



# SAR4Change: Deforestation Detection Using Dual-polarimetric SAR Information

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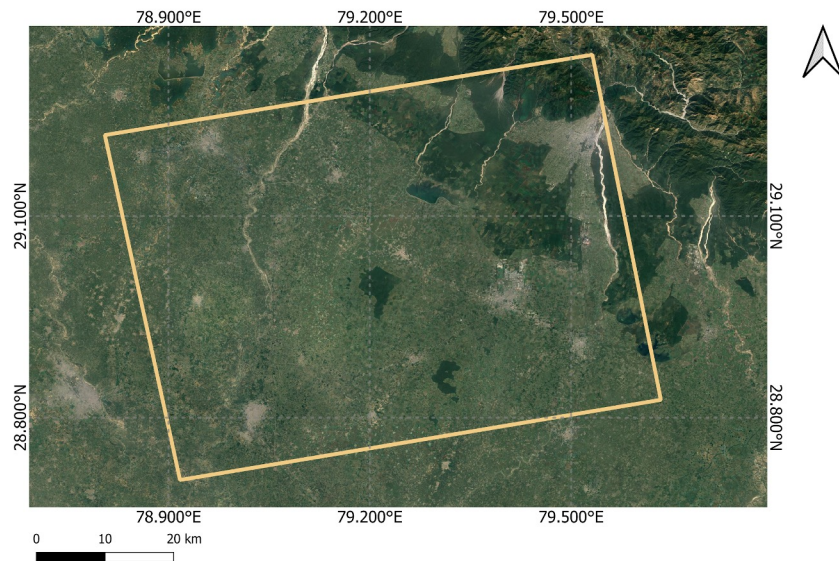
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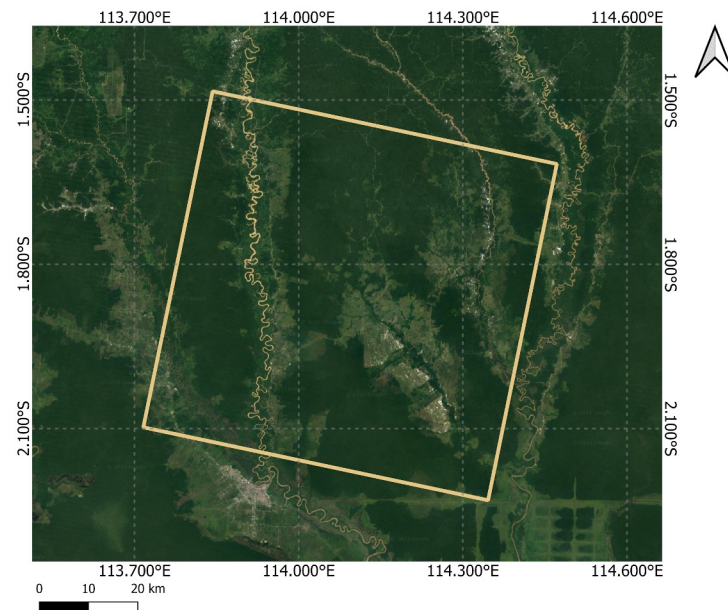
# Motivation

- Deforestation detection using SAR time-series is important
- Focus on Sentinel-1 and NISAR missions, we aim to study the impact of
  - Frequency
  - Polarization
  - Resolution
  - Forest eco-regions

## Haldwani, India



## Kalimantan, Indonesia



## ❑ C-band

- ❑ Sentinel-1, Dual-pol, SLC
- ❑ Timeline: 2016-2020
- ❑ Number of acquisitions
  - ❑ Haldwani: 71
  - ❑ Kalimantan: 134

## ❑ L-band

- ❑ ALOS-2, Dual-pol, Stripmap SLC
- ❑ Timeline: 2016-2020
- ❑ Number of acquisitions
  - ❑ Haldwani: 23
  - ❑ Kalimantan: 28

### Kalimantan

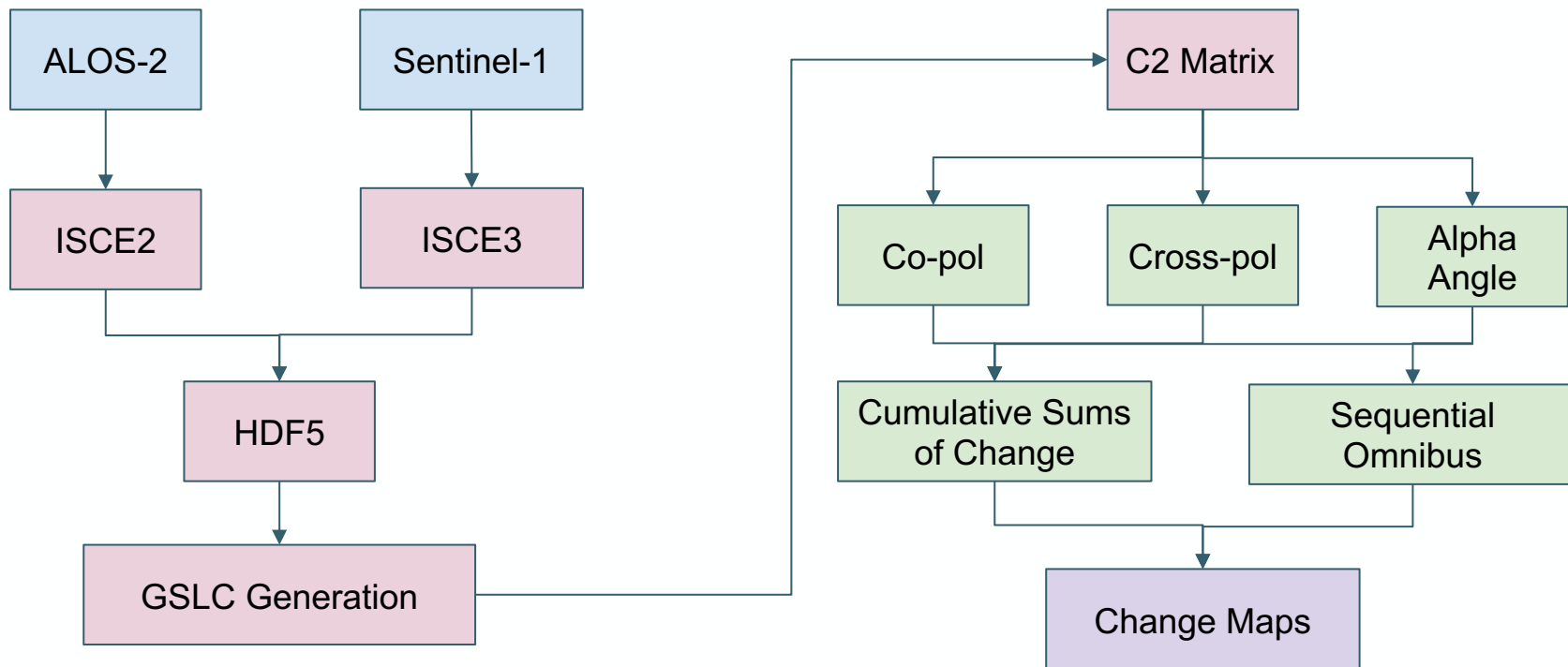
2016	2017	2018	2019	2020
14	29	29	28	24

### Haldwani

2016	2017	2018	2019
8	27	26	10



# Processing Workflow



## Cumulative Sums of Change

Initialize cumulative sum  
 $S_0=0$

Calculate cumulative sums  
 $S_i = S_{i-1} + (I_i - I_{\text{mean}})$

Calculate  
 $S_{\text{max}} = \max_{i=0,1,\dots,n} S_i$

Calculate  
 $S_{\text{min}} = \min_{i=0,1,\dots,n} S_i$

Calculate  
 $S_{\text{diff}} = S_{\text{max}} - S_{\text{min}}$

### *Bootstrapping*

Generate a bootstrap sample  
by randomly reordering the  
original values

Calculate bootstrap CuSum  
 $S_i^o$  for  $i=1,2,\dots,n$  ( $S_0^o$ ,  
 $S_1^o,\dots,S_n^o$ )

Calculate max, min and  
difference of bootstrap  
CuSum denoted  $S_{\text{max}}^o$ ,  $S_{\text{min}}^o$   
and  $S_{\text{diff}}^o$

Calculate number of samples  
X where  $S_{\text{diff}}^o < S_{\text{diff}}$   
Total number of sample = N

Calculate Confidence Level =  
 $100 * X/N \%$

## Sequential Omnibus

Test for equality of several complex  
covariance matrices  
Null Hypothesis:  
 $H_0 = \Sigma_1 = \Sigma_2 = \dots = \Sigma_k$

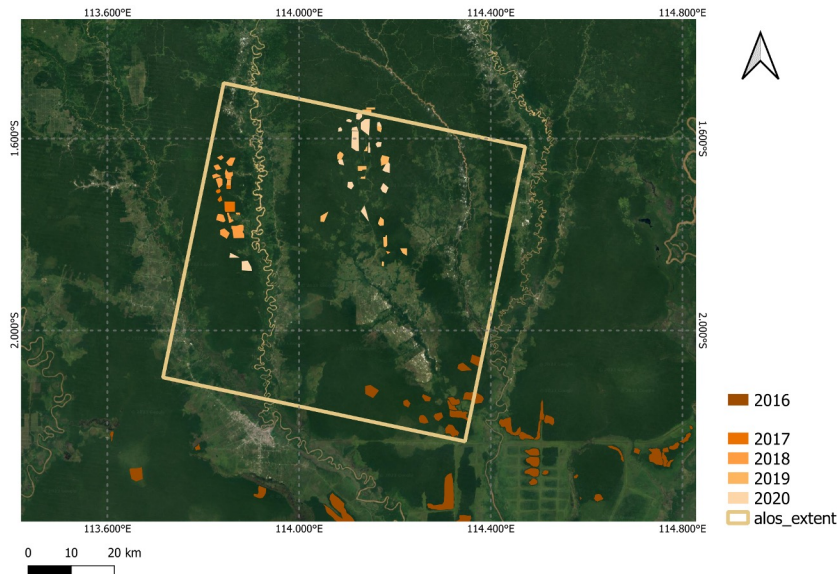
Omnibus Test Statistic  
 $Q = \{k^p \prod_{i=1}^k |X_i| / |X|^{kn}\}$

Complex Wishart  
Distribution followed by  
 $X_i = n \sum_i = n \langle C \rangle_i$   
N = equivalent number of  
looks  
 $X = \prod_{i=1}^k X_i$

p = 3: full pol  
2: dual pol  
1: single pol

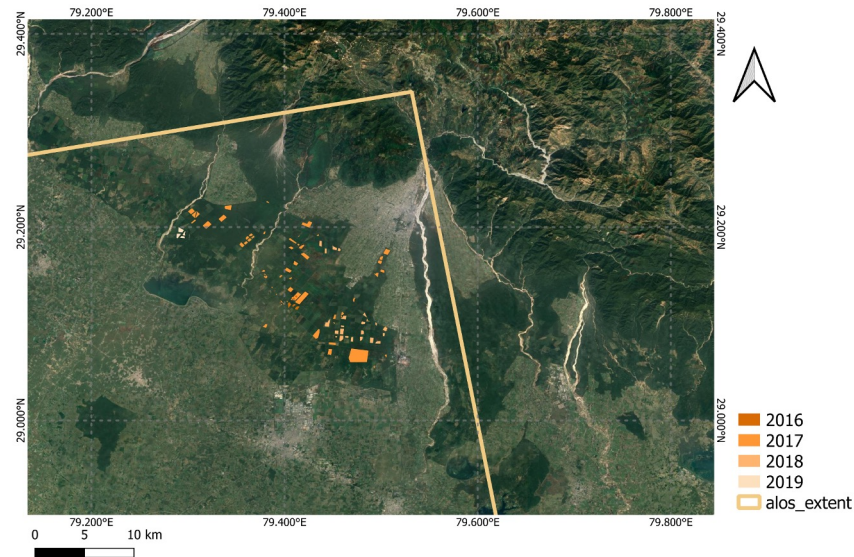


## Kalimantan



Minimum Area: 4 ha  
Maximum Area: 79 ha

## Haldwani



Minimum Area: 1 ha  
Maximum Area: 220 ha

- **Overall Accuracy:** Proportion of correctly classified instances over the total number of instances.

$$\text{Overall Accuracy} = (TP + TN) / (TP + TN + FP + FN)$$

- **User's accuracy:** Proportion of correctly classified positive instances out of all actual positive instances.

$$\text{User's Accuracy} = TP / (TP + FN)$$

- **Producer's Accuracy (Precision):** Proportion of correctly classified positive instances out of all predicted positive instances.

$$\text{Producer's Accuracy} = TP / (TP + FP)$$

- **F1 Score:** Harmonic mean of precision and recall, providing a single metric that balances both metrics.

$$\text{F1 Score} = 2 * (\text{Producer's Accuracy} * \text{User's Accuracy}) / (\text{Producer's Accuracy} + \text{User's Accuracy})$$

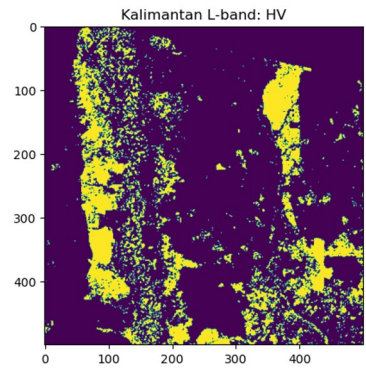
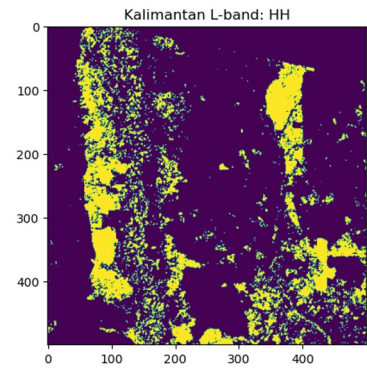
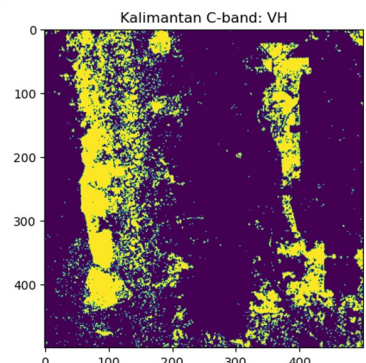
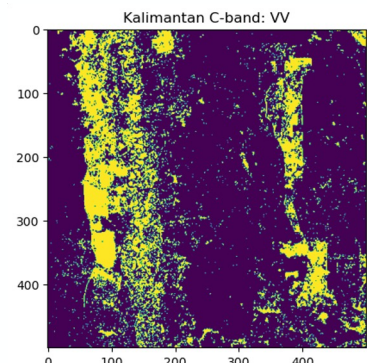
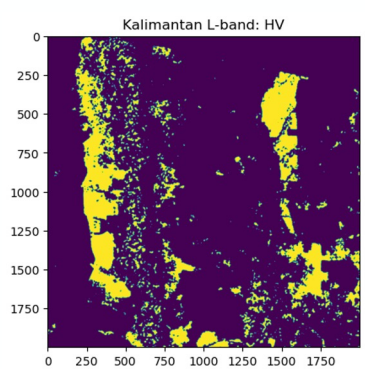
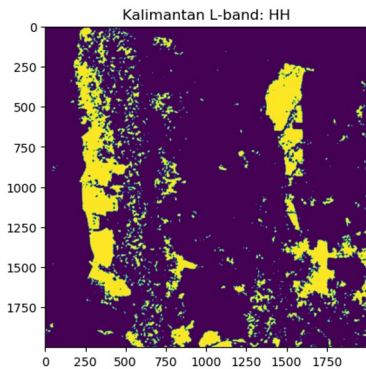
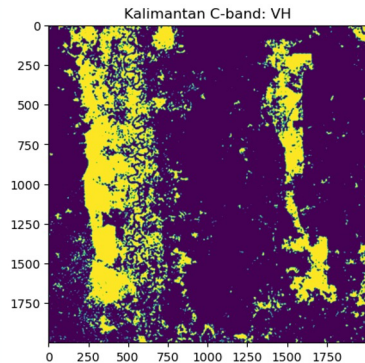
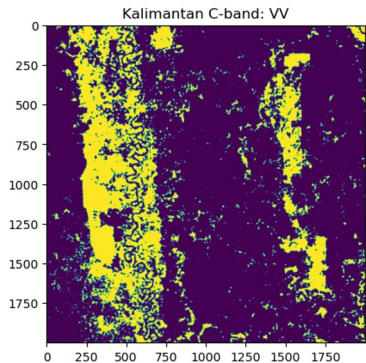
- **Kappa Coefficient:** Agreement between the predicted and actual labels, adjusted for the agreement occurring by chance.

$$\text{Kappa} = (\text{Overall Accuracy} - \text{Expected Accuracy}) / (1 - \text{Expected Accuracy})$$

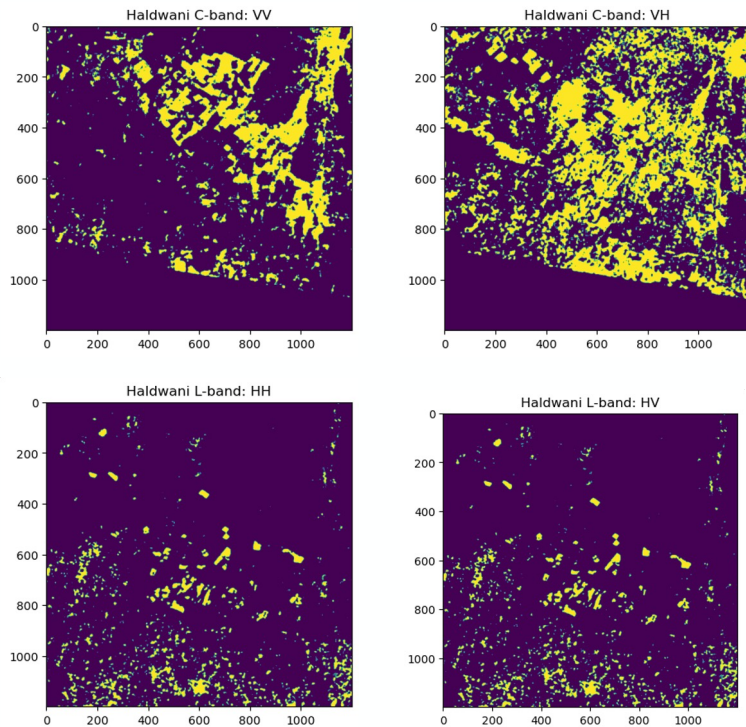


25m

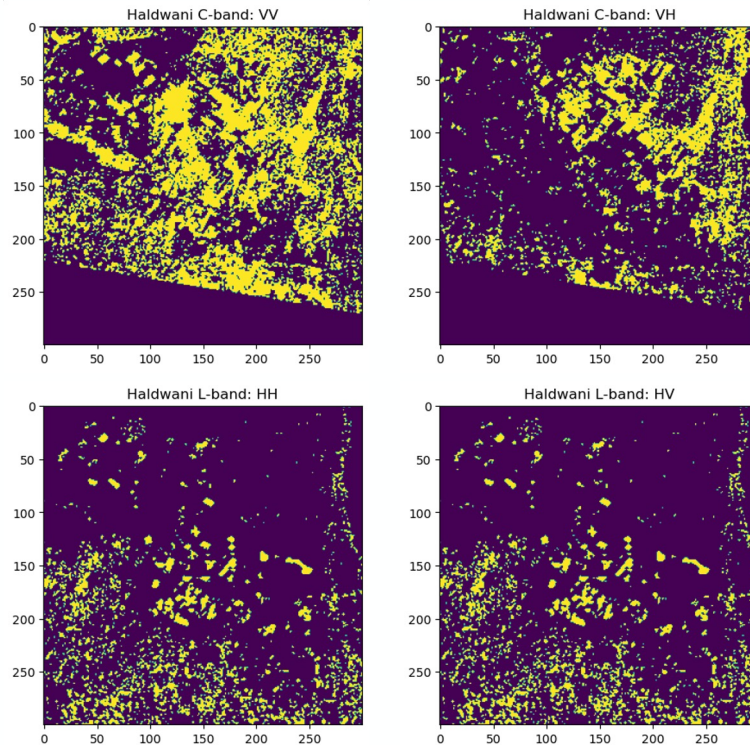
100m



25m

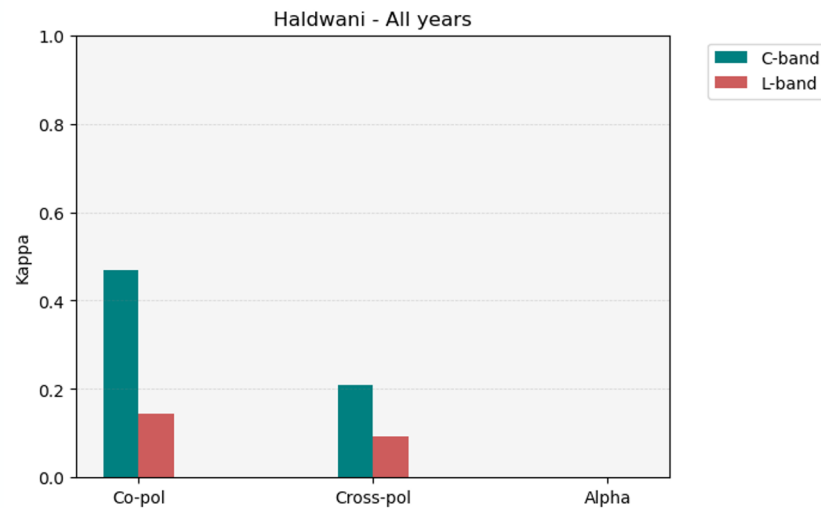
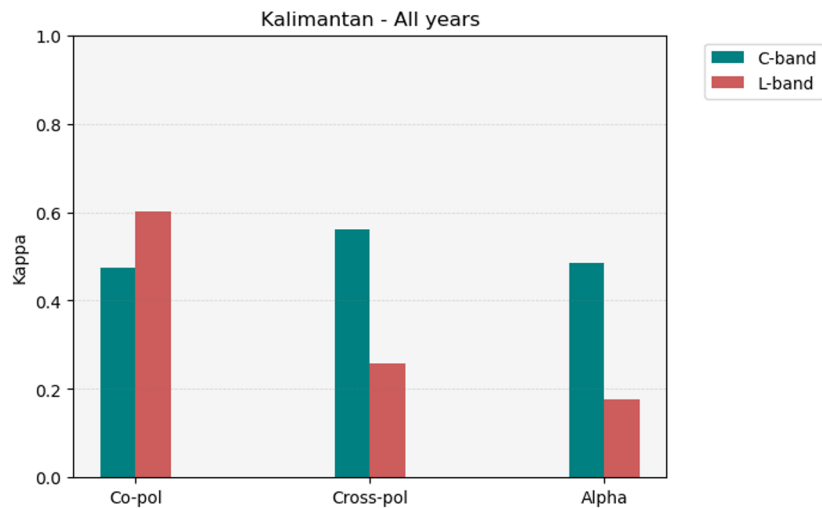


100m

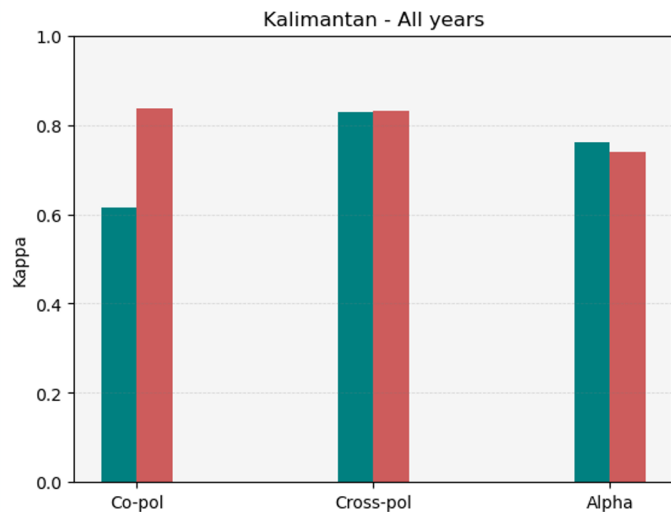




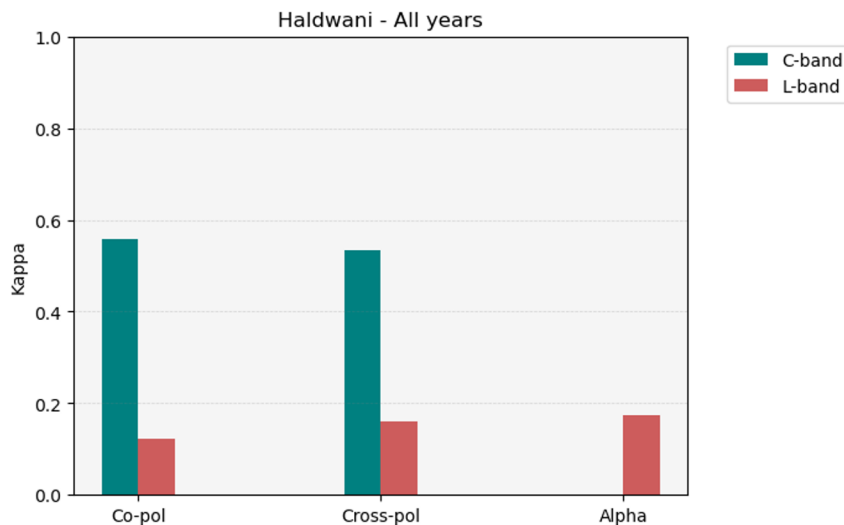
At 25m resolution, Cumulative Sums of Change



At 100m resolution, Cumulative Sums of Change



C-band  
L-band



C-band  
L-band

## Overall Accuracy

### Kalimantan

		25m	100m
Sentinel-1	Co-pol	0.737	0.802
	Cross-pol	0.783	0.915
ALOS-2	Co-pol	0.796	0.919
	Cross-pol	0.635	0.915

### Haldwani

		25m	100m
Sentinel-1	Co-pol	0.745	0.782
	Cross-pol	0.608	0.770
ALOS-2	Co-pol	0.625	0.577
	Cross-pol	0.563	0.585



## Kappa

### Kalimantan

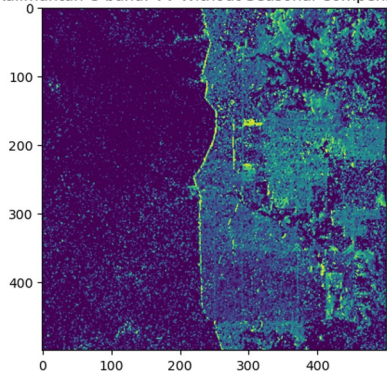
		25m	100m
Sentinel-1	Co-pol	0.474	0.614
	Cross-pol	0.561	0.828
ALOS-2	Co-pol	0.602	0.837
	Cross-pol	0.258	0.830

### Haldwani

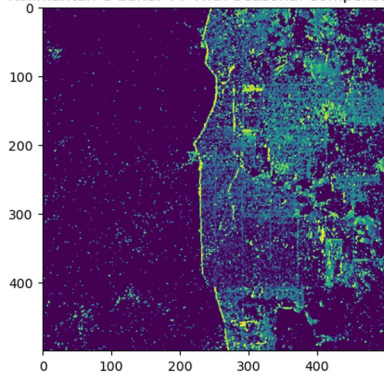
		25m	100m
Sentinel-1	Co-pol	0.745	0.559
	Cross-pol	0.608	0.533
ALOS-2	Co-pol	0.144	0.122
	Cross-pol	0.093	0.161

# Compensating for seasonality

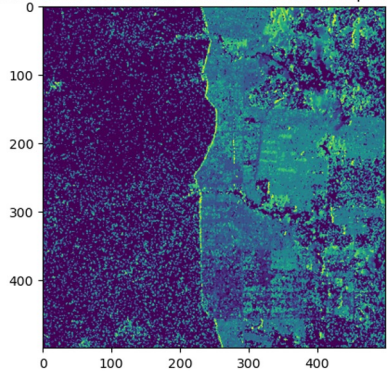
Kalimantan C-band: VV Without Seasonal Compensation



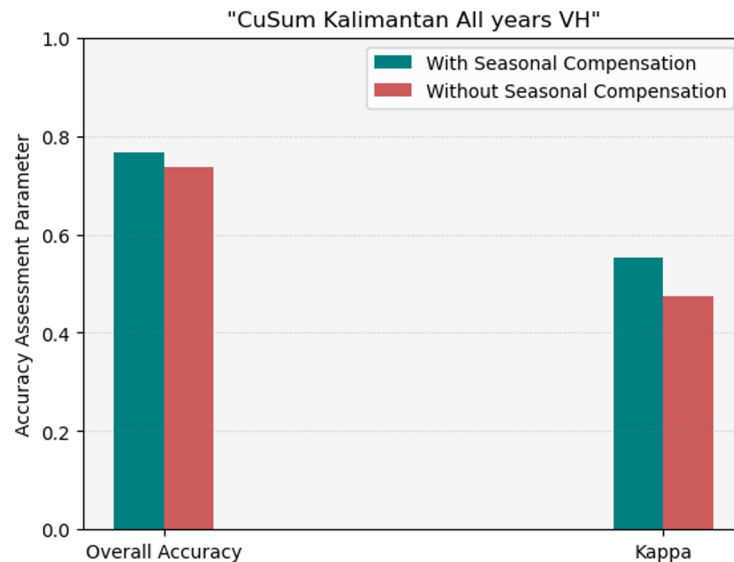
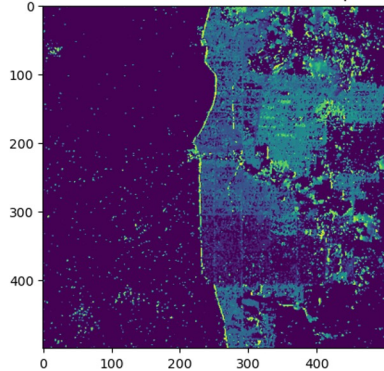
Kalimantan C-band: VV With Seasonal Compensation



Kalimantan C-band: VH Without Seasonal Compensation



Kalimantan C-band: VH With Seasonal Compensation



# Effect of logged area size

Kalimantan

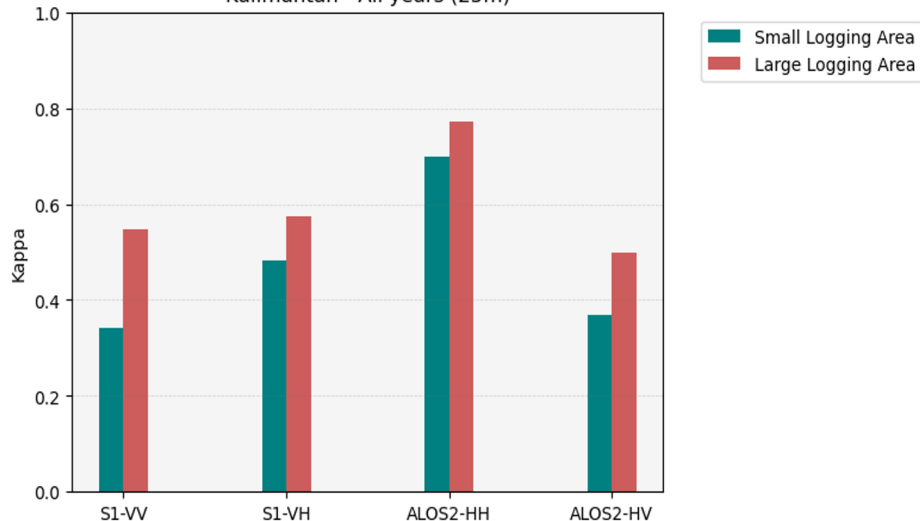


Two subsets chosen with different sizes of logged area:

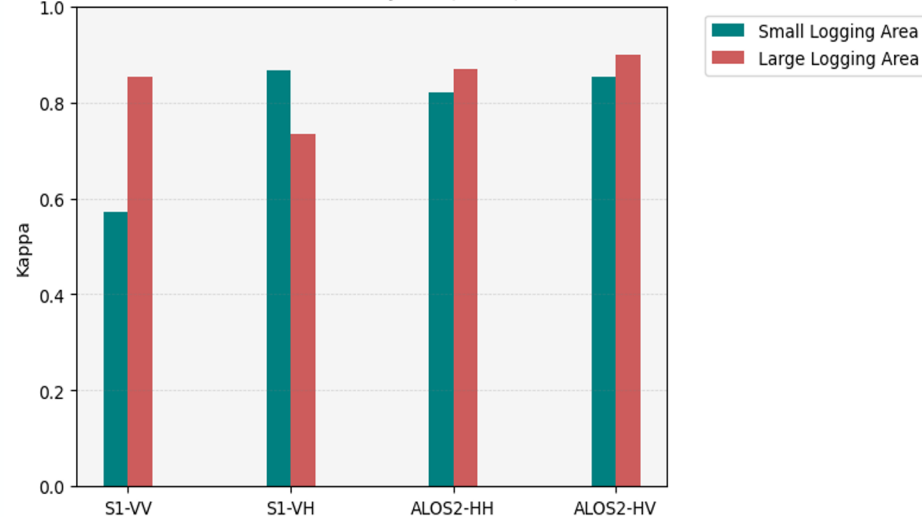
Small logging areas: 4 ha to 20 ha

Large logging areas: 52 ha to 796 ha

Kalimantan - All years (25m)



Kalimantan - All years (100m)





# Effect of logged area size

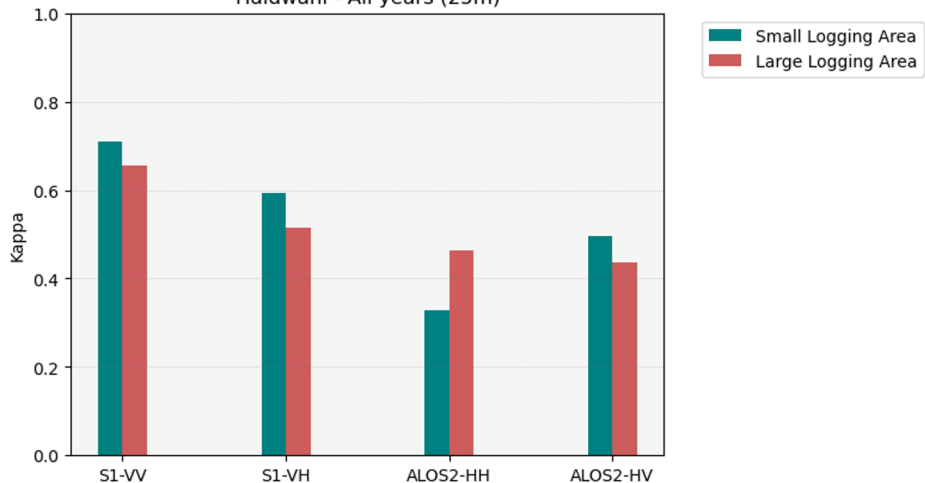
Haldwani

Two subsets chosen with different sizes of logged area

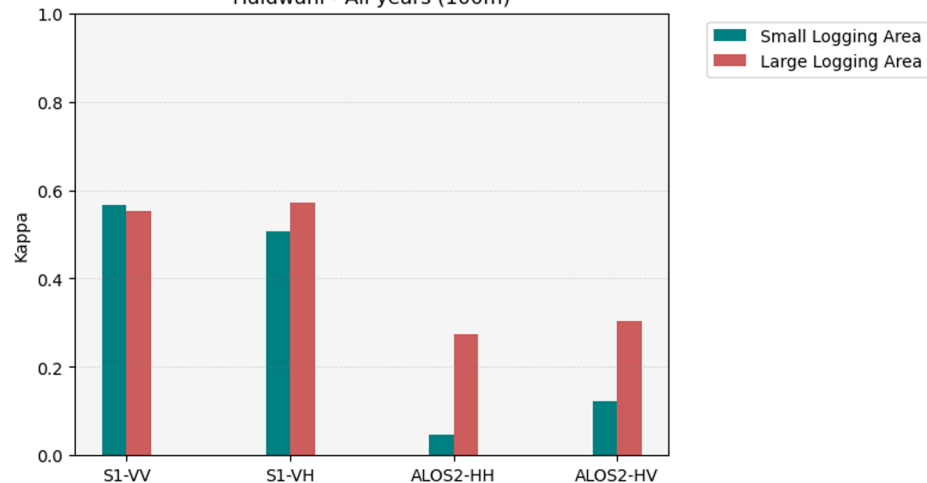
Small logging areas: 1 ha to 20 ha

Large logging areas: 30 ha to 90 ha

Haldwani - All years (25m)



Haldwani - All years (100m)



- Forest- and management-type affects the detection of deforestation/logging
  - Logged area size
  - Frequency of logging activity
- Smaller logged regions are not captured (Haldwani)
- Impact of logged area much higher in case of L-band than at C-band
- Both polarizations perform equally in case of Kalimantan
- Cross-pol backscatter performs better in managed forests



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Bengaluru, India  
December 10-13, 2023



10 – 13 December 2023, Bengaluru

Abstract Submission Deadline: 15th August 2023

<https://ingarss.org/>







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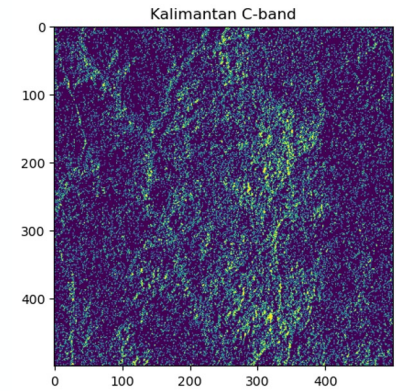
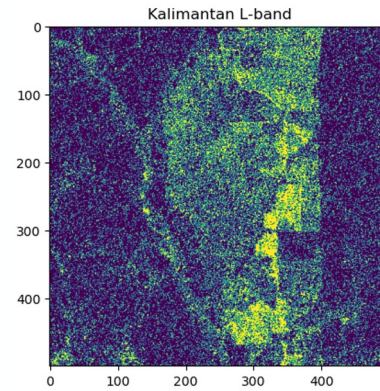
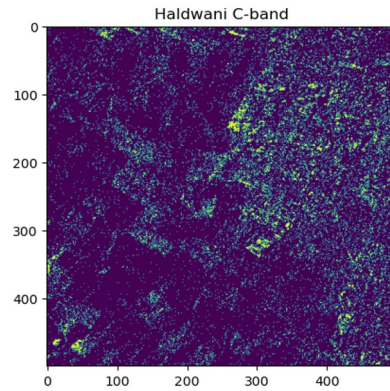
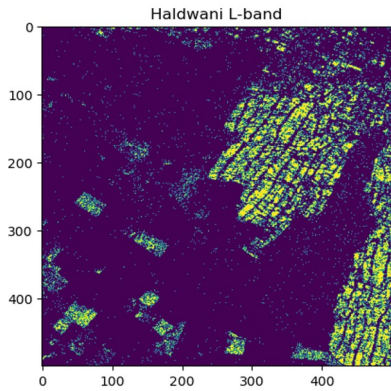
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### Sequential Omnibus



## Sequential Omnibus

