



CZECH
HYDROMETEOROLOGICAL
INSTITUTE



DEPARTMENT OF
ATMOSPHERIC PHYSICS
CHARLES UNIVERSITY

CONVECTIVE STORM NOWCASTING CAPABILITIES OF REMOTE SENSING IN CENTRAL EUROPE

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3rd European Nowcasting Conference

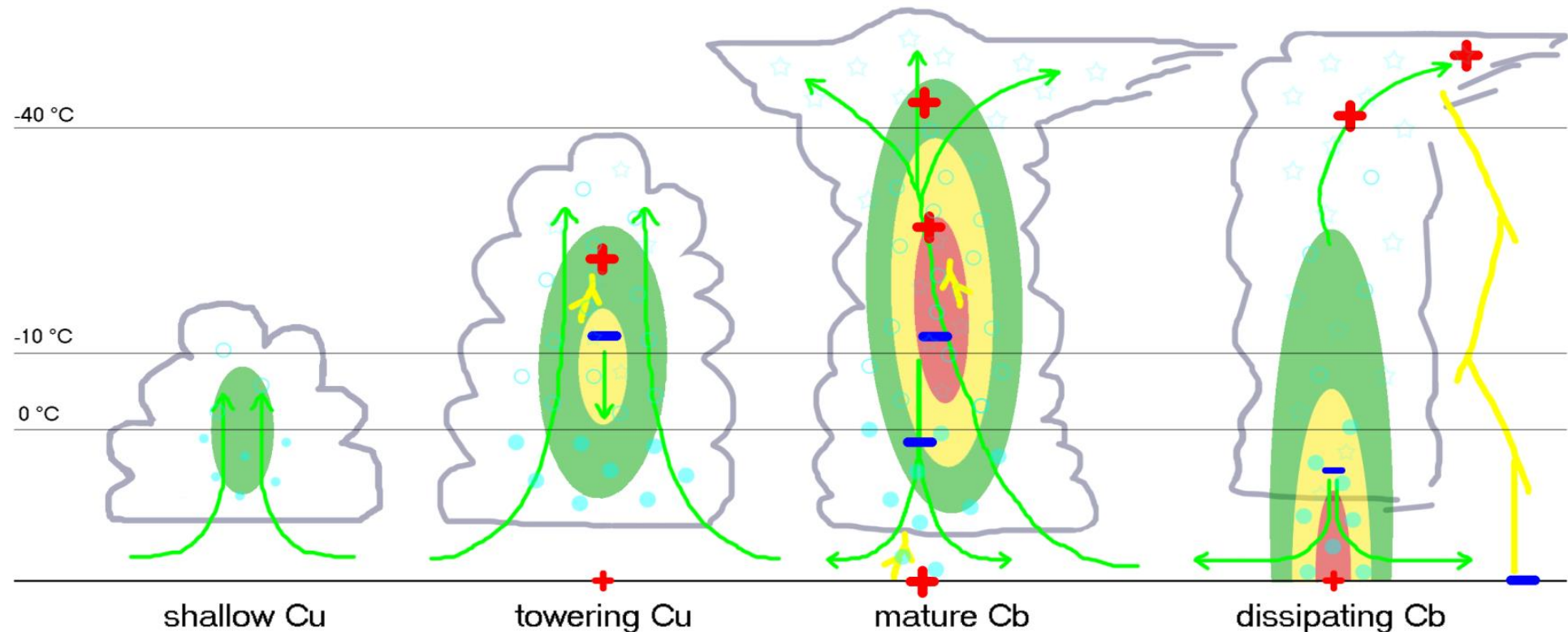
Madrid, Spain, 25 April 2019

www.chmi.cz

Central Forecasting Office, Prague
michaela.valachova@chmi.cz

POSSIBLE UTILIZATION

- electrification, dynamics and microphysics connected → changes visible in all remote sensing data → **NOWCASTING**





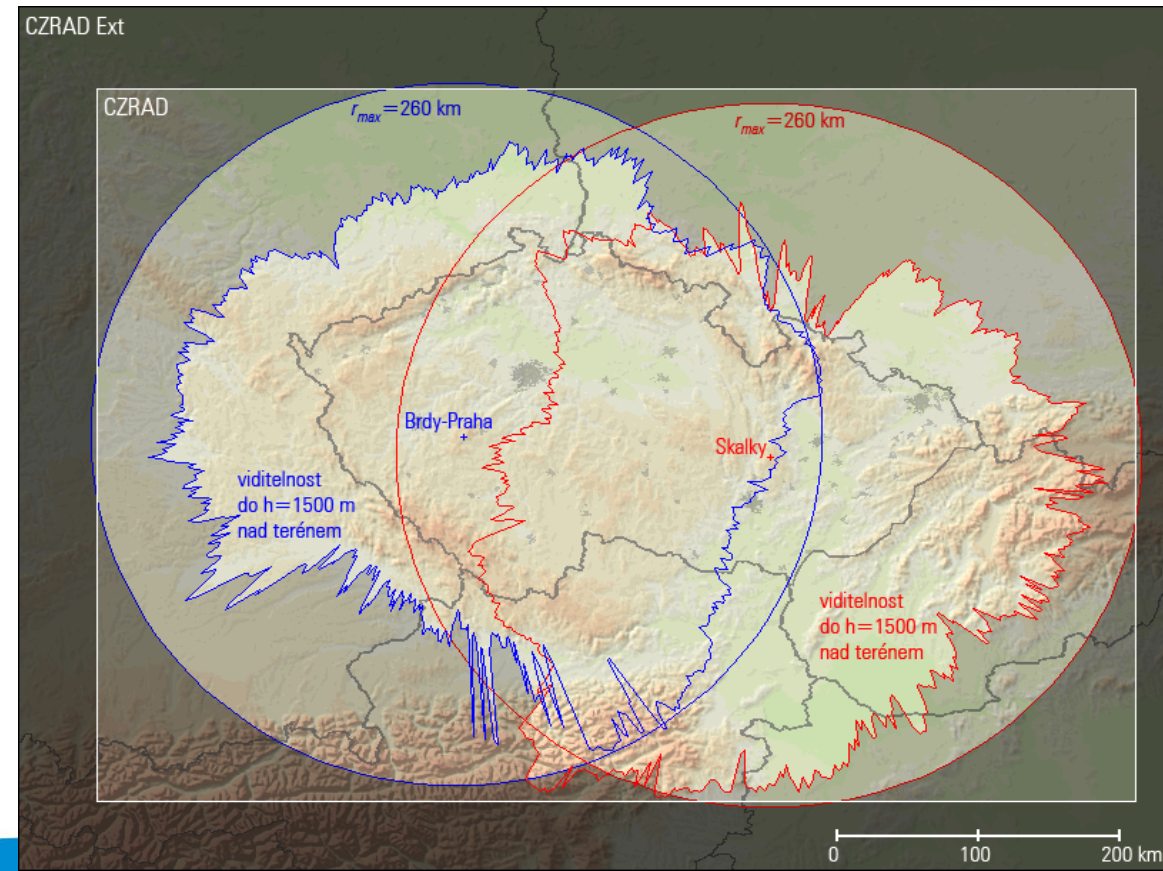
REMOTE SENSING

RADARS

- microphysical properties and dynamics

Polarimetric Doppler radars (upgrade in 2015)

- C band ($\lambda \sim 5$ cm), 12 elevations
- resolution 1×1 km (whole domain)
- many useful applications:
 - CELLTRACK, COTREC
 - CELDN strokes
 - PrecipView, WarnView

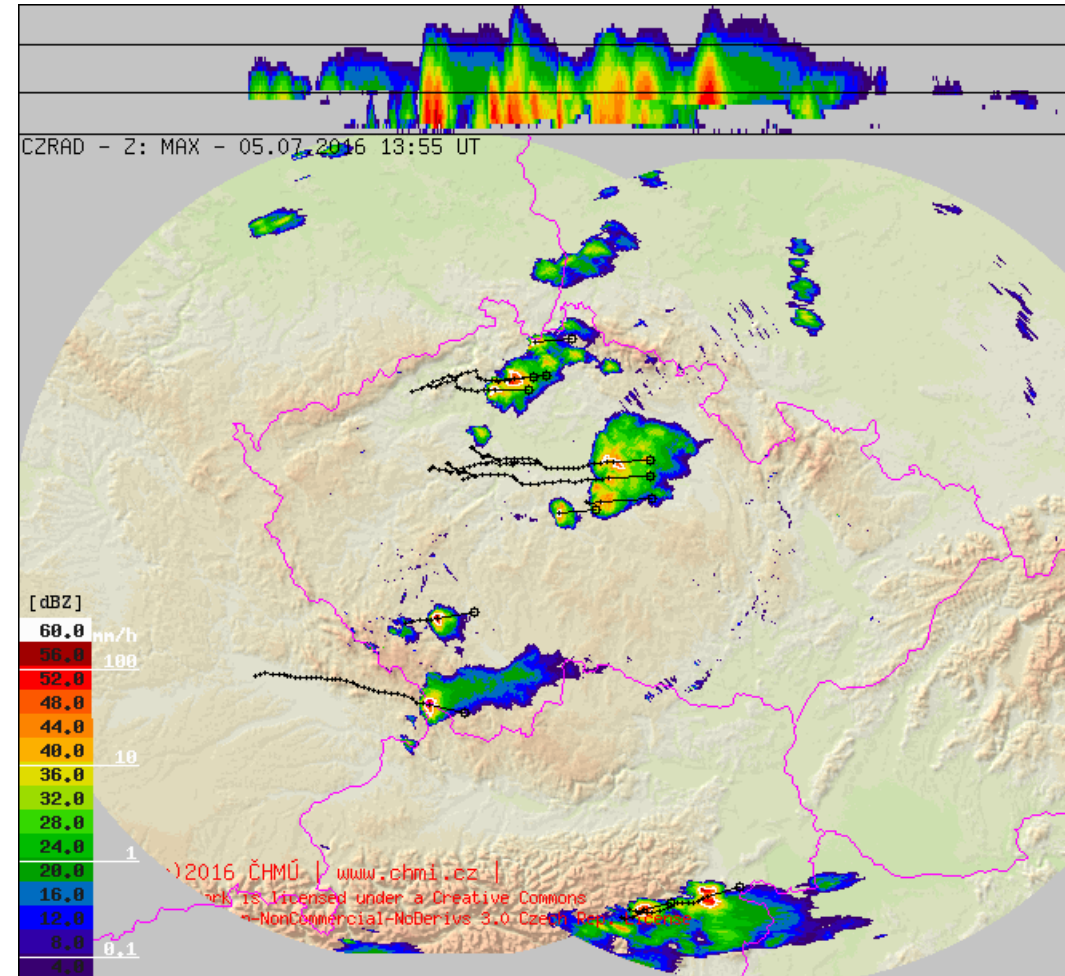


RADARS - CELLTRACK

- reflectivity cores tracking algorithm
- developed in CHMI (Hana Kyznarová)

For the presented study:

- no tracking, just identification of cores
- characteristics of cells:
 - threshold of 44 → 30 dBZ (isolated storms)
 - parameters:
 - AREA, VOL, VOL44, MAX_R, TOP
 - VIL, POH, SHI, POSH, MESH



Operational output of CELLTRACK
(JSMeteoView)

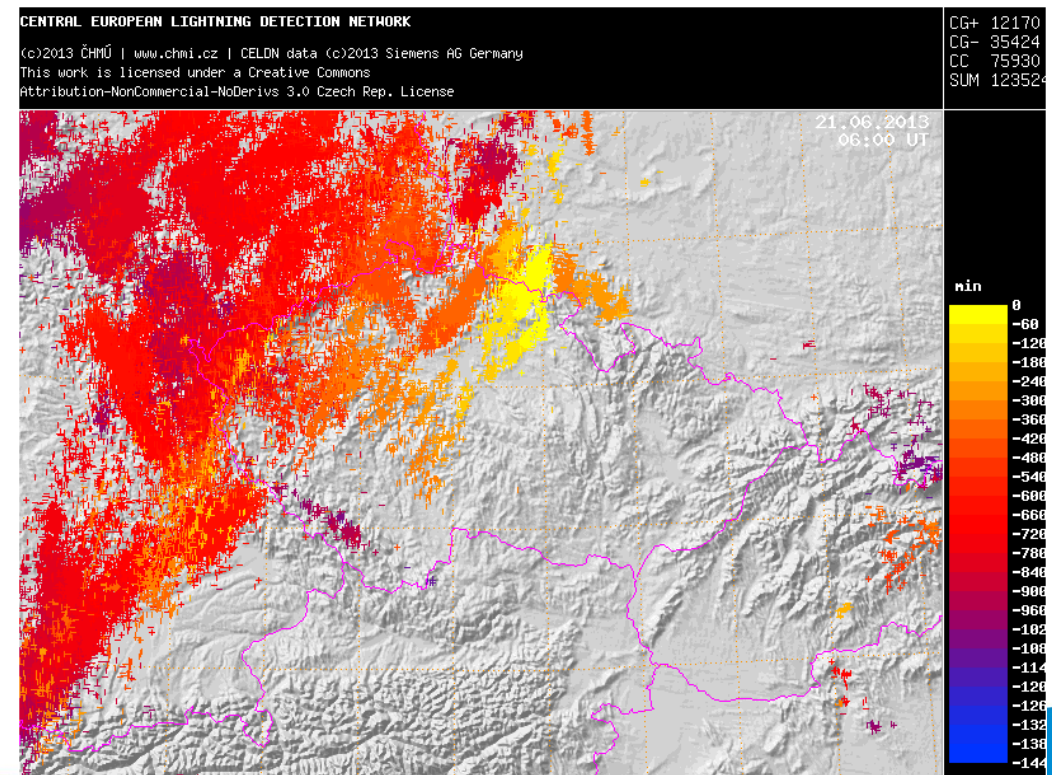
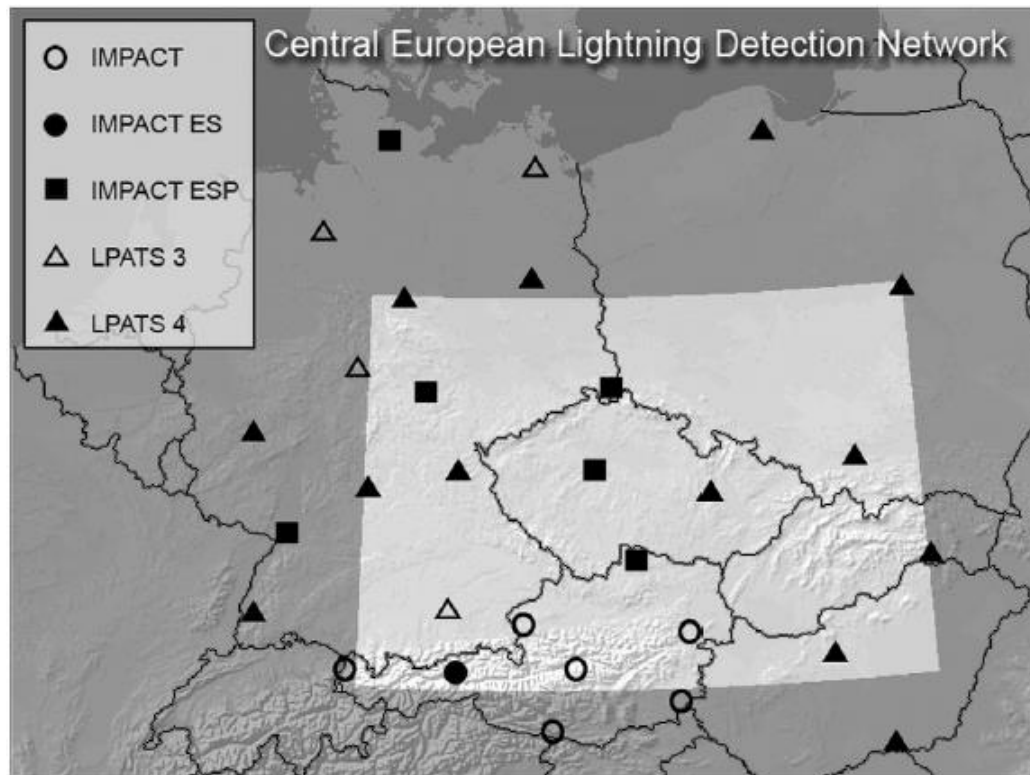
Kyznarová H., Novák P. (2009): CELLTRACK – Convective cell tracking algorithm and its use for deriving lifecycle characteristics, Atmospheric Research, vol. 93



LIGHTNING DETECTION

CELDN (Central European Lightning Detection Network)

- part of EUCLID, operated by Siemens AG
- operatively used in CHMI until 30 Sep 2017



LIGHTNING DETECTION

- microphysical properties, strength of updraft

every stroke: type (CC, CG) and polarity, time [ms], location (LAT and LON), stroke peak current estimation [kA]

- detection efficiency: 85 % or higher for CG
- location accuracy: better than 500 m
- estimate of peak current: error ~ tens of %
- no stroke clustering into flashes



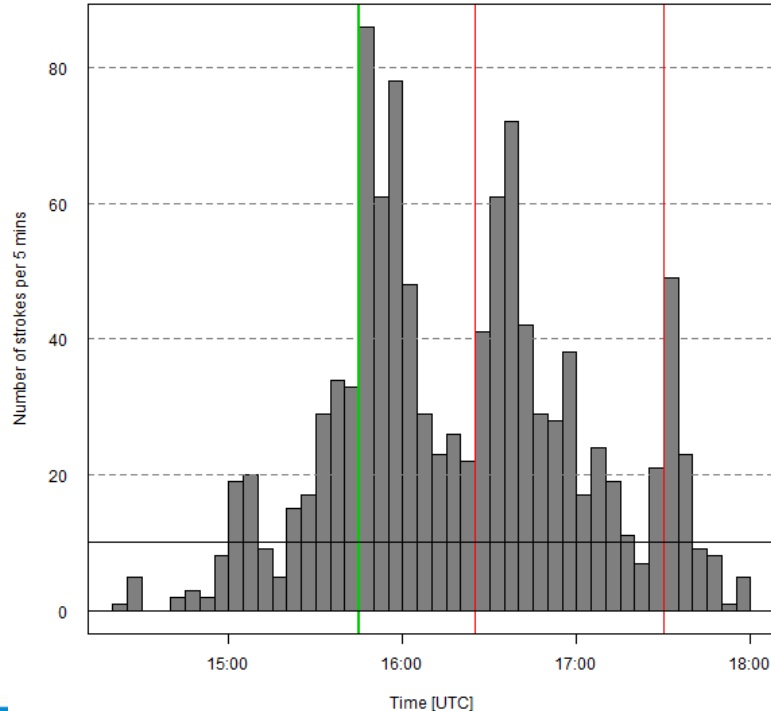
* Adarsh Hatwar



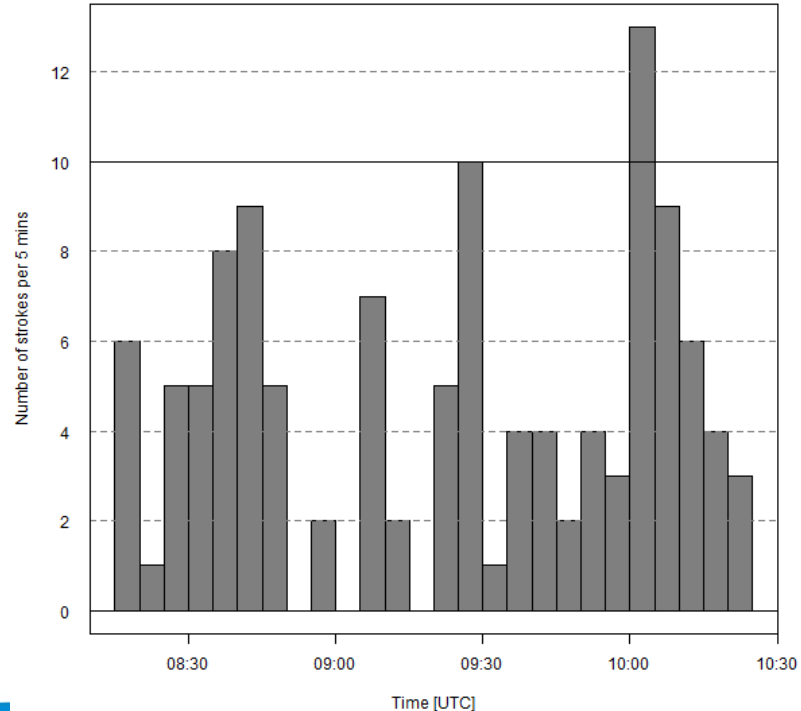
LIGHTNING JUMP

- lightning jump algorithm
 - threshold for activation: 10 strokes per 5 min (median for non-severe storms)
 - difference is higher than 2σ of previous 15 min (Schultz et al., 2009)
 - normalized difference between the amount of strokes = value of LJ

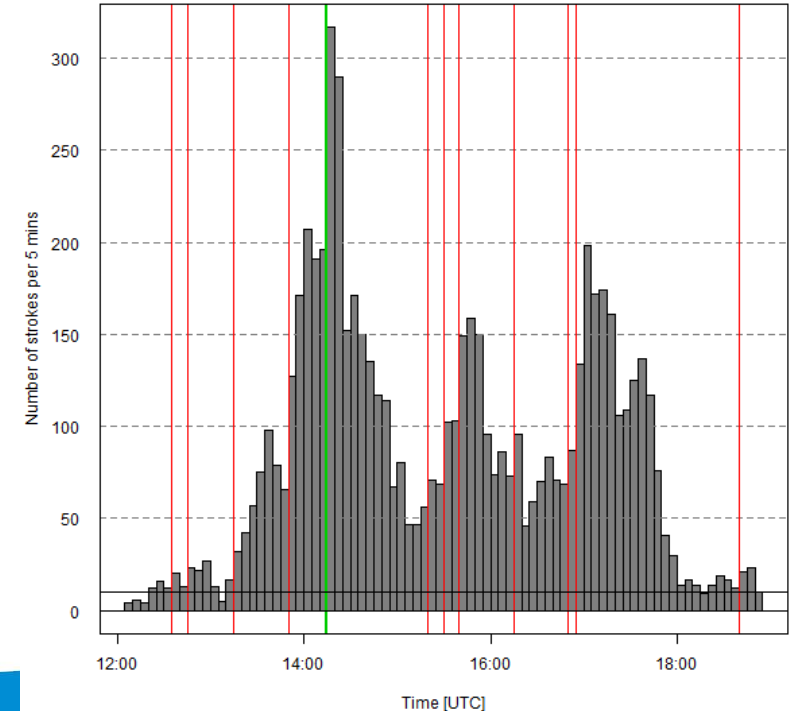
Lightning jumps on 2017-07-22 of the storm CZ-Luhacovice



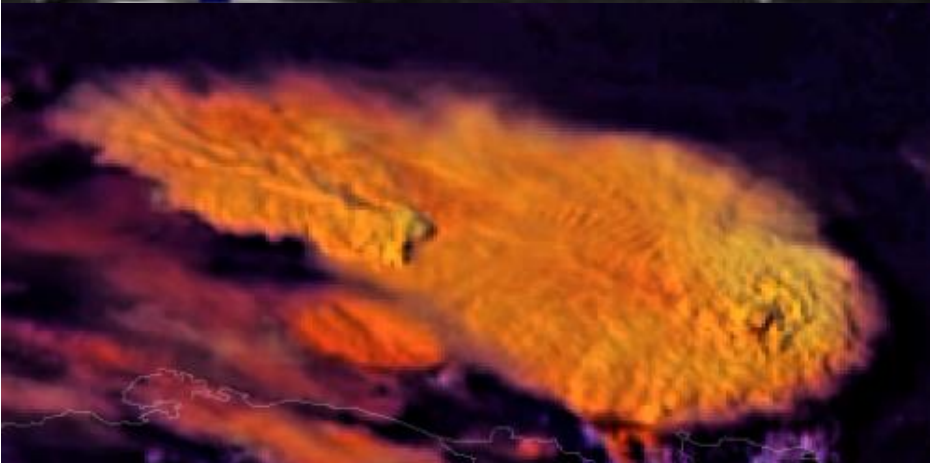
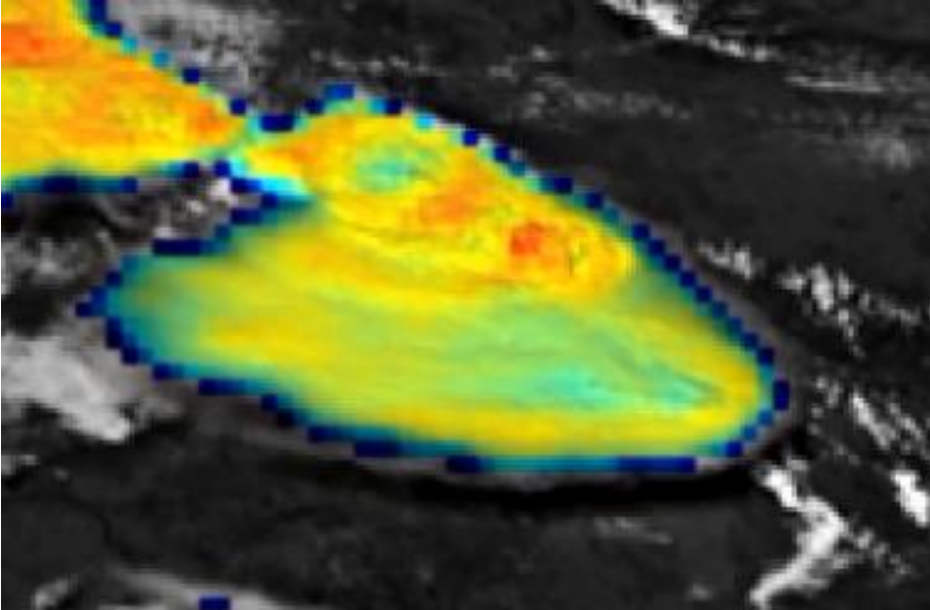
Lightning jumps on 2017-07-08 of the storm CZ-Opava



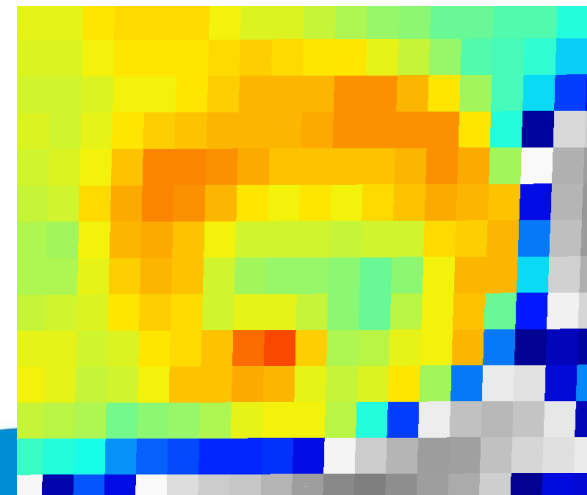
Lightning jumps on 2017-05-19 of the storm D-Erding



SATELLITES

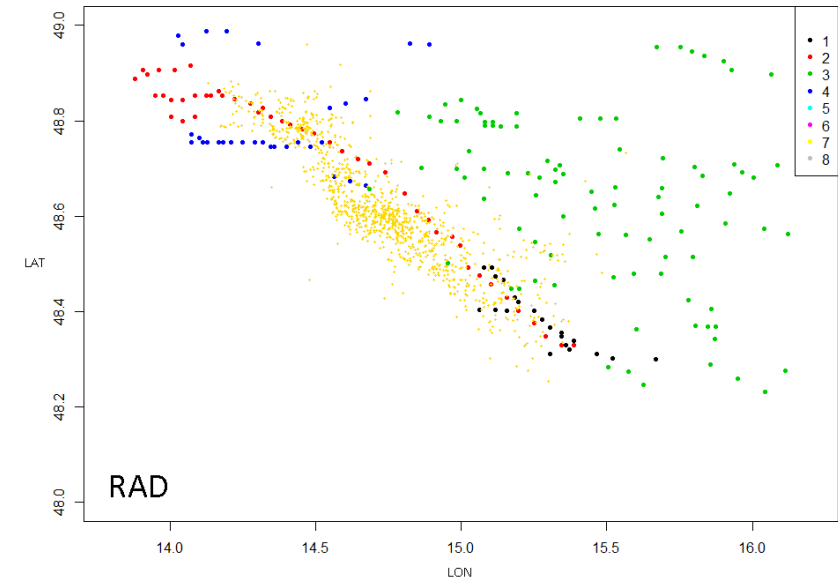
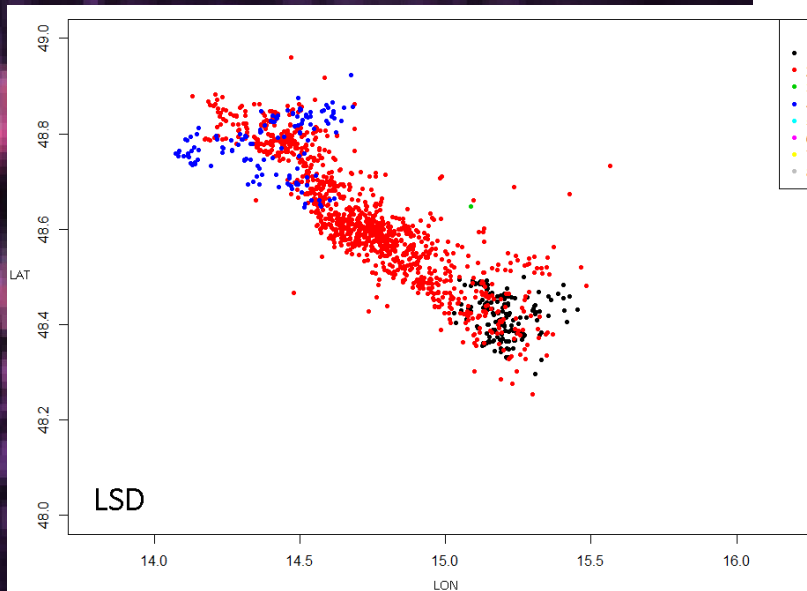
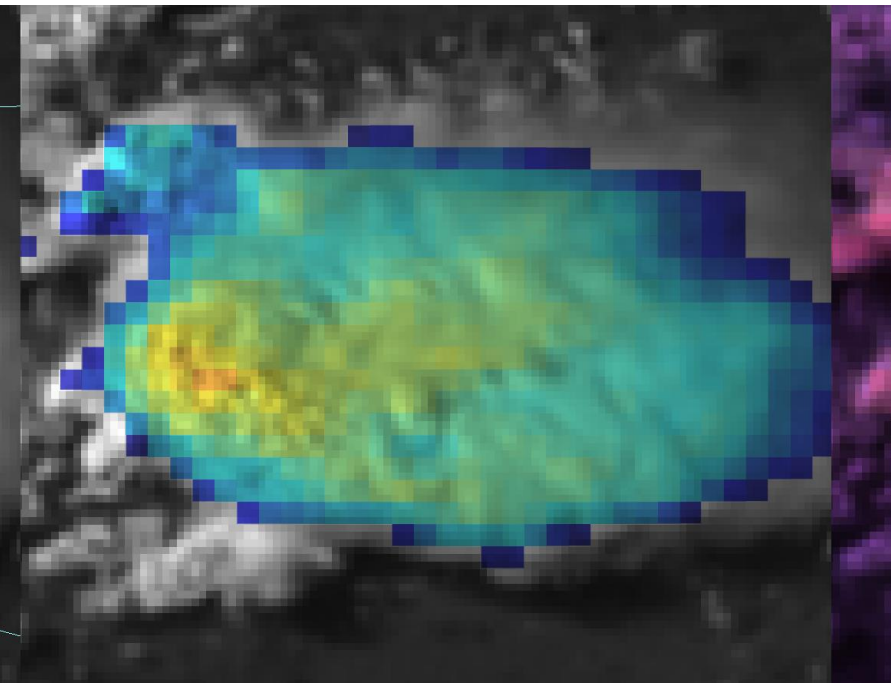
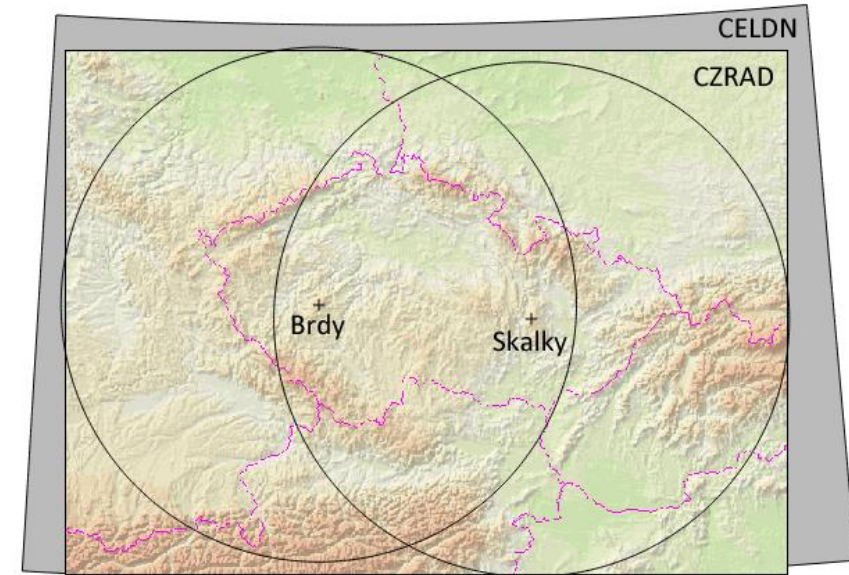


- microphysical properties and dynamics
- MSG/SEVIRI 5 min RSS or 15 min data
 - resolution 3×6 km (1×2 km in HRV)
 - individual channels:
 - » IR 10.8 and IR-BT, IR 3.9 and HRV
 - RGB products
 - » Storm, VIS-IR, Snow
 - sandwich products



ISOLATED STORM SELECTION

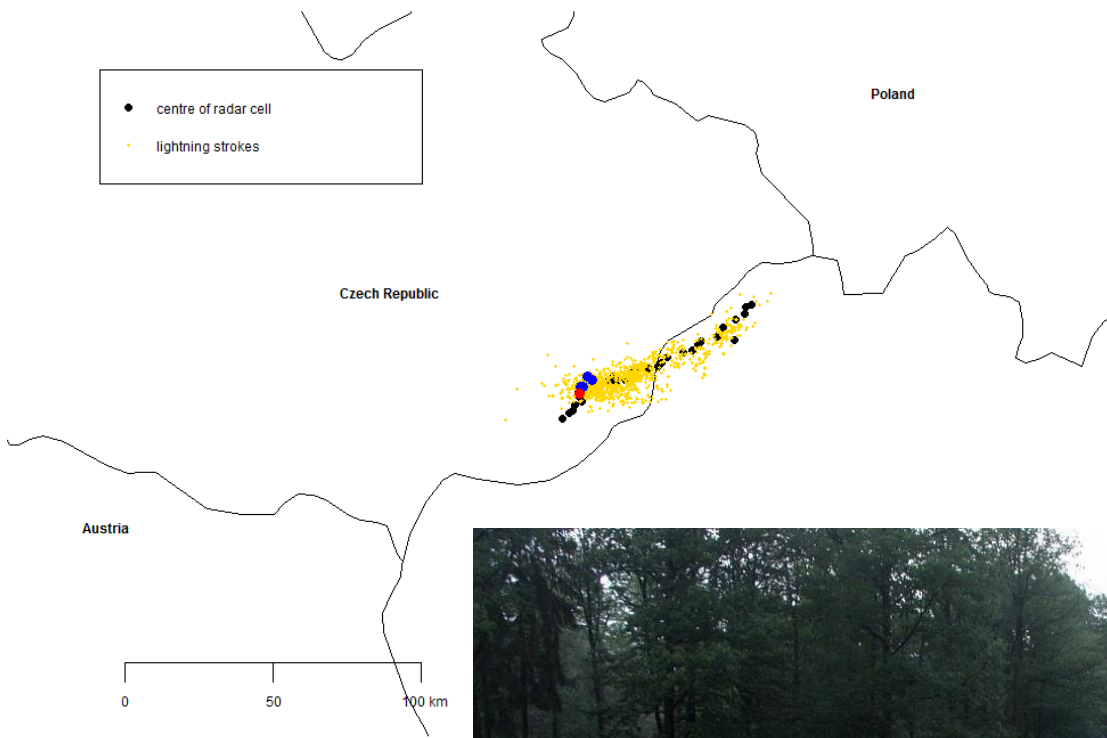
- at least 5 strokes over the Czech republic
- CELLTRACK algorithm (30 dBZ threshold)
- remote sensing data available



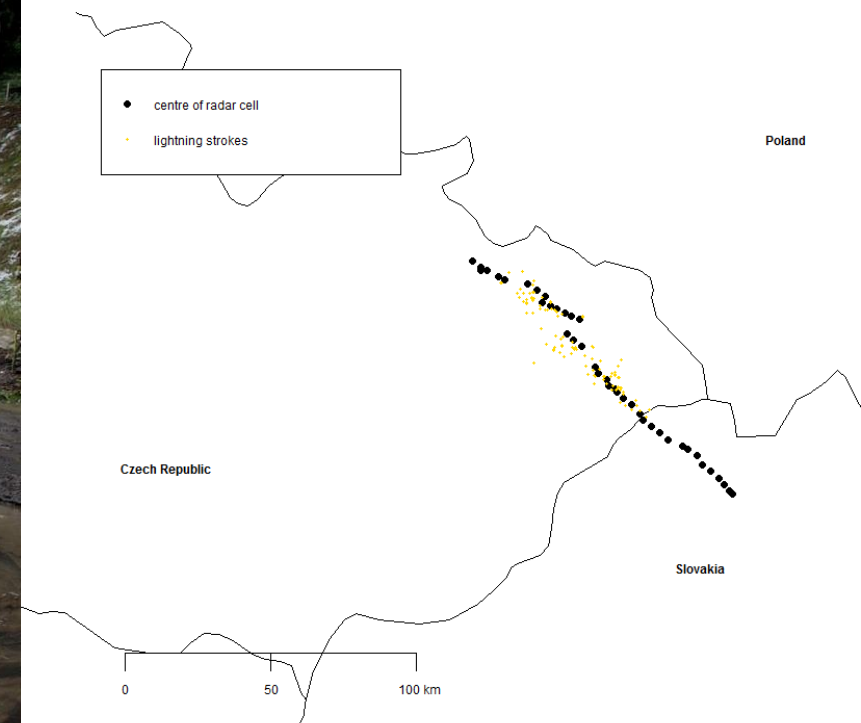


SEVERE OR NON-SEVERE ?

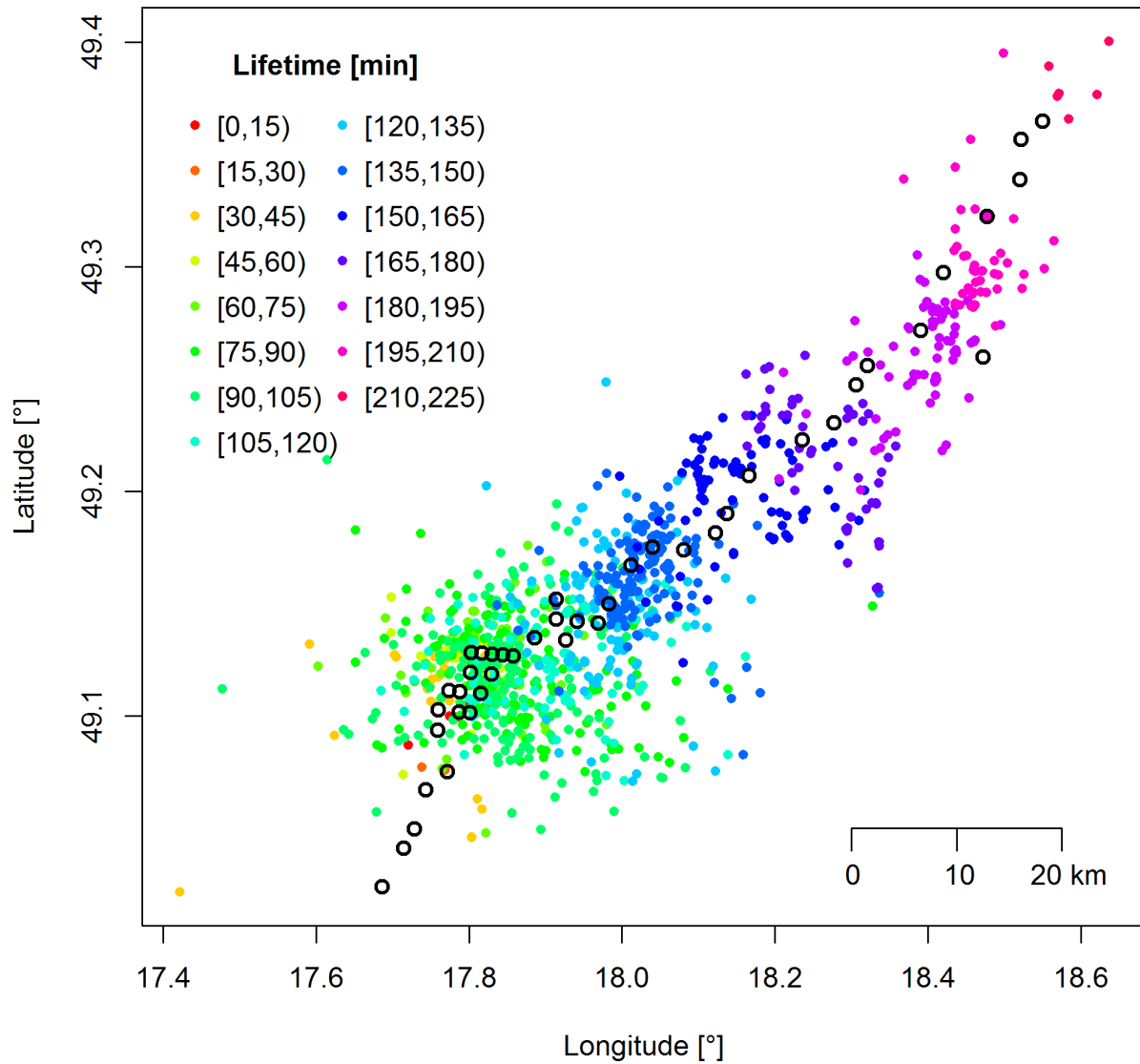
Radar and lightning detections on 2017-07-22 at 15:57 UTC (CZ-Luhacovice)



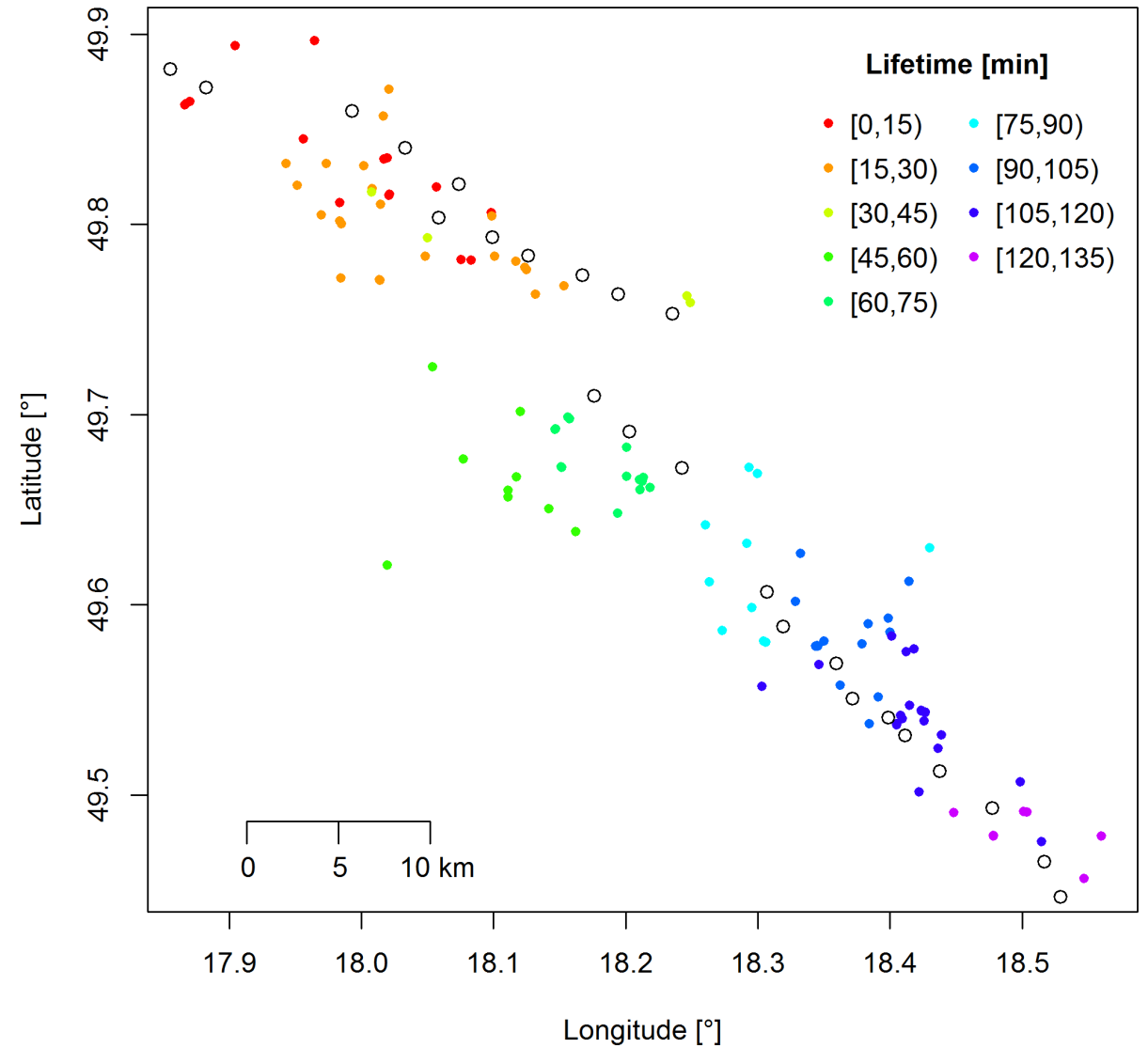
Radar and lightning detections on 2017-07-08 (CZ-Opava)



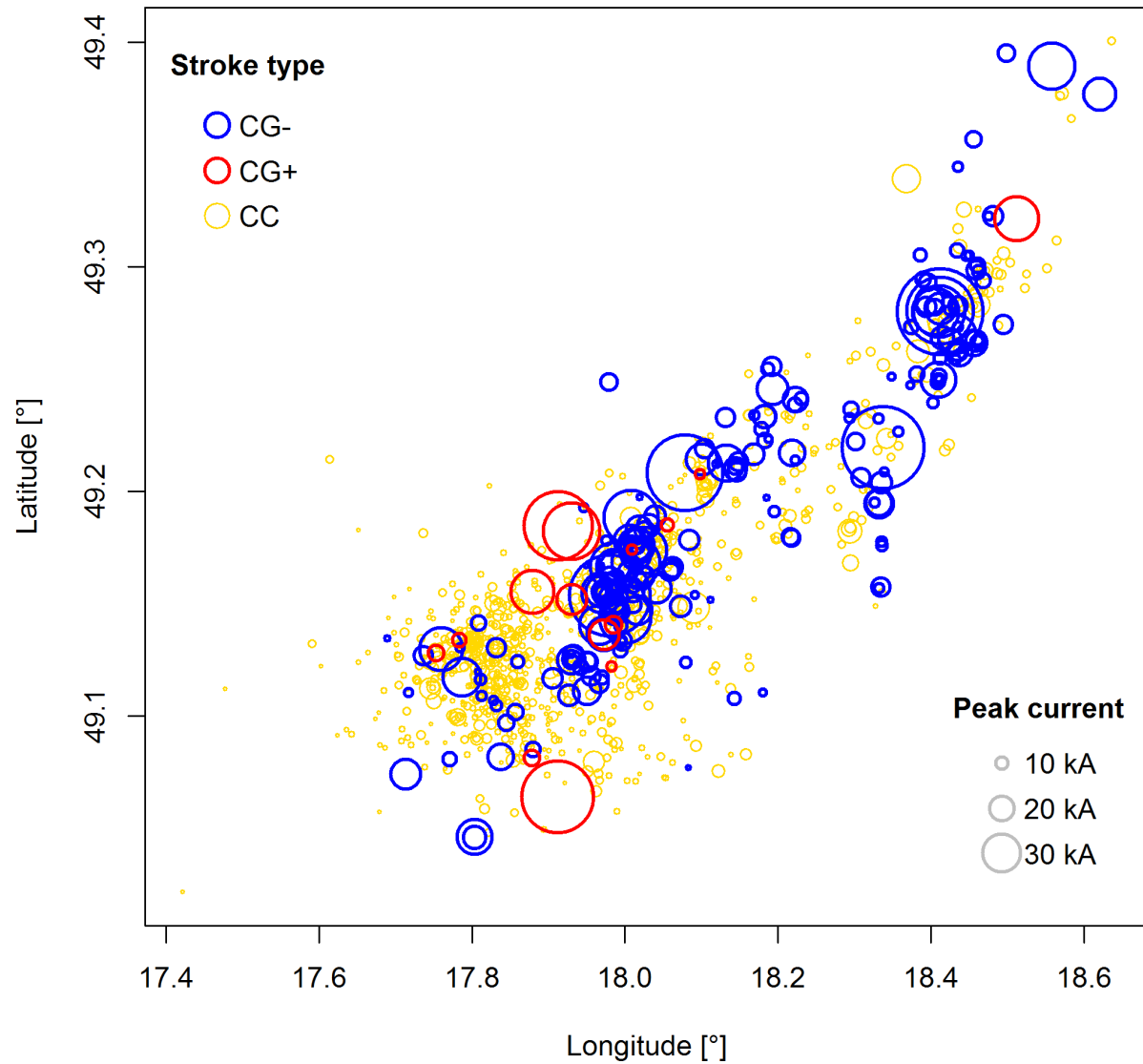
Time evolution of all strokes on 2017-07-22 from 14:20 UTC (CZ-Luhaco)



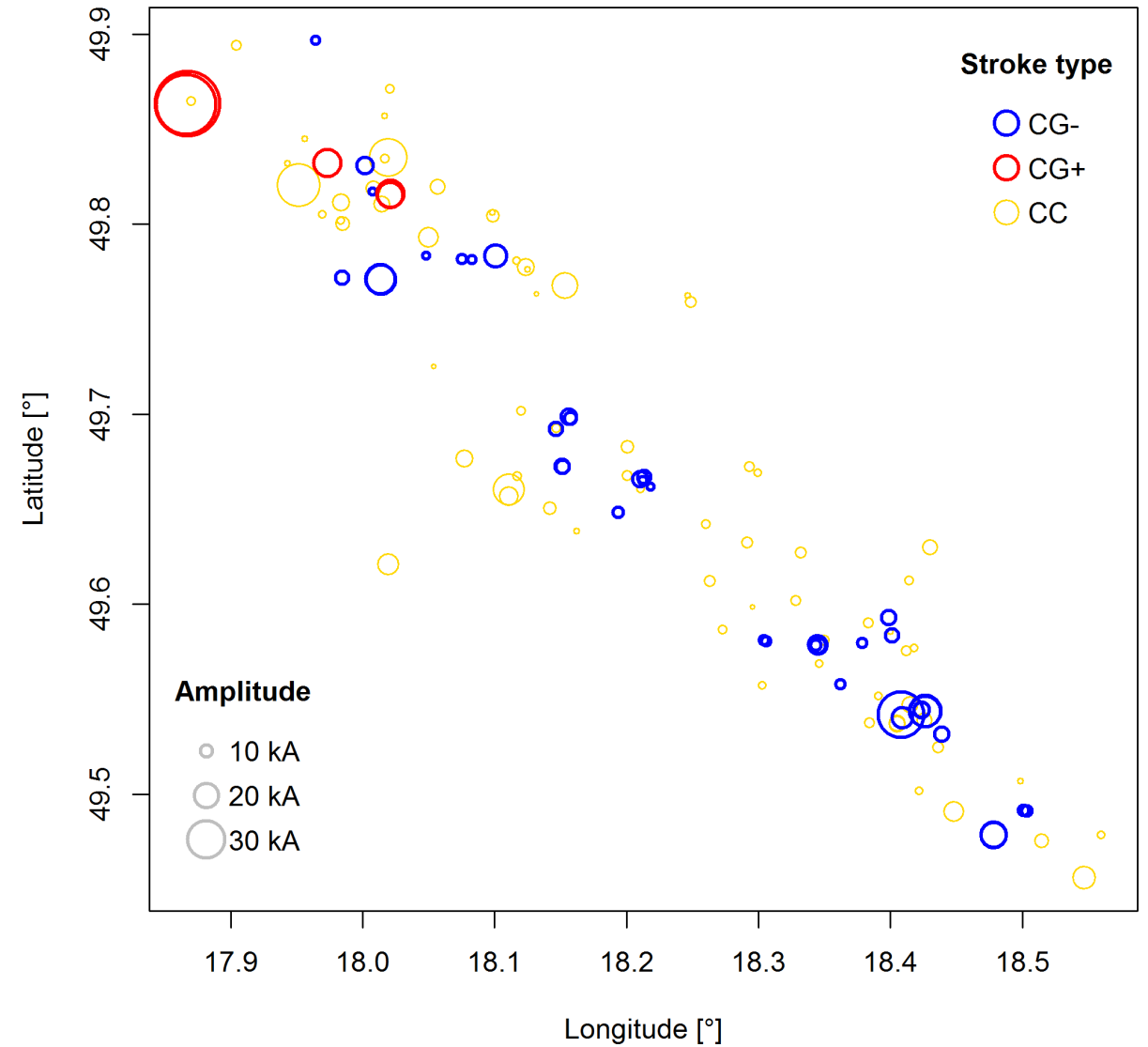
Time evolution of all strokes on 2017-07-08 from 08:17 UTC (CZ-Opav)



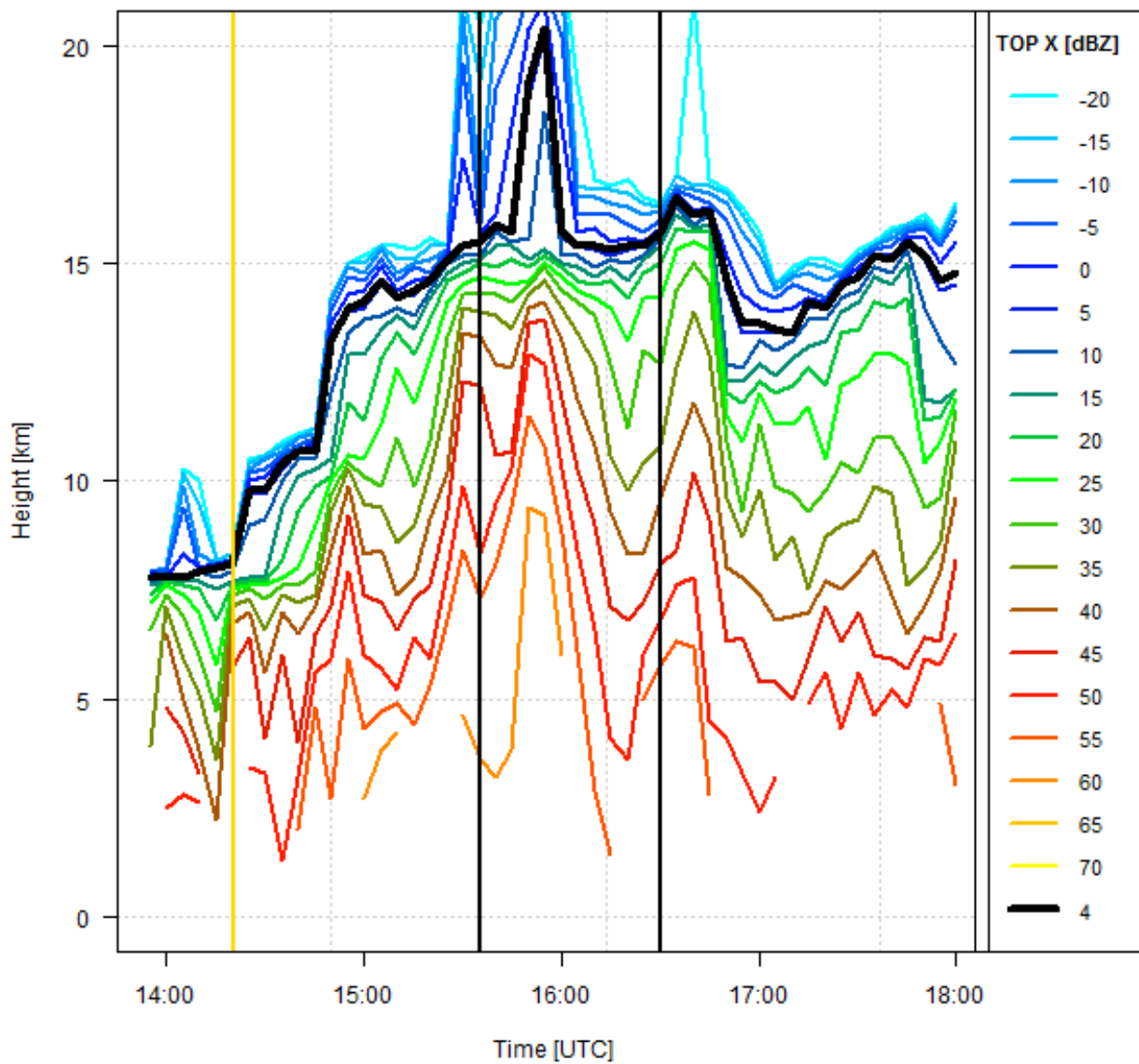
Stroke type distribution on 2017-07-22 from 14:20 UTC (CZ-Luhacovic)



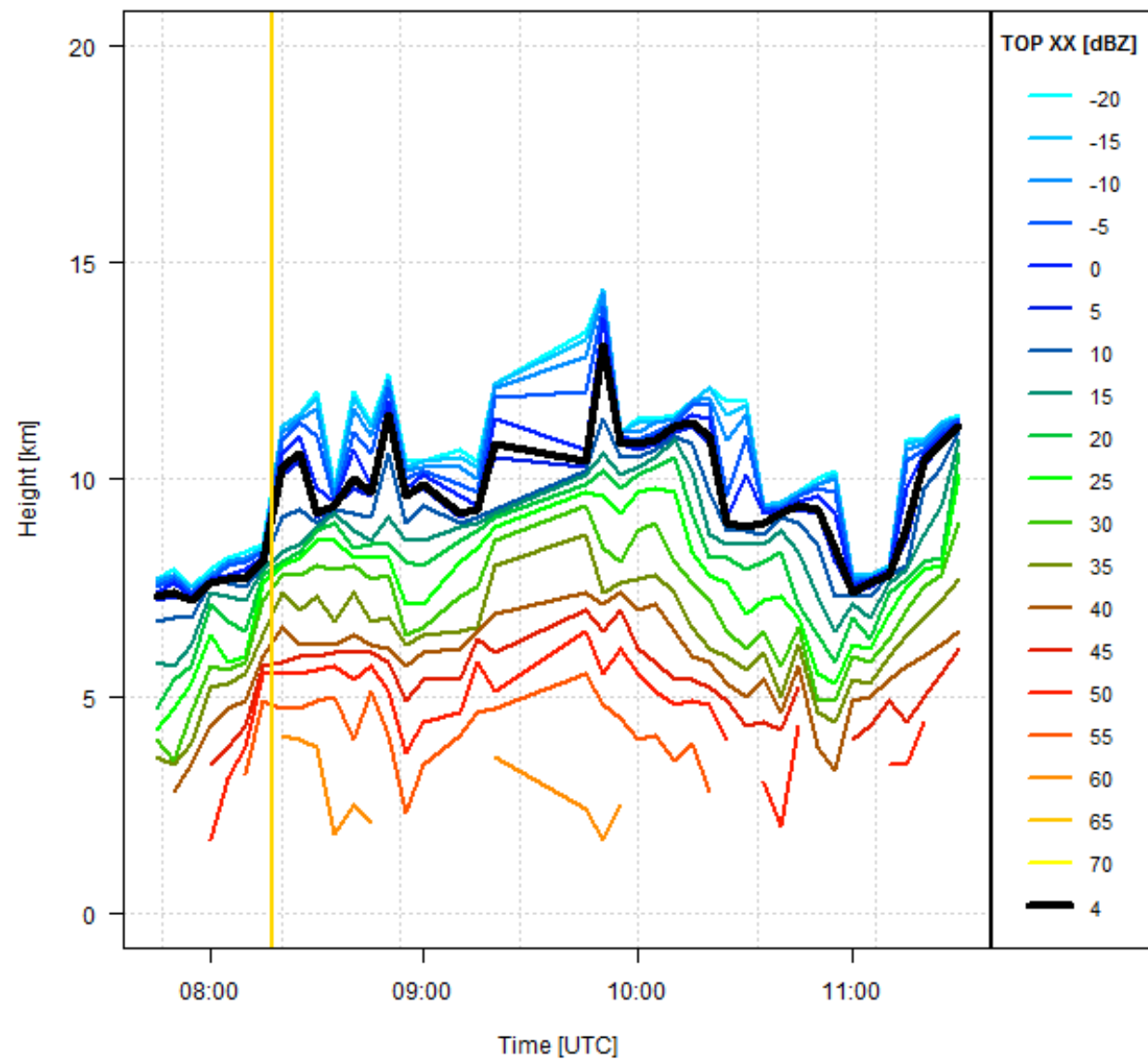
Stroke type distribution on 2017-07-08 from 08:17 UTC (CZ-Opava)



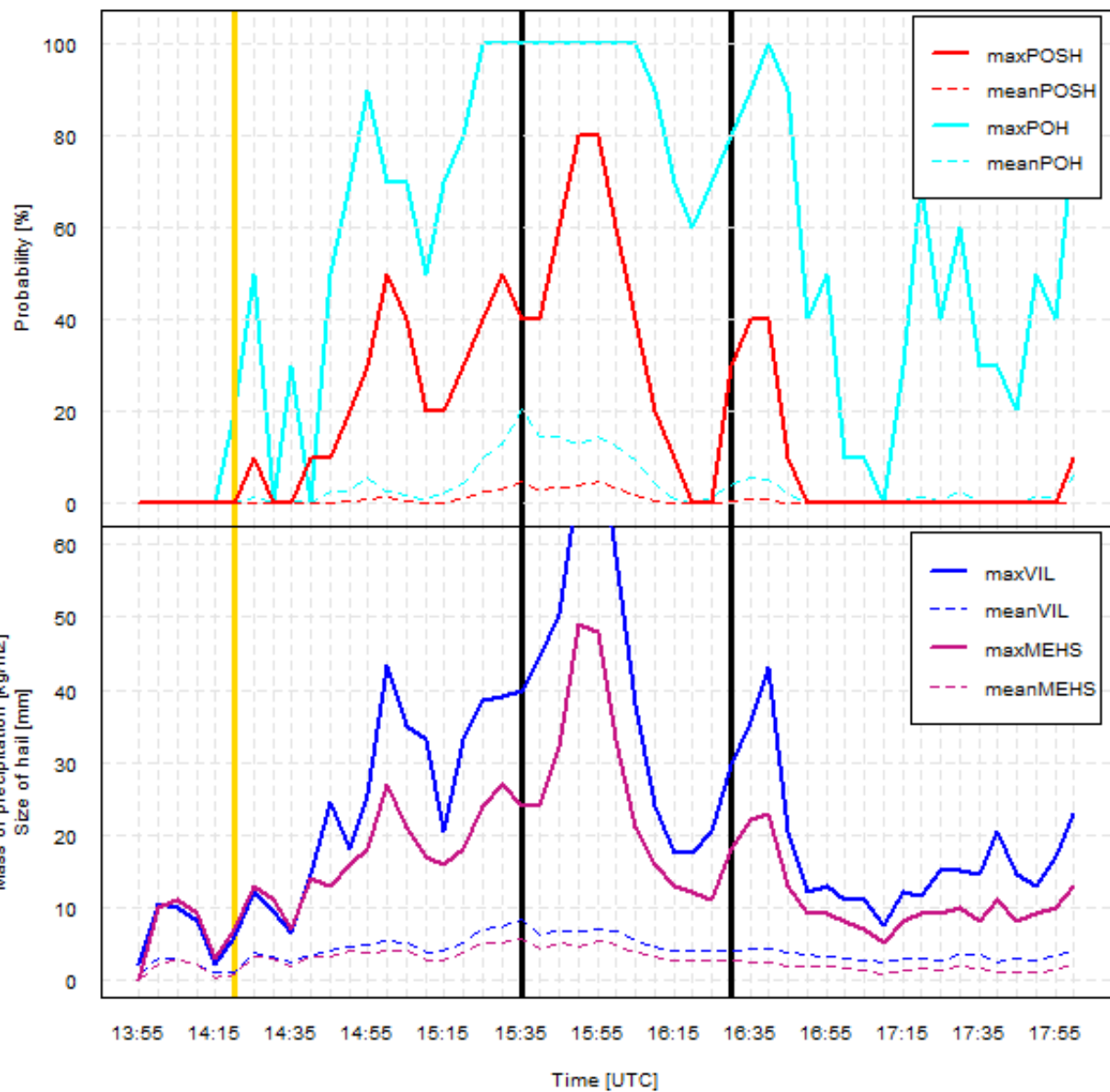
EchoTOPs of (CZ-Luhacovice) on 2017-07-22 by CELLTRACK



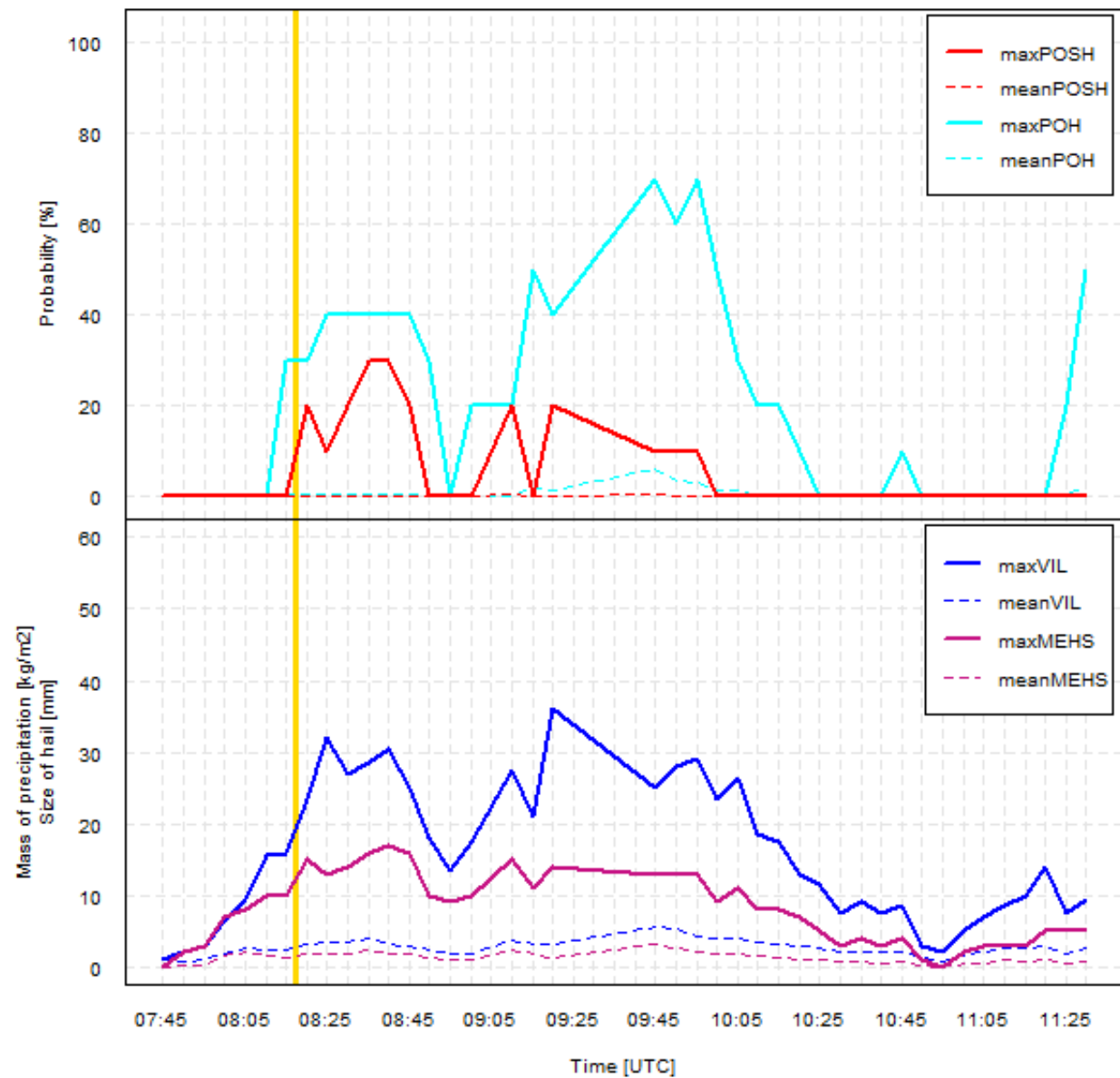
EchoTOPs of (CZ-Opava) on 2017-07-08

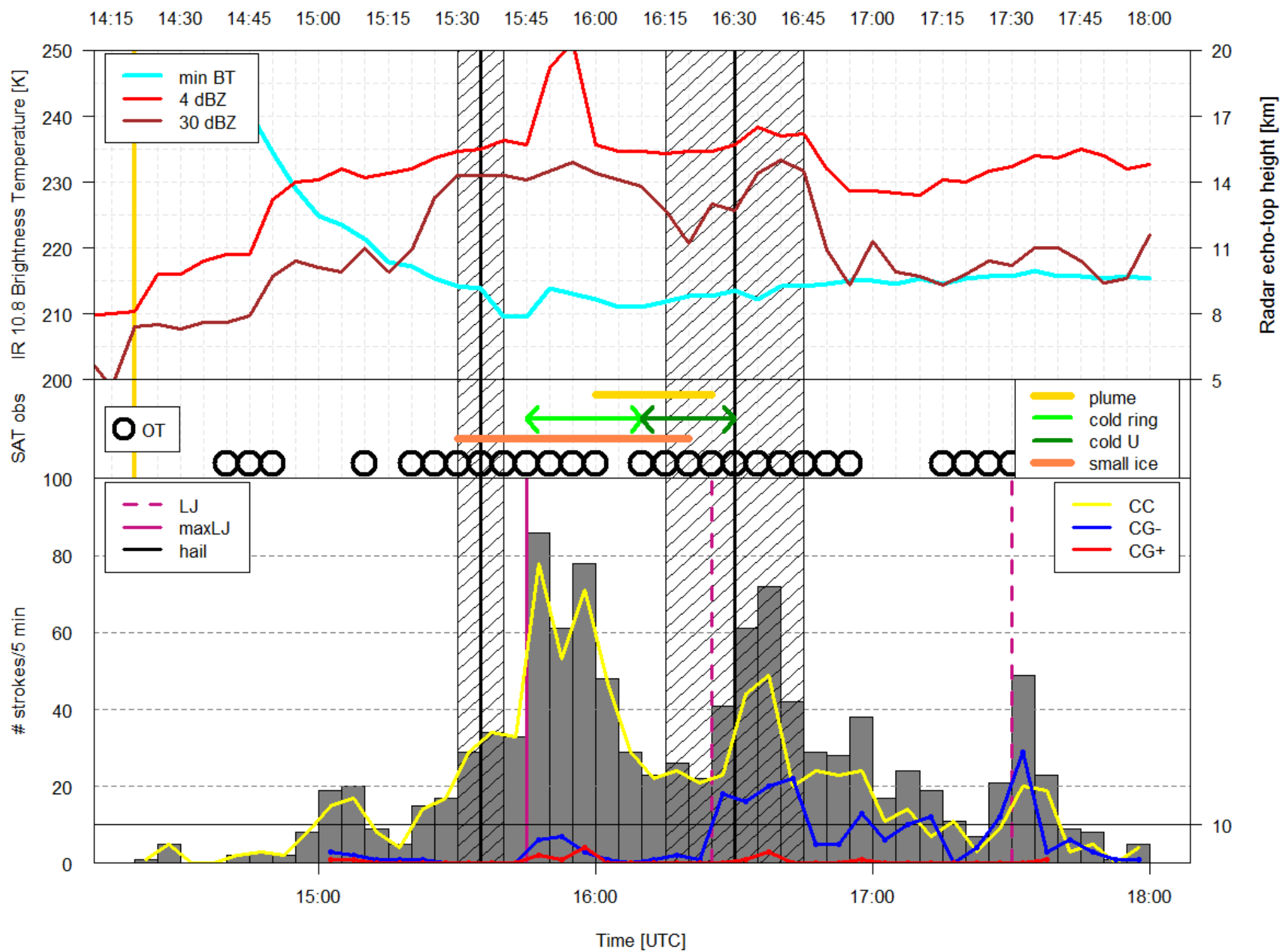


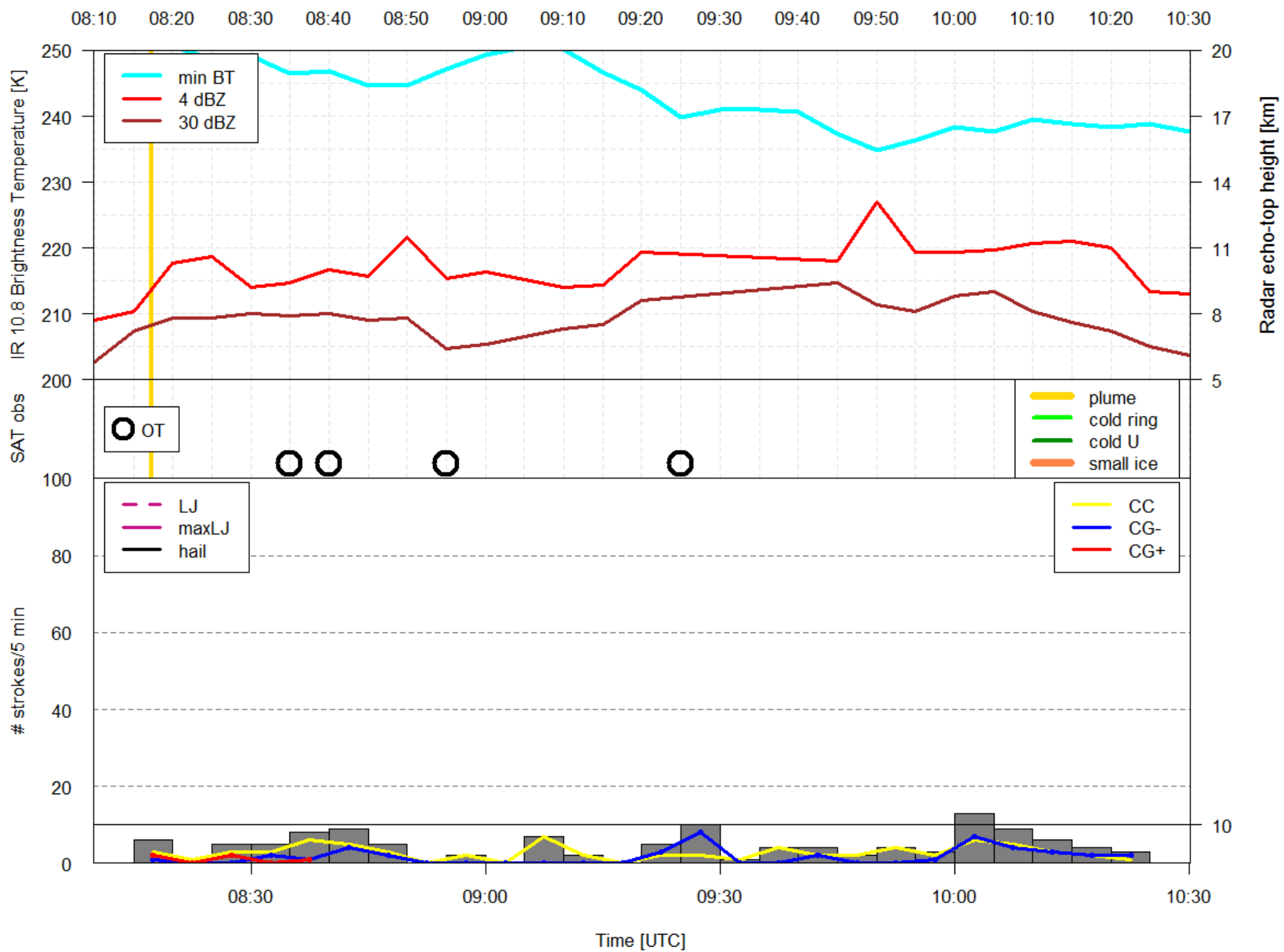
Storm POSH, POH, MEHS and VIL on 2017-07-22 by CELLTRACK (CZ-Luhacovice)



Storm POSH, POH, MEHS and VIL on 2017-07-08 by CELLTRACK (CZ-Opava)



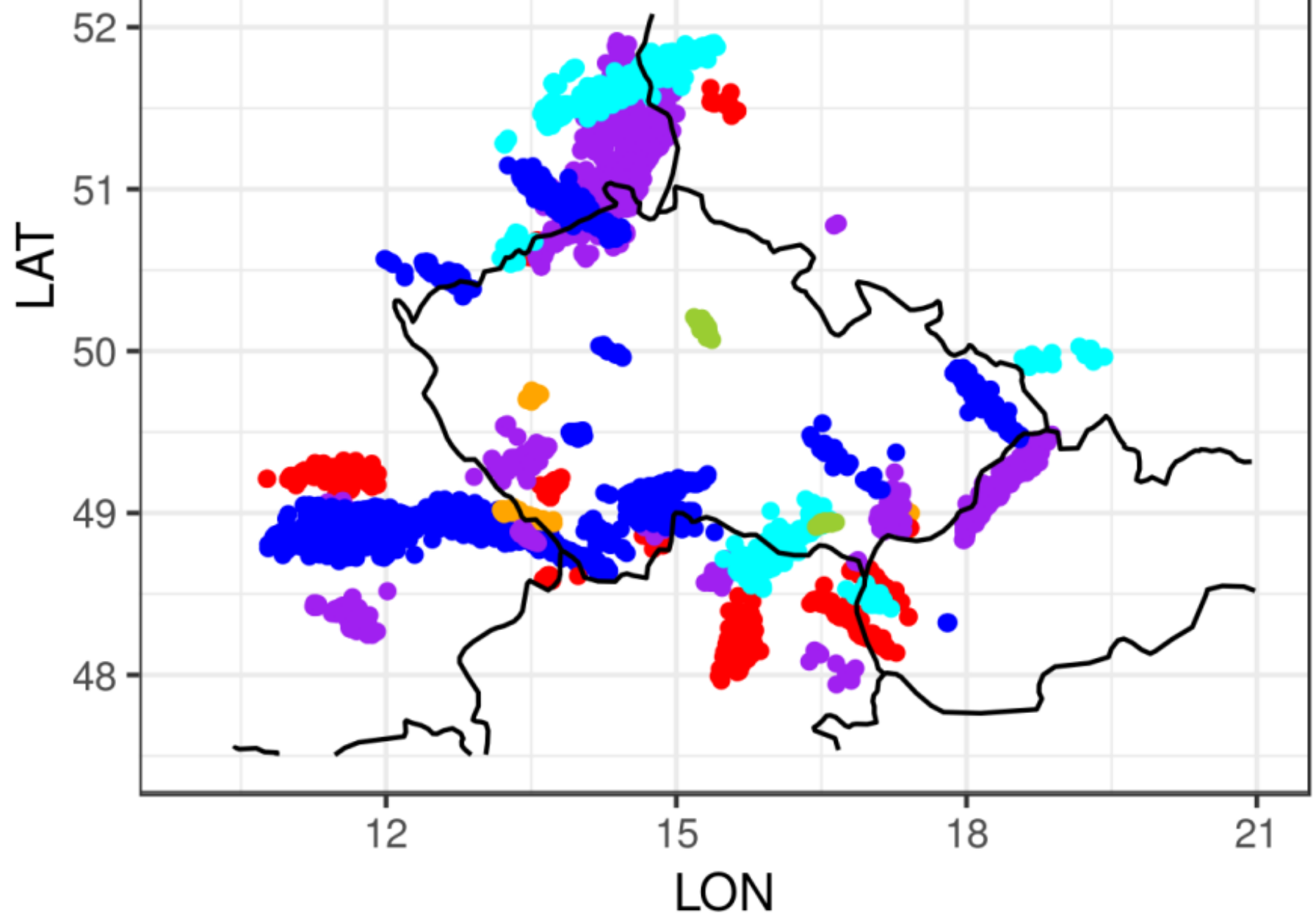






STORM DATABASE

Storm without ESWD report

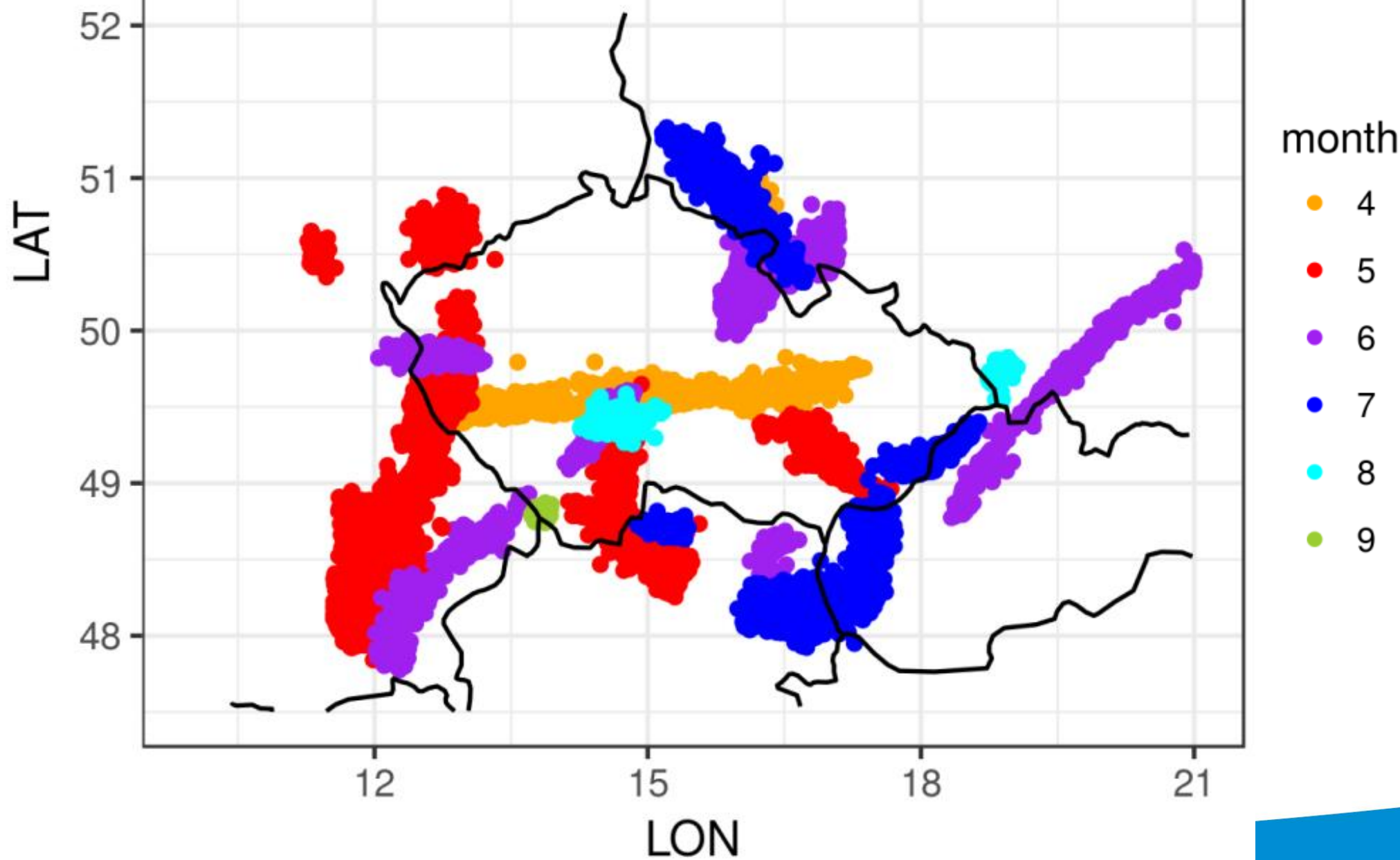


month

- 4
- 5
- 6
- 7
- 8
- 9



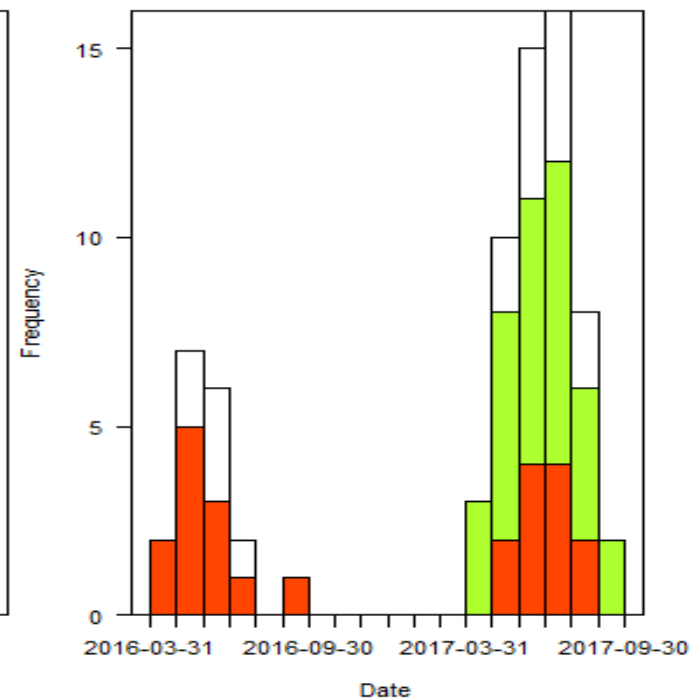
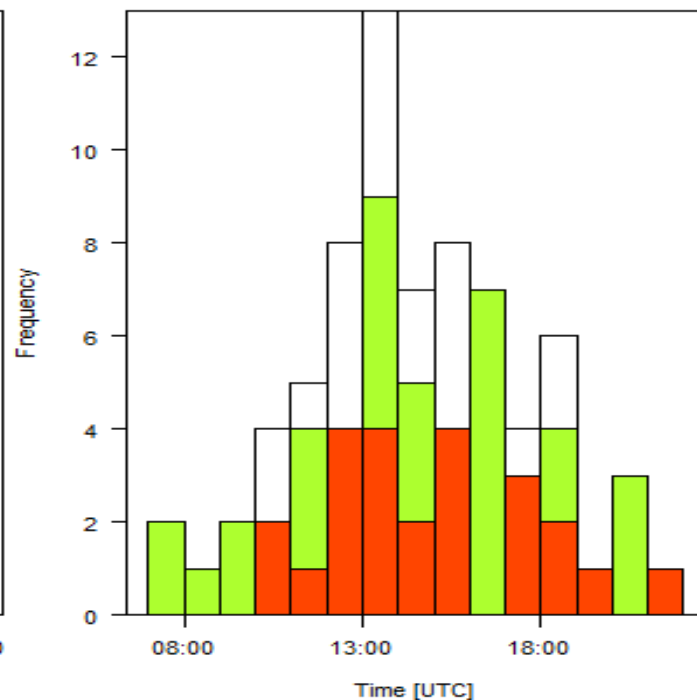
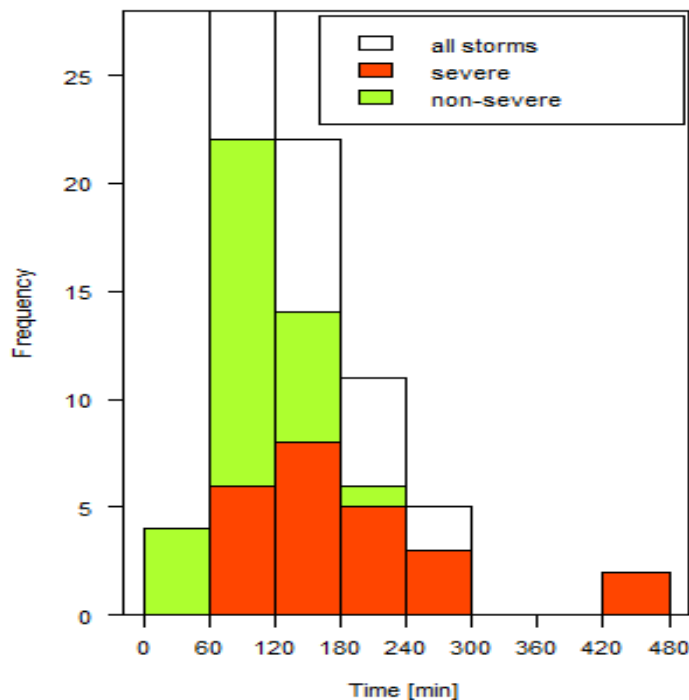
Storm with ESWD report



STORM DATABASE

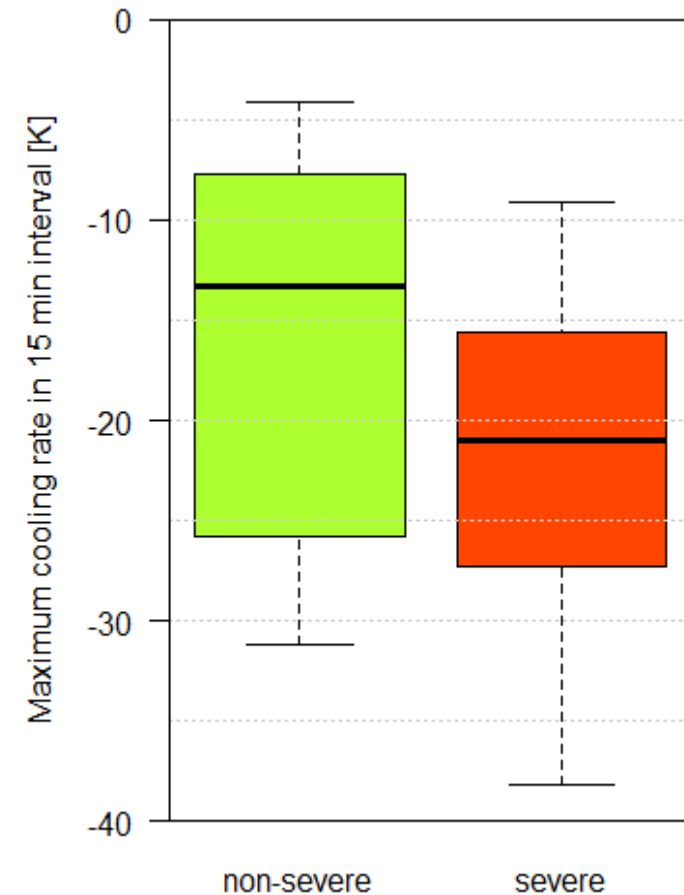
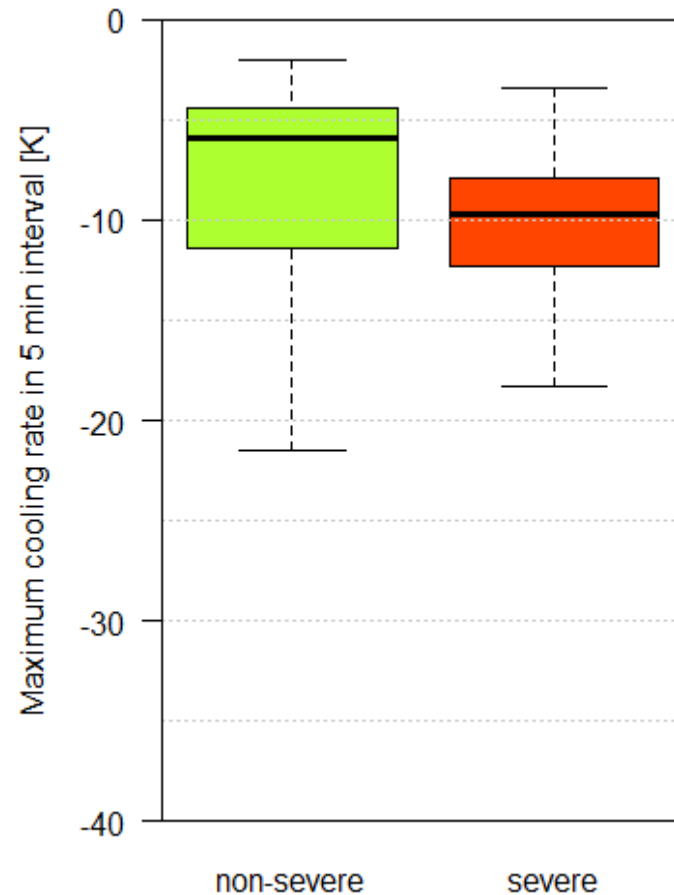
72 storm cases from April to September 2016 and 2017

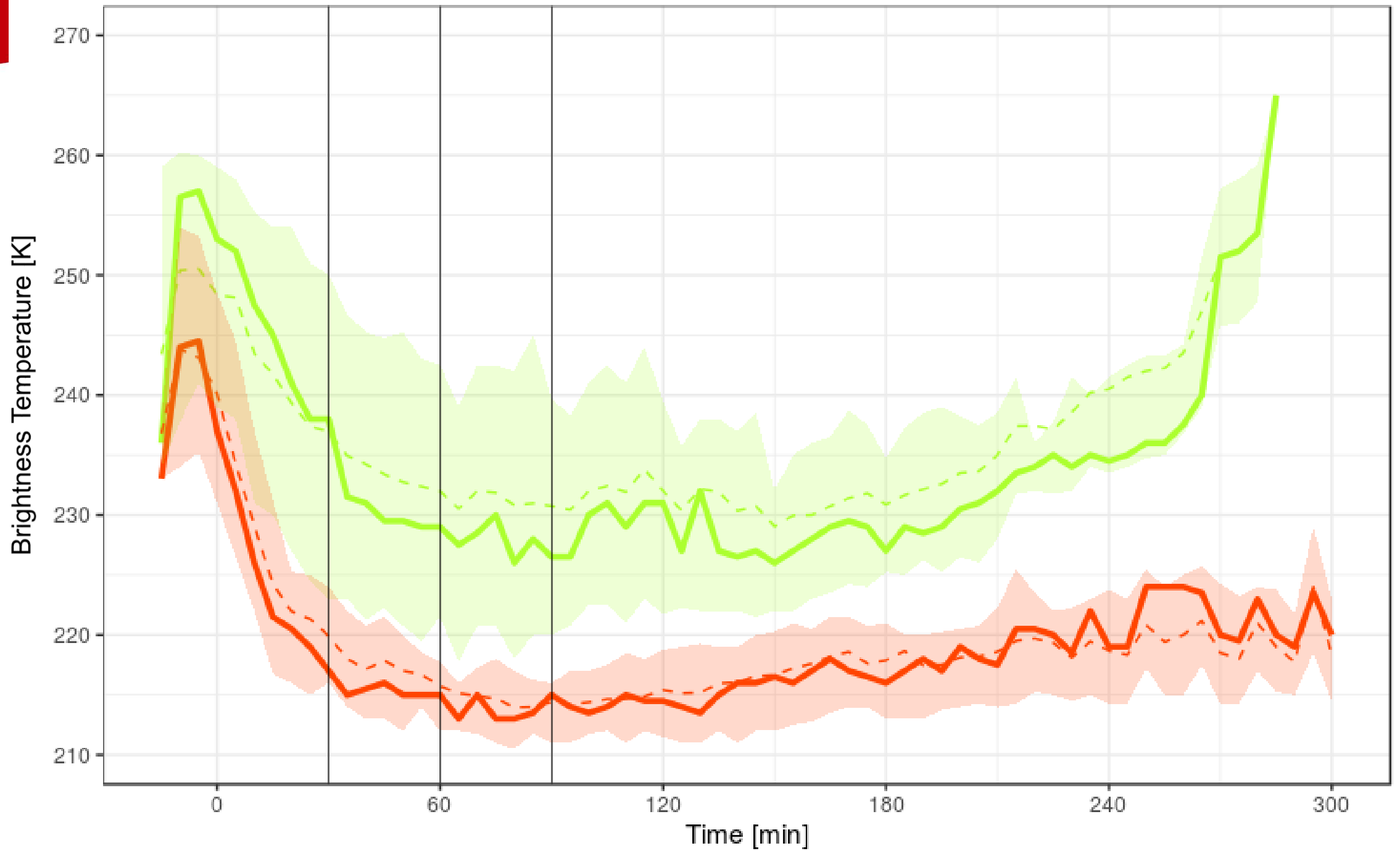
- 24 severe, 48 non-severe storms
- 19 supercells, 19 multicells, 34 single cells

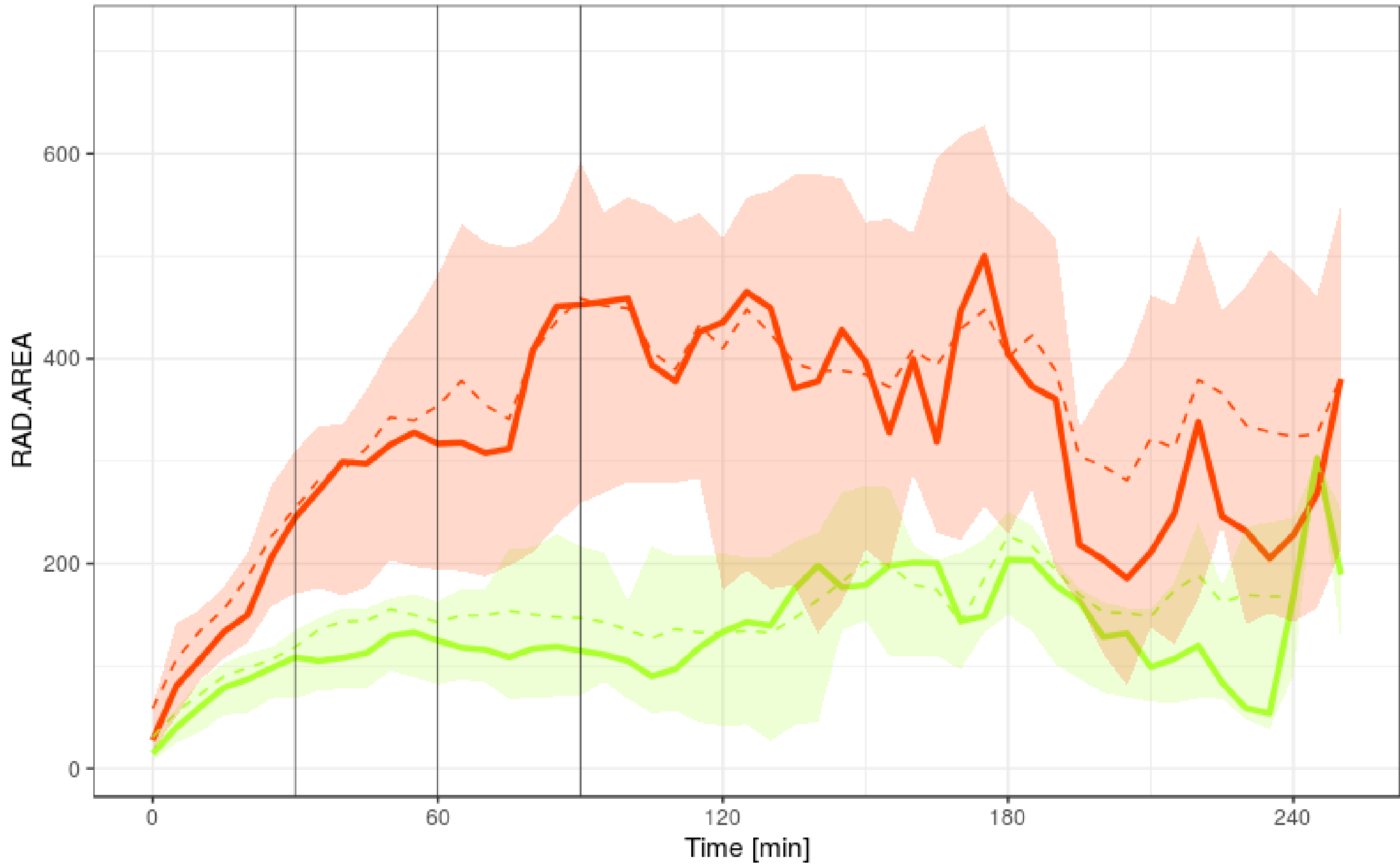


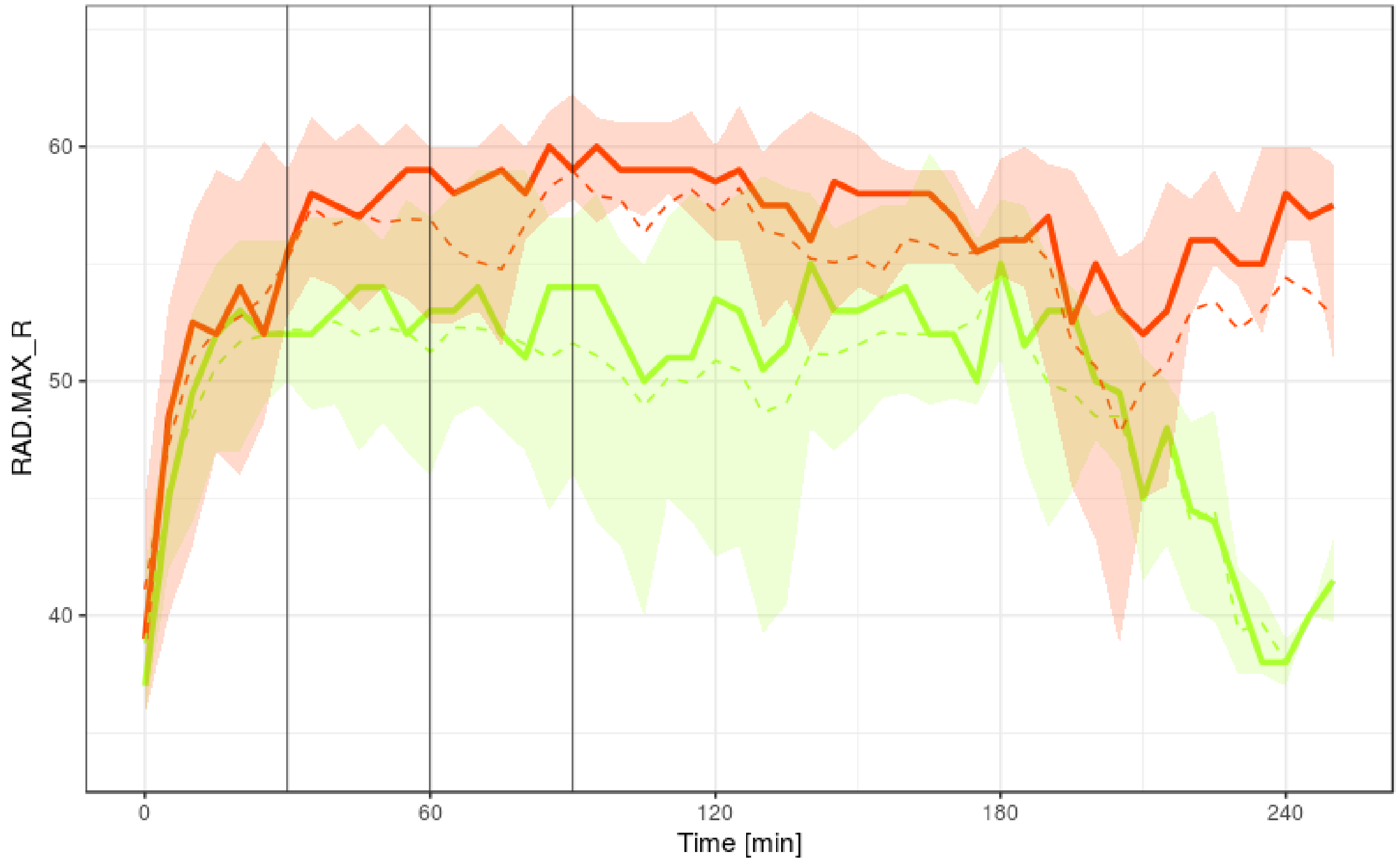
SATELLITES

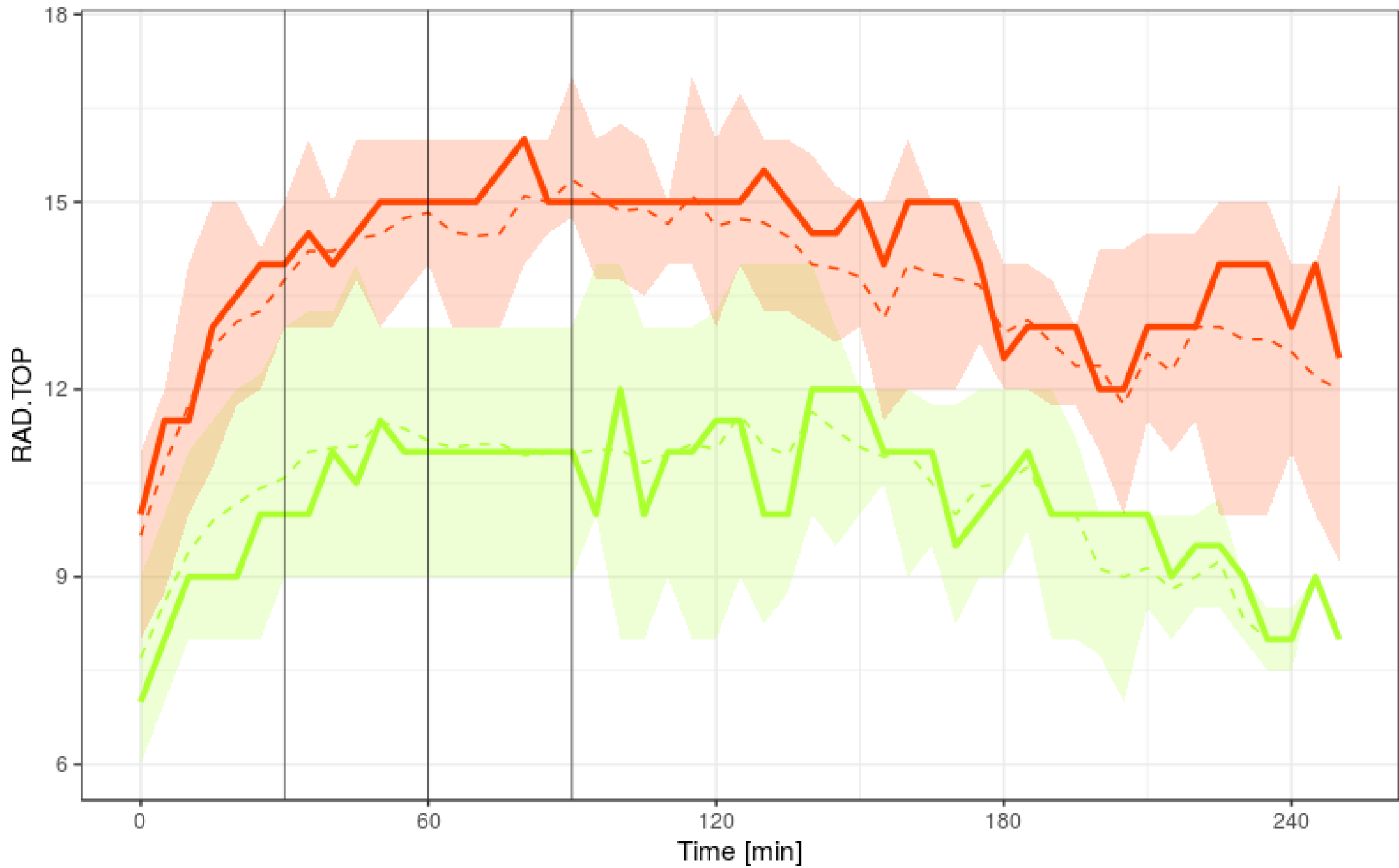
- minimum BT in IR 10.8
- cooling rate:
in 5, 15 and 30 min
- cloud-top features:
OTs, cold-U or cold ring
shapes, plume, small ice
particles

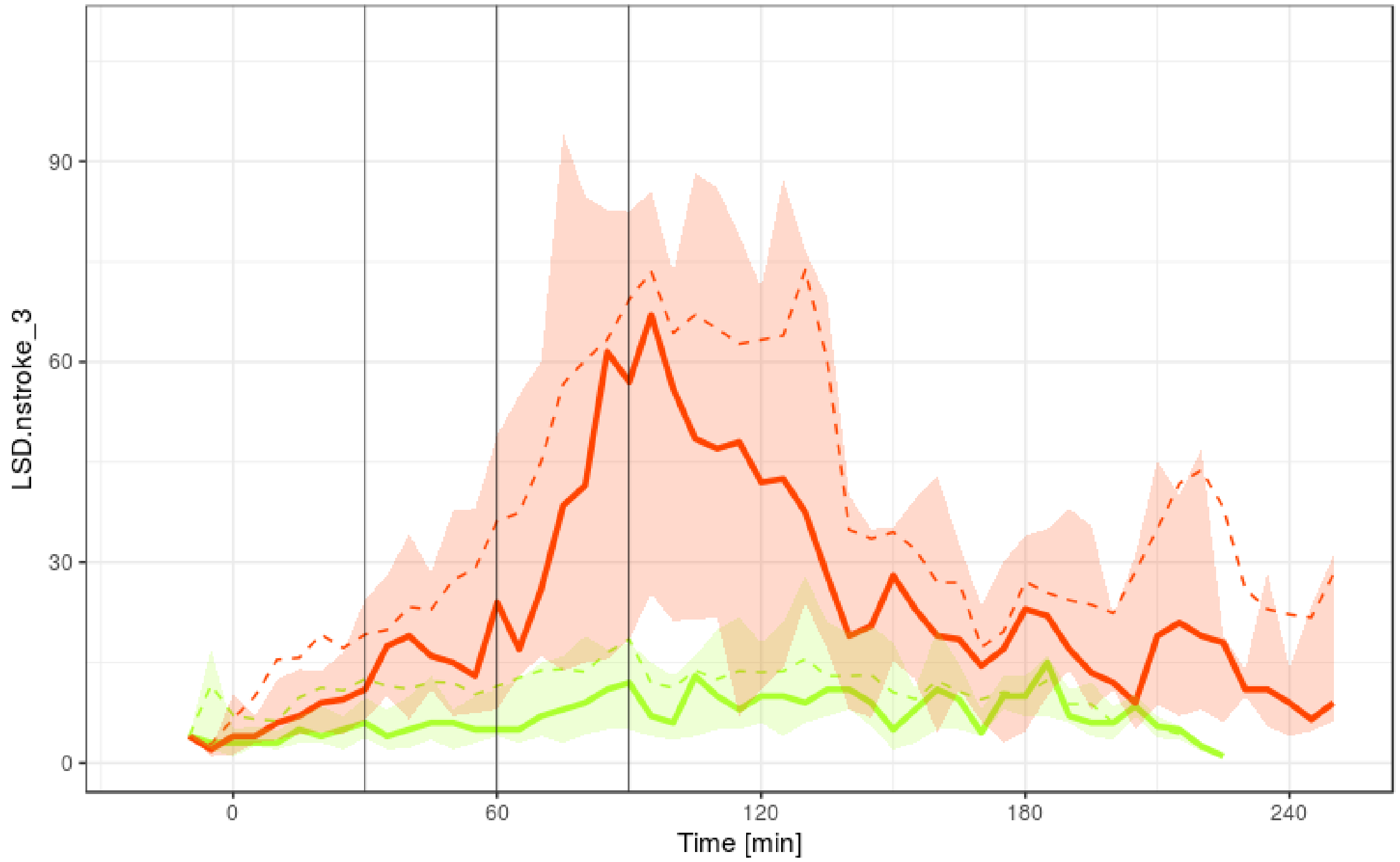


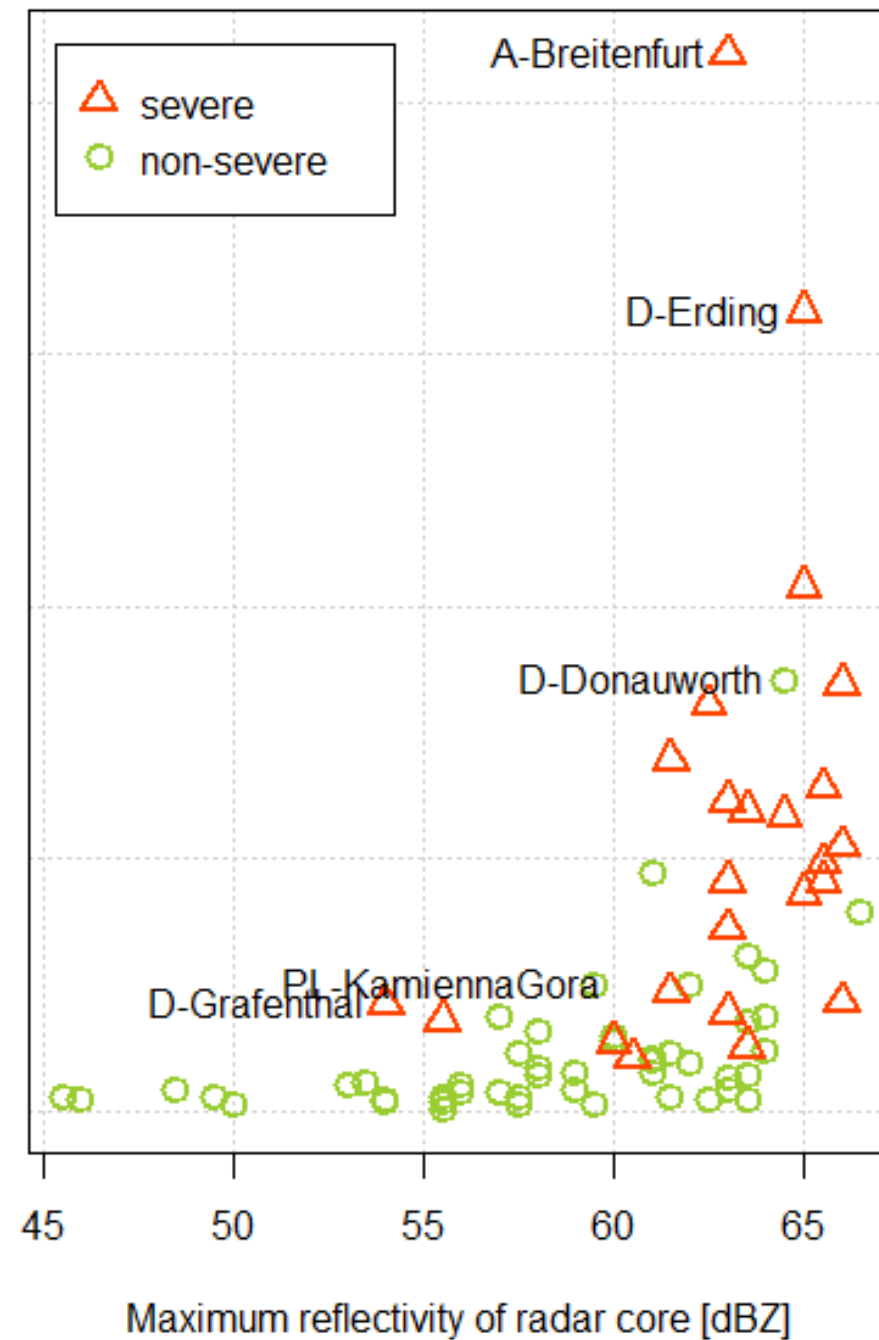
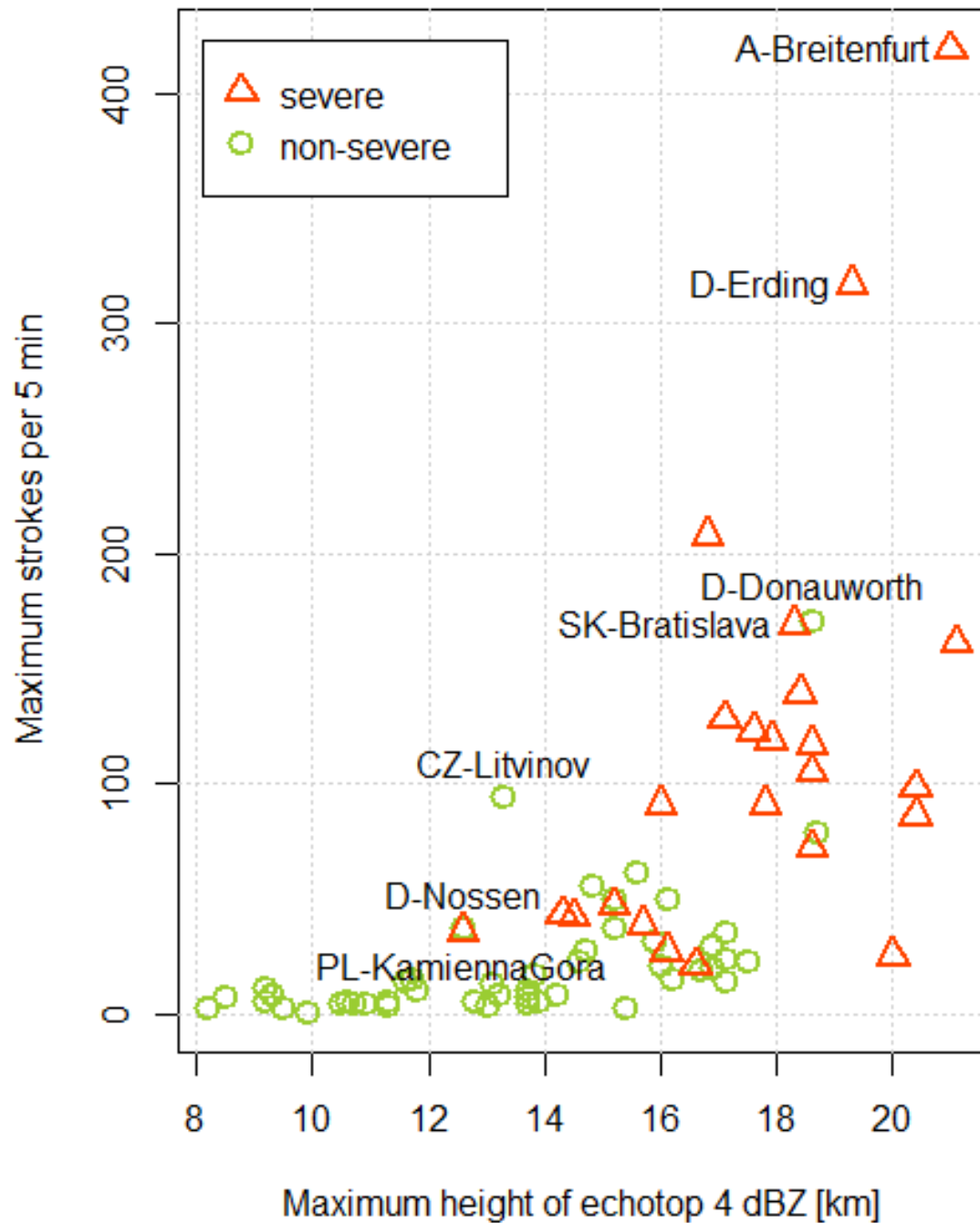




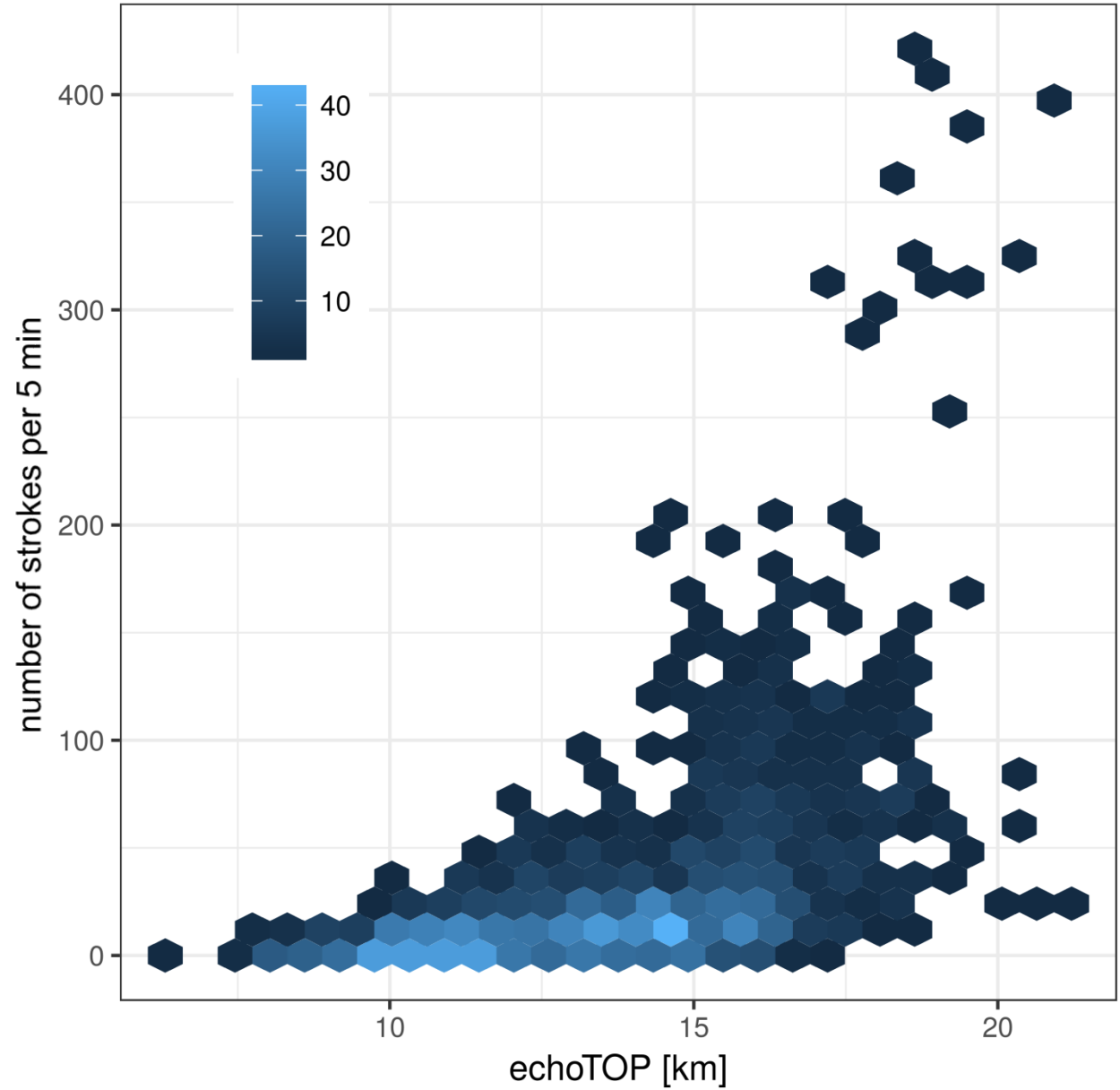
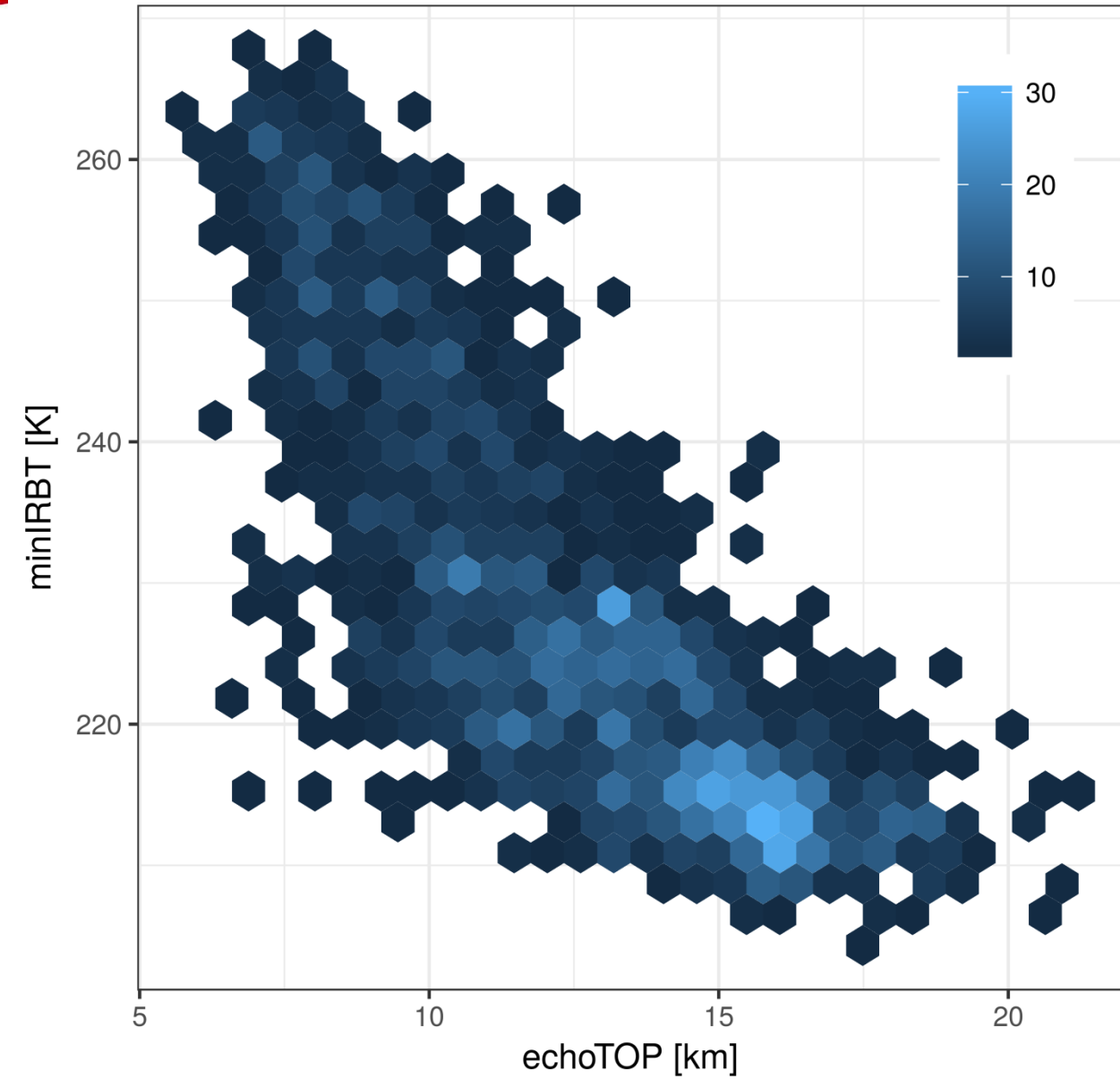








n = 2536





REGRESSION MODELS

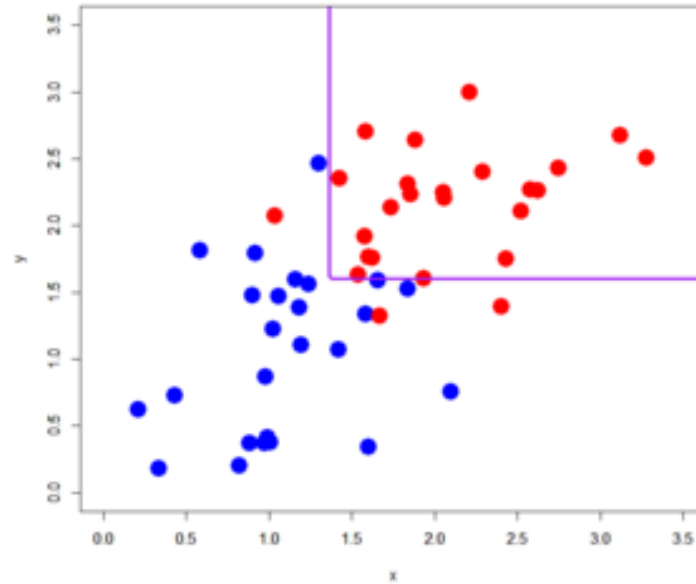
REGRESSION

- machine learning classifiers
- probability of the storm severity

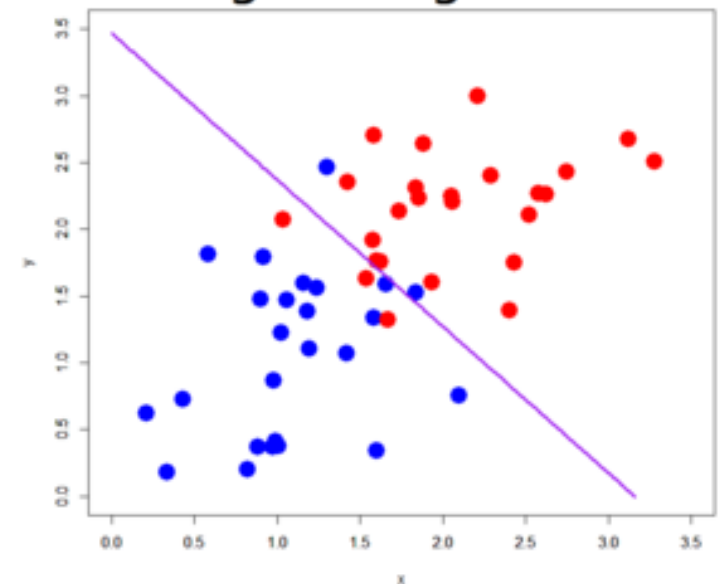
When?

- three time intervals
30, 60 and 90 minutes

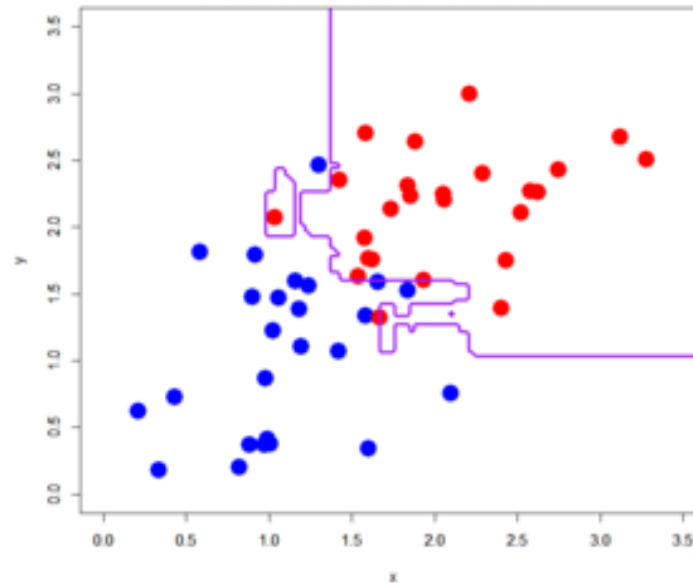
Decision Tree



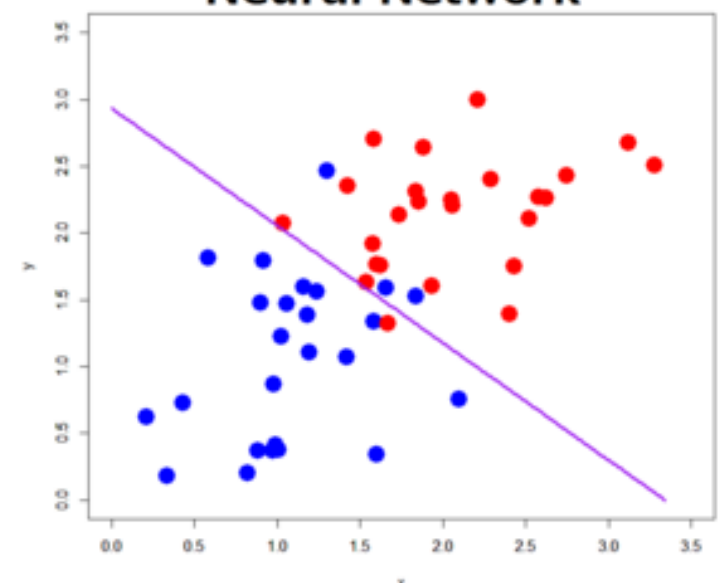
Logistic Regression



Random Forest



Neural Network



LOGISTIC REGRESSION MODEL

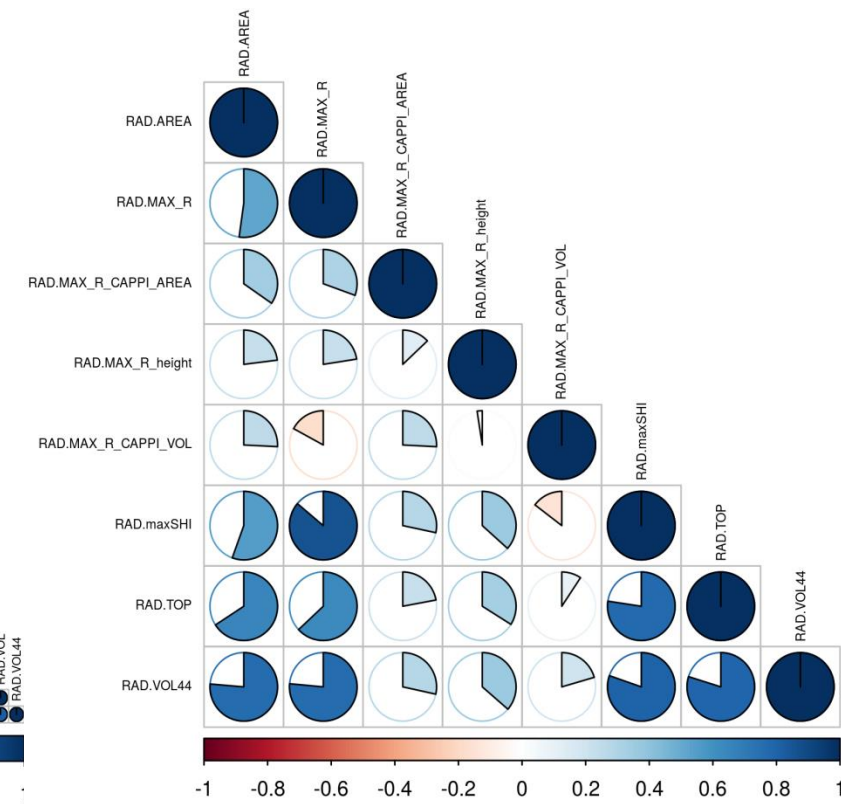
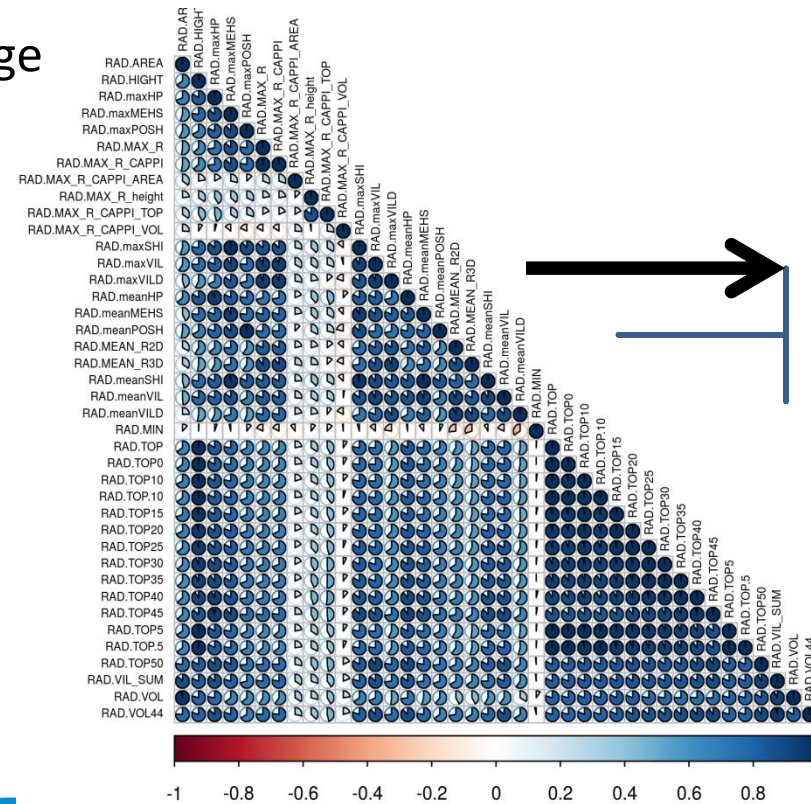
- explain relationships between one dependent dichotomous variable (0 or 1) and one or more independent variables →
 - odds of the storm being severe based on predictors from remote sensing measurements
- probability of the storm being severe
- conditions:
 - no high correlations among the predictors !!!
 - about 1 predictor per 10 cases to make model converge



VARIABLE SELECTION

- multicollinearity

- detected based on Variance Inflation Factor (VIF)
- pre-selection of predictors based on:
 - scientific knowledge
 - $VIF < 4$



VARIABLE SELECTION

- **sign OK**

- remove predictors, if its multivariable (MVA) sign is different from univariable (UVA) sign
- repeat, until all selected predictors have a correct sign


	UVA	MVA
RAD.AREA	0.0134	0.0138
RAD.MAX_R	0.0991	-0.0464
RAD.MAX_R_CAPPI_AREA	0.3579	-0.1183
RAD.MAX_R_height	0.0004	0.0002
RAD.MAX_R_CAPPI_VOL	0.0677	0.0199
RAD.maxSHI	0.0088	-0.0029
RAD.TOP	0.0004	0.0004
RAD.VOL44	0.0043	-0.0030



VARIABLE SELECTION

- **stepwise backward method**

- remove the most insignificant predictors ($p\text{-value} > 0.157$) and reestimate
- repeat until all predictors are significant



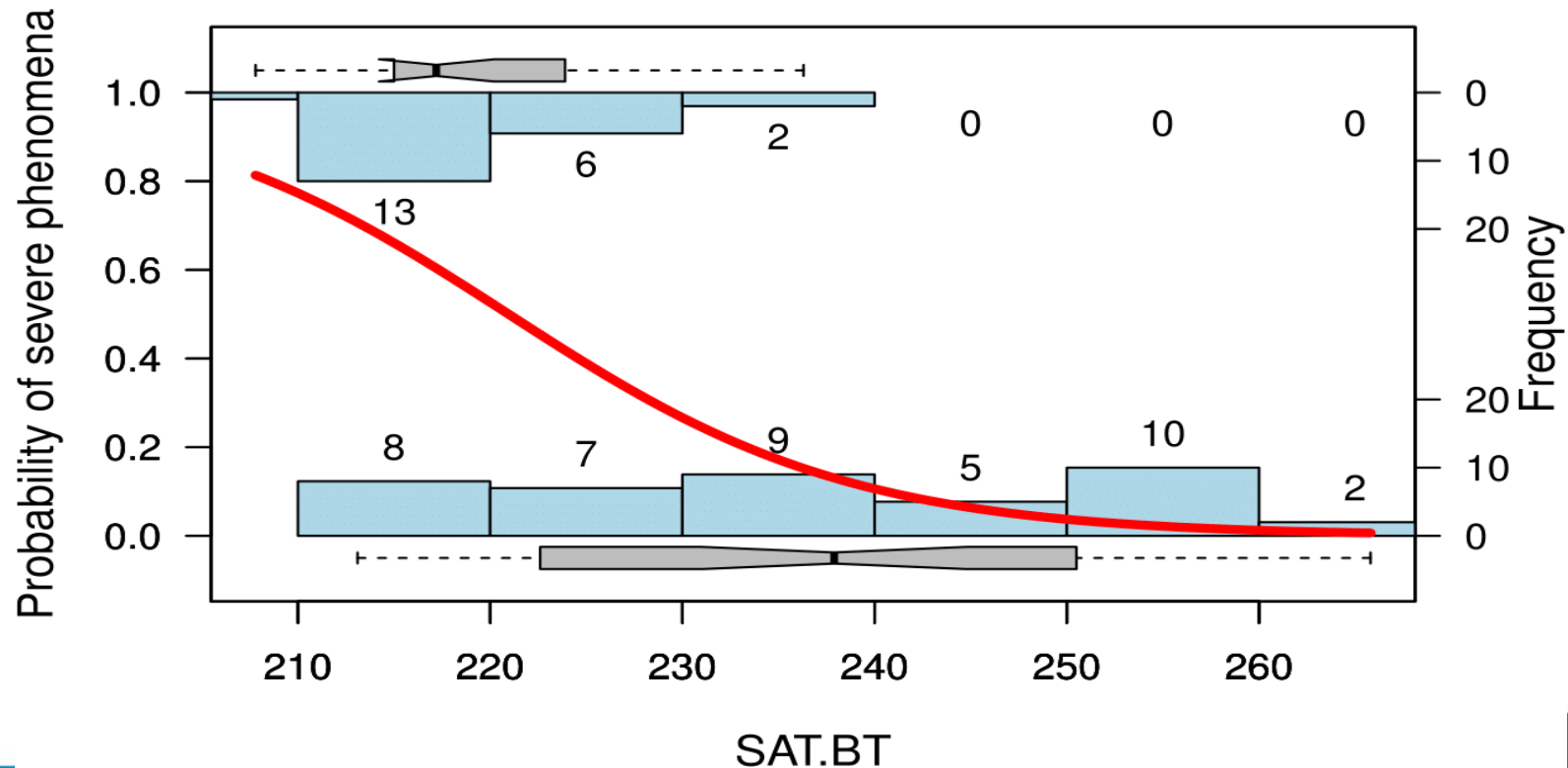
```
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -5.3353710  1.5954852  -3.344 0.000826 ***
## RAD.AREA      0.0090681  0.0038440   2.359 0.018324 *
## RAD.MAX_R_height 0.0001049  0.0002060   0.509 0.610690
## RAD.MAX_R_CAPPI_VOL 0.0321039  0.0498861   0.644 0.519871
## RAD.TOP       0.0001863  0.0001286   1.449 0.147423
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -4.9580578  1.5187031  -3.265  0.0011 **
## RAD.AREA     0.0096687  0.0039140   2.470  0.0135 *
## RAD.TOP      0.0001957  0.0001260   1.553  0.1203
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```



LOGISTIC REGRESSION MODEL

- individual model for:
 - remote sensing methods (RAD, LSD and SAT)
 - the first 30, 60 and 90 minutes of the storm lifecycle



RESULTS OF MODELS

	<i>Dependent variable:</i>		
	sev.phenomena		
	RAD30	RAD60	RAD90
	(1)	(2)	(3)
RAD.AREA	1.010 p = 0.014**	1.007 p = 0.001***	1.007 p = 0.0005***
RAD.TOP	1.000 p = 0.121		
RAD.maxHP		1.037 p = 0.010***	
RAD.MAX_R_height			1.001 p = 0.007***
Observations	72	72	72
Log Likelihood	-33.145	-29.612	-28.251
Akaike Inf. Crit.	72.291	65.225	62.501

Note:

*p<0.1; **p<0.05; ***p<0.01



RESULTS OF MODELS

	<i>Dependent variable:</i>		
	LSD30 (1)	sev.phenomena LSD60 (2)	LSD90 (3)
LSD.sum_curr_3		1.004 p = 0.090*	
LSD.LJ	2.105 p = 0.007***	1.463 p = 0.019**	
LSD.nstroke_3			1.035 p = 0.001***
LSD.sum_curr_0_neg			0.996 p = 0.043**
Observations	72	72	72
Log Likelihood	-38.568	-36.036	-30.952
Akaike Inf. Crit.	81.136	78.073	67.905

Note:

*p<0.1; **p<0.05; ***p<0.01



RESULTS OF MODELS

	<i>Dependent variable:</i>		
	sev.phenomena		
	SAT30	SAT60	SAT90
	(1)	(2)	(3)
SAT.BT	0.894	0.820	0.773
	p = 0.0005***	p = 0.002***	p = 0.002***
Observations	63	63	63
Log Likelihood	-29.744	-27.657	-25.068
Akaike Inf. Crit.	63.487	59.314	54.136

Note:

*p<0.1; **p<0.05; ***p<0.01



ELASTIC NET



- a penalized regression technique
 - number of features ($k=81/69$) exceeds the number of observations ($n=63$)
 - presence of highly correlated predictors
- two penalty terms (α, λ) in the maximum likelihood formula:
 - objective selection of relevant predictors
 - shrink regression coefficients to reduce the model over-fitting
- *cv.glmnet* in R tests the performance of each λ by using the cross validation
 - small size of dataset \rightarrow Leave-One-Out Cross Validation method (LOOCV)

RESULTS OF MODELS

Table 1: Elastic model coefficients for 30 min

Predictors	Coefficient	Odd_ratio
(Intercept)		
RAD.AREA		
LSD.sum_curr_0_pos		
LSD.LJ		
SAT.BT		

Table 2: Elastic model coefficients for 60 min

Predictors	Coefficient	Odd_ratio
(Intercept)		
RAD.AREA		
LSD.LJ		
SAT.BT		

Predictors	Coefficient	Odd_ratio
(Intercept)	5.537	253.937
RAD.AREA	0.002	1.002
RAD.MAX_R_height	0	1
LSD.nstroke_3	0.001	1.001
LSD.sum_curr_0_neg	-0.001	0.999
SAT.BT	-0.035	0.966

Table 3: Elastic model coefficients for 90 min

EVALUATION OF MODELS

- Recall, Precision and F1 Score by LOOCV

<https://towardsdatascience.com/accuracy-precision-recall-or-f1-331fb37c5cb9>

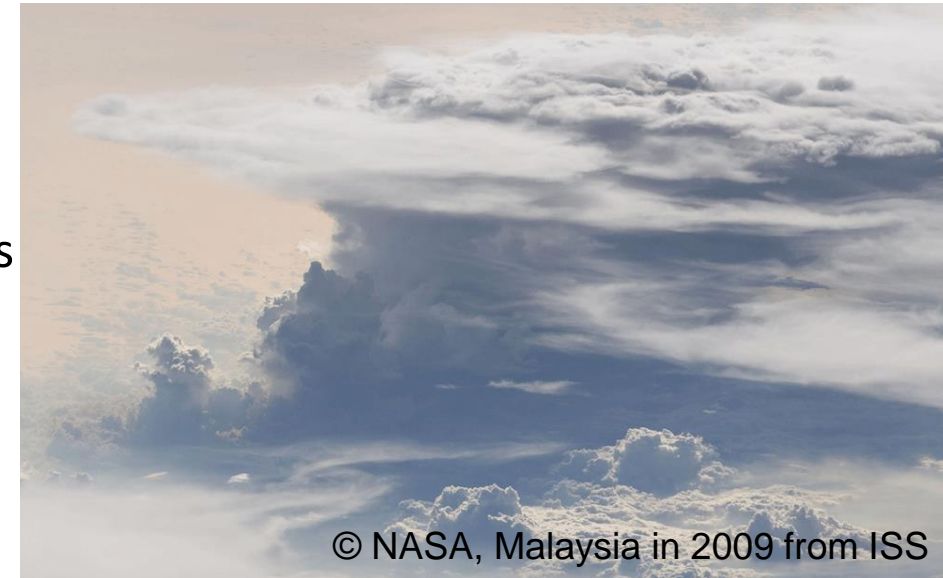
	Recall [%]	Precision [%]	F1 Score
LRM-RAD	64	78	0.70
LRM-LSD	36	89	0.52
LRM-SAT	68	62	0.65
ENet	75	94	0.83

Performance of the models for the first 30 minutes of the storm lifecycle



SUMMARY

- most of studied remote sensing parameters are dependent on the storm severity → useful information for nowcasting
- regression models were employed
 - high precision of the models (over 70 %)
 - similar predictors for logistic regressions and elastic nets
- predictors of the storm severity:
 - SAT.BT, RAD.AREA, LSD.LJ
- future steps:
 - improve the LJ algorithm, find relations for new data sources
 - adaptations for the operation in CHMI
 - probability of the storm severity, thresholds



ACKNOWLEDGEMENT

Support, motivation, inspiration:

Katrin Wapler (DWD)

André Simon (OMSZ)

Justin Sieglaff (CIMSS UW)

John Cintineo (CIMSS UW)

David Rýva (CHMI)

Marian Rybář (MATSTAT)

Data source:

CHMI, EUMETSAT, Siemens AG

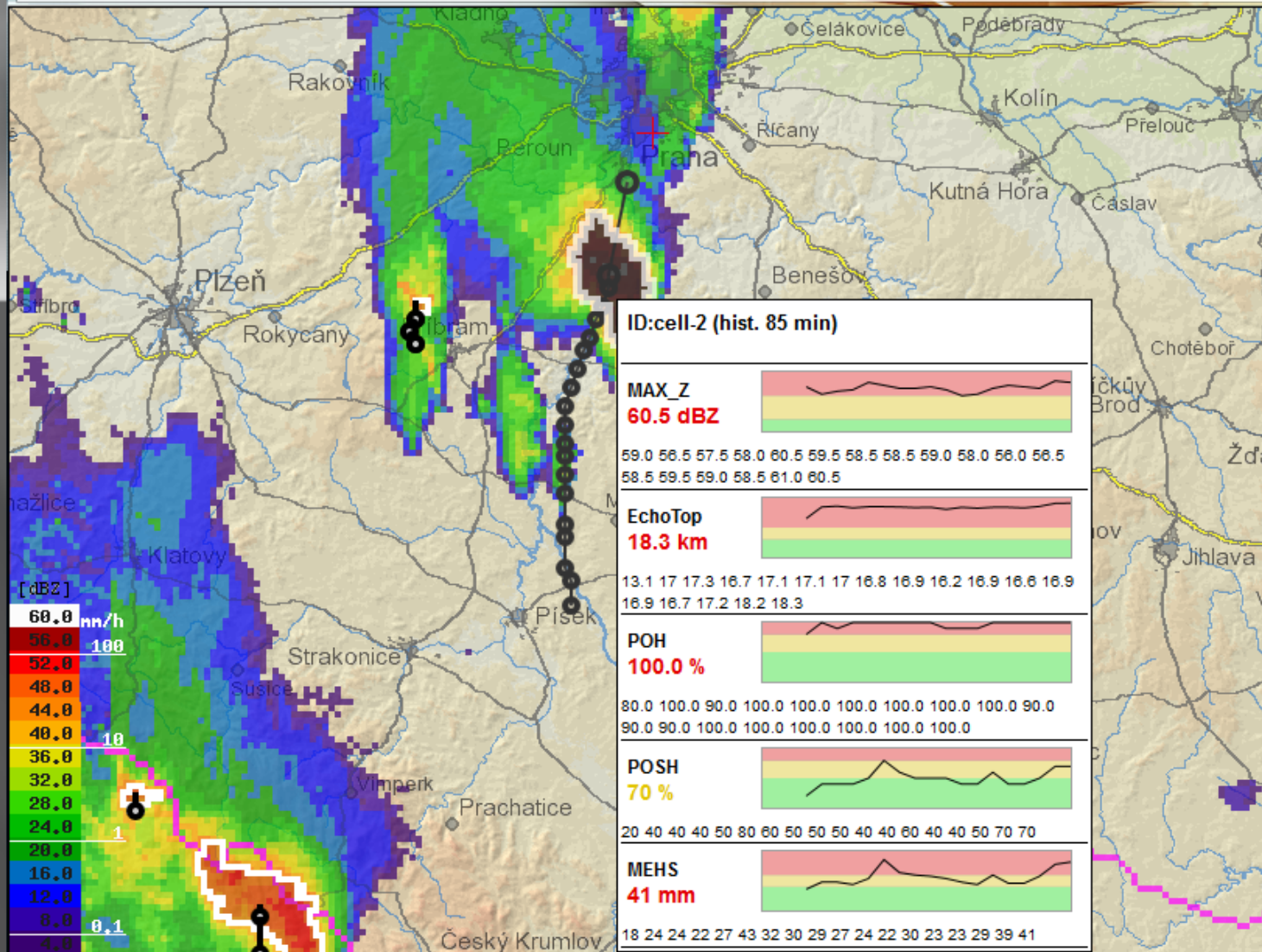
ESSL and all active spotters



REFERENCES

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ID: cell-2 (hist. 85 min)

MAX_Z
60.5 dBZ

59.0 56.5 57.5 58.0 60.5 59.5 58.5 58.5 59.0 58.0 56.0 56.5
58.5 59.5 59.0 58.5 61.0 60.5

EchoTop
18.3 km

13.1 17 17.3 16.7 17.1 17.1 17 16.8 16.9 16.2 16.9 16.6 16.9
16.9 16.7 17.2 18.2 18.3

POH
100.0 %

80.0 100.0 90.0 100.0 100.0 100.0 100.0 100.0 100.0 90.0
90.0 90.0 100.0 100.0 100.0 100.0 100.0 100.0

POSH
70 %

20 40 40 40 50 80 60 50 50 40 40 60 40 40 50 70 70

MEHS
41 mm

18 24 24 22 27 43 32 30 29 27 24 22 30 23 23 29 39 41

JSMeteoView

15.08.2010 18:30 UT

Extrapol. fct: +00 min

Every 6th 3rd

- 15.08.2010 18:30 XX
 - 15.08.2010 18:25 XX
 - 15.08.2010 18:20 XX
 - 15.08.2010 18:15 XX
 - 15.08.2010 18:10 XX
 - 15.08.2010 18:05 XX
 - 15.08.2010 18:00 XX
 - 15.08.2010 17:55 XX
 - 15.08.2010 17:50 XX
 - 15.08.2010 17:45 XX
- LOAD (24 / 24)

ORO barva

POD0 mesta 50

POD1 toky 60

POD2 silnice 60

SAT IR BT 100

RAD Z MAX 70

BLESK CELDN 100

CELLTRACK

CELLTRACK - 15.08.2010 18:30

100

Vyber POHYB

Vyber ALADIN

Vyber SYNOP

Vyber HZS 100

NAD0 mesta 20

NAD1 silnice 30

NAD2 nic 100

STUPNICE 100

ANIM: 1 s/img LAST: +2 s AUTO UPDATE Do not update UPDATE NOW

KRÍŽ red LON 14.447 LAT 50.008 Vyber předdefinovanou polohu

ADV.INFO cursor position is [289,244.25] = [14.268,49.771] 29km to the S-SW(206deg.) from NAVIG cross ZOOM COLOR black

ZOOM 1x 2x 4x POSUN

#

