

Estimation of hail frequency and its trends under climate change using ML





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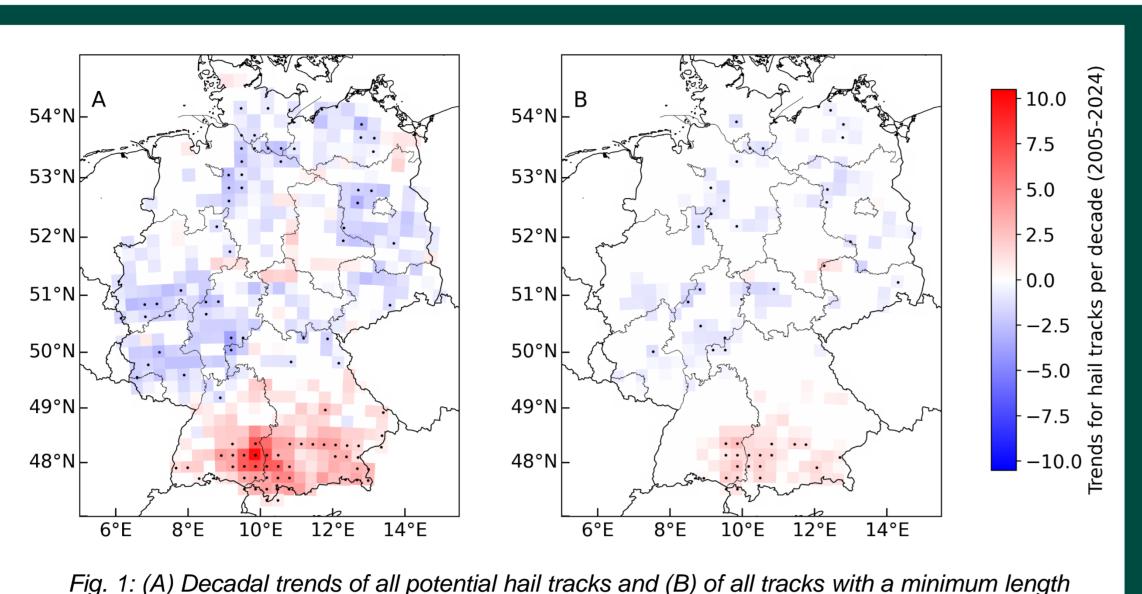
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Motivation

Why do radaridentified hail events show negative trends over large parts of **Germany during the** last 20 years?

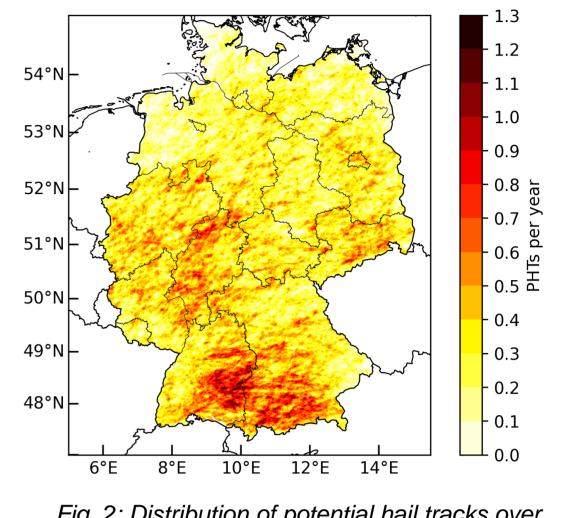
How can a proxybased ML hail model be designed that reproduces observed hail trends?



of 50 km for 20 years of hail data (2005 – 2024, April – September), Mohr et al. (2025).

Data basis

- Potential hail tracks identified from 3D radar network of German Weather Service (DWD; April – September, 2005 – 2024)
- Cloud-to-ground lightning data (EUCLID; summer half-years 2005 – 2023)
- Convective parameters of ERA5 (0.25° x 0.25° grid, hourly resolution) calculated by thunder package (available for 1990 – 2024; Taszarek et al., 2024)



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Fig. 2: Distribution of potential hail tracks over Germany on a 1 km grid (2005 – 2024, April – September).

Proxy based hail model

Table 1: Convective parameters of ERA5 used as predictors in logistic regression and XGBoost hail day model

MU_CAPE	most-unstable convective available potential energy	
MU5_E_LI	most-unstable entrainment lifted index at 500 hPa	
MU_LI	most-unstable lifted index	
RH_500850	mean rel. humidity between 500 & 850 hPa	
MU_MIXR	mixing ratio at height of MU parcel	
MU_cold_cloud	layer depth between -10°C and equilibrium level	
LR_16km	temperature lapse rate between 1 and 6 km AGL	
PRCP_WATER_eff	effective precipitable water acounting for RH	
BS_06km	bulk wind shear between surface and 6 km	
SRH_03km_RM	storm-relative helicity between surface and 3 km	

Better performance of **XGBoost model** compared to logistic regression:

Table 2: Optimum splitting threshold and quality measures for both logistic regression and XGBoost model

	Logistic Regression	XGBoost
Best threshold for splitting	0.73	0.79
Area under precision-recall curve	0.399	0.440
Matthew's correlation coefficient	0.441	0.460

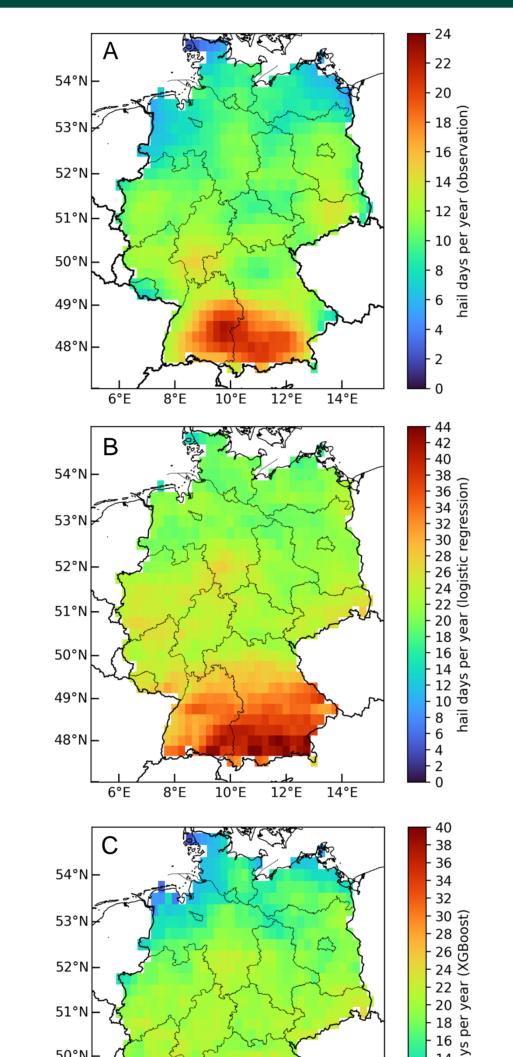


Fig. 3: Hail days per year (A) observed, (B) modeled with logistic regression, and (C) XGBoost on 0.25° grid resolution.

10°E

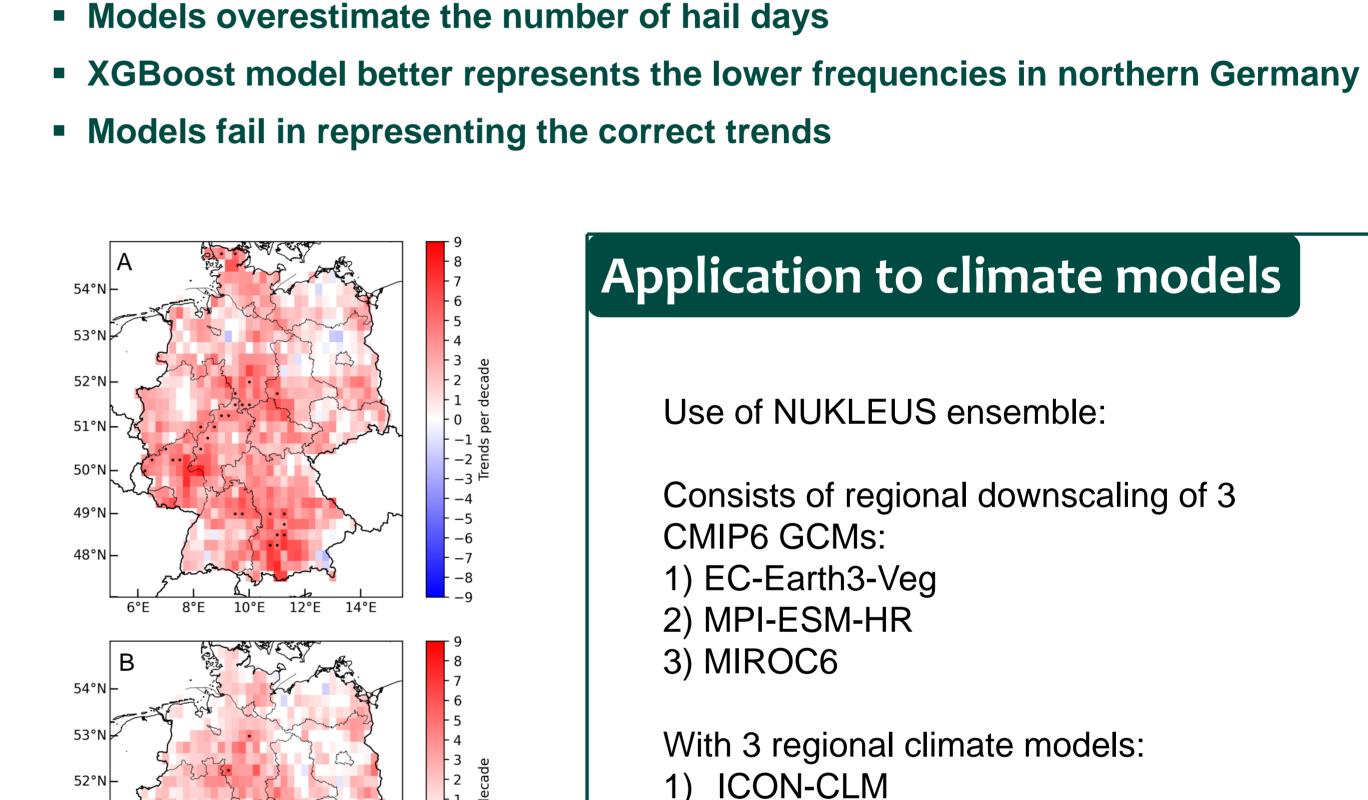


Fig. 4: Trends per decade of hail days per year modeled with (A) **logistic regression** and (B) XGBoost on 0.25° grid resolution.

Application to climate models

Use of NUKLEUS ensemble:

Consists of regional downscaling of 3

CMIP6 GCMs: 1) EC-Earth3-Veg

2) MPI-ESM-HR

3) MIROC6

Models represent overall structure of hail distribution with hotspots in hail

frequency in southern Germany and a strong gradient from north to south

With 3 regional climate models:

1) ICON-CLM

2) COSMO-CLM

3) REMO

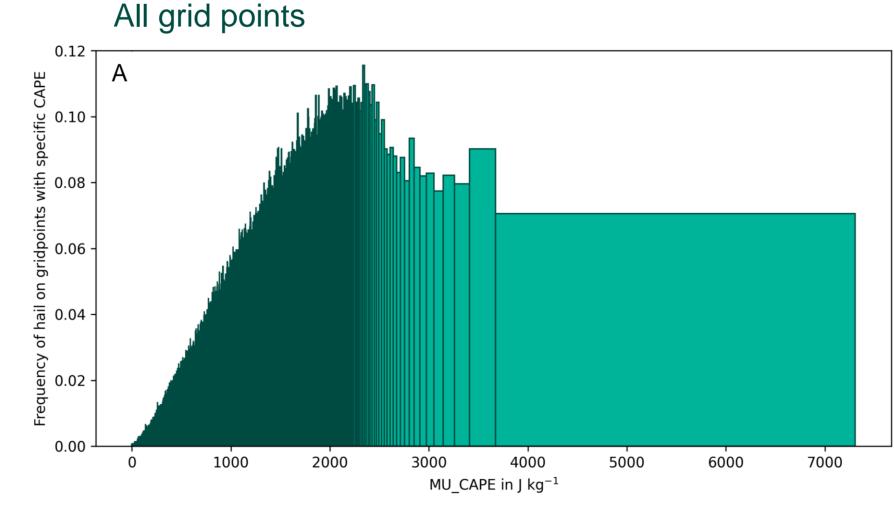
Data available for 3 and 12.5 km grid spacing for historical period and +2K and +3K global warming level

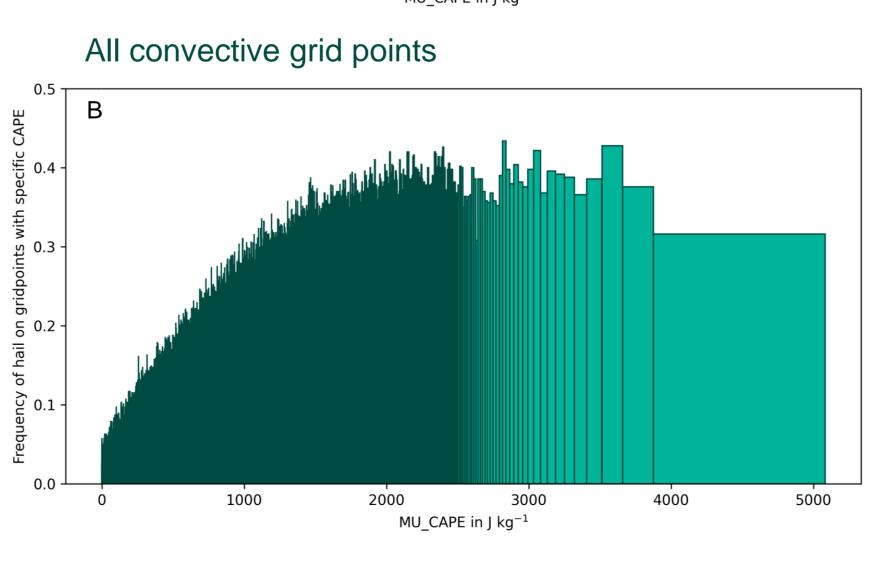
Higher instability – higher hail occurrence?

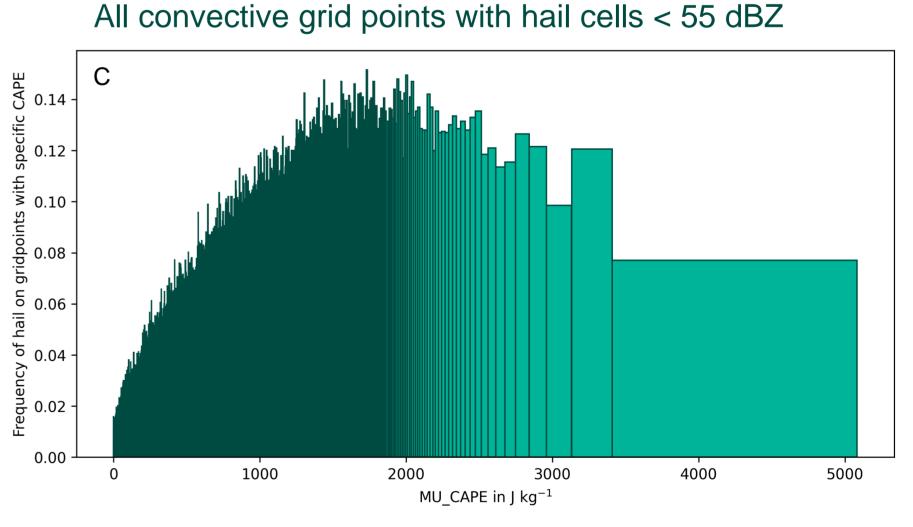
Fig. 5: Hail frequency for grid points with specific most-unstable CAPE over Germany according to ERA5 parameters (thundeR; Taszarek, 2024) and potential hail tracks. A grid point is counted as hail gridpoint if there is a cell detection within a radius of 50km from 30min before to 30min after the time step. Each bar of a plot contains the same number of grid points (for A: 5000, for B, C

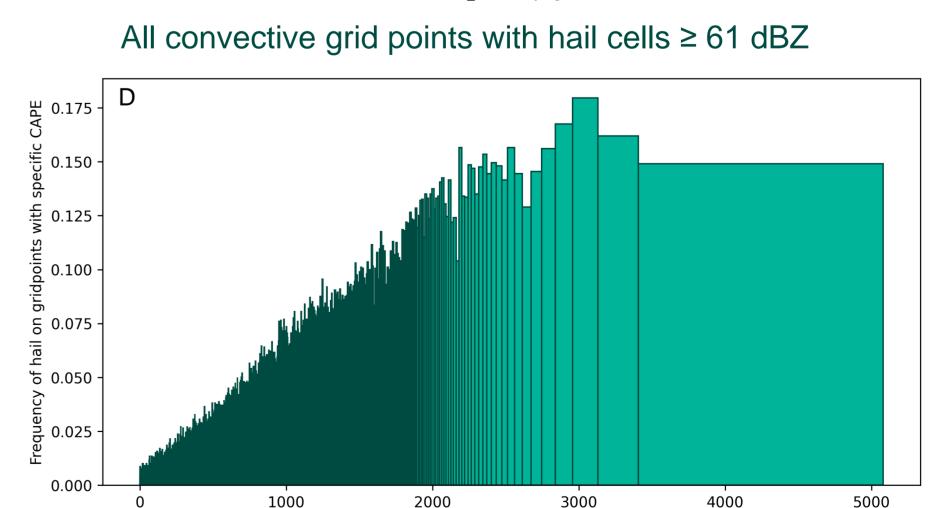
(A) All grid points are considered (hourly resolution; April – September 2005 –

- (B) Only grid points with convection are considered (hourly resolution; April September 2005 – 2023); convective grid points are defined by 5 cloud-toground lightning strokes within a radius of 50km from 30min before to 30min after the time step
- (C) Same as (B), but identified hail cells have only low reflectivities in the precipitation scan (< 55 dBZ)
- (D) Same as (B), but identified hail cells have high reflectivities in the precipitation scan (≥ 61 dBZ)
- No further increase in hail potential for increasing CAPE above ~ 2000 J/kg
- For lower reflectivity: maximum hail potential is reached at lower CAPE values and higher CAPE reduces hail potential clearly
- For higher reflectivity: maximum hail potential is reached at higher CAPE values









MU_CAPE in J kg⁻¹

Literature

- Mohr, S., M. Tonn, M. Augenstein, C. Sperka, G. Kavil Kambrath, M. Kunz (2025): A 20-year spatio-temporal analysis of 3D radar-based hail tracks in Germany: Trends and regional differences. Handed in in Frontiers in Environmental Science.
- Taszarek, M., B. Czernecki, P. Szuster (2024): ThundeR A rawinsonde package for processing convective parameters and visualizing atmospheric profiles.