

# #TEAM- DISENTANGLEMENT



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# 01 Exploring The Data

What can we expect from the data

# 02 Research Question 1

Is it possible to disentangle volcanic and solar forcing by region selection?

# 03 Research Question 2

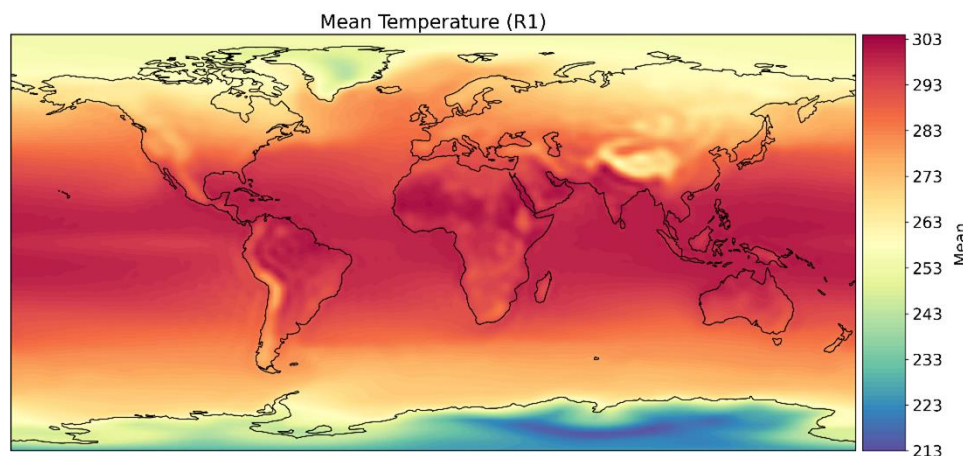
Is it possible to boost the prediction of volcanic and solar forcing by relying solely on the identified regions?



# Agenda

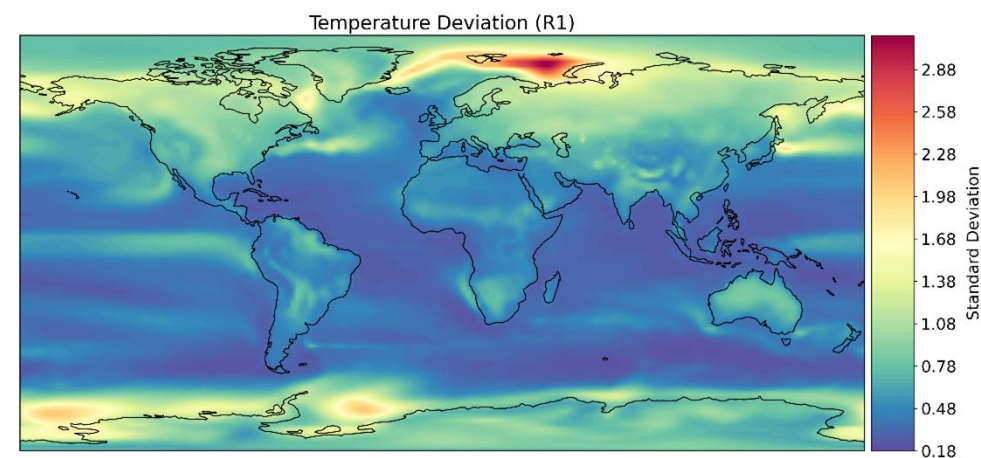


# Temperature variation



- Standard deviation from simulation results R1 over the whole simulation time
- Highest temperature variations around the north pole

- Mean temperature from simulation results R1 over the whole simulation time
- Region-dependent temperature



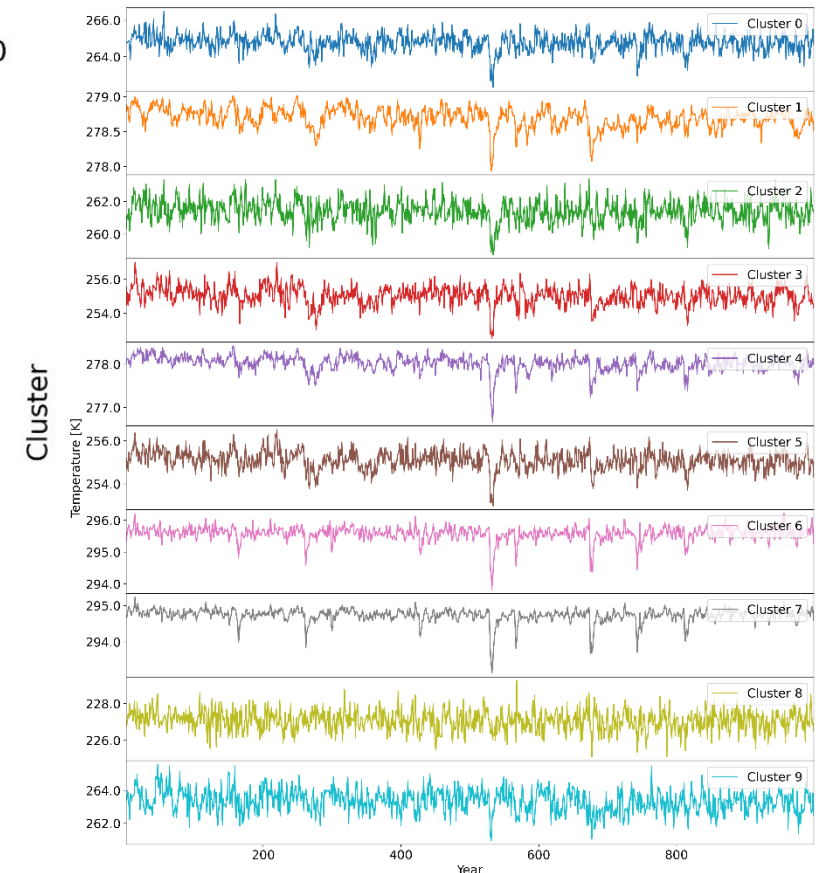
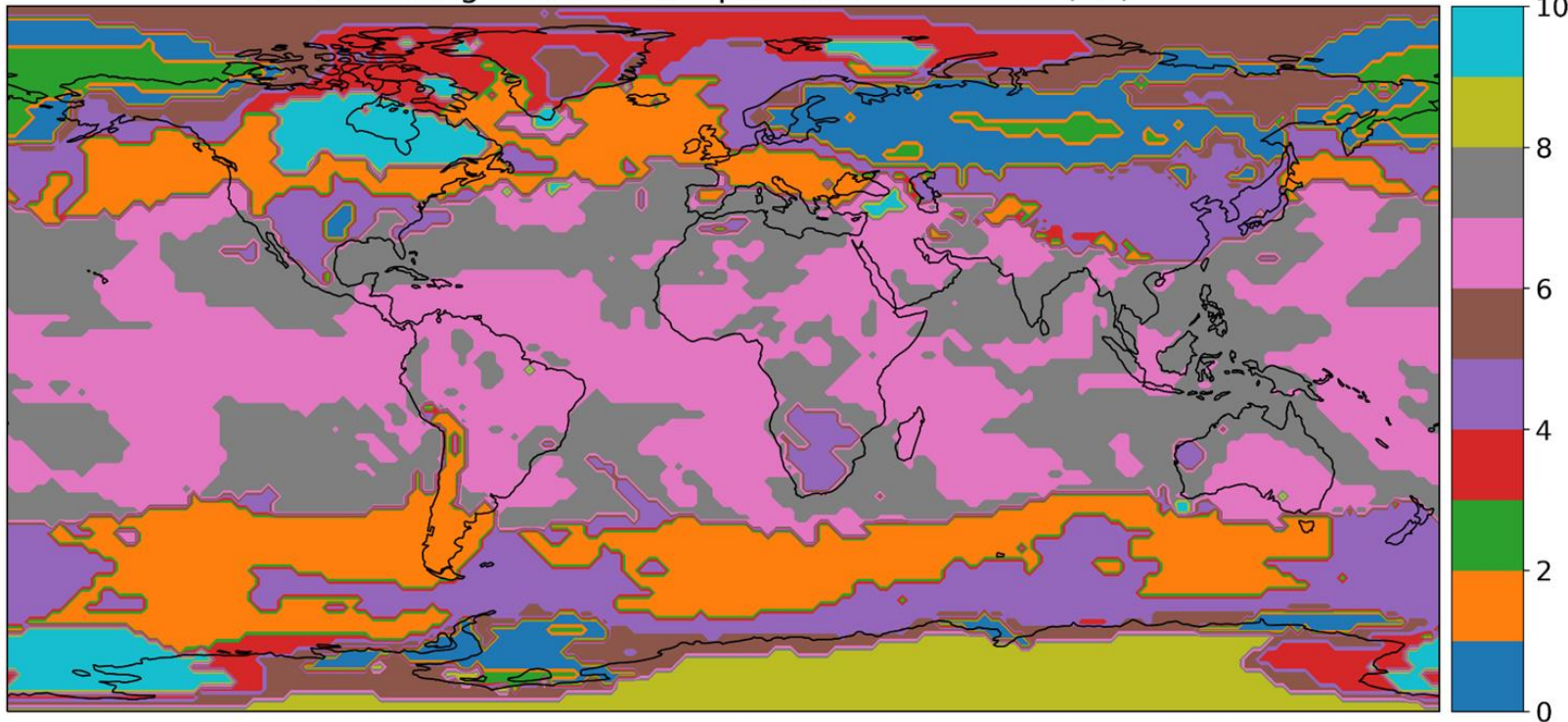
# RQ1: Can we disentangle volcanic and solar forcing by region selection?

Hypothesis: We can disentangle volcanic and solar forcing through region selection

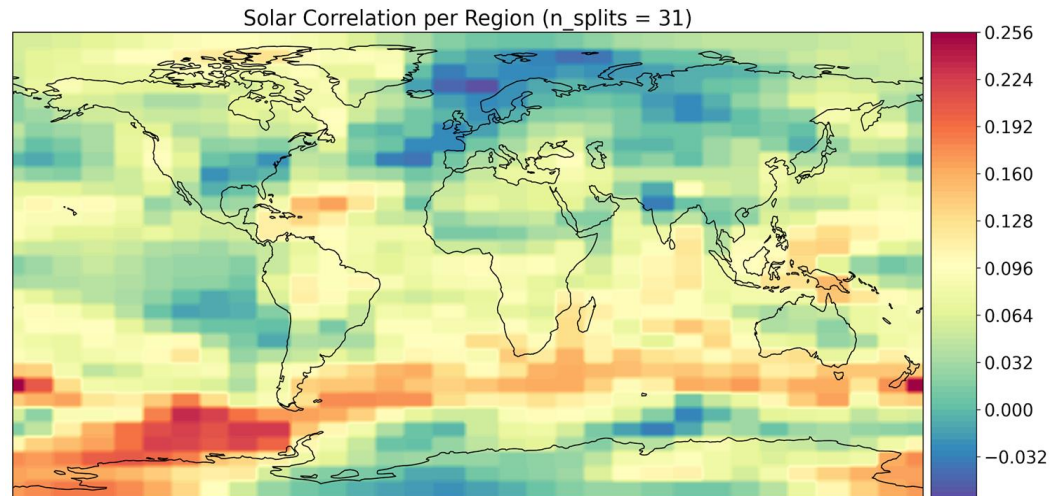
# RQ1.1: Disentanglement based on temperature time series data

Selection of regions via k-means clustering based on the temperature differences  $T(\text{yr}) - T(\text{yr}-1)$  time series

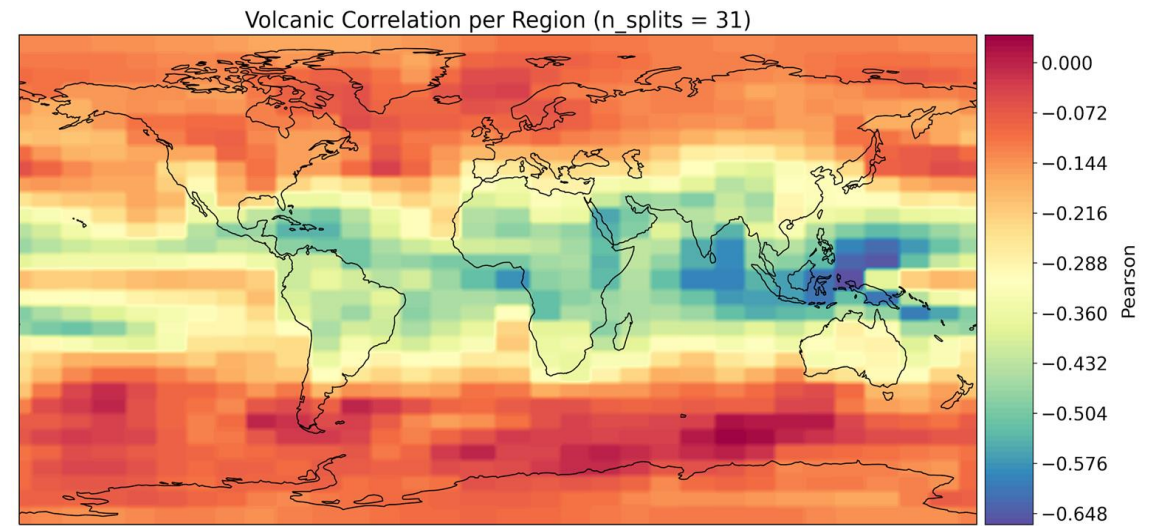
Clustering Based on Temperature Differences (R1)



# RQ 1.2: Disentanglement including prior knowledge of volcanic/solar activity



Pearson correlation coefficient to disentangle volcanic and solar activity.

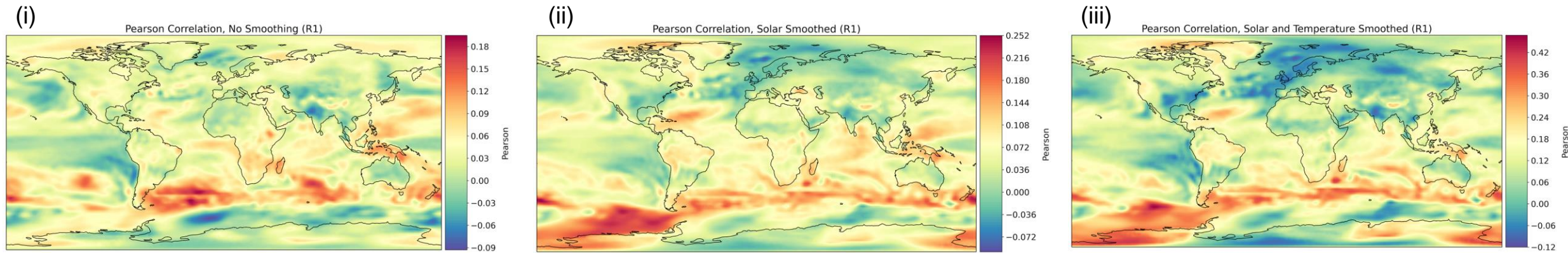




# RQ 1.2: Disentanglement including prior knowledge of volcanic/solar activity

Impact of a filter for disentangling solar activity. Measured by pearson correlation coefficient between

- i) Solar force and average temperature
- ii) filtered solar force using a butterworth filter of 8th order and average temperature
- iii) filtered solar force and average temperature using a butterworth filter of 8th order



The filter method changes the regional areas with high correlation.

# RQ2: Selecting the appropriate regions, can we detect volcanic activities / regress solar activities?

Hypothesis: we can detect volcanic activities as outliers in the temperature time series data of a selected region

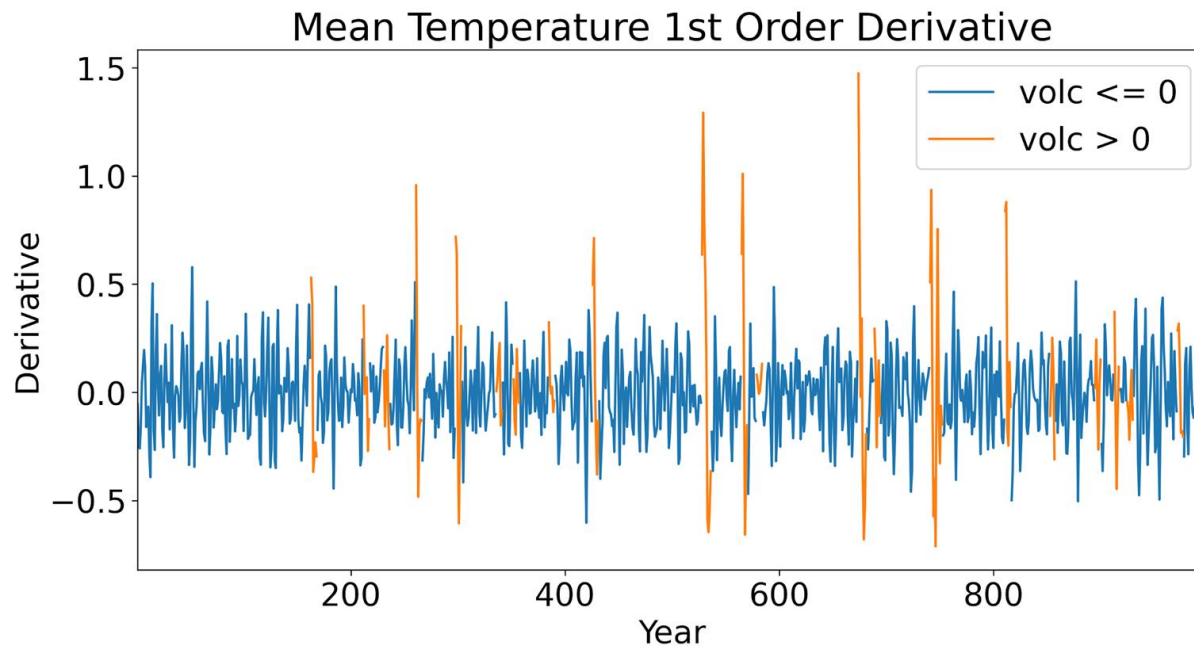


# RQ 2.1: Detecting volcanic activities as outliers

Approach:

- Selecting clusters 6 & 7
- Generating the first derivative of the temperature time series
- Outlier detection via Isolation Forest

Results:



		Predicted	
		out	in
True	out	26	112
	in	4	857

# RQ 2.2: Can we expect a correlation based on Pearson?

Experiment to check, whether Pearson correlation coefficient is actually meaningful:

	GT & temperature data	GT & 1000 simulated temperature time series (based on an autoregressive model AR(1))
GT = TSI	0.42	0.40
GT = AOD	0.68	0.56

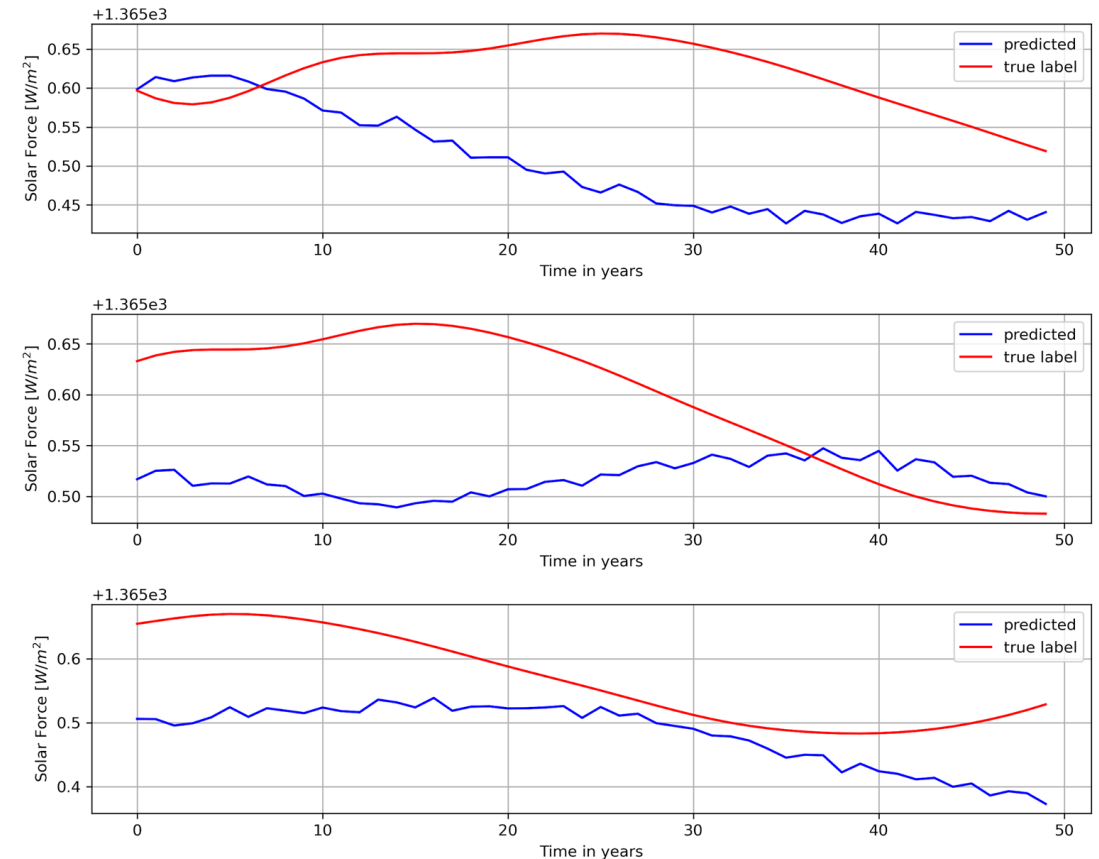
# RQ 2.3: Predicting solar force using CNN

## Approach:

- Selection progress of input data based on the disentangled high correlated & filtered solar forces between the average temperature
- 3-layered CNN designed for automatic feature engineering and a two layered dense network architecture for solving the regression task of predicting 50 years of solar force

## Result:

- Low correlation between input and output data results in low correlation between predicted and true curves





# Summary

- Regions to disentangle the impact of volcanic and solar forcing could be identified
- Phases of volcanic activity were detected using Isolation Forest on the first derivative of temperature data in a selected region
- Disentangled regions with highest correlation coefficient between solar forces and temperature were not enough to lead to acceptable results

