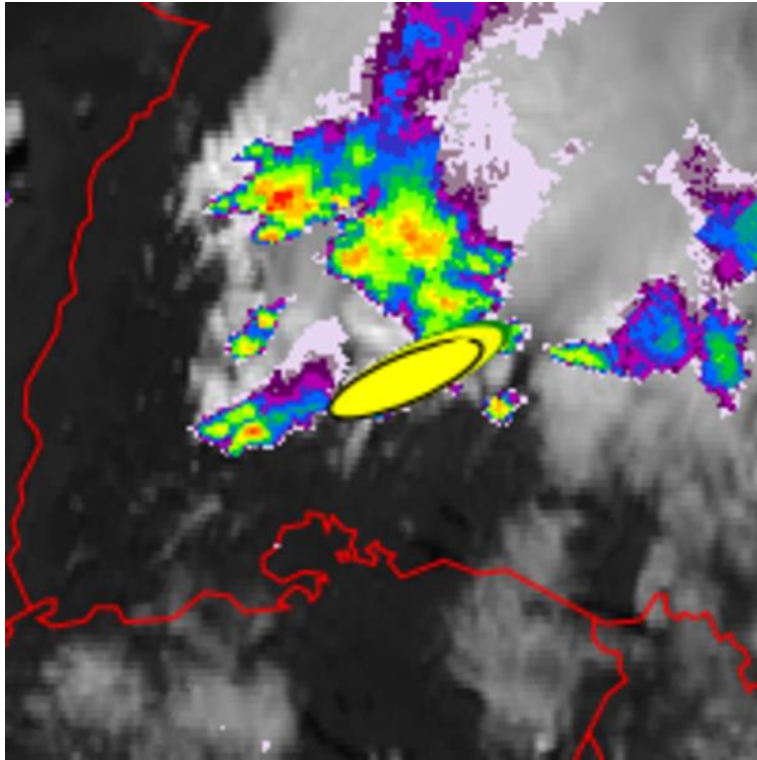


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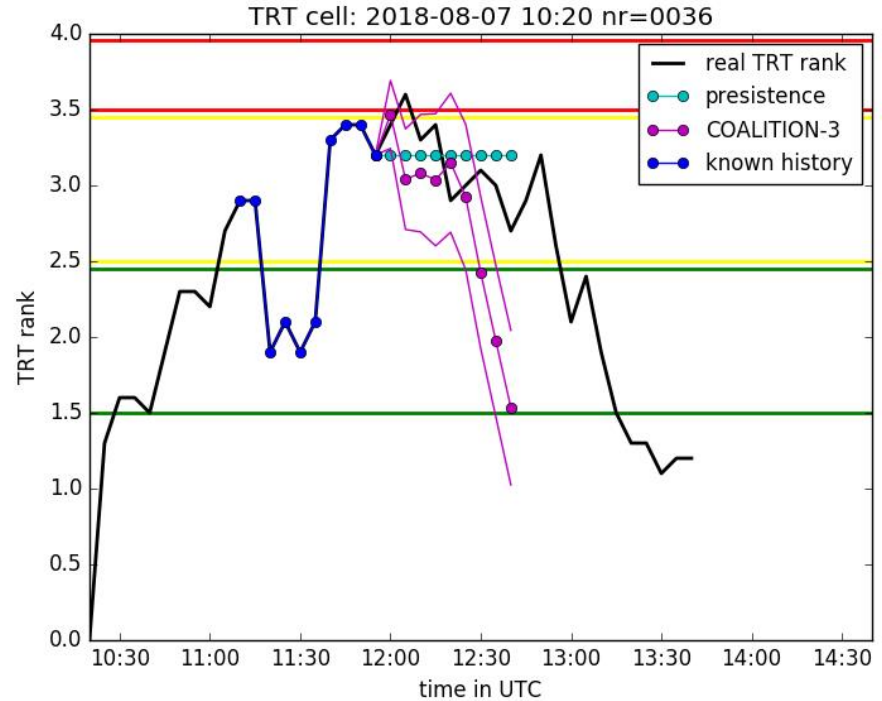
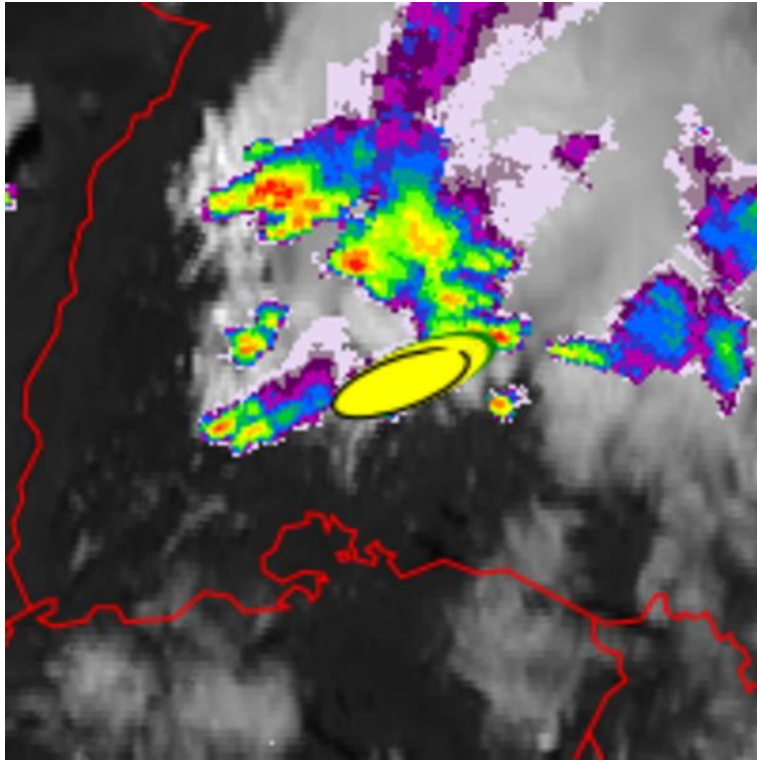


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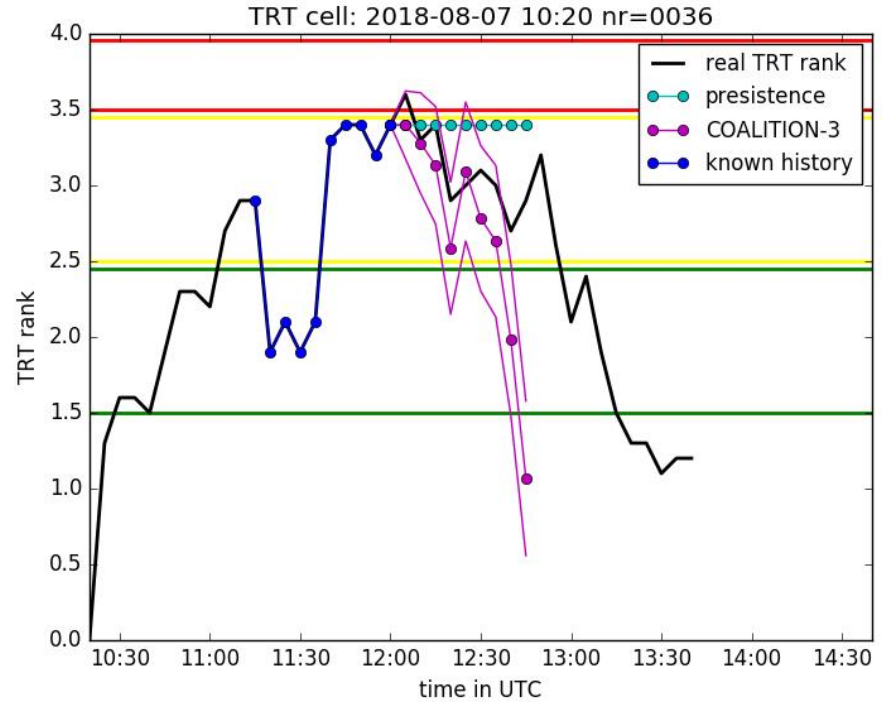
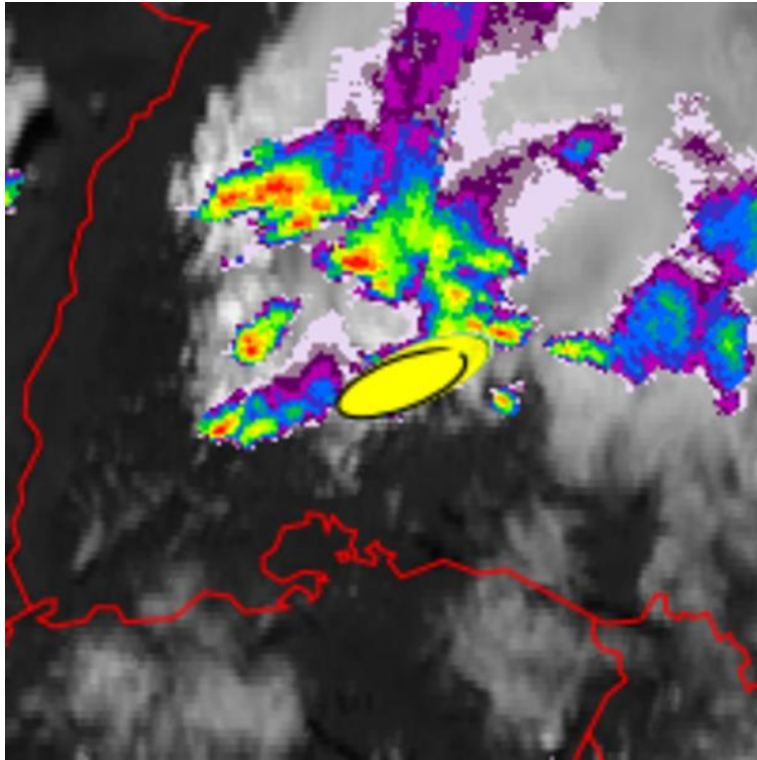


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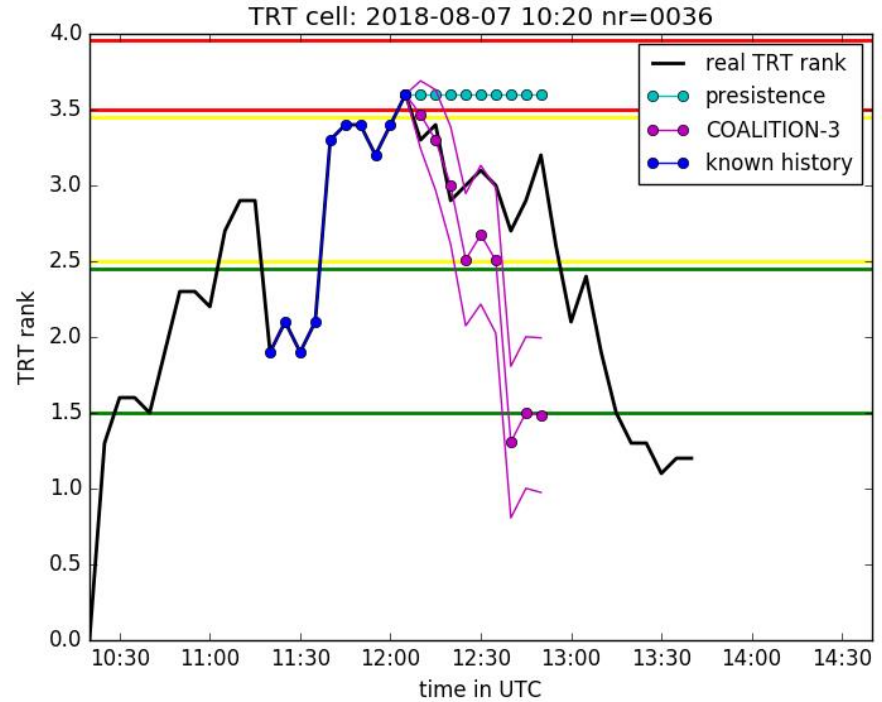
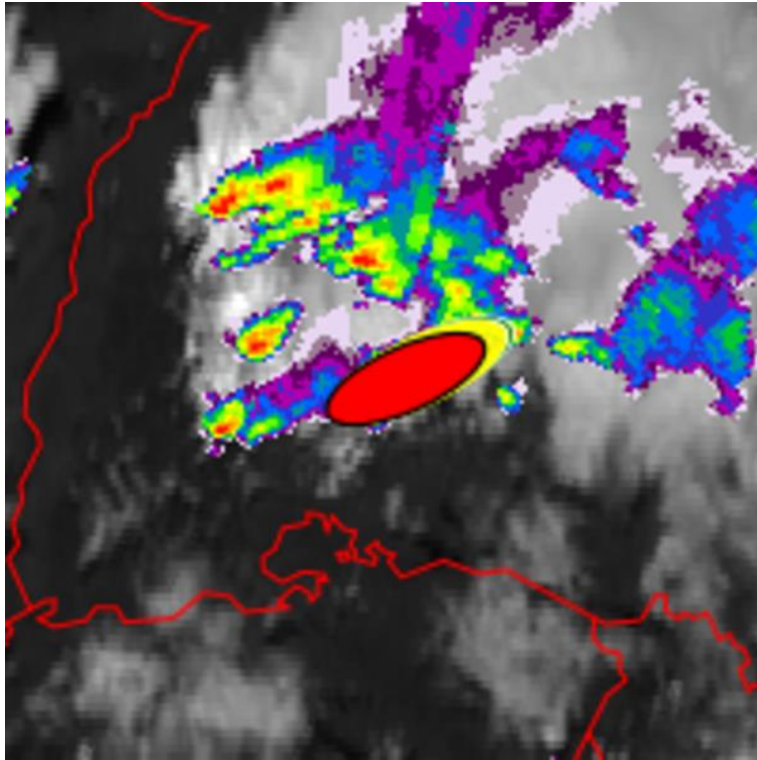


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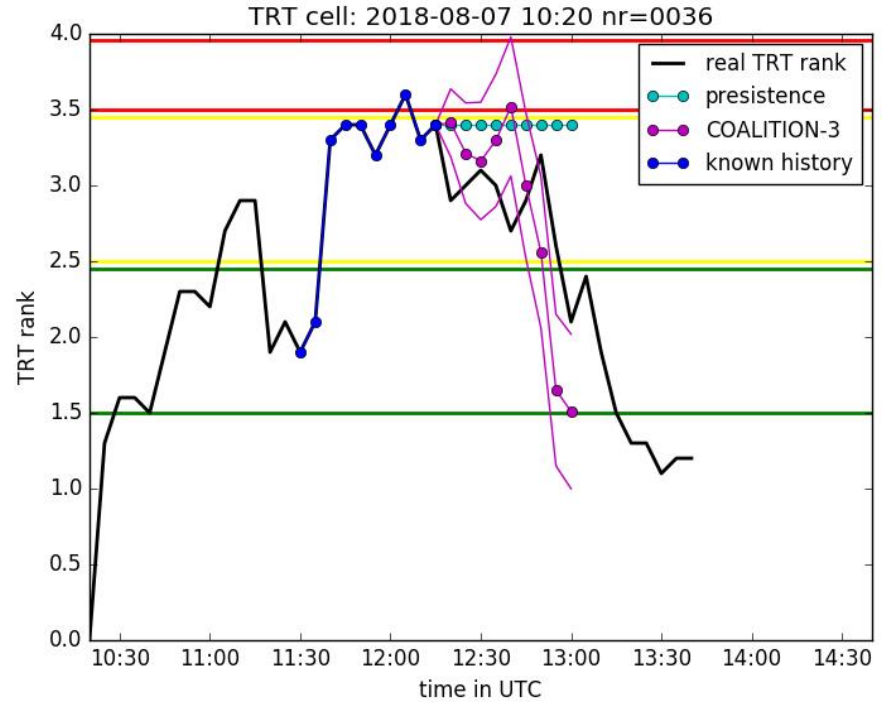
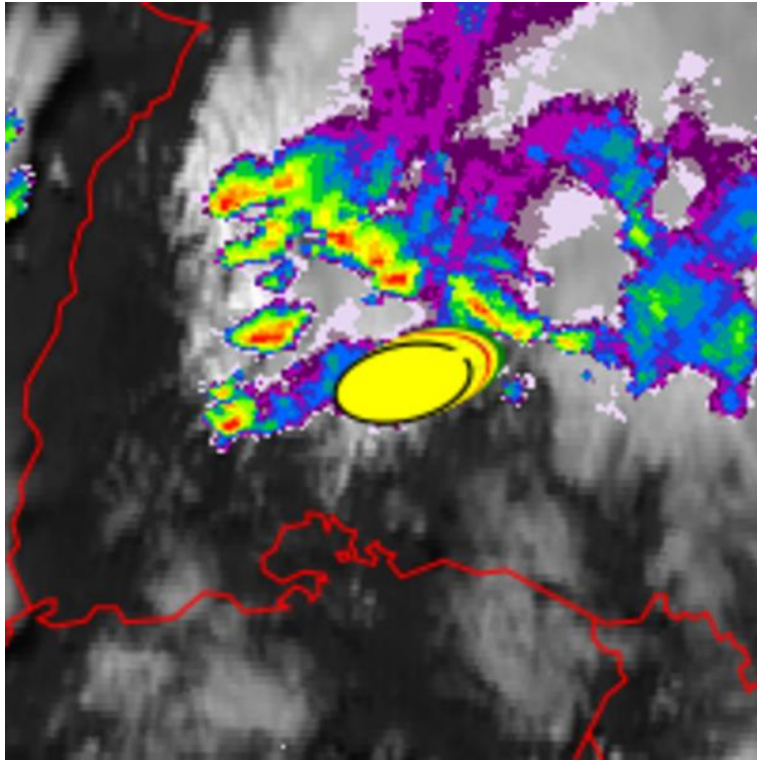


Visualization of the nowcasting results



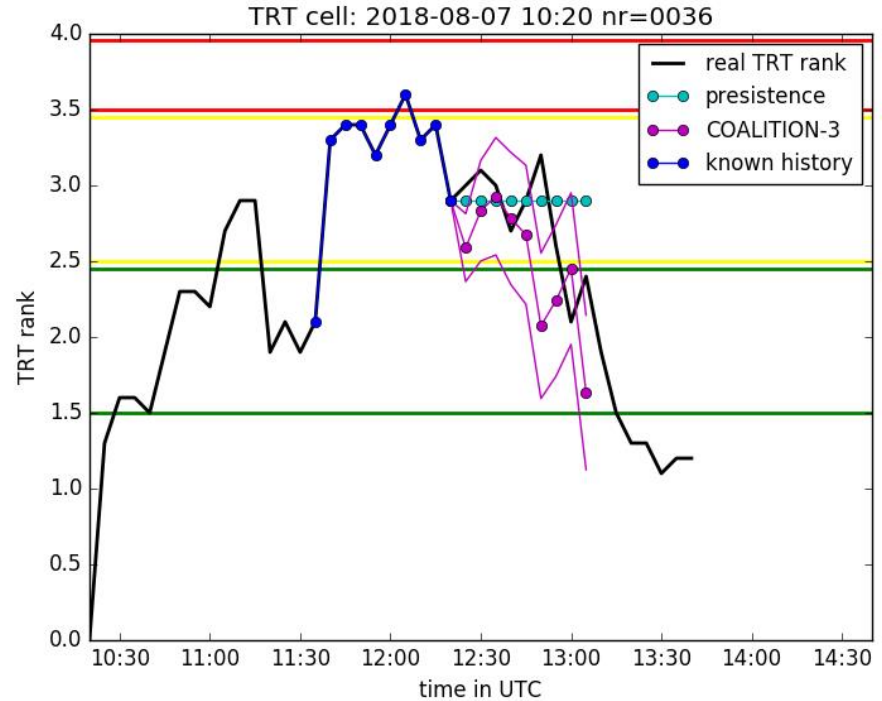
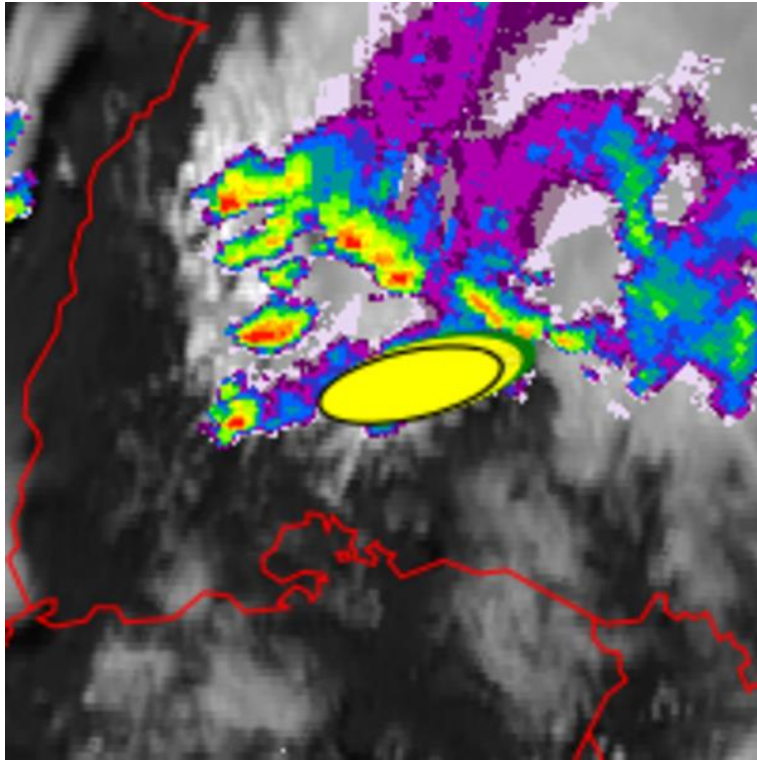


Visualization of the nowcasting results



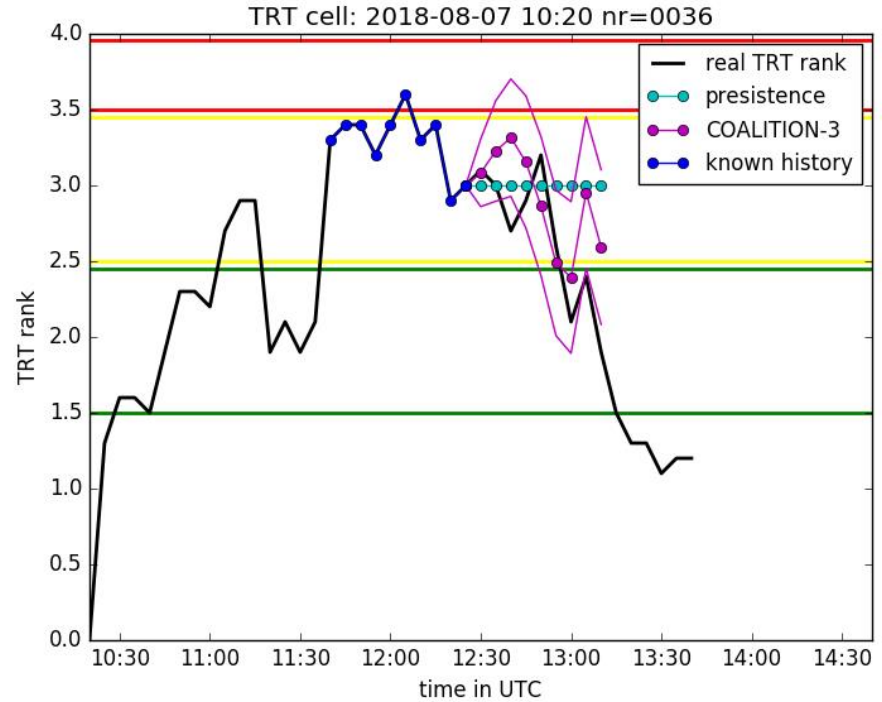
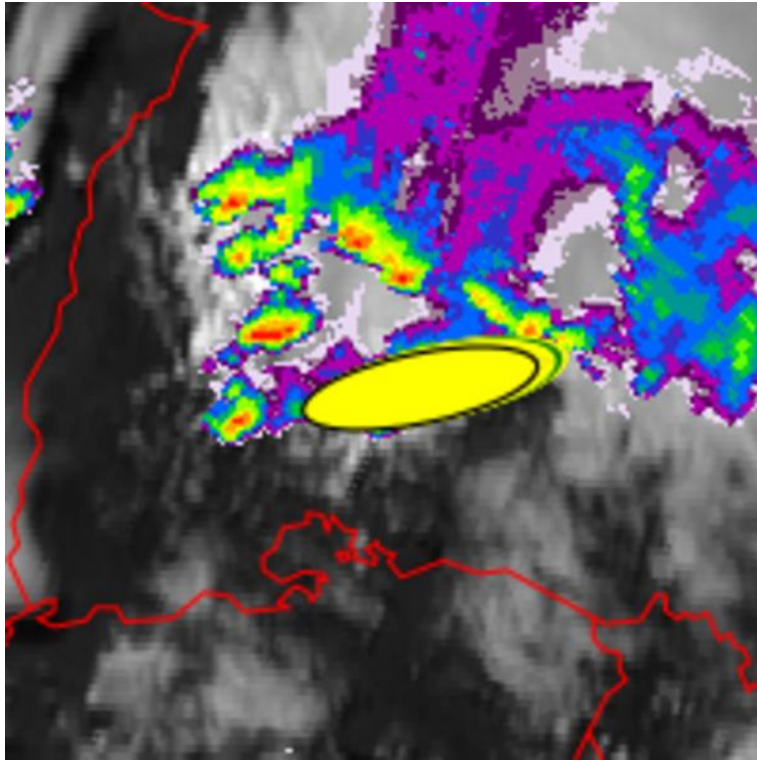


Visualization of the nowcasting results



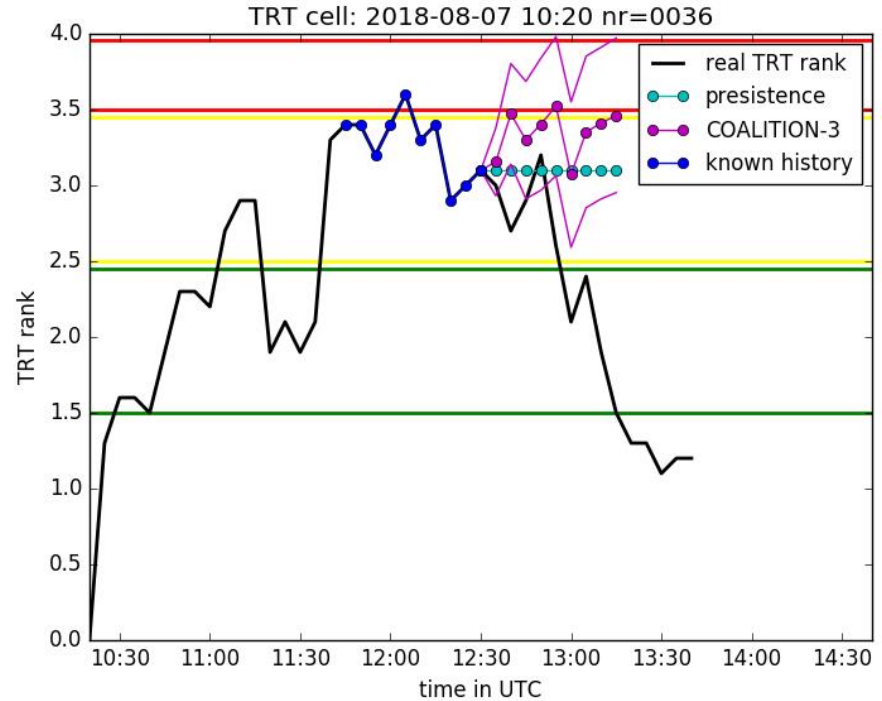
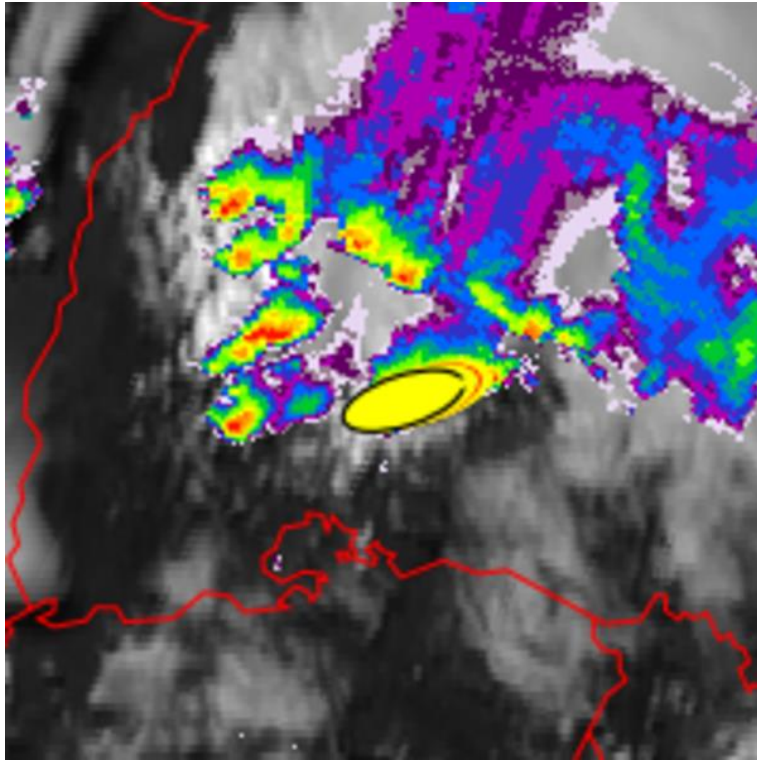


Visualization of the nowcasting results



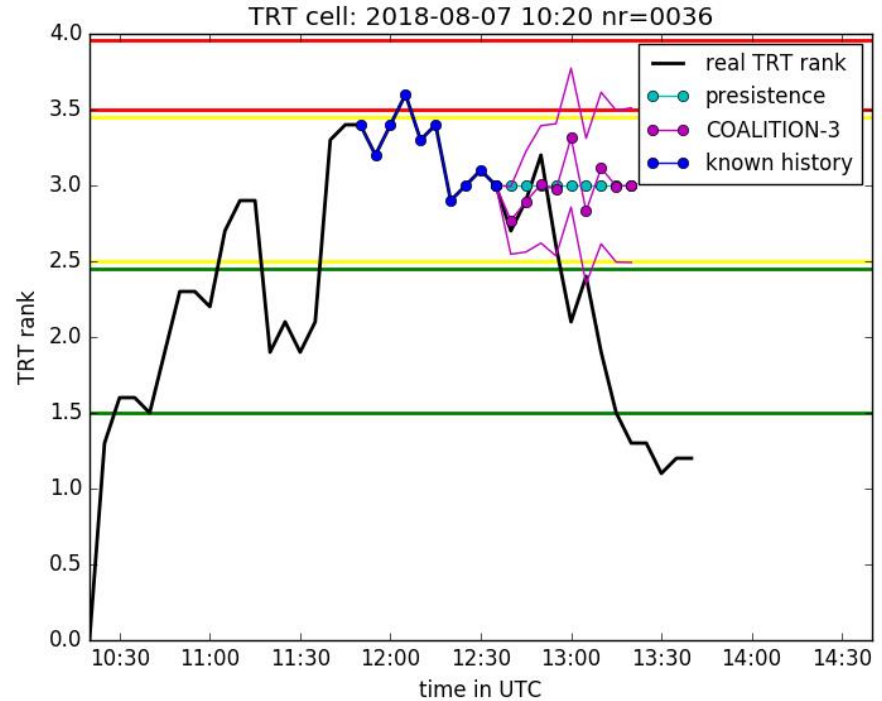
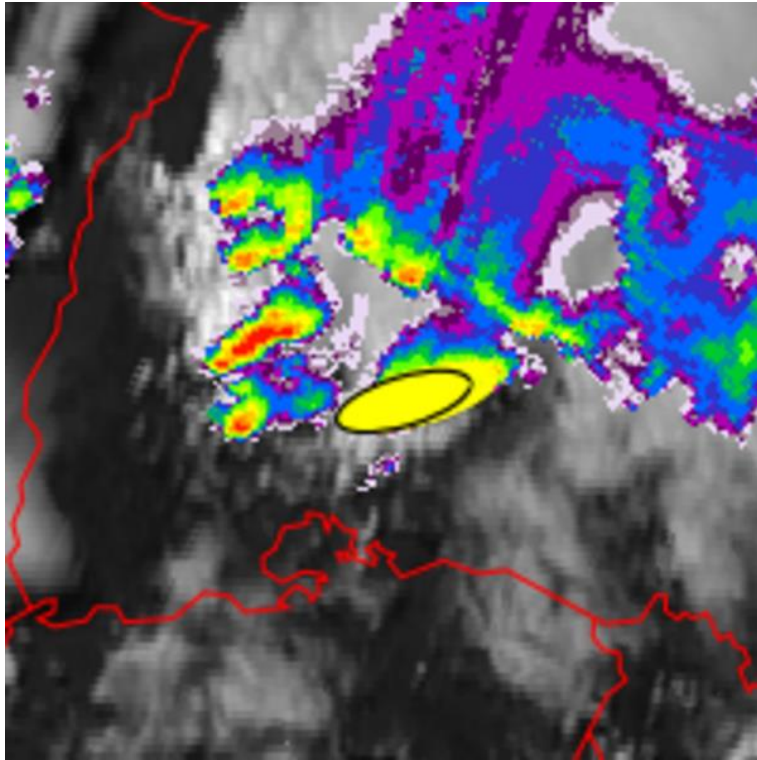


Visualization of the nowcasting results



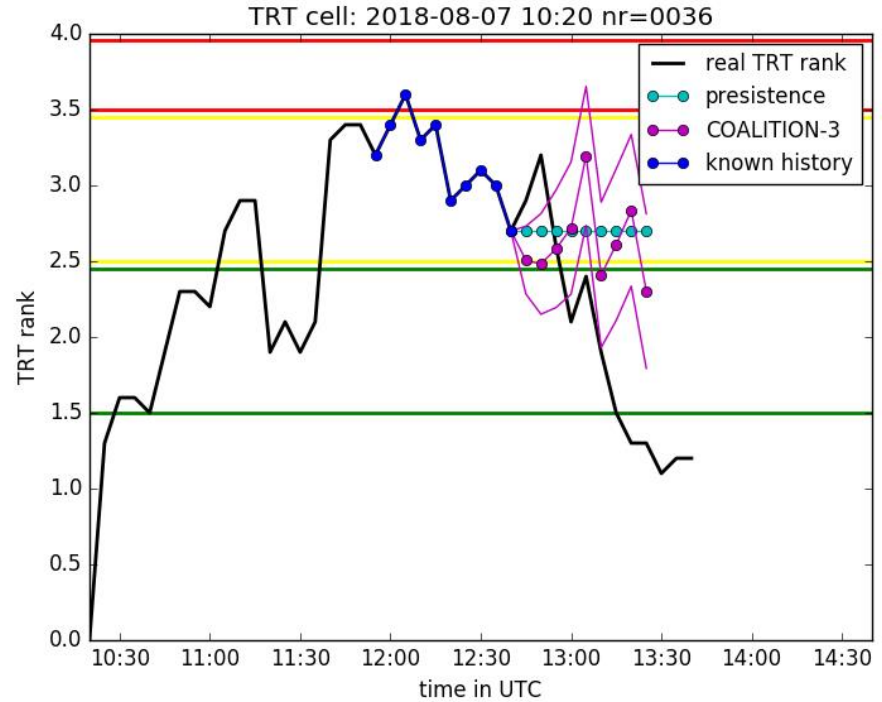
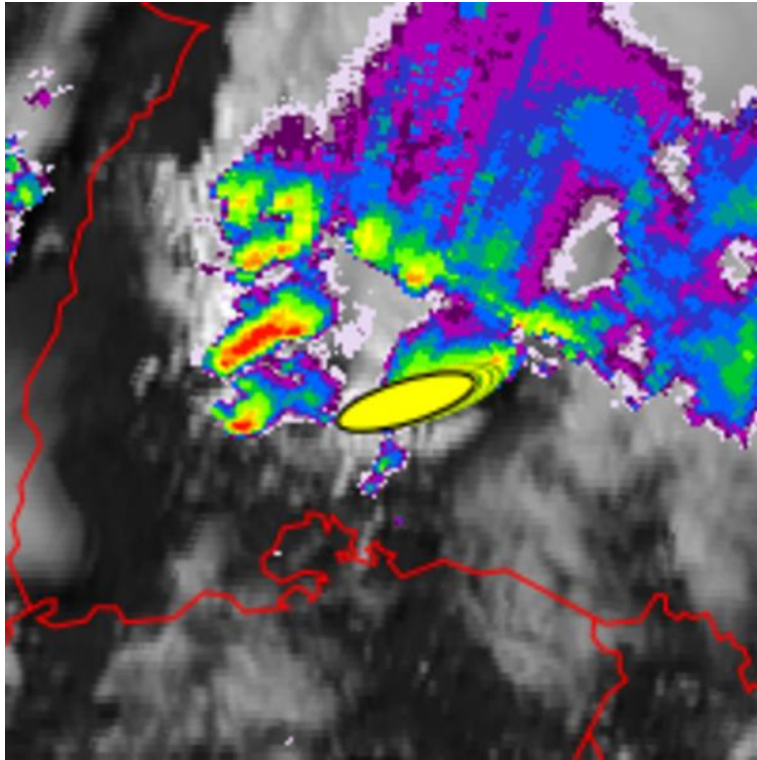


Visualization of the nowcasting results



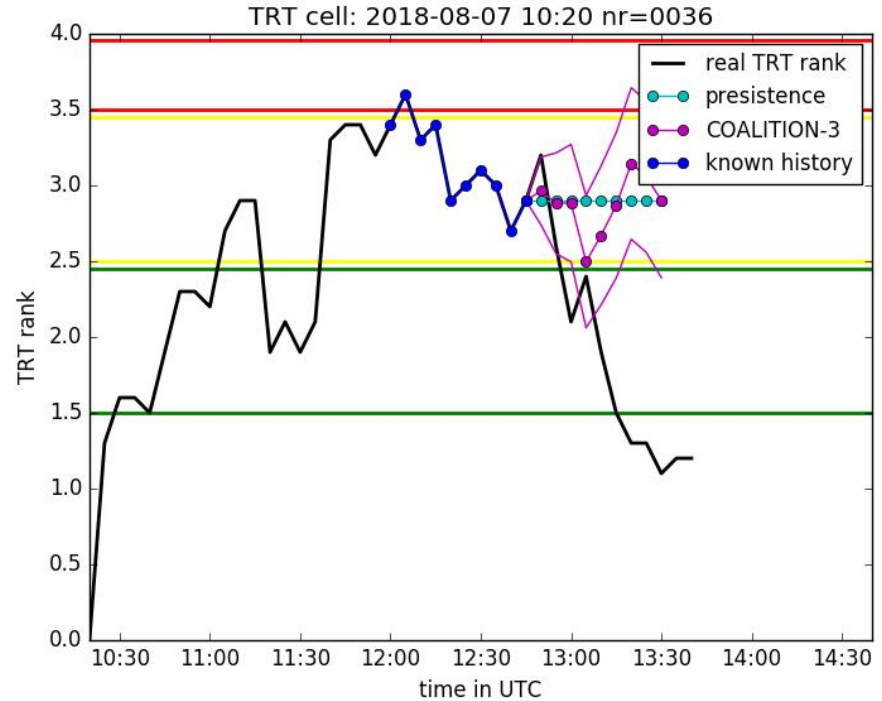
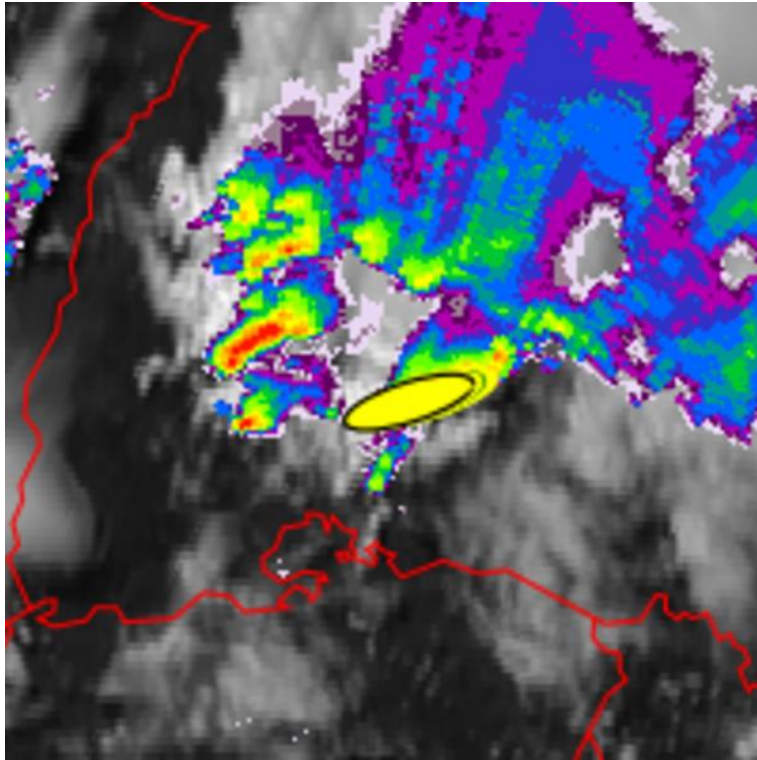


Visualization of the nowcasting results



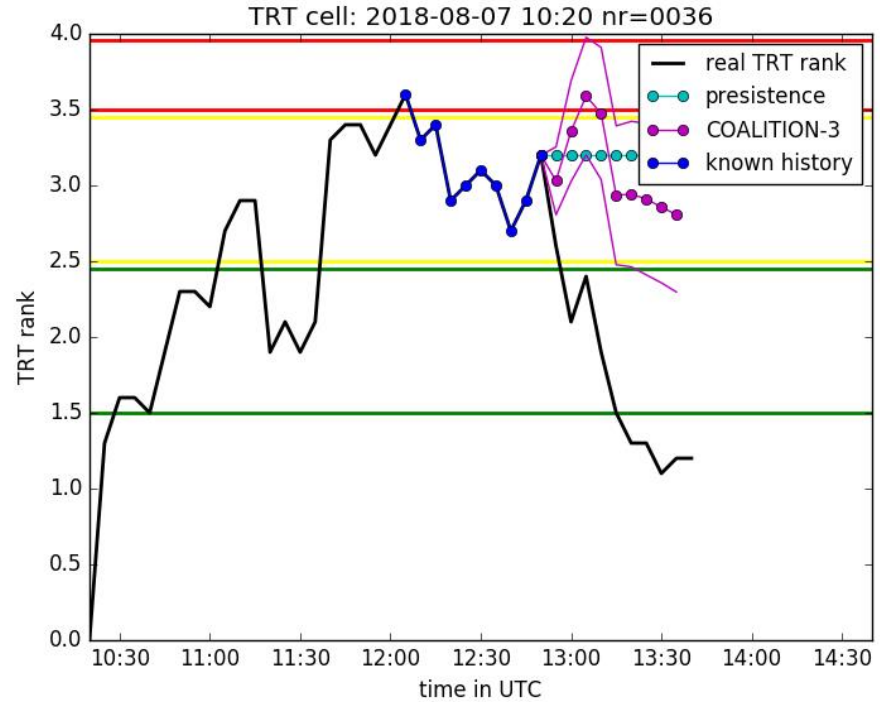
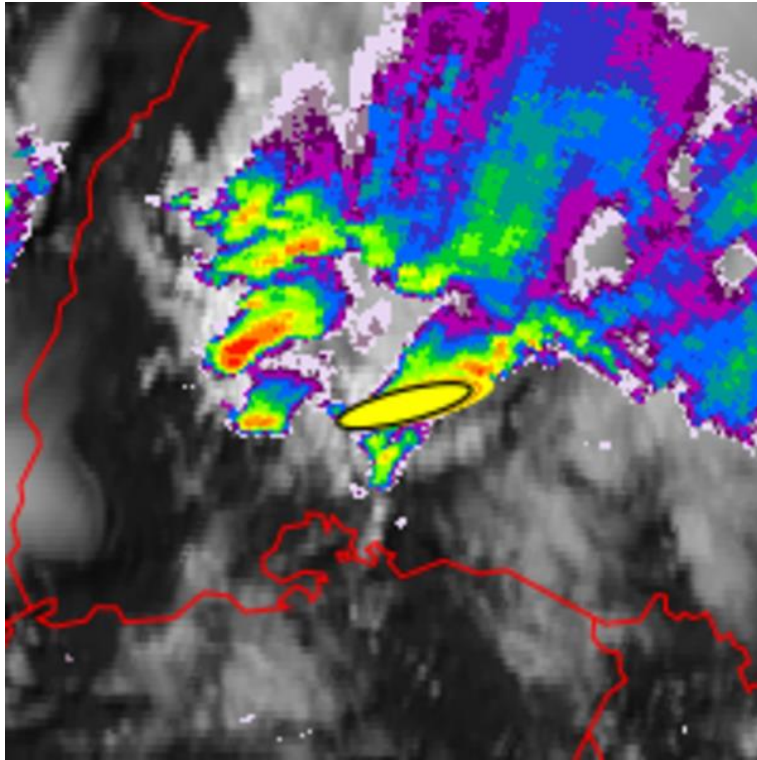


Visualization of the nowcasting results



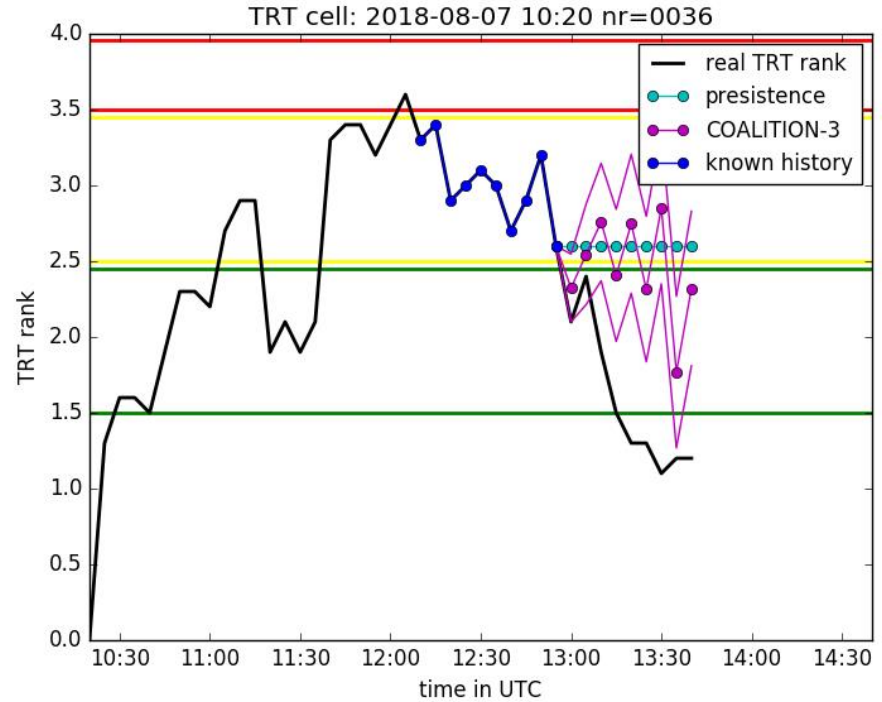
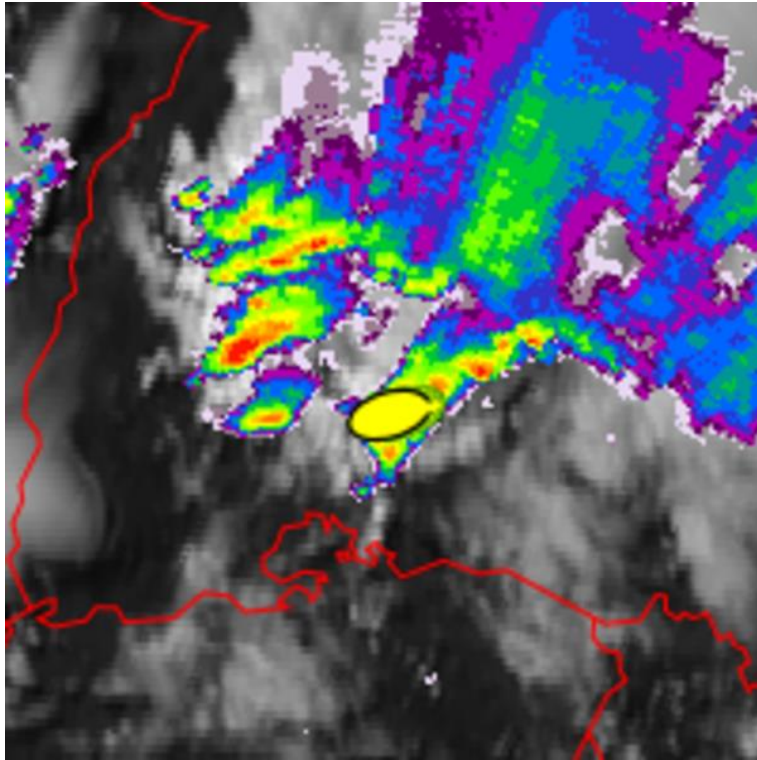


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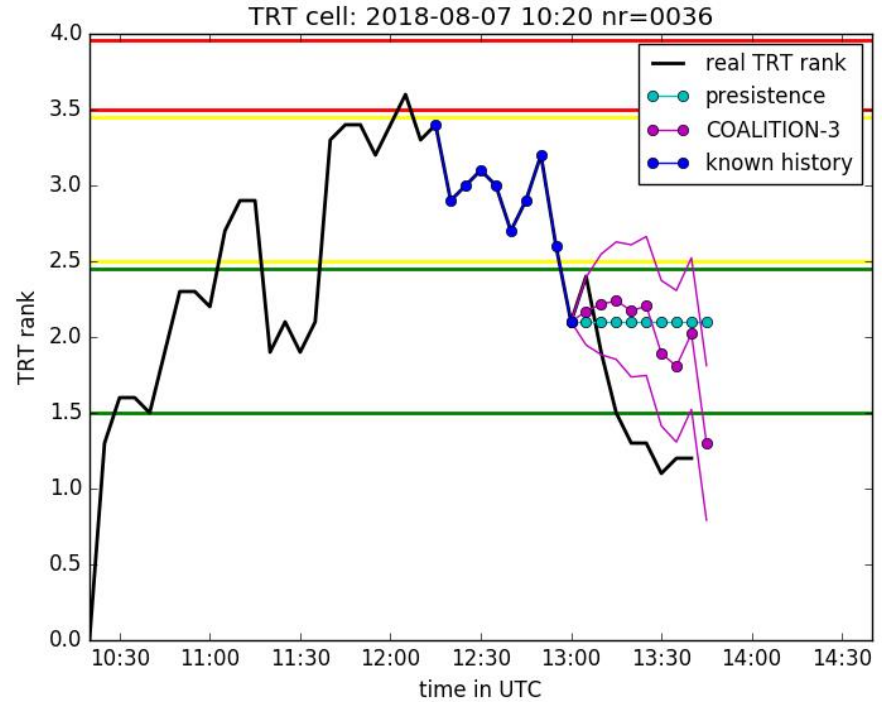
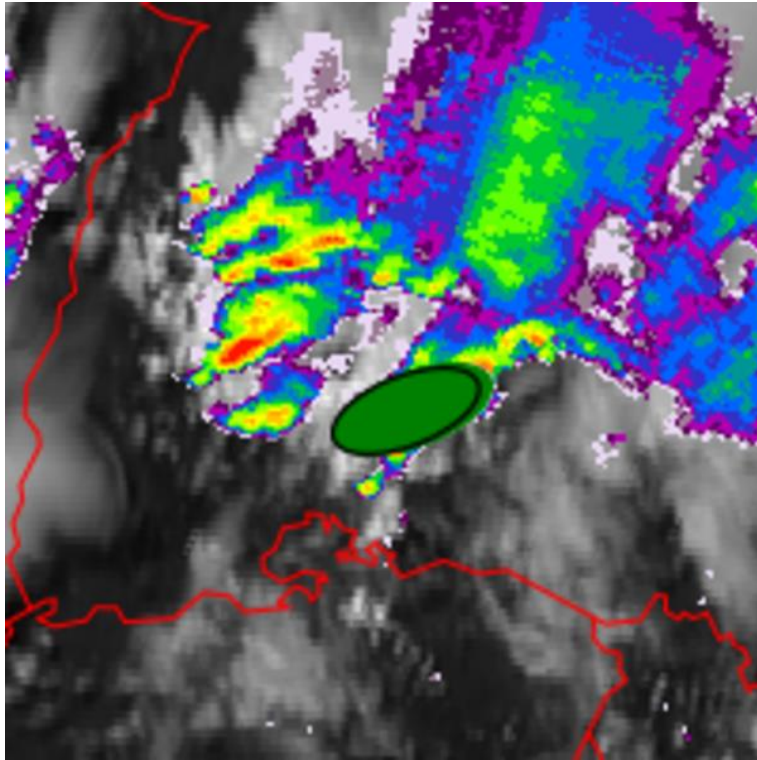


Visualization of the nowcasting results



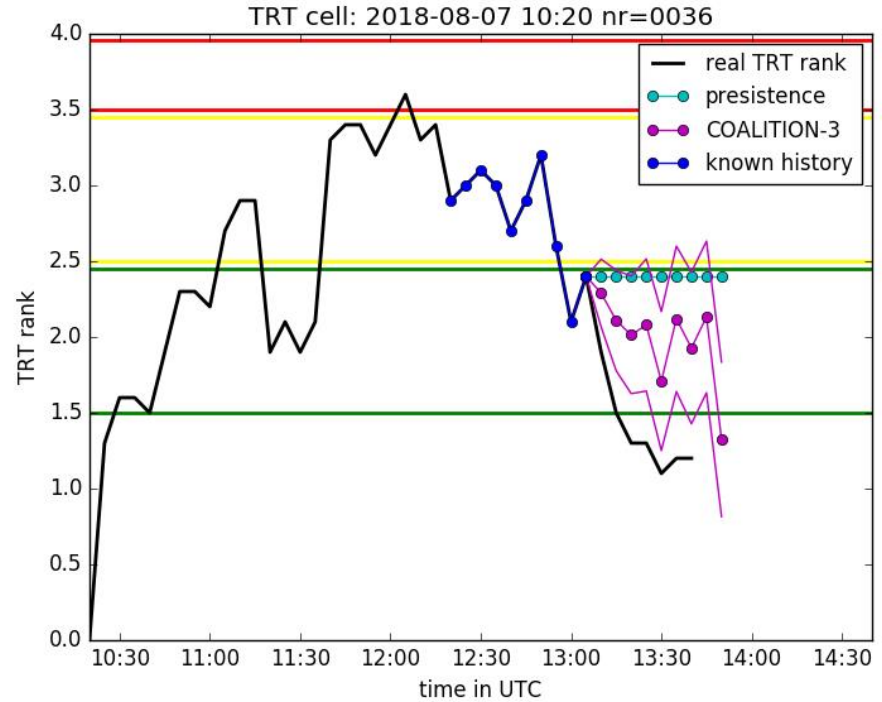
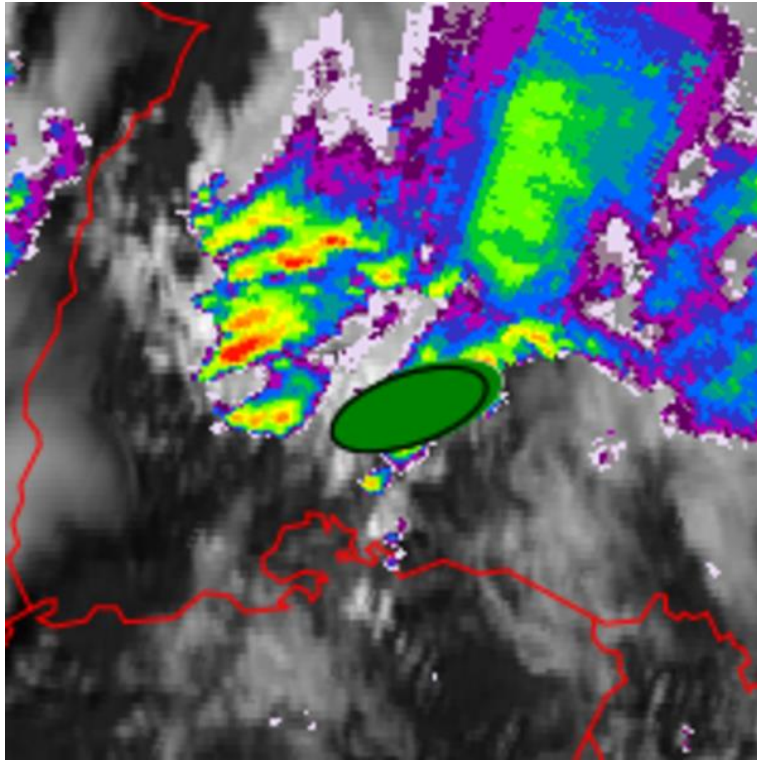


Visualization of the nowcasting results



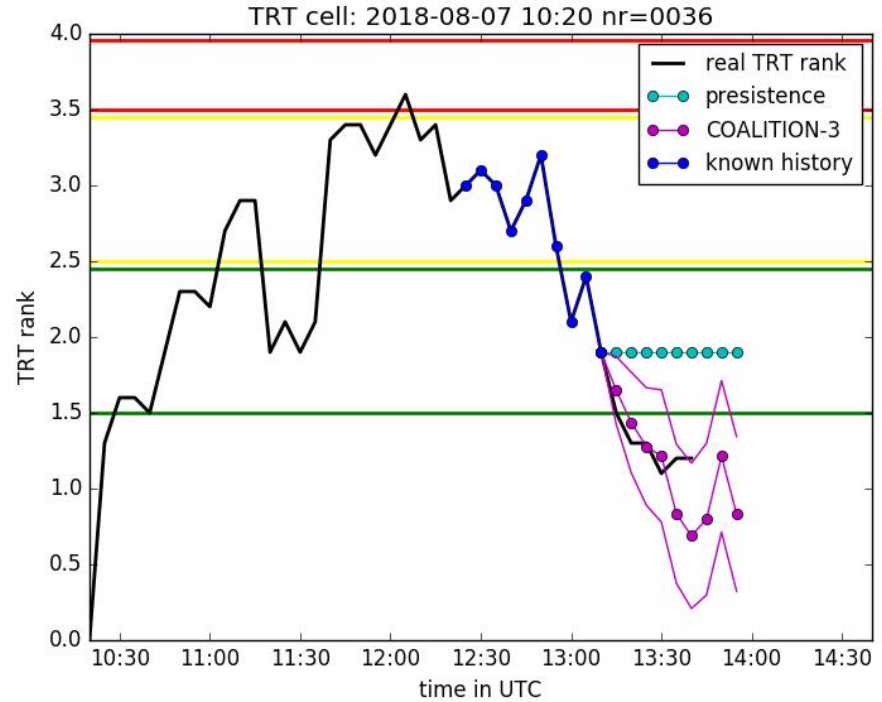
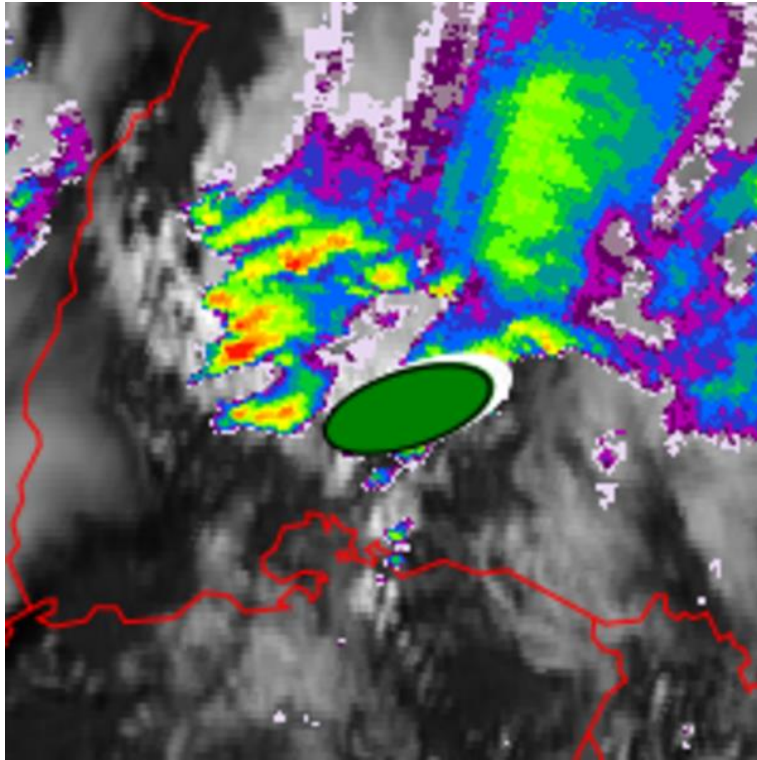


Visualization of the nowcasting results





Visualization of the nowcasting results





Summary

- A multi-sensor thunderstorm nowcasting was developed (COALITION-3)
- Thunderstorm cell are tracked with motion vectors
- Multi-sensor cell history is monitored
- Gradient boosted trees (XGBoost) is used to nowcast thunderstorm intensity
- Feature importance ranking enables reduction to 750 predictors
- Nowcasted TRT ranks for all forecast times better than persistence
- Nowcasted TRT ranks skilful up to 45 min into the future
- Probability matching is used to preserve standard deviation
- Easily expandable to more input variables
- Strait forward to train ML to nowcast other variables, e.g. lighting activity



Future Outlook

Future work COALITION-3

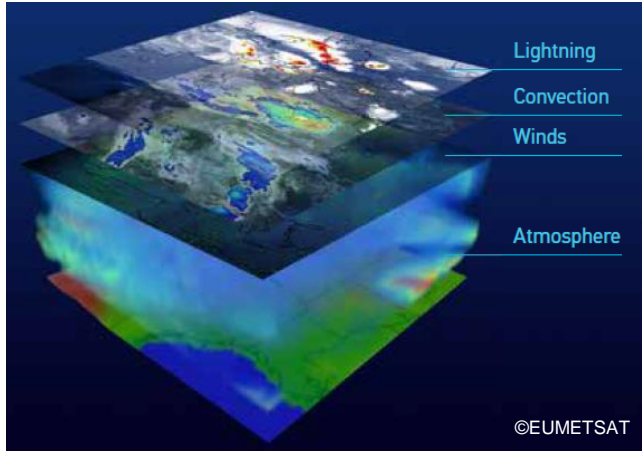
- Improved TRT rank **uncertainty estimates**
- **Feedback** from **forecasters**
- **Validation** of COALITION-3 in comparison to TRT and operational thunderstorm warnings (POD, FAR, SS for different warning levels)
- **Operationalisation** of COALITION-3 - Implement into automatic warning suggestion system of NinJo
- Explore **applicability for aviation** with European coverage

Future work COALITION-4

- Forecast of **specific thunderstorm hazards**: heavy precipitation, lightning, hail and wind gusts
- Improved **thunderstorm motion** (right movers, topographic steering)
- Exploiting **GPU-powered deep learning** technology, e.g. convolutional neural networks to improve further the nowcast quality and lead time
- **Prioritize warning** suggestions by risk & uncertainty

Future Outlook – Adaptation to MTG

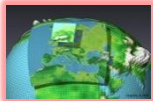
1. MTG 4D Weather Cube



Radar



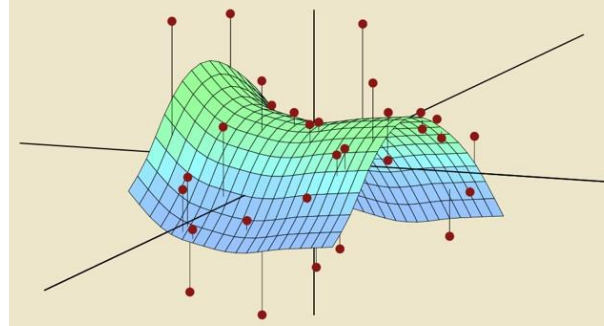
NWP model



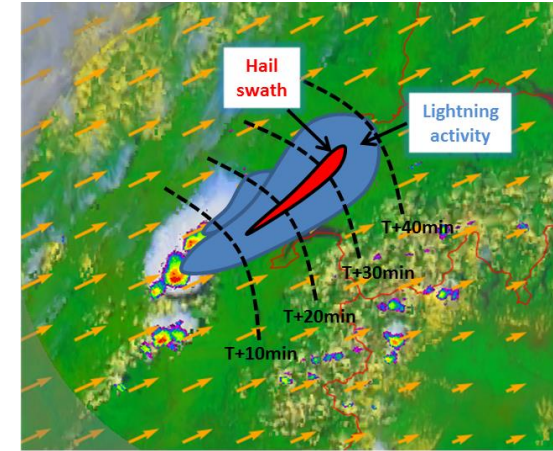
Lightning



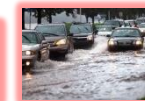
2. Advanced Machine Learning



3. Thunderstorm Nowcast



Rain



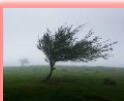
Hail



Lightning



Wind





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