

# Exploratory analysis of Recurrent deforestation warnings in the Brazilian Amazon

Alber Sanchez  
alber.ipia@inpe.br



September 22, 2023

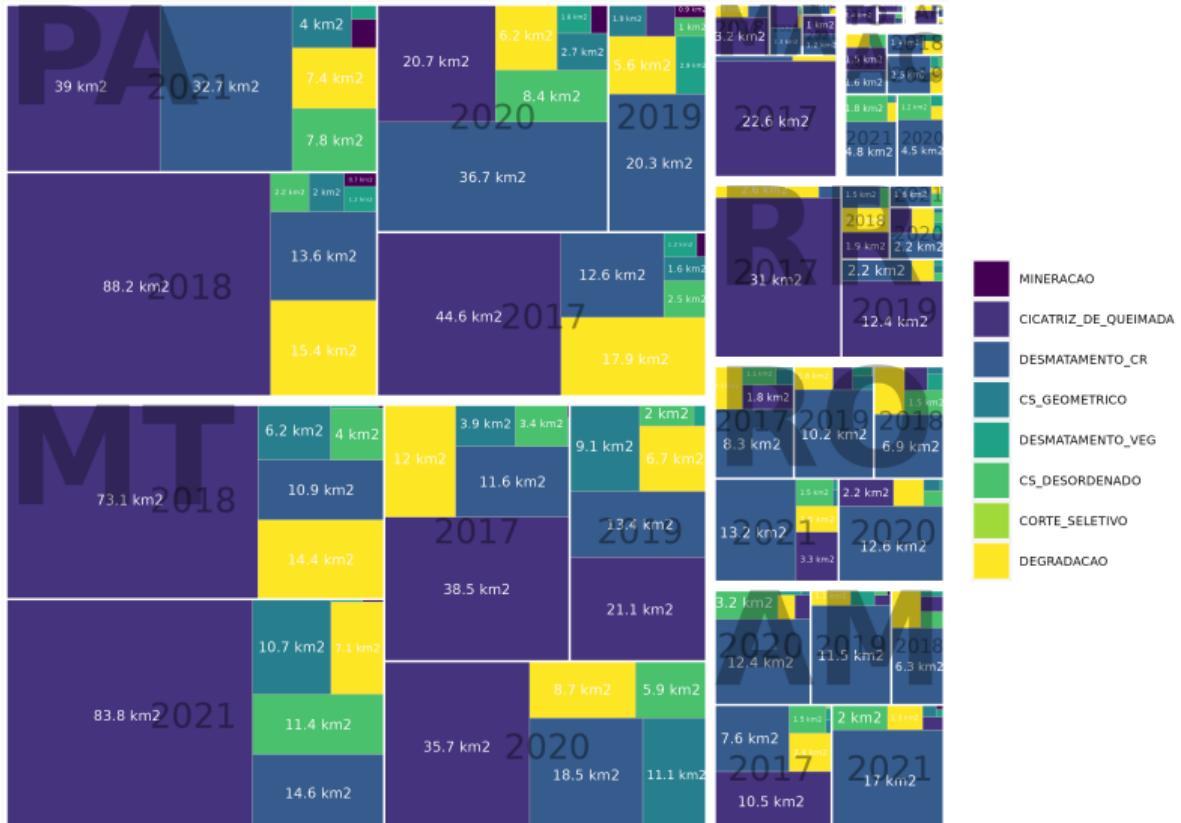
# Introduction

- ▶ Deforestation by successive degradation remains a challenging question in the scientific literature.
- ▶ We think an answer to this question lies down in DETER data.
- ▶ This answer could play an important role, for example, in the brazilian estimation of greenhouse gases.
- ▶ We used DETER data from 2016 to 2021 of the Amazon Biome in Brazil.

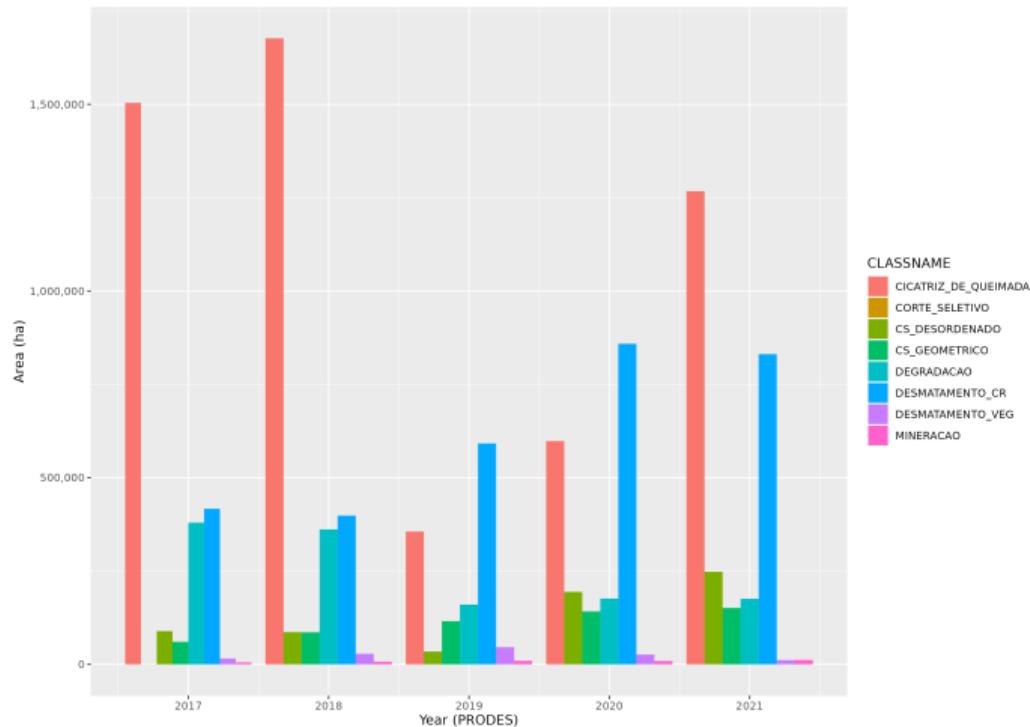
## What is DETER?

- ▶ DETER is a GIS which produces a fast assessment of deforestation and forest degradation in the Brazilian Amazon [Shimabukuro et al., 2006].
- ▶ DETER employs Linear Mixture Models of CBERS imagery and human experts to deter and issue warnings of deforested (or degraded) areas larger than 3 ha [De Almeida et al., 2022].
- ▶ Annually, DETER takes from PRODES the current forested area, starting anew issuing warnings.

# DETER warnings

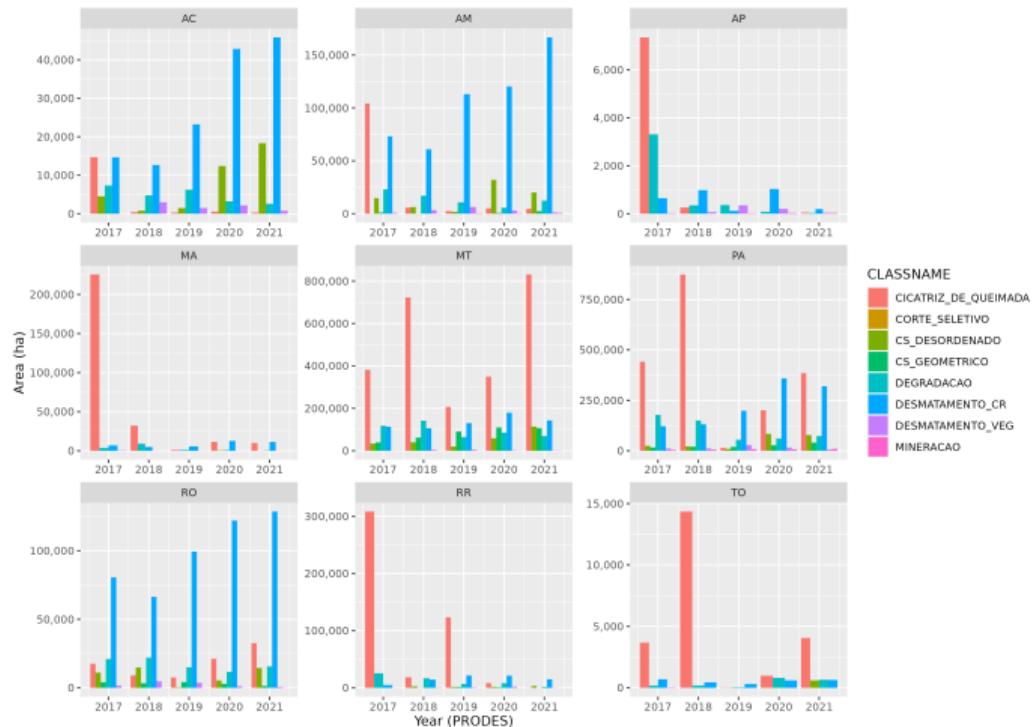


# DETER warnings by class



Burn scars and clear cut are the most common warnings.

# DETER warnings by class and state

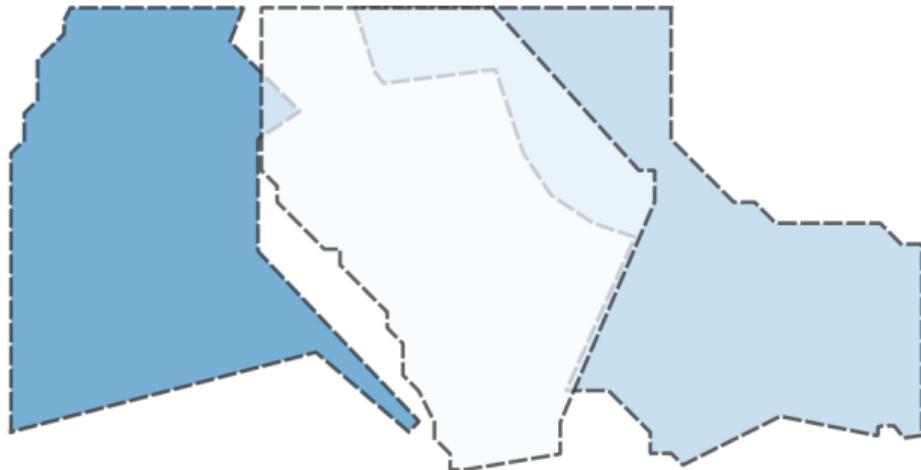


Burn scars and clear cut are the most common warnings.

## DETER warnings and time

- ▶ The spatial properties of DETER warning are inconsistent along time (shape, size, position, orientation).

## Warnings are inconsistent along time

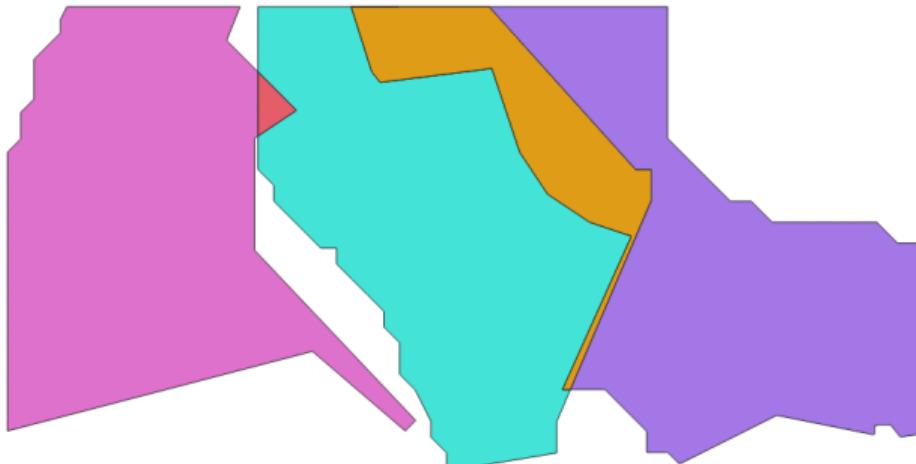


DETER warnings don't fit along time.

## DETER subareas

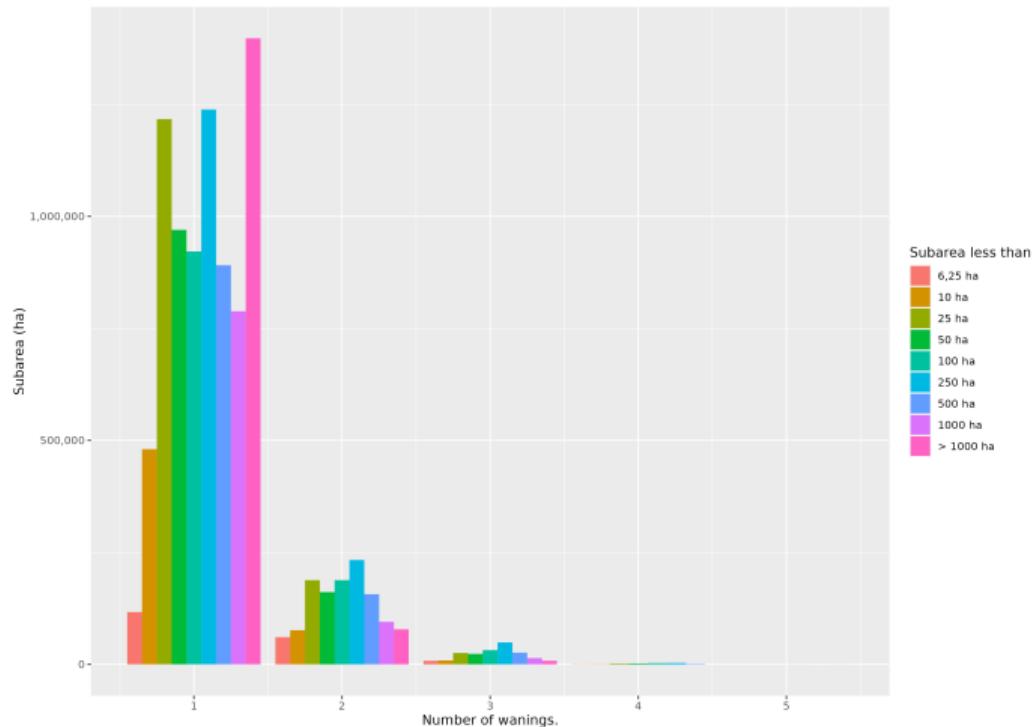
- ▶ The spatial properties of DETER warning are inconsistent along time (shape, size, position, orientation).
- ▶ DETER subareas maintain their spatial properties along time.

## DETER subareas



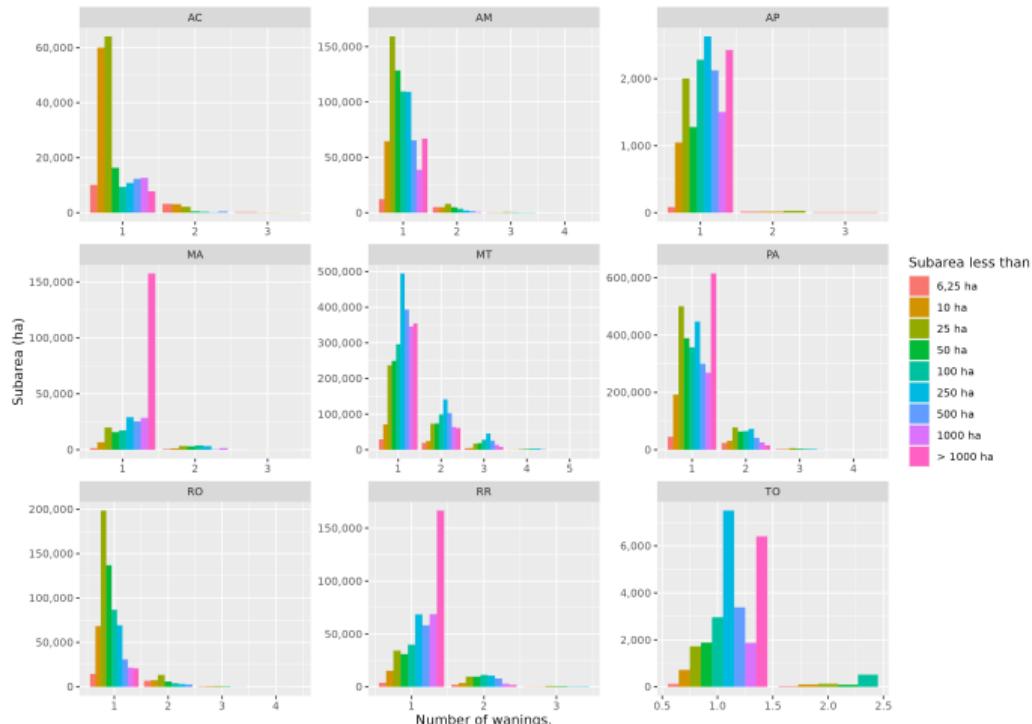
From 3 DETER warnings, we get 7 subareas!

# DETER subareas



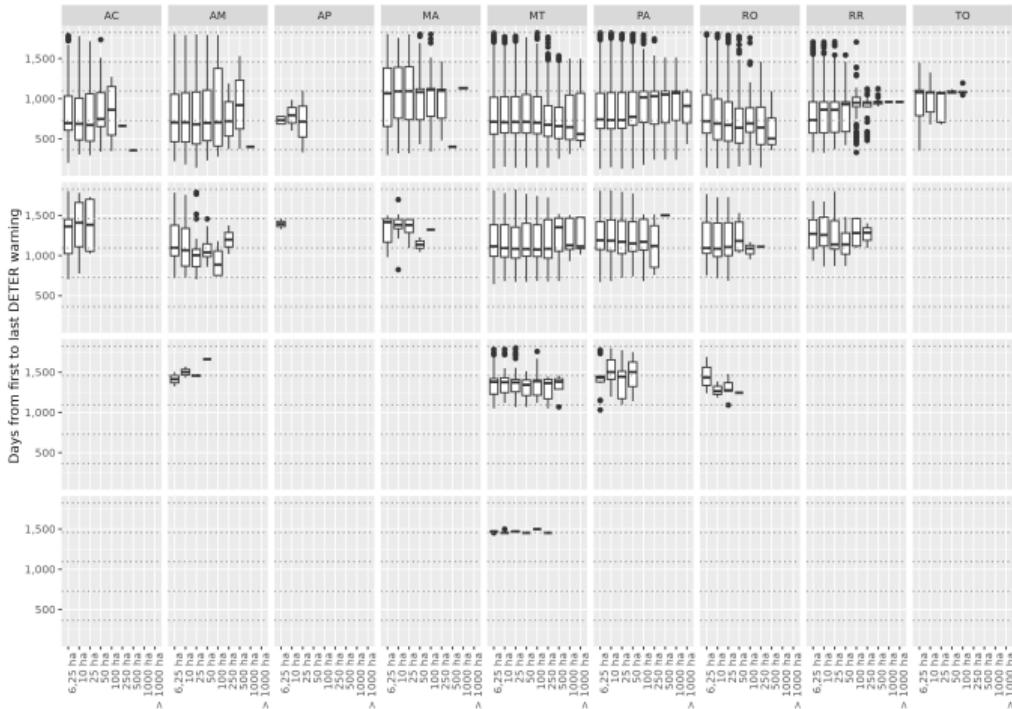
There are subareas with up to 5 recurrent warnings.

# DETER subareas



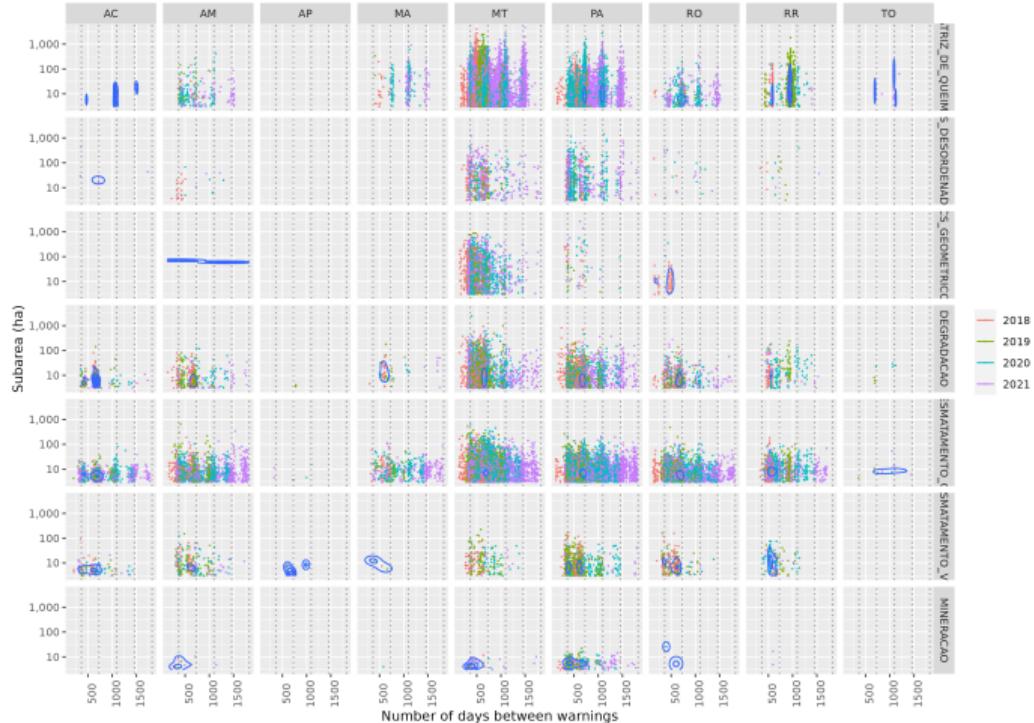
The warning recurrence changes by brazilian state.

# DETER subareas



Number of days between first and last warning.

# DETER subareas

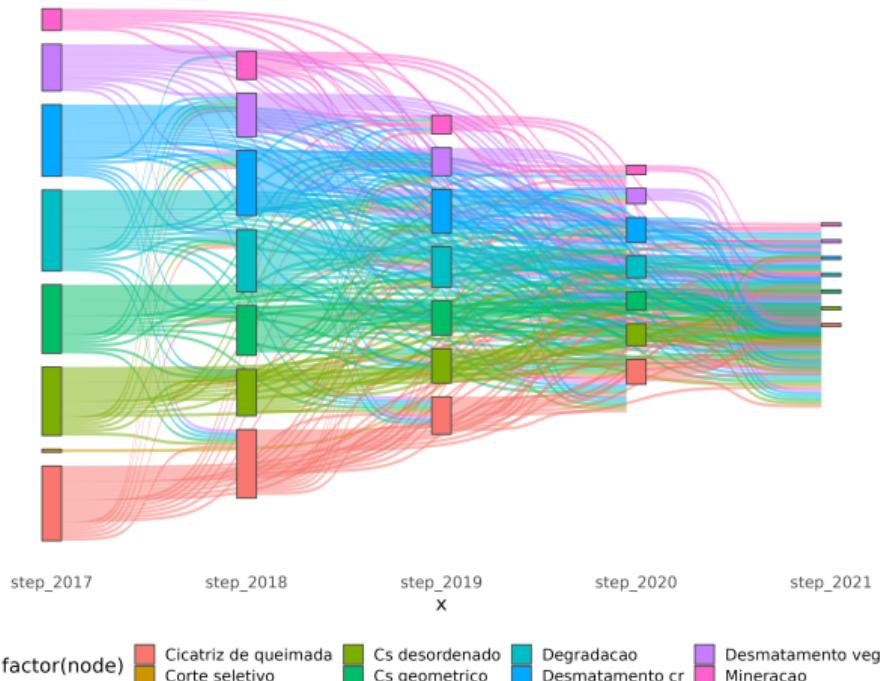


The number of days between warnings behaviour in space and time.

## Subarea trajectories

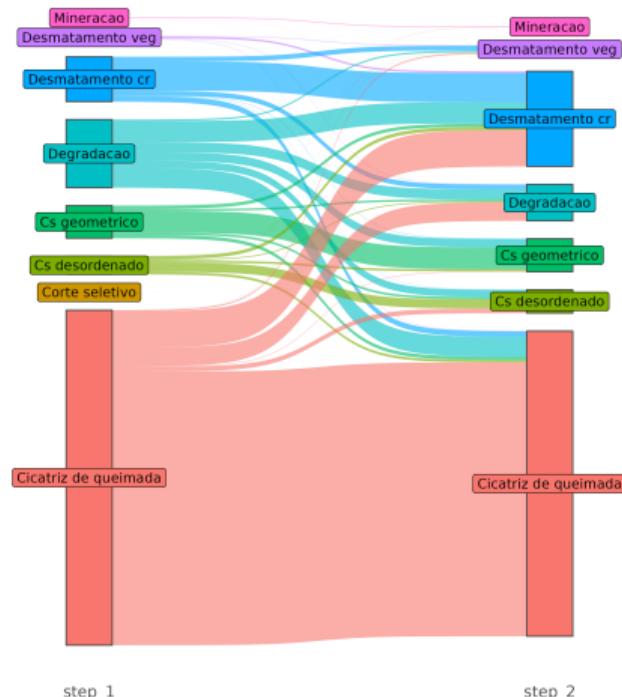
- ▶ Overlaped subareas organized along time describe trajectories of change.

# DETER subareas



Trajectory of subareas along the years.

# DETER subareas (2 warnings)

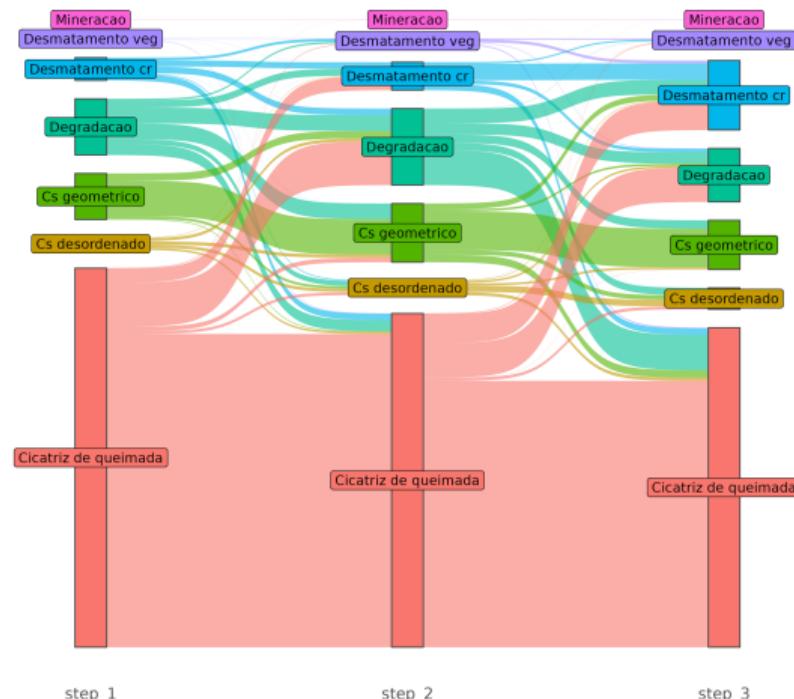


Trajectory of subareas with 2 wanings.

## DETER - Top 5 trajectories (2 warnings) |

position_1	position_2	subarea_ha	n_traj	p_area	p_traj
Cicatriz de queimada	Cicatriz de queimada	673806.1	15015	54.5	33.5
Cicatriz de queimada	Desmatamento cr	89540.6	5493	7.2	12.3
Desmatamento cr	Desmatamento cr	71670.7	7882	5.8	17.6
Cs geometrico	Cs geometrico	53594.9	623	4.3	1.4
Degradacao	Desmatamento cr	52004.1	3935	4.2	8.8
Total	-	1236329.4	44831	100.0	100.0

# DETER subareas (3 warnings)

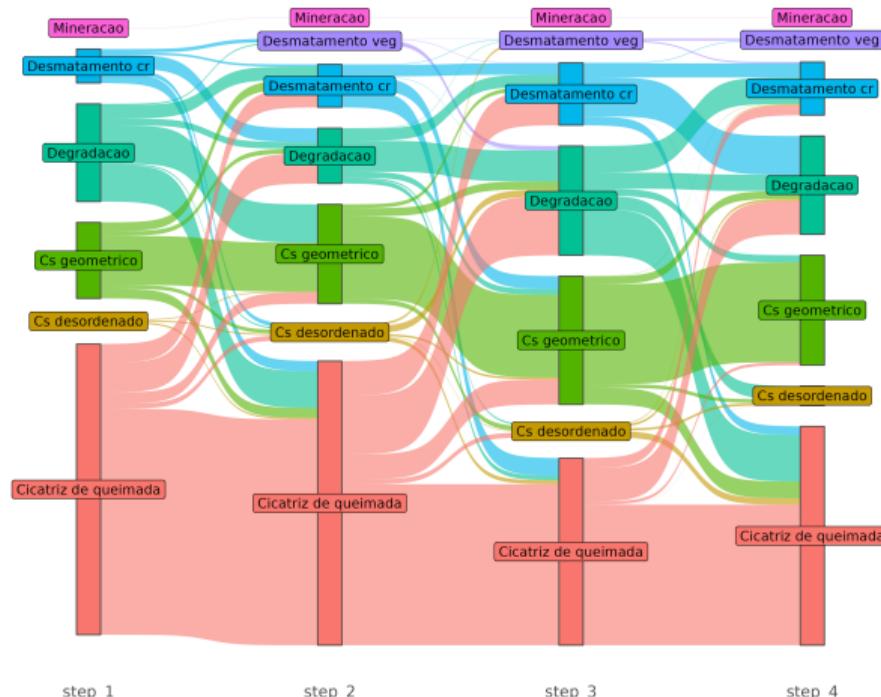


Trajectory of subareas with 3 wanings.

## DETER - Top 5 trajectories (3 warnings) |

position_1	position_2	position_3	subarea_ha	n_traj	p_area	p_traj
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	96013.6	1789	49.5	29.6
Cicatriz de queimada	Cicatriz de queimada	Degradacao	11700.9	345	6.0	5.7
Cicatriz de queimada	Degradacao	Cicatriz de queimada	11374.3	336	5.9	5.6
Cs geometrico	Cs geometrico	Cs geometrico	10345.3	145	5.3	2.4
Cicatriz de queimada	Cicatriz de queimada	Desmatamento cr	8944.6	423	4.6	7.0
Total	-	-	193940.4	6041	100.0	100.0

# DETER subareas (4 warnings)

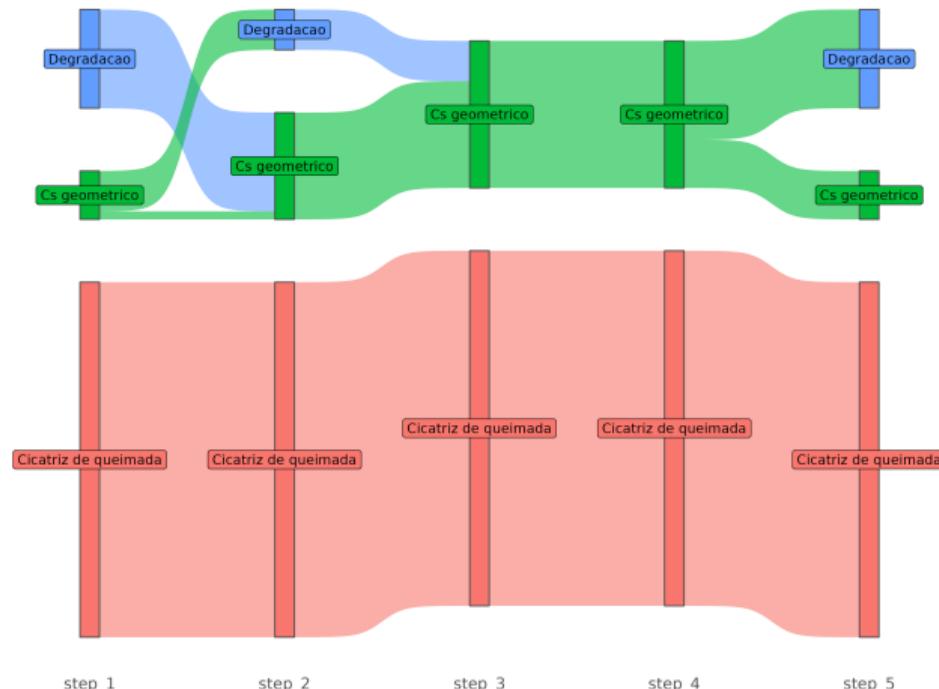


Trajectory of subareas with 4 wanings.

## DETER - Top 5 trajectories (4 warnings) |

position_1	position_2	position_3	position_4	subarea_ha	n_traj	p_area	p_traj
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	3003.6	92	24.4	20.5
Cs geometrico	Cs geometrico	Cs geometrico	Cs geometrico	892.2	16	7.3	3.6
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Degradação	725.4	12	5.9	2.7
Cicatriz de queimada	Cicatriz de queimada	Desmatamento cr	Degradação	525.2	10	4.3	2.2
Degradação	Cs geometrico	Cs geometrico	Cs geometrico	515.6	16	4.2	3.6
Total	-	-	-	12296.1	449	100.0	100.0

# DETER subareas (5 warnings)



Trajectory of subareas with 5 wanings.

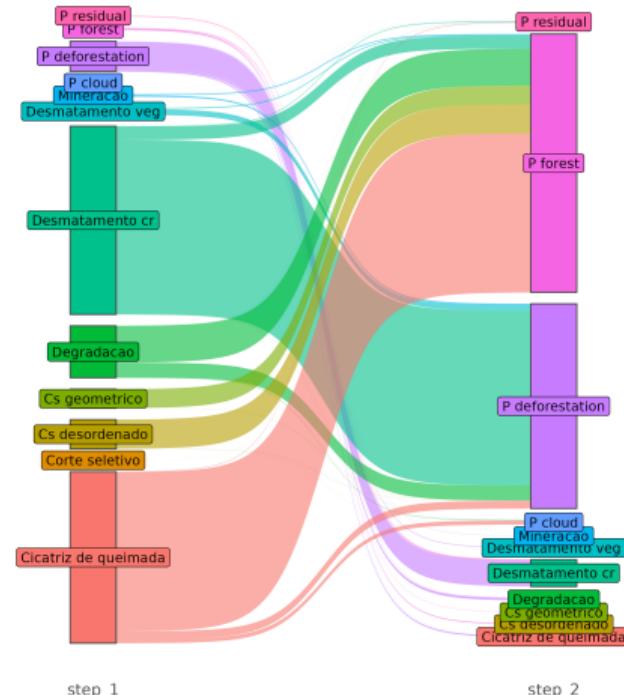
## DETER - Top 5 trajectories (5 warnings) |

position_1	position_2	position_3	position_4	position_5	subarea_ha	n_traj	p_area	p_traj
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	289.7	11	70.7	61.1
Degradacao	Cs geo- metrico	Cs geo- metrico	Cs geo- metrico	Degradacao	80.5	2	19.7	11.1
Cs geo- metrico	Degradacao	Cs geo- metrico	Cs geo- metrico	Cs geo- metrico	32.9	3	8.0	16.7
Cs geo- metrico	6.5	2	1.6	11.1				
Total	-	-	-	-	409.7	18	100.0	100.0
Total	-	-	-	-	409.7	18	100.0	100.0

## DETER & PRODES

- ▶ Add to the trajectories the PRODES class corresponding to each DETER subarea.
- ▶ The selected class corresponds to the mode of PRODES' pixels in each DETER subarea.
- ▶ Use PRODES' view date to sort trajectories.

# DETER & PRODES subareas (2 warnings)

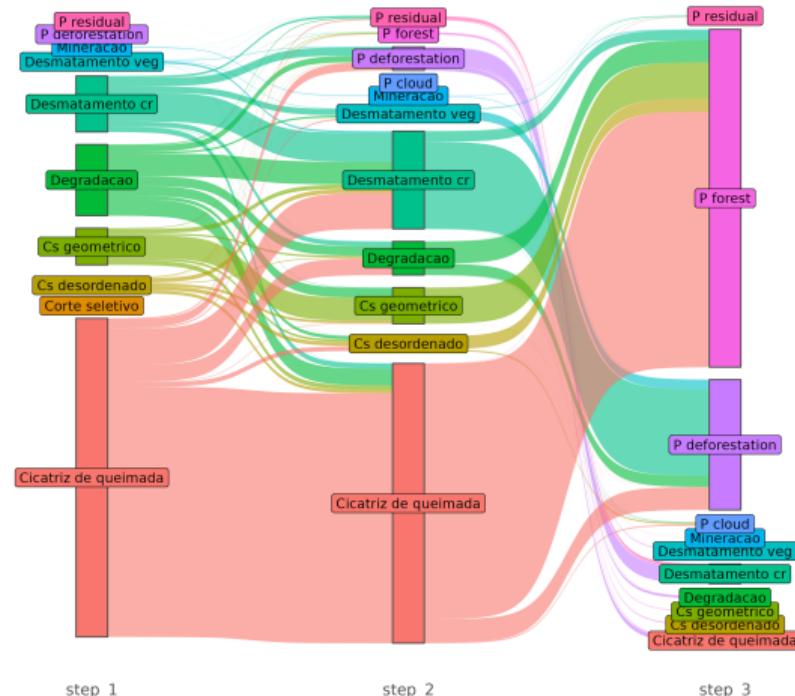


Trajectory of subareas with 2 wanings.

## DETER & PRODES - Top 5 trajectories (2 warnings) I

position_1	position_2	subarea_ha	n_traj	p_area	p_traj
Desmatamento cr	P deforestation	1862279.5	81144	34.8	49.1
Cicatriz de queimada	P forest	1695317.2	20680	31.7	12.5
Degradacao	P forest	390180.5	6985	7.3	4.2
Cs desordenado	P forest	310209.7	1892	5.8	1.1
P deforestation	Desmatamento cr	274060.6	16931	5.1	10.3
Total	-	5353742.5	165168	100.0	100.0

# DETER & PRODES subareas (3 warnings)

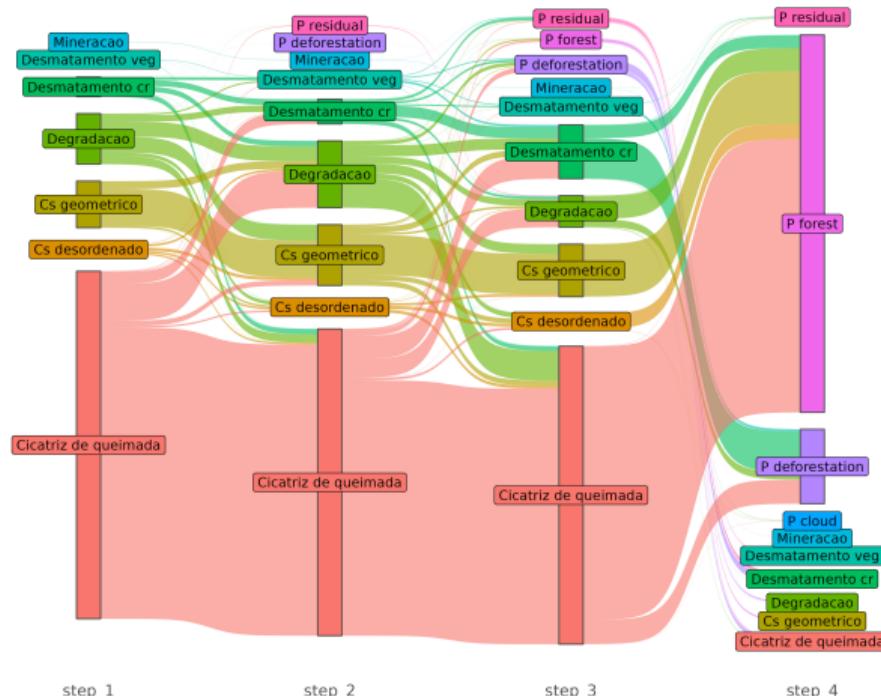


Trajectory of subareas with 3 wanings.

## DETER & PRODES - Top 5 trajectories (3 warnings) |

position_1	position_2	position_3	subarea_ha	n_traj	p_area	p_traj
Cicatriz de queimada	Cicatriz de queimada	P forest	386457.3	7570	46.0	23.4
Cicatriz de queimada	Desmatamento cr	P deforestation	51487.8	2626	6.1	8.1
Desmatamento cr	Desmatamento cr	P deforestation	43495.4	4322	5.2	13.4
Cs geometrico	Cs geometrico	P forest	38239.2	512	4.6	1.6
Degradacao	Desmatamento cr	P deforestation	33926.2	2272	4.0	7.0
Total	-	-	839718.8	32290	100.0	100.0

## DETER & PRODES subareas (4 warnings)

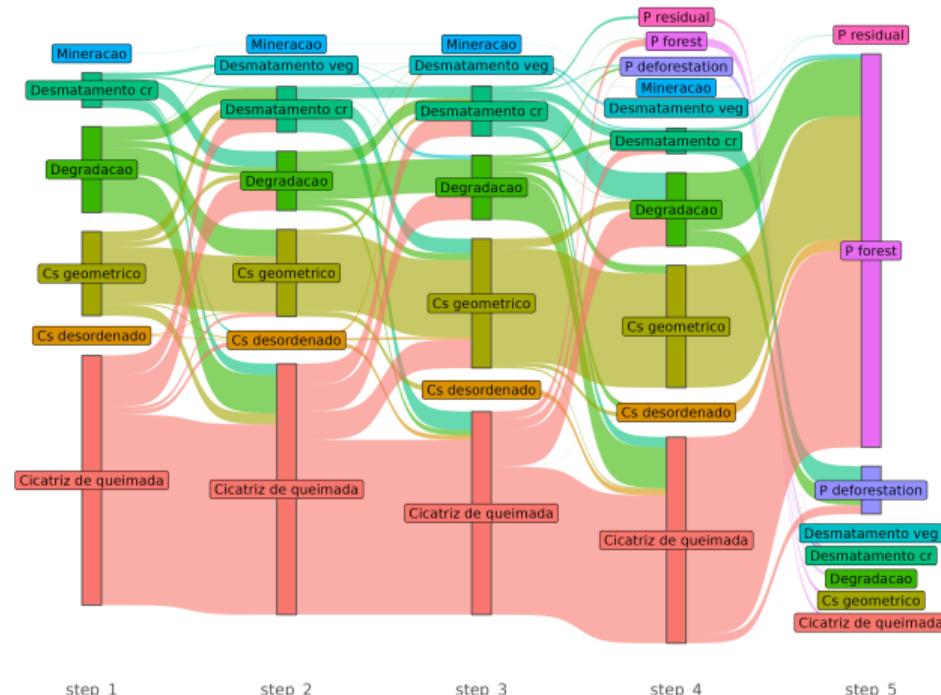


Trajectory of subareas with 4 wanings.

## DETER & PRODES - Top 5 trajectories (4 warnings) I

position_1	position_2	position_3	position_4	subarea_ha	n_traj	p_area	p_traj
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	P forest	75345.9	1203	49.5	27.0
Cs geometrico	Cs geometrico	Cs geometrico	P forest	8873.6	121	5.8	2.7
Cicatriz de queimada	Degradação	Cicatriz de queimada	P forest	6911.1	206	4.5	4.6
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	P deforestation	4408.4	132	2.9	3.0
Cicatriz de queimada	Cicatriz de queimada	Degradação	P forest	4280.1	136	2.8	3.1
Total	-	-	-	152156.4	4454	100.0	100.0

# DETER & PRODES subareas (5 warnings)

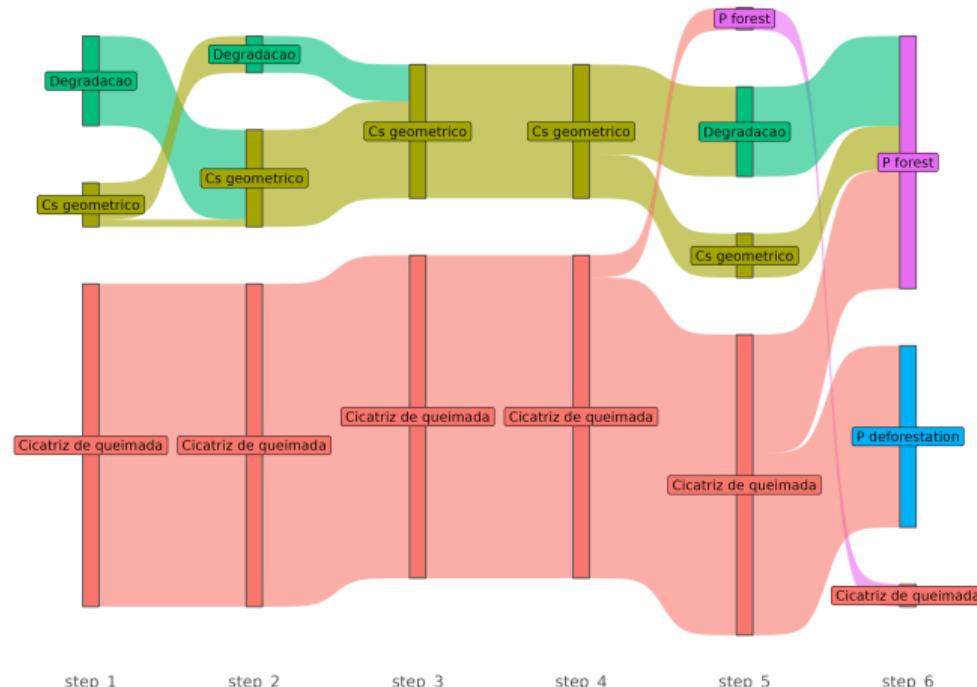


Trajectory of subareas with 5 wanings.

## DETER & PRODES - Top 5 trajectories (5 warnings) |

position_1	position_2	position_3	position_4	position_5	subarea_ha	n_traj	p_area	p_traj
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	P forest	2616.3	66	26.8	17.8
Cs geo- metrico	Cs geo- metrico	Cs geo- metrico	Cs geo- metrico	P forest	892.2	16	9.1	4.3
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Degradação	P forest	706.7	11	7.2	3.0
Degradação	Cs geo- metrico	Cs geo- metrico	Cs geo- metrico	P forest	515.6	16	5.3	4.3
Cicatriz de queimada	Degradação	Degradação	Cicatriz de queimada	P forest	424.8	20	4.3	5.4
Total	-	-	-	-	9773.7	370	100.0	100.0

# DETER & PRODES subareas (6 warnings)



Trajectory of subareas with 6 wanings.

## DETER & PRODES - Top 5 trajectories (6 warnings) I

position_1	position_2	position_3	position_4	position_5	position_6	subarea_ha	n_traj	p_area	p_traj
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	P defor- estation	163.1	6	39.8	33.3
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	P forest	106.6	4	26.0	22.2
Degradacao	Cs geo- metrico	Cs geo- metrico	Cs geo- metrico	Degradacao	P forest	80.5	2	19.7	11.1
Cs geo- metrico	Degradacao	Cs geo- metrico	Cs geo- metrico	Cs geo- metrico	P forest	32.9	3	8.0	16.7
Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	Cicatriz de queimada	P forest	Cicatriz de queimada	20.0	1	4.9	5.6
Total	-	-	-	-	-	409.7	18	100.0	100.0

## DETER & PRODES - proximity in time I

- ▶ There is a potential DETER-PRODES overlap in our analysis.
- ▶ DETER warnings and PRODES deforestation polygons could refer to the same event.
- ▶ The next slide reports the DETER warnings closest in time to PRODES polygons.

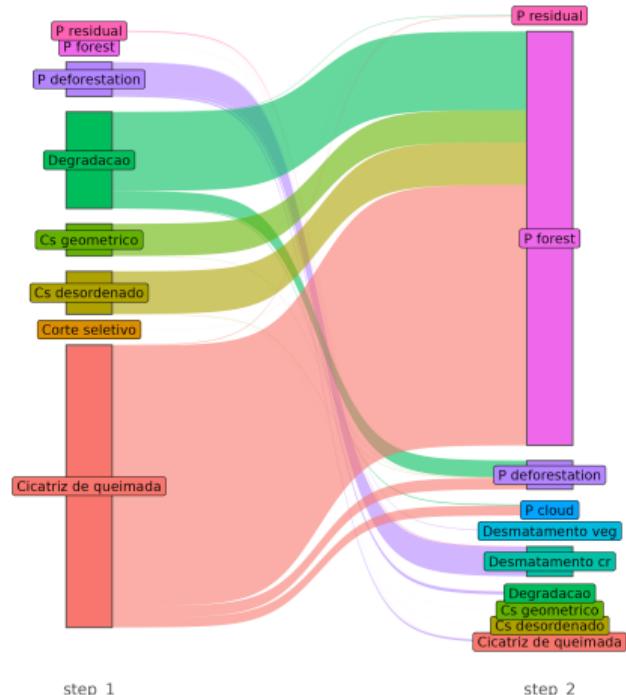
## DETER & PRODES - Top 5 proximity in time I

CLASSNAME	closest_class	total_ha	n	median_days	median_days_abs	sd_days	sd_abs
P deforestation	Desmatamento	2324959.2	111688	59	67	190.9	179.9
	cr	2215819.5	31574	677	677	514.4	514.3
P forest	Cicatriz de queimada	439190.5	8739	1009	1009	582.4	581.8
P forest	Degradacao	348185.6	2419	446	446	569.1	568.7
P forest	Cs desordenado	285442.5	2953	376	376	461.8	461.6
	Cs geometrico						

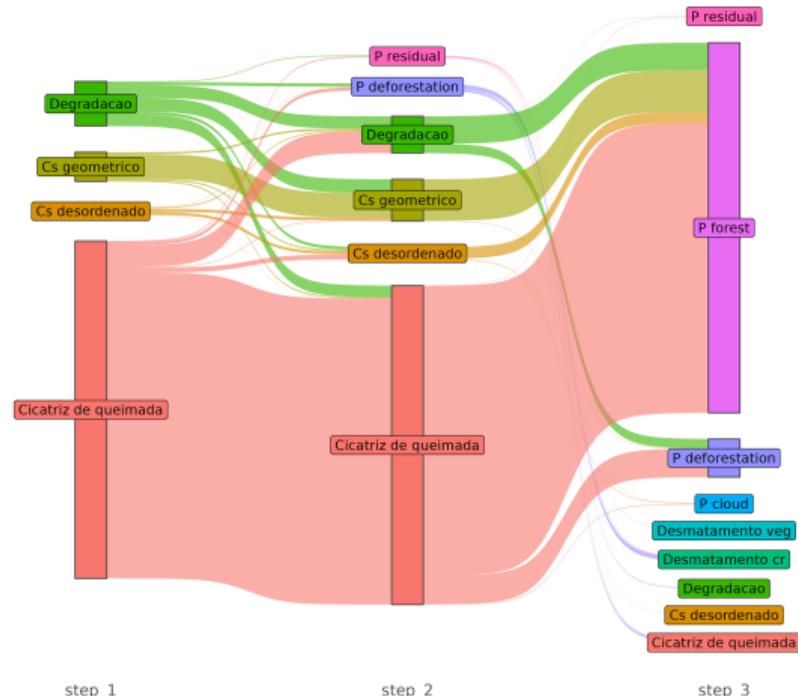
## Analysis 1

- ▶ Trajectories have one event in each PRODES year. There were 70/517059 with more than one.
- ▶ Trajectories related to mining were excluded.
- ▶ Trajectories end as soon as they reach deforestation.
- ▶ Trajectories include at least one PRODES event.

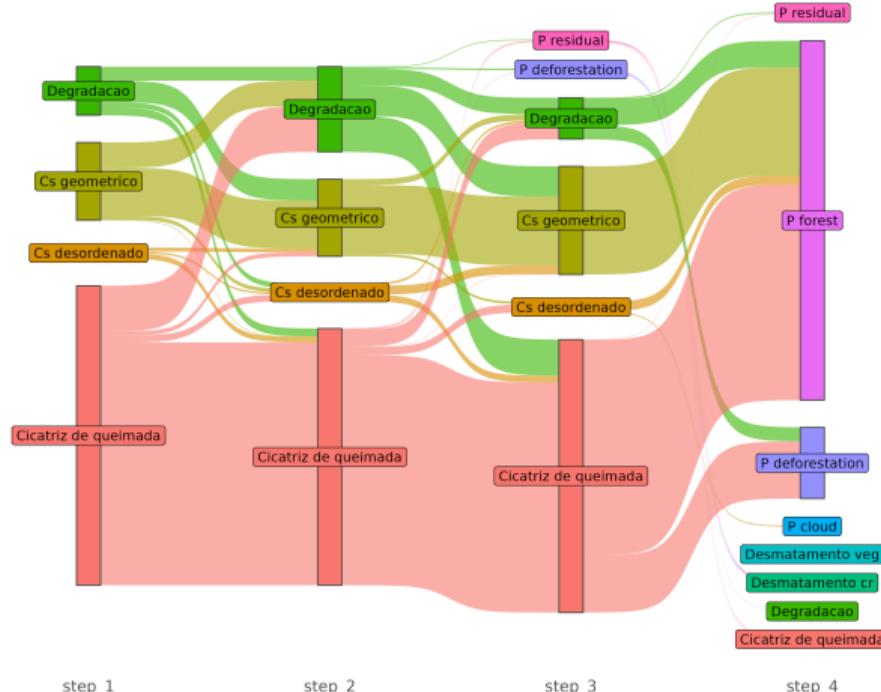
# Analysis 1 (2 warnings)



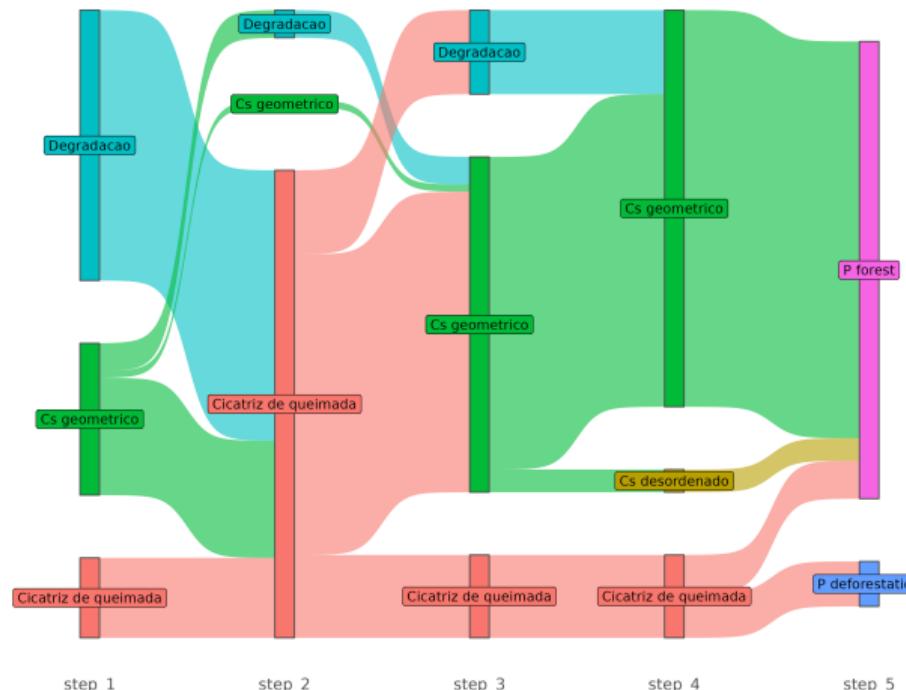
# Analysis 1 (3 warnings)



## Analysis 1 (4 warnings)



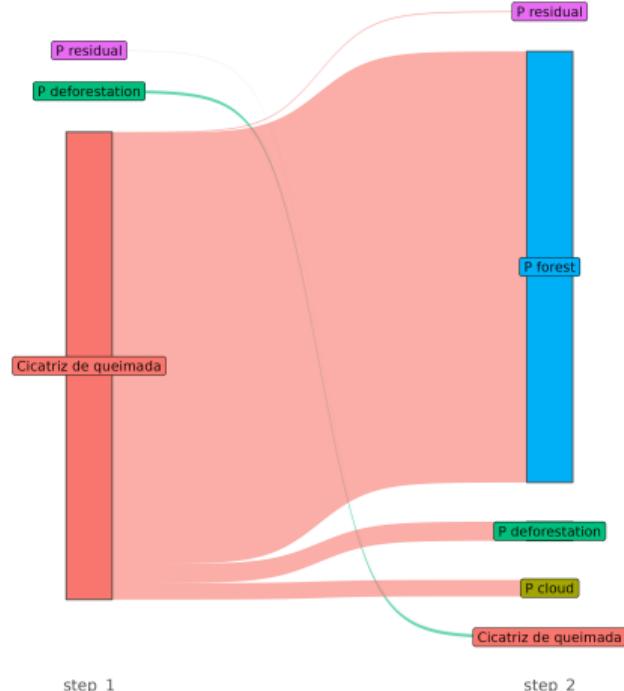
# Analysis 1 (5 warnings)



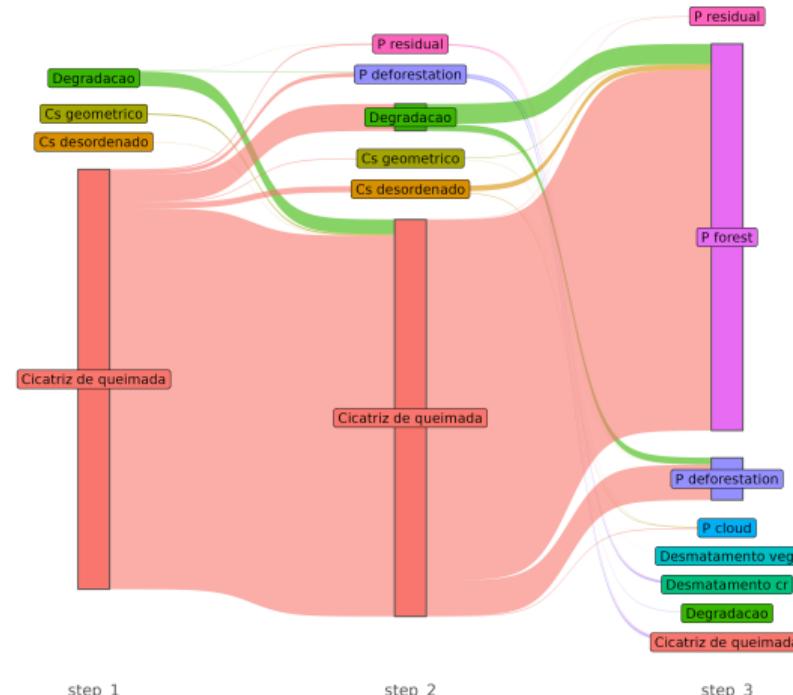
## Analysis 2

- ▶ Same as Analysis 1, but only using DETER's burn scars.

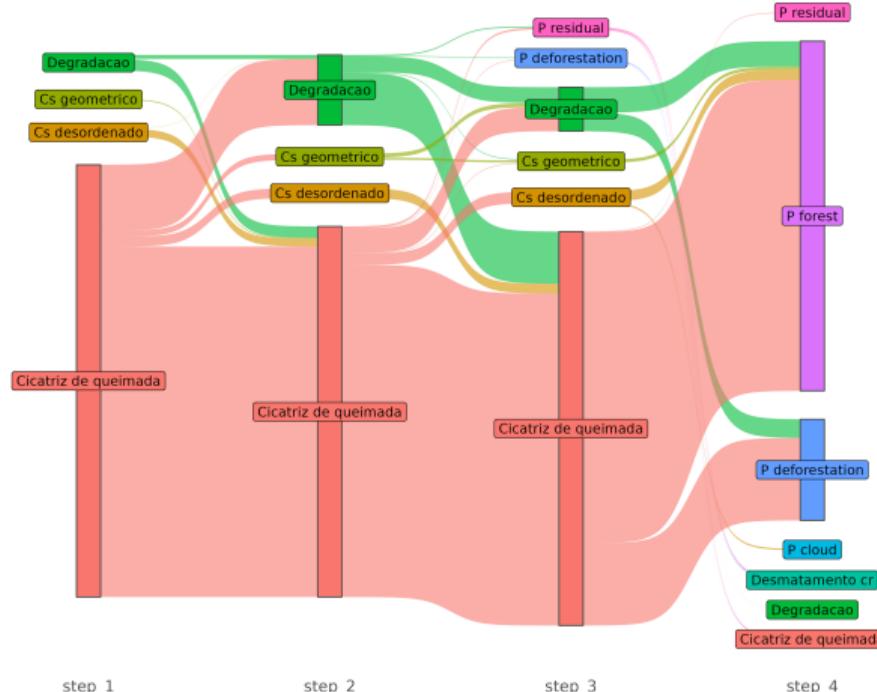
## Analysis 2 (2 warnings)



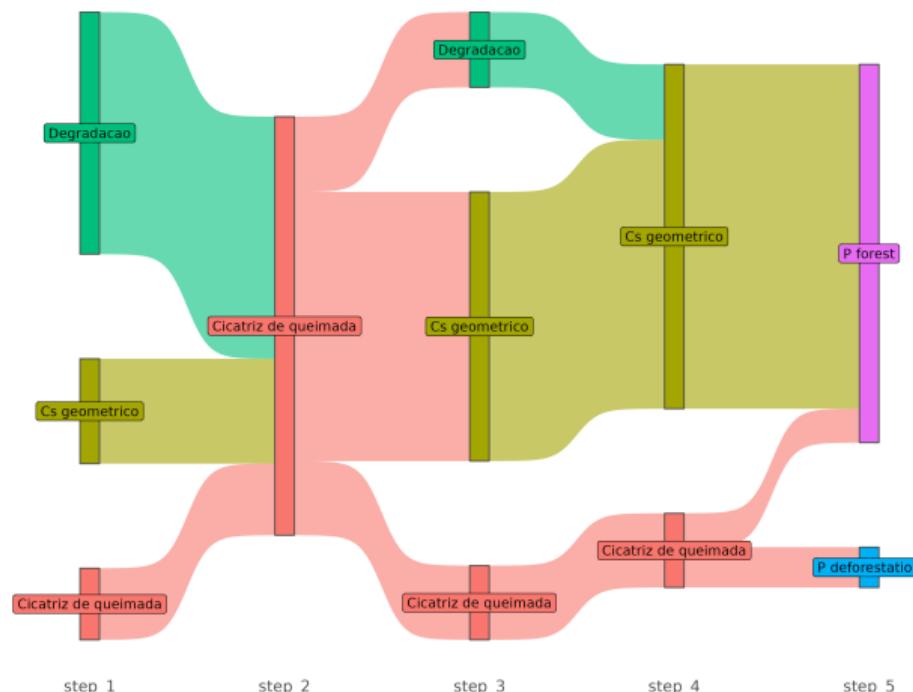
## Analysis 2 (3 warnings)



## Analysis 2 (4 warnings)



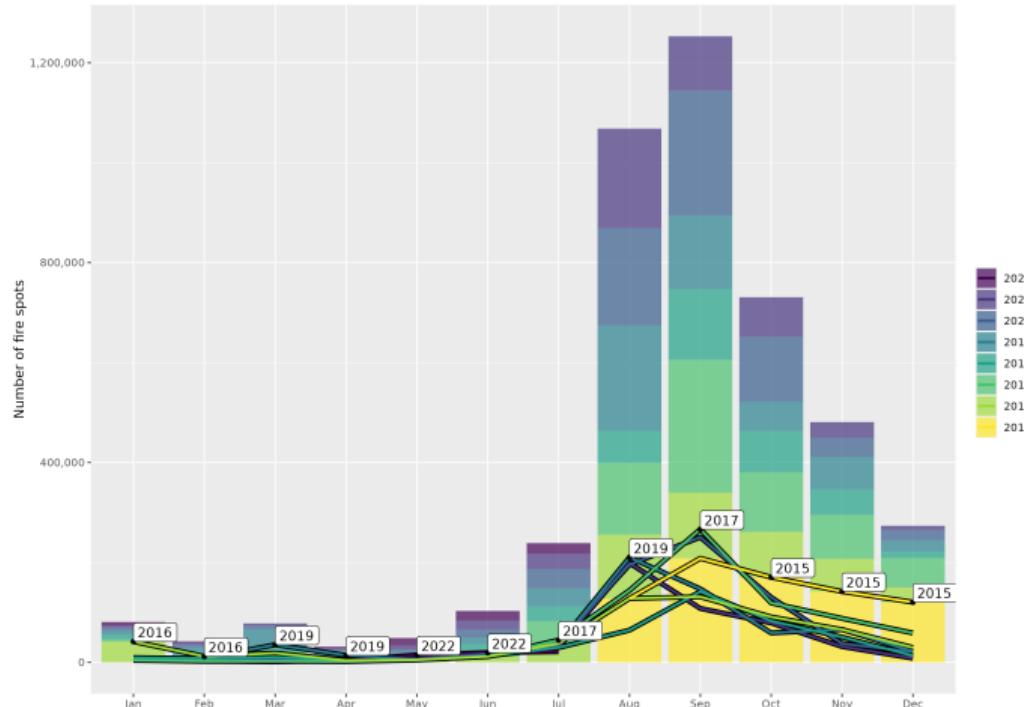
## Analysis 2 (5 warnings)



## Fire spots

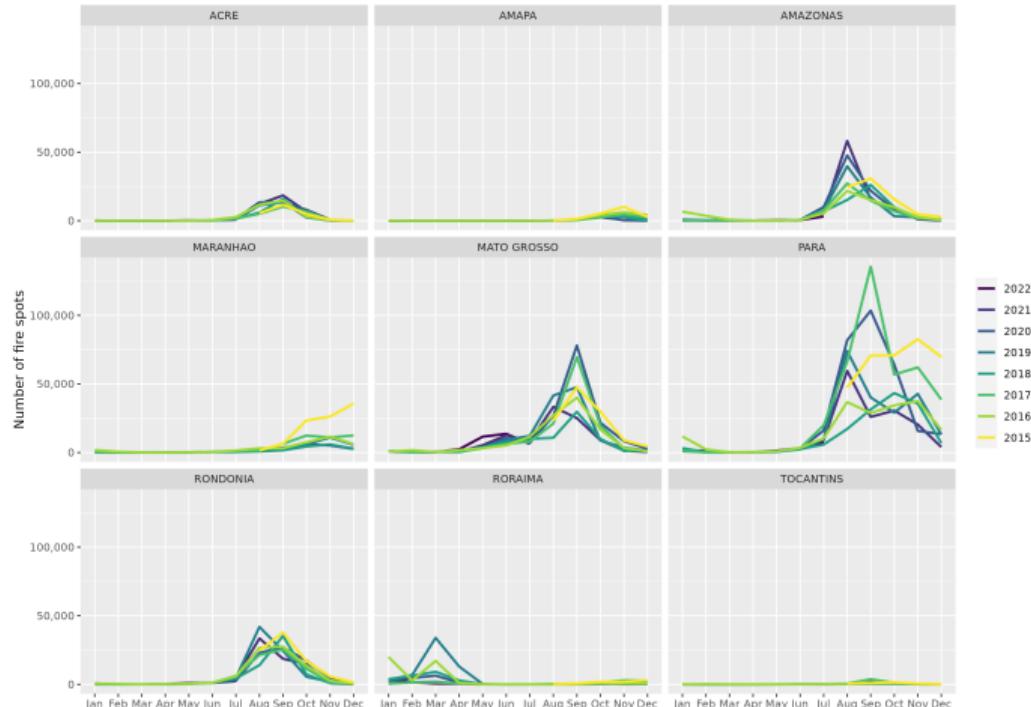
- ▶ We downloaded fire spots from INPE's *Queimadas* project.
- ▶ The data corresponds to VIIRS satellite (NPP-375 morning and afternoon).

## Fire spots by month



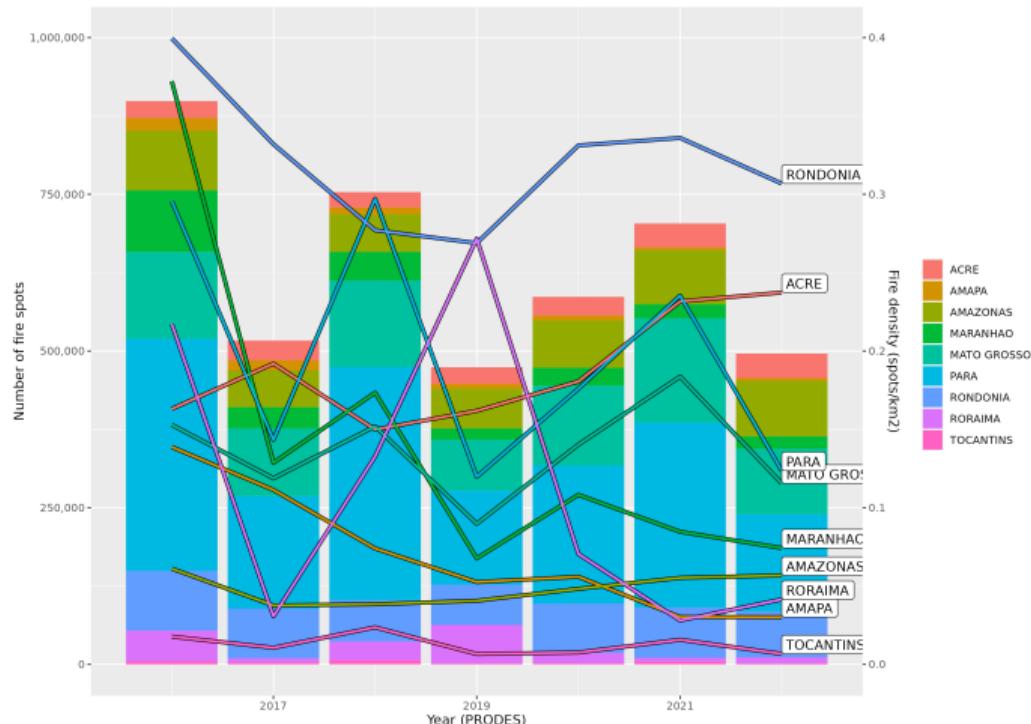
Aug (2019) & Sep (2017) top fire spots. Note 2015's last trimester.

# Fire spots by month and state



Note the increasing trend in Amazonas and its move towards August.

# Fire spots density (by area overlap with BLA)



Number of fire spots by state and their density (by area in BLA).

## Current analysis venue

- ▶ How to convert fires spots into fire events?
- ▶ For example, [Andela et al., 2022] had the same problem.
- ▶ DETER data constrains the problem's spatial domain.

## Final remarks

- ▶ The analysis of DETER warning subareas along time could improve the characterization of forest degradation along time.
- ▶ Potential applications of our work are:
  - ▶ Improve estimation of emissions of greenhouse gases, i.e. our data could help avoiding double counting.
  - ▶ Identify spatio-temporal areas which could help training Machine-Learning algorithms for automatic identification of forest degradation.
- ▶ Code available at <https://github.com/albhasan/treesburnareas>

## References I

-  Andela, N., Morton, D. C., Schroeder, W., Chen, Y., Brando, P. M., and Randerson, J. T. (2022).  
Tracking and classifying Amazon fire events in near real time.  
*Science Advances*, 8(30):eabd2713.
-  De Almeida, C. A., Maurano, L., Valeriano, D. M., Câmara, G., Vinhas, L., Da Motta, M., Gomes, A. R., Monteiro, A. M. V., Souza, A. A. D. A., Messias, C. G., Rennó, C. D., Adami, M., Escada, M. I. S., De Souza Soler, L., and Amaral, S. (2022).  
Metodologia Utilizada nos Sistemas PRODES e DETER - 2a Edição (atualizada).  
Technical report, Instituto Nacional de Pesquisas Espaciais (INPE).

## References II

-  Shimabukuro, Y., Duarte, V., Anderson, L., Valeriano, D., Arai, E., Freitas, R., Rudorff, B., and Moreira, M. (2006).  
Near real time detection of deforestation in the Brazilian Amazon using MODIS imagery.  
*Ambiente e Agua - An Interdisciplinary Journal of Applied Science*, 1(1):37–47.