

# Creation of 10-m LULC Maps for Viet Nam from 2017 to 2022 Using a Time-Feature Convolutional Neural Network

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2024/01/31

# Our group's activities at University of Tsukuba, Japan

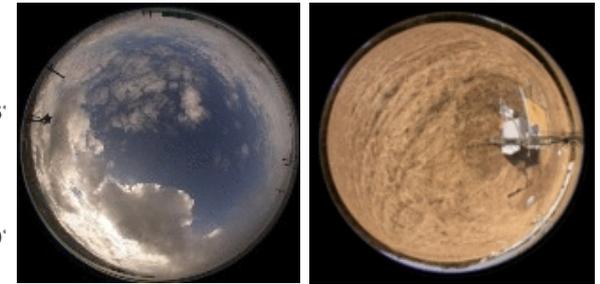
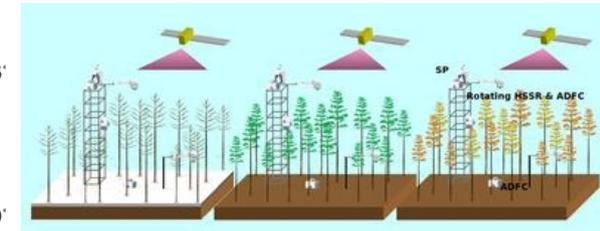
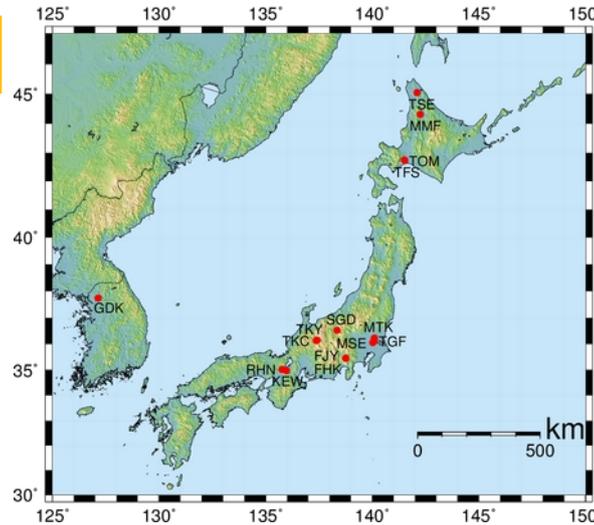
## Research topics

- Phenological Eyes Networks (PEN)
- Sensor validation tasks (GCOM-C project, JAXA)
- Land Use Land Cover mapping
  - + Viet Nam
  - + Japan
  - + Thailand
  - + Bangladesh

Our  
group



JAXA  
EORC



<http://www.pheno-eye.org/>

# LULC changes are main causes of many environmental issues

Conversion of forests to urban/built-up or cropland



Biodiversity loss, habitat disruption, exacerbated climate change, natural disasters  
(Betts et al., Nature, 2017)

Conversion of cropland to urban/built-up



Agriculture productivity, food security,  
(Anwar et al., Remote Sensing, 2023)

Decrease open land due to urbanization > increase household waste water



Reduction of ground water quality  
(Ahmad et al. Scientific Reports, 2021)

Increase of solar panel



Change the earth surface's albedo and heat exchange with atmosphere  
(brennan et al., Scientific Reports, 2014)

**LULC mapping is an important but a challenging task**

- Temporal coverage
- Spatial coverage
- Spatial resolution
- Accuracy
- Number of categories
- ...

# Existing global LULC maps

Old products

recent products



	MCD12Q1	GLC2000	GLCNMO	GlobCover	ESA-CCI	FROM_GLC	CGLS-LC100	GLC_FS30	ESRI	ESA	DW
<b>Time of Maps</b>	2001 -2020	2000	2003, 2008, 2013	2004-2006, 2009	1992 -2015	2001, 2010, 2015, 2017	2015 - 2019	2015	2017 - 2021	2020, 2021	2015 – current
<b>Number of Category</b>	17	22	20	22	22	7, 11	23	9	10	11	9
<b>Spatial Resolution (m)</b>	500	1000	1000, 500, 500	300	300	250, 250, 30, 10	100	30	10	10	10
<b>Max. Overall Acc. (%)</b>	73.6	-	77.9	67.5	75.4	75.2	80.5	82.5	75	80.7	-

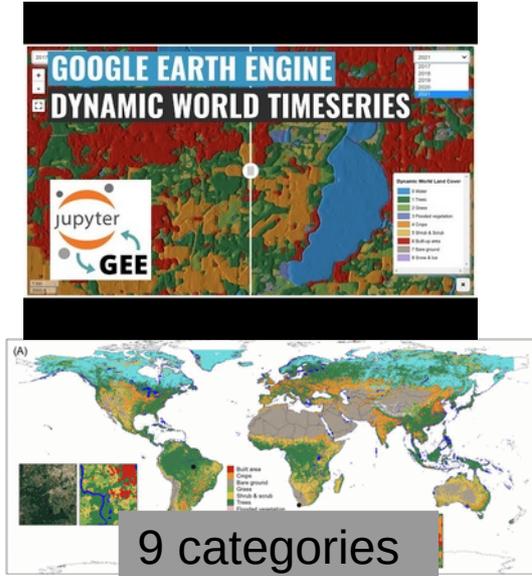
➤ Increase of spatial resolution (1000 m >>> 10 m)

A big contribution of Sentinel-2 and Sentinel-1

- MCD12Q1: MODIS Land Cover Type Product
- GLC2000: Global Land Cover
- GLCNMO: Land Cover GLCNMO Global
- GlobCover: GlobCover land Cover Maps
- ESA-CCI: ESA Climate Change Initiative Land Cover
- FROM\_GLC: Finer Resolution Observation and Monitoring-Global Land Cover

- CGLS-LC100: Copernicus Global Land Service (CGLS-LC100)
- GLC\_FS30: Global Land Cover Product with Fine Classification System
- DW: Dynamic World
- ESA: ESA World Cover
- ESRI: ESRI Land Cover

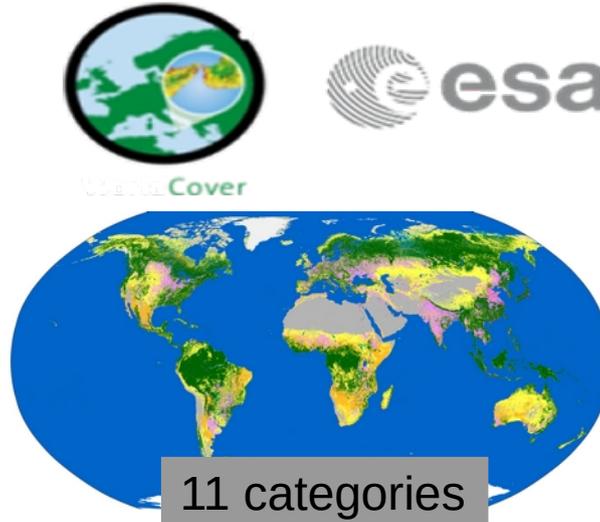
# Recent global LULC maps



9 categories

## LEGEND

- Water
- Trees
- Grass
- Flooded vegetation
- Crops
- Shrub and scrub
- Built
- Bare
- Snow and ice



11 categories

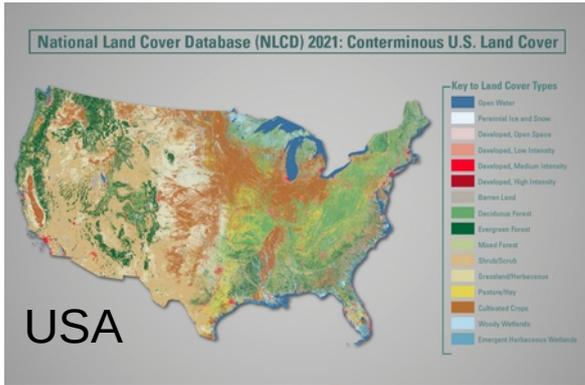
- Tree cover
- Shrubland
- Grassland
- Cropland
- Built-up
- Bare / sparse vegetation
- Snow and ice
- Permanent water bodies
- Herbaceous wetland
- Mangroves
- Moss and lichen



10 categories

- Water
- Trees
- Grass
- Flooded Vegetation
- Crops
- Scrub/Shrub
- Built Area
- Bare Ground
- Snow/Ice
- Clouds

# Some countries have produced their own LULC maps



National Land Cover Database  
30m, 16 categories



2022  
10m, 14 categories  
JAXA-EORC



2017, 10 m, 13 categories,  
<https://s2glc.cbk.waw.pl/>

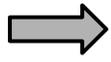
# The importance of LULC mapping for Vietnam

- Total area: **330,000 km<sup>2</sup>**
- Forest cover: **42 %** total area
- Economic growth rate: **8.0 % (2022)**
- **One of ten** countries with the richest biodiversity

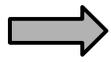


- To 2030: **reduce greenhouse emissions by 70%** and **increase carbon sinks by 20%** compare to business-as-usual scenario (BAU)

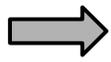
LULC maps



Land use planning and management



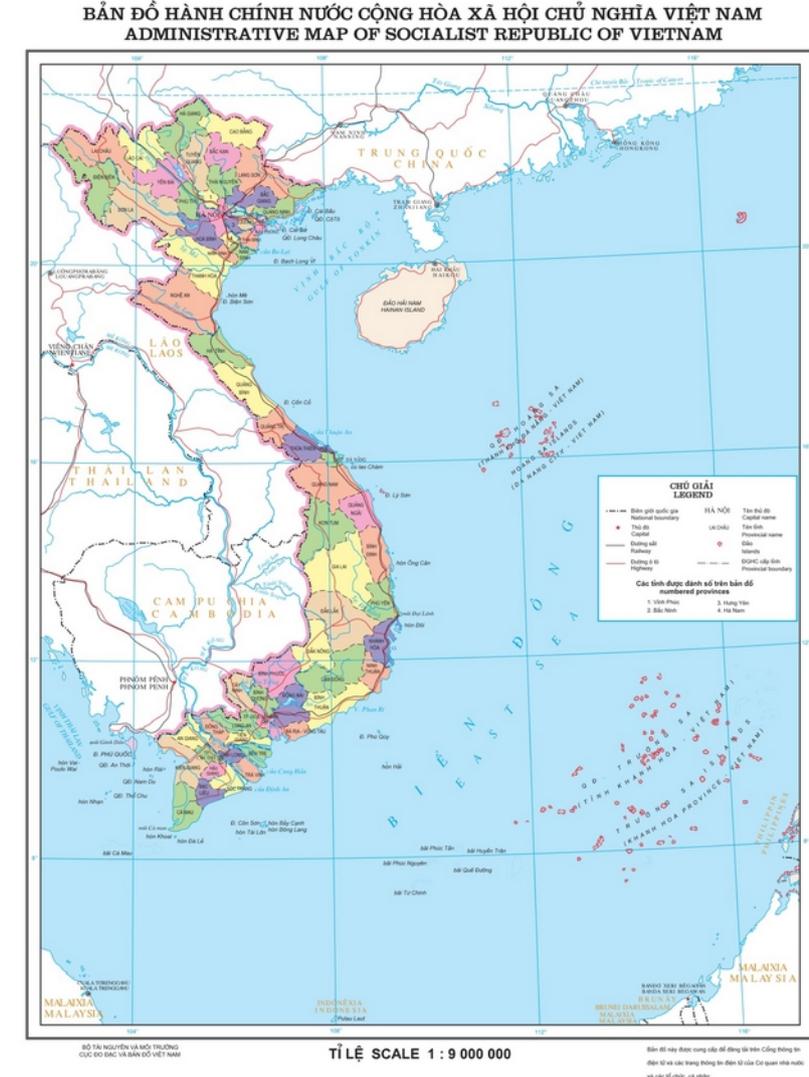
Biodiversity evaluation and conservation



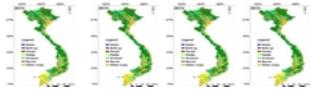
SDG of Forest and Land use sector



Various applications: biomass estimation, disaster countermeasure, REDD+, etc.

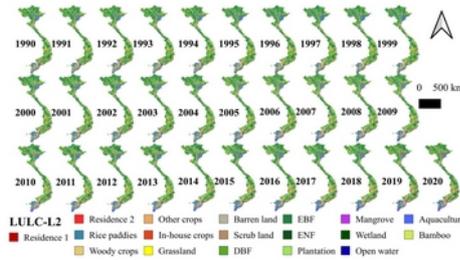


# Existing Nation-wide LULC maps of Vietnam



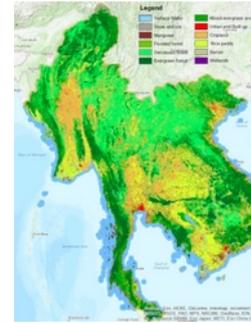
50 m (2015 - 2018)

7 categories,  
OA: 79%

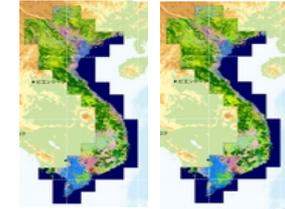


30 m (1990 - 2020)

17 categories,  
OA: 77.6% -  
84.7%



30 m (2000 - 2017)  
18 categories  
OA: 94%



10 m (2007 & 2016)  
12 categories,  
OA: ~85%

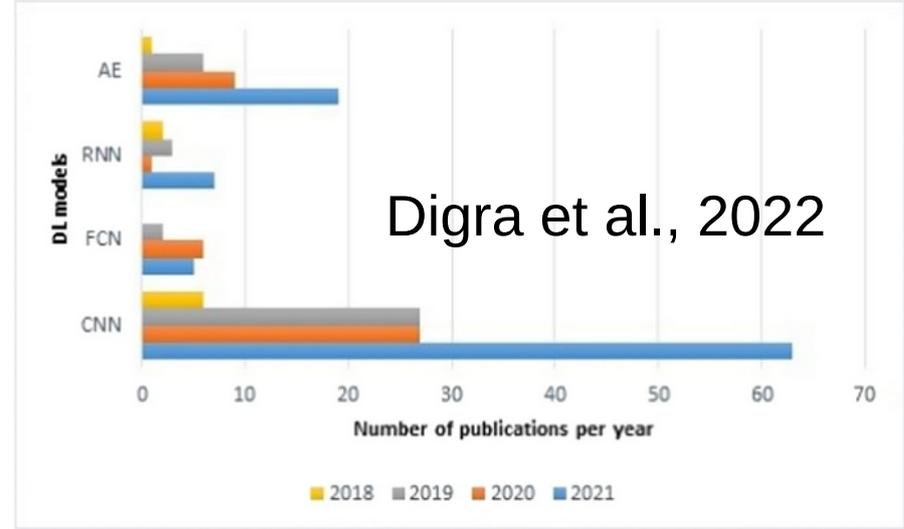
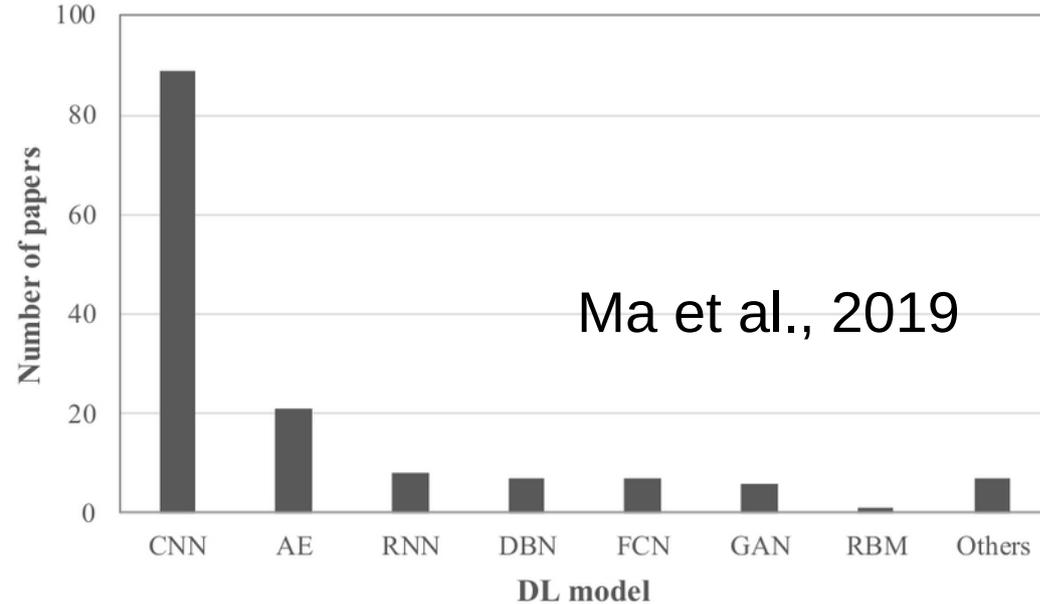
**This study's plan**

10 m (2017- 2022)

> 12 categories,  
OA: > 90%

# The selection of classification algorithm

Convolutional Neural Network (CNN) has been the most frequently used method for remote sensing applications



CNN: Convolutional Neural Network

AE: Auto-encoder

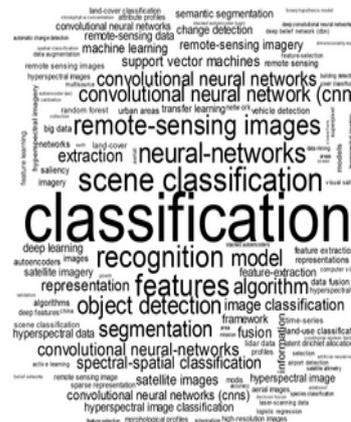
RNN: Recurrent Neural Network

DBN: Deep Belief Network

FCN: Fully Convolutional Network

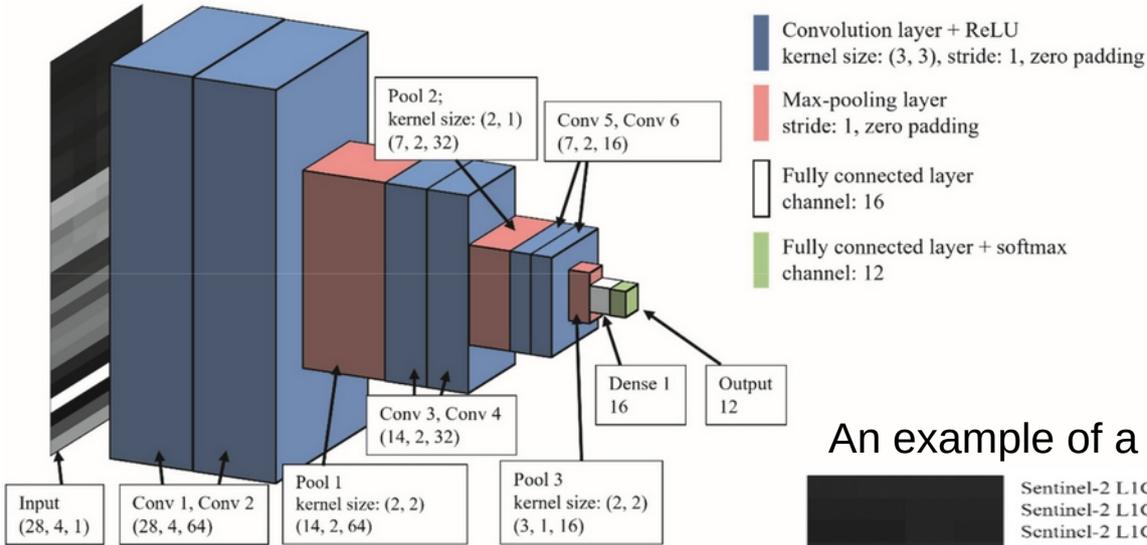
GAN: Generative Adversarial Network

RBM: Restricted Boltzmann Machine



(Ma et al., 2019)  
**Classification task is the main application of CNN**

# JAXA has been developing a new CNN for using multi-temporal data



Structure of CNN model used for producing LULC map of Japan v.21.11

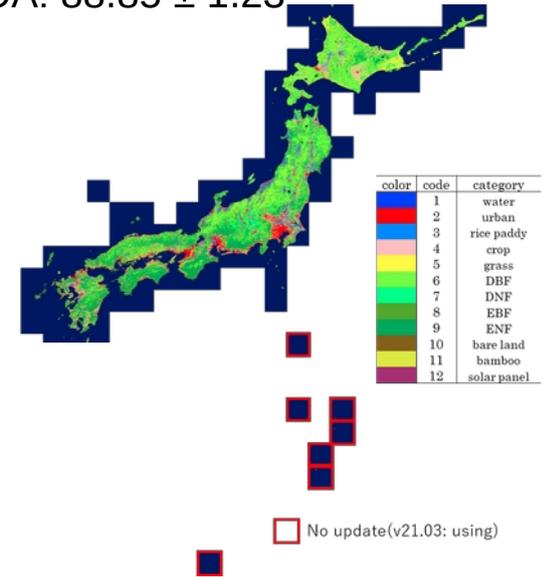
(Hirayama et al., 2022)

An example of a training data

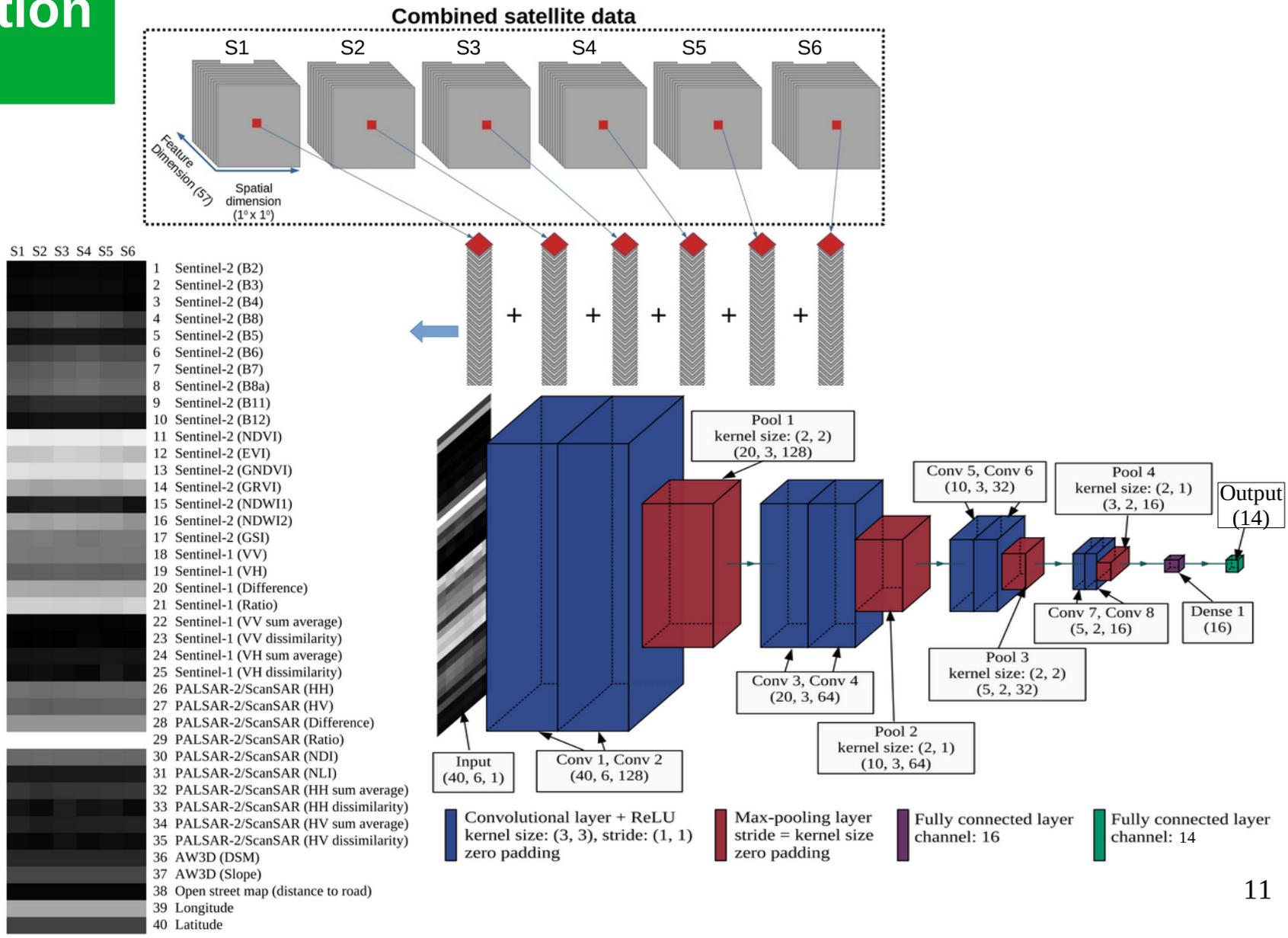


LULC map of Japan ver. 21.11  
12 categories

OA:  $88.85 \pm 1.23$



# Classification Algorithm

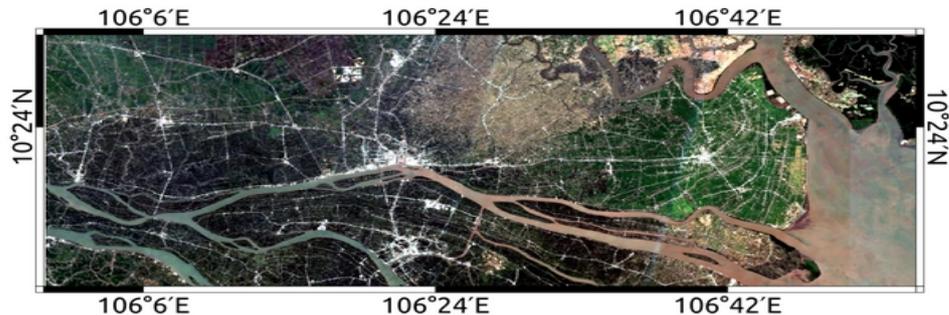


# Satellite data used to produce time-series LULC maps

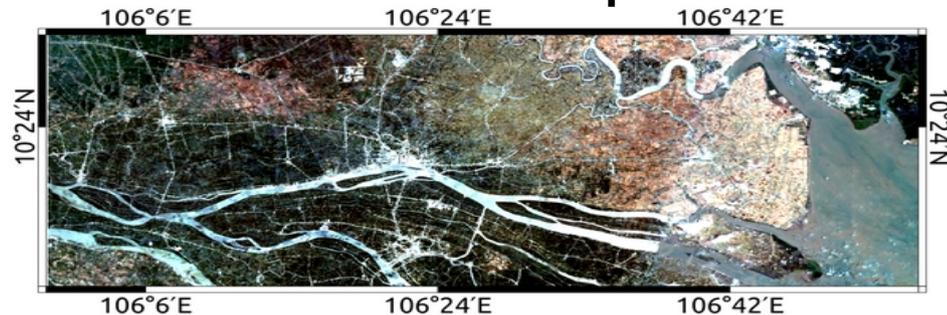
No	Data	Spatial resolution / bands	Time/Seasons
1	Sentinel-2 (level 2A, level 1C)	10 m (B2, B3, B4, B8) 20 m (B5, B6, B7, B8A, B11, B12) NDVI, EVI, GNDVI, GRVI, NDWI1, NDWI2, GSI	S1. 20XX/01/01-20XX/02/29 S2. 20XX/03/01-20XX/04/30 S3. 20XX/05/01-20XX/06/30 S4. 20XX/07/01-20XX/08/31 S5. 20XX/09/01-20XX/10/31 S6. 20XX/11/01-20XX/12/31 (XX: 17-22)
2	Sentinel-1 GRD	10 m (VV, VH, VV-VH, VV/VH, VV_savg, VV_diss, VH_savg, VH_diss)	
3	PALSR-2/ScanSAR	25 m (HH, HV, HH-HV, HH/HV, NDI, NLI, HH_savg, HH_diss, HV_savg, HH_diss)	
4	ALOS AW3D DSM	30 m (DSM, Slope)	-
5	OpenstreetMap	Distance to Road	2017-2022

# The problem of cloud in Sentinel-2 images in Vietnam

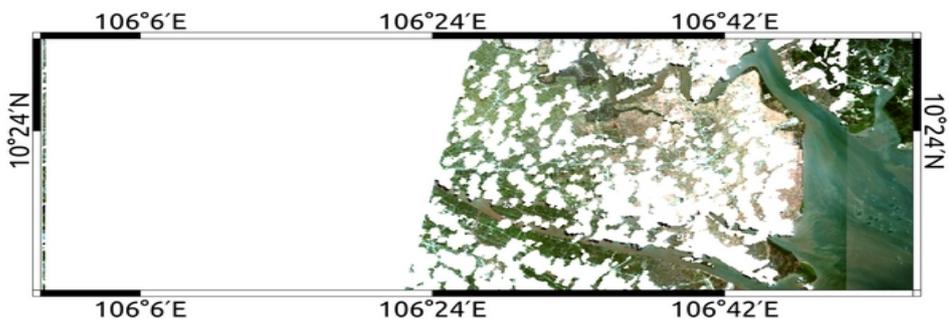
**S1: Jan. & Feb.**



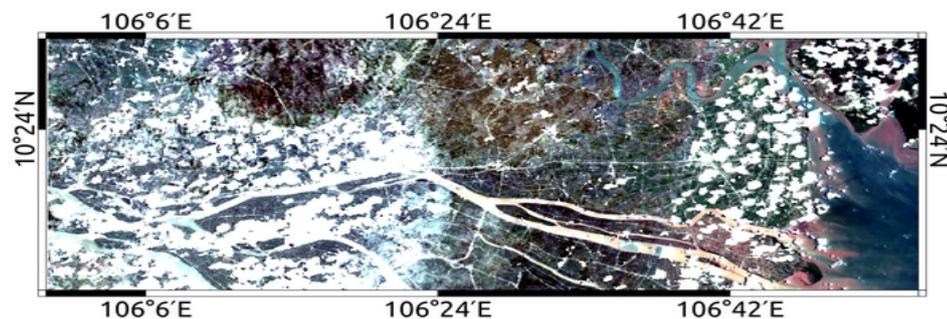
**S2: Mar. & Apr.**



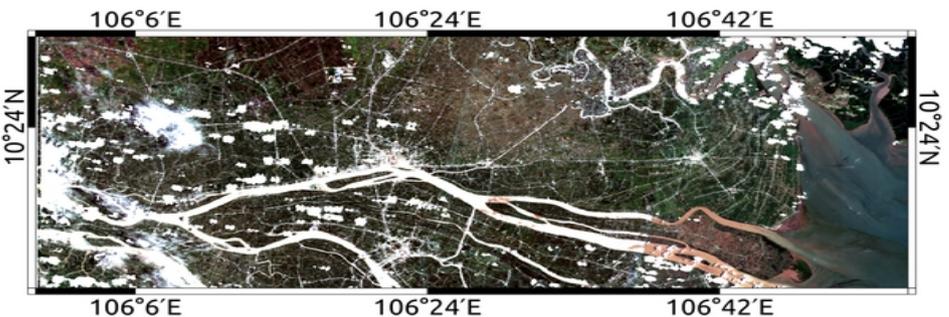
**S3: May & June**



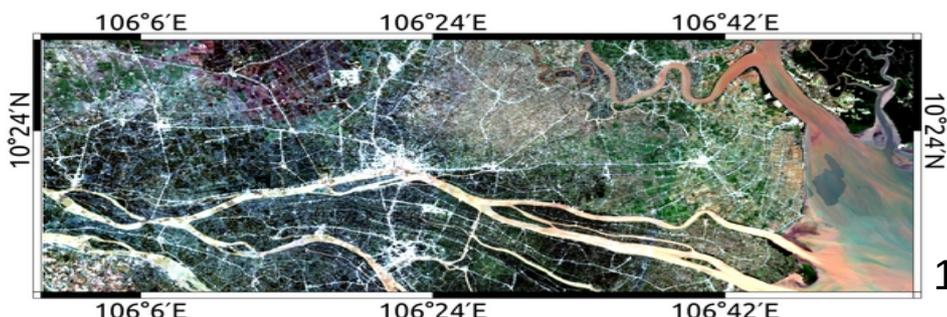
**S4: July & Aug.**



**S5: Sep. & Oct.**



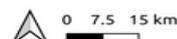
**S6: Nov. & Dec.**



# Example of making cloud-free time-series images for 2020

Images used

Normal approach

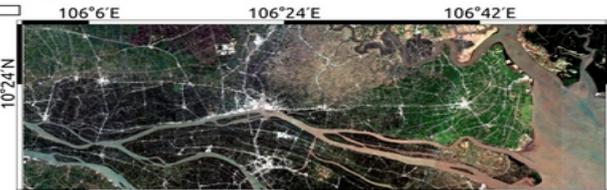


Our approach

Images used

2020/01  
2020/02

S1



2020/01 x 2, 2020/02 x 2  
2020/03 x 1  
2021/01 x 1, 2021/02 x 1

2020/03  
2020/04

S2



2020/03 x 2, 2020/04 x 2  
2020/02 x 1, 2020/05 x 1  
2021/03 x 1, 2021/04 x 1

2020/05  
2020/06

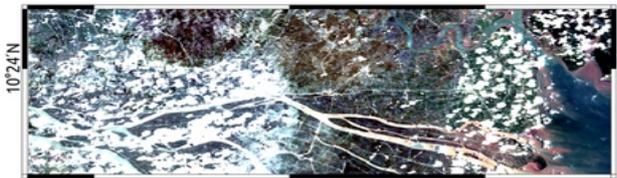
S3



2020/05 x 2, 2020/06 x 2  
2020/04 x 1, 2020/07 x 1  
2021/05 x 1, 2021/06 x 1  
2021/04 x 1, 2021/07 x 1

2020/07  
2020/08

S4



2020/07 x 2, 2020/08 x 2  
2020/06 x 1, 2020/09 x 1  
2021/07 x 1, 2021/08 x 1  
2021/06 x 1, 2021/09 x 1

2020/09  
2020/10

S5



2020/09 x 2, 2020/10 x 2  
2020/08 x 1, 2020/11 x 1  
2021/09 x 1, 2021/10 x 1  
2021/08 x 1, 2021/11 x 1

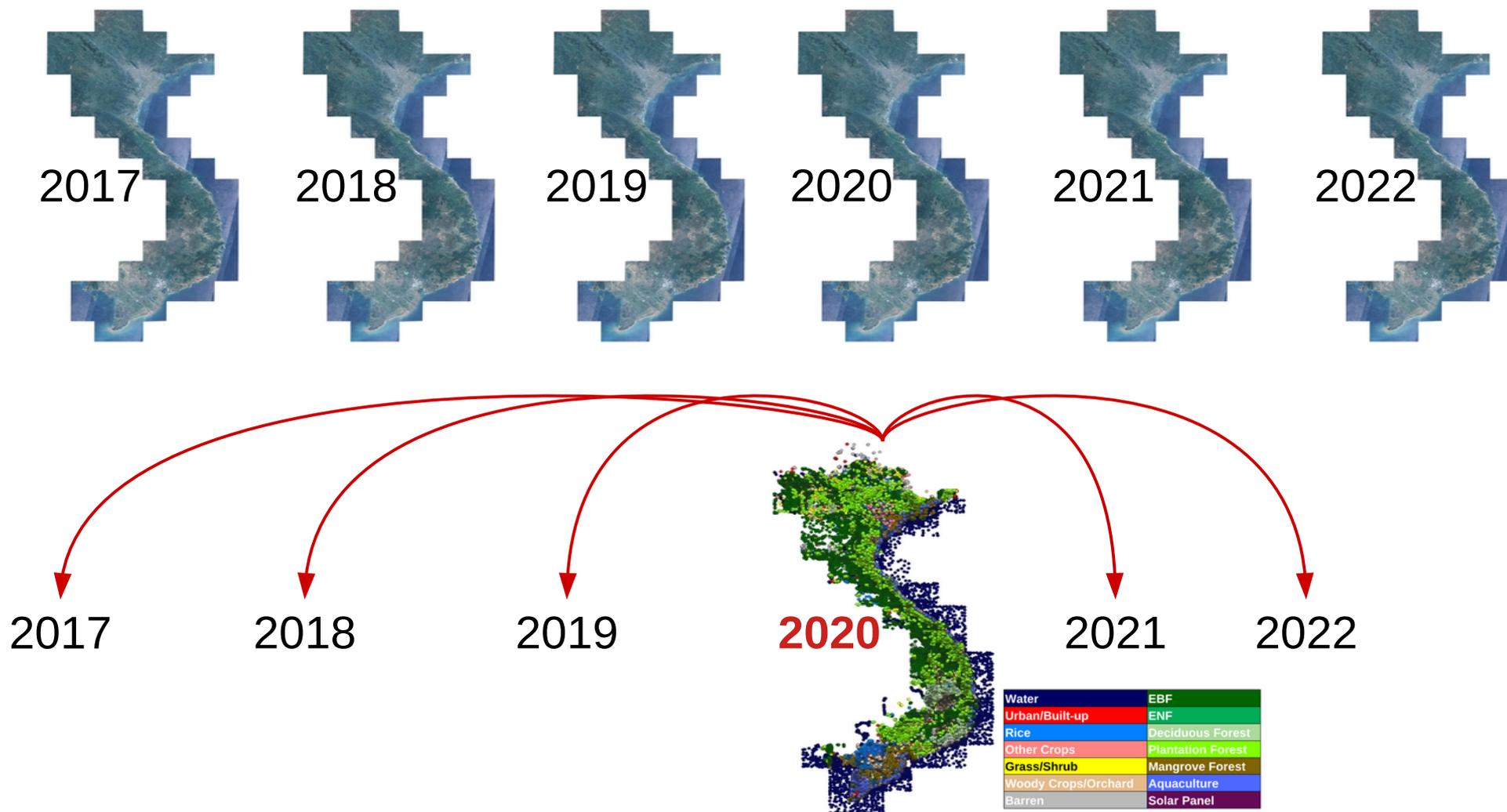
2020/11  
2020/12

S6



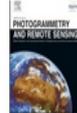
2020/11 x 2, 2020/12 x 2  
2020/10 x 1  
2021/11 x 1, 2021/12 x 1

# Producing time series of reference data



170,000 reference data for 2020 have been collected by visual interpretation on Google earth, Sentinel-2, Planet's scope

# Method for reference data migration



## The migration of training samples towards dynamic global land cover mapping

Huabing Huang<sup>a,b,c,\*</sup>, Jie Wang<sup>b,d</sup>, Caixia Liu<sup>b</sup>, Lu Liang<sup>e</sup>, Congcong Li<sup>f</sup>, Peng Gong<sup>d,\*</sup>

<https://doi.org/10.1016/j.isprsjprs.2020.01.010>

Received 21 October 2018; Received in revised form 6 December 2019; Accepted 9 January 2020

2020 ( $t_1$ )



$X_{i(t_1)}$

$X_1$

$X_2$

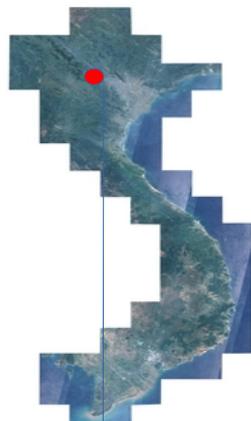
⋮

⋮

⋮

$X_N$

Target year ( $t_2$ )



$Y_{i(t_2)}$

$Y_1$

$Y_2$

⋮

⋮

⋮

$Y_N$

$$\theta = \cos^{-1} \frac{\sum_{i=1}^N X_{i(t_1)} Y_{i(t_2)}}{\sqrt{\sum_{i=1}^N (X_{i(t_1)})^2 \sum_{i=1}^N (Y_{i(t_2)})^2}},$$

$$SAD = \cos(\theta) \quad (1)$$

$$ED = \sqrt{\sum_{i=1}^N (X_{i(t_1)} - Y_{i(t_2)})^2} \quad (2)$$

SAD: Spectral similarity

SAD  $\rightarrow$  1 : LULC no change

ED: Spectral distance

ED  $\rightarrow$  0 : LULC no change

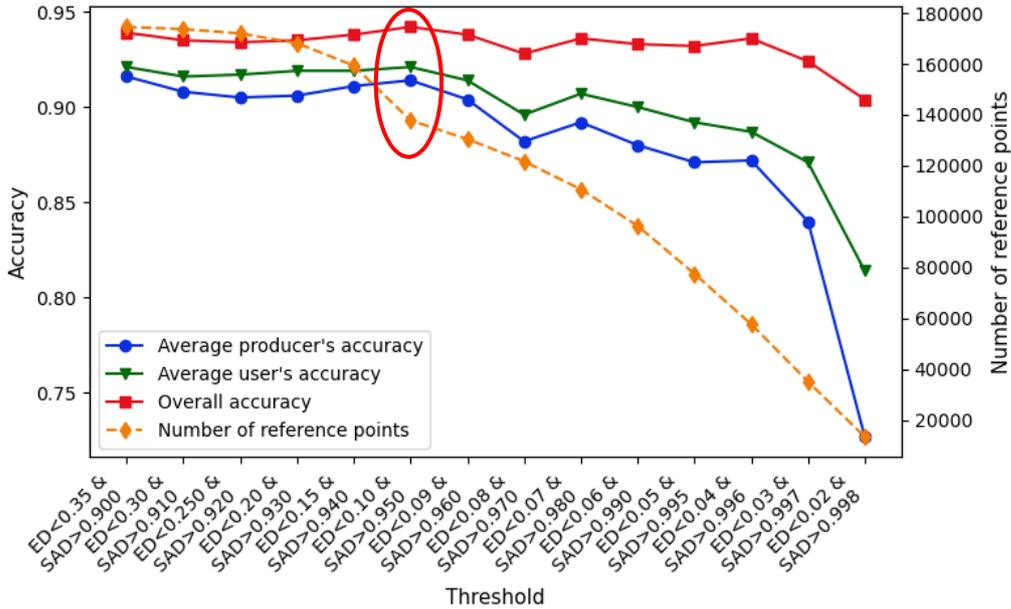
$i$ : is the target pixel (red point)

$X_{i(t_1)}$  : is reference spectra at time  $t_1$

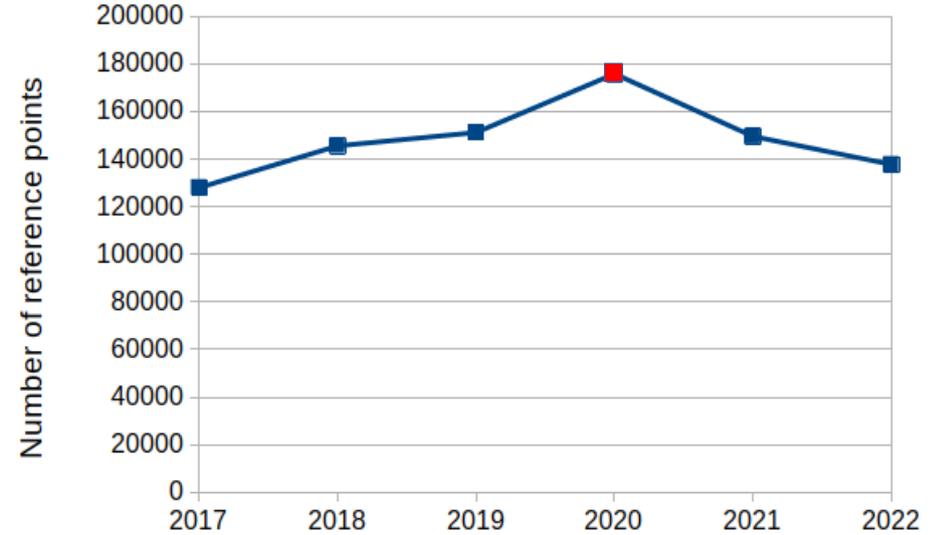
$Y_{i(t_2)}$  : is the target spectra at time  $t_2$

$N$ : is number of bands

# The selection of threshold value for reference data migration

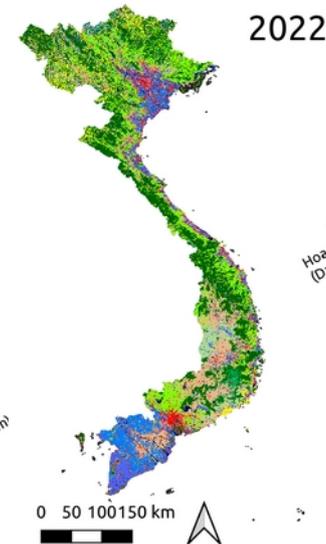
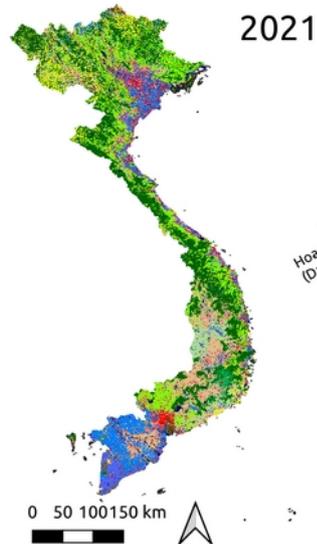
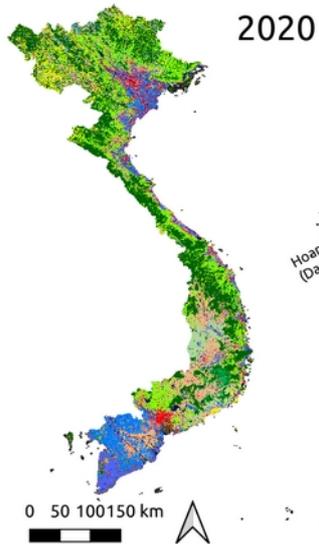
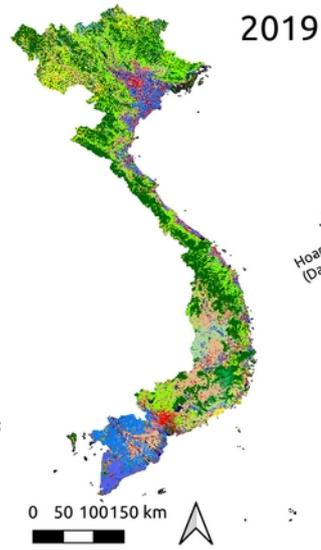
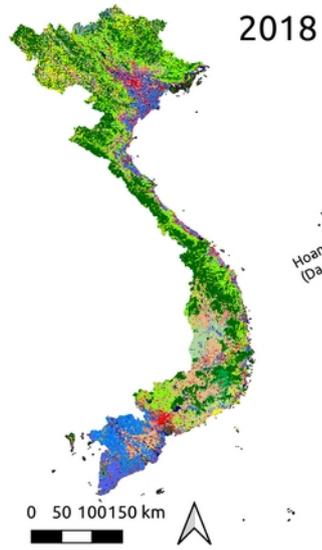
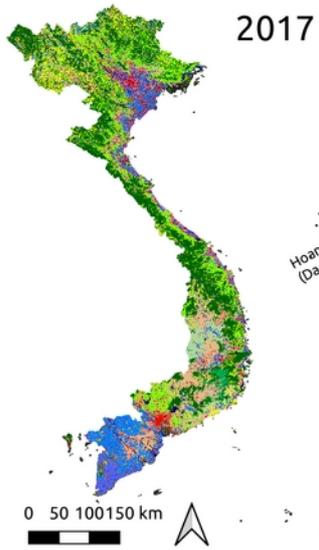


Variation of accuracy by thresholds value sets of ED and SAD



Number of reference data by years  
Selected thresholds: ED < 0.1 & SAD > 0.95

# 14-category LULC maps for Vietnam from 2017-2022



	Overall Acc.	User's Acc.	Prod.'s Acc
<b>2017</b>	92.3 ± 0.2	92.9 ± 0.3	86.9 ± 0.8
<b>2018</b>	92.4 ± 0.2	92.7 ± 0.9	86.6 ± 0.7
<b>2019</b>	92.7 ± 0.2	93.2 ± 0.2	88.8 ± 0.4
<b>2020</b>	92.4 ± 0.1	93.0 ± 0.2	89.3 ± 0.5
<b>2021</b>	92.4 ± 0.2	93.0 ± 0.2	91.2 ± 0.3
<b>2022</b>	91.5 ± 0.2	92.0 ± 0.2	87.3 ± 0.9

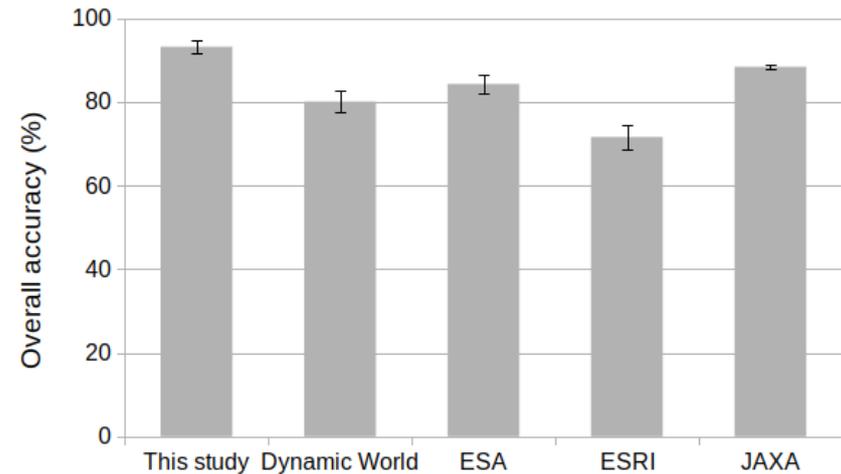
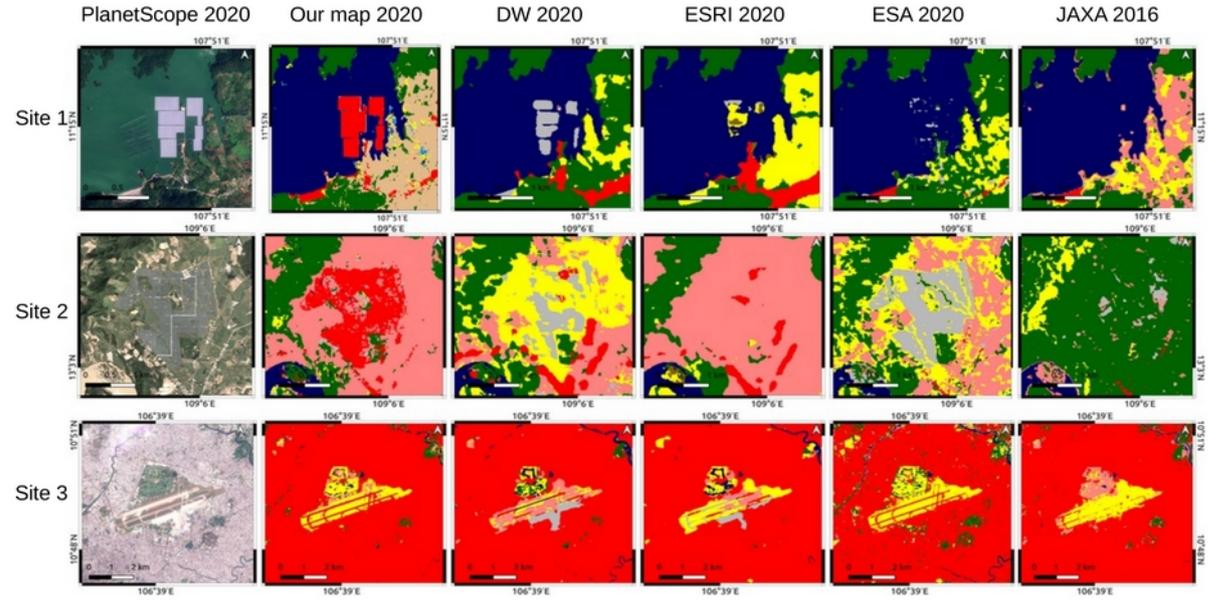
Water	EBF
Urban/Built-up	ENF
Rice	Deciduous Forest
Other Crops	Plantation Forest
Grass/Shrub	Mangrove Forest
Woody Crops/Orchard	Aquaculture
Barren	Solar Panel

# A comparison with other LULC products



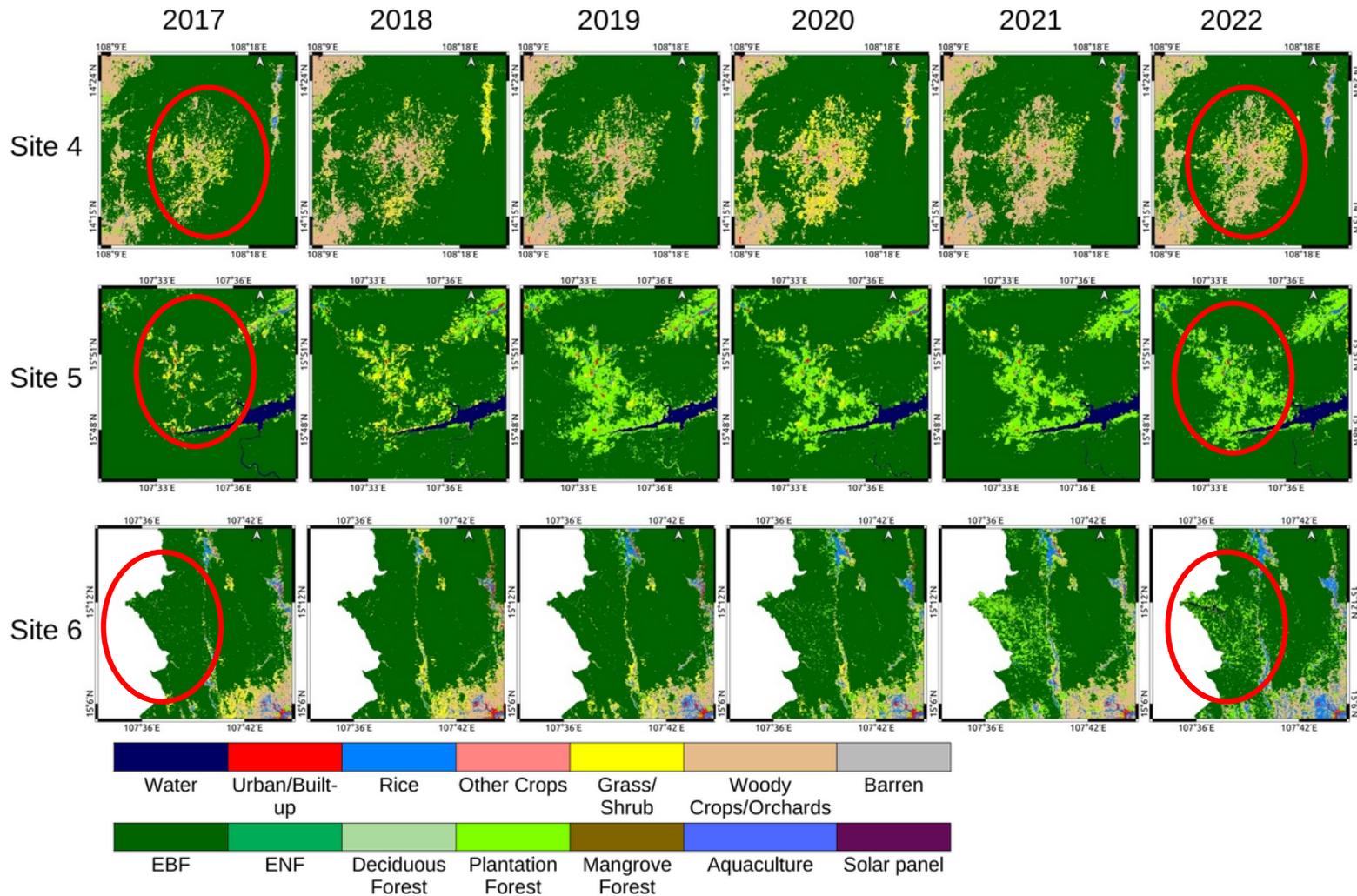
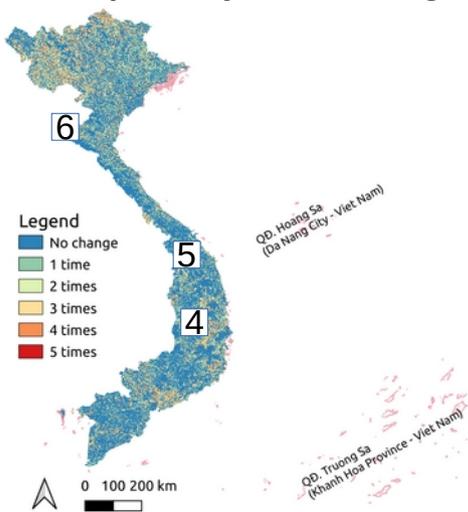
Hoang Sa Islands  
(Da Nang City - Viet Nam)

Truong Sa Islands  
(Khanh Hoa Province - Viet Nam)

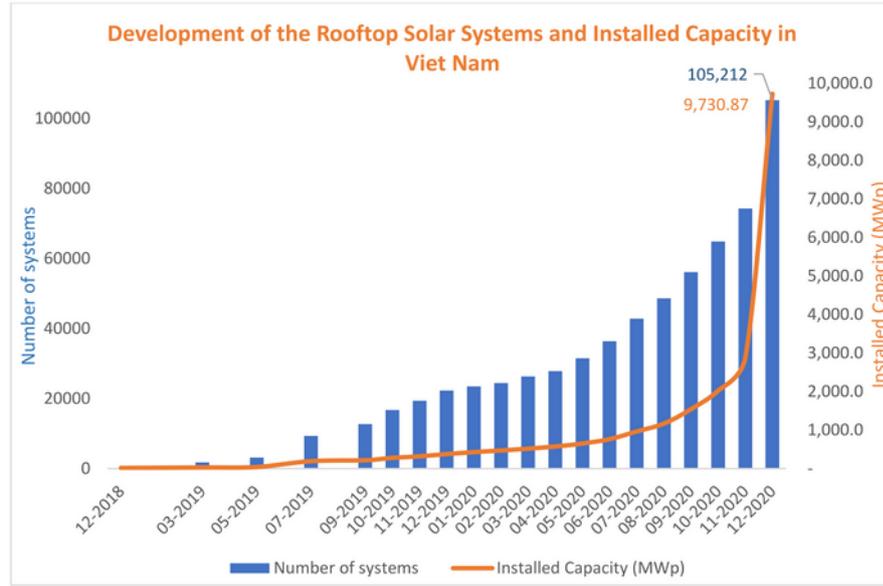
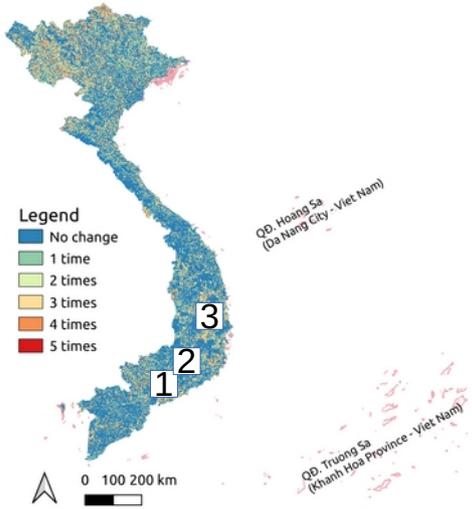


# Examples of LULC changes in Vietnam from 2017-2022

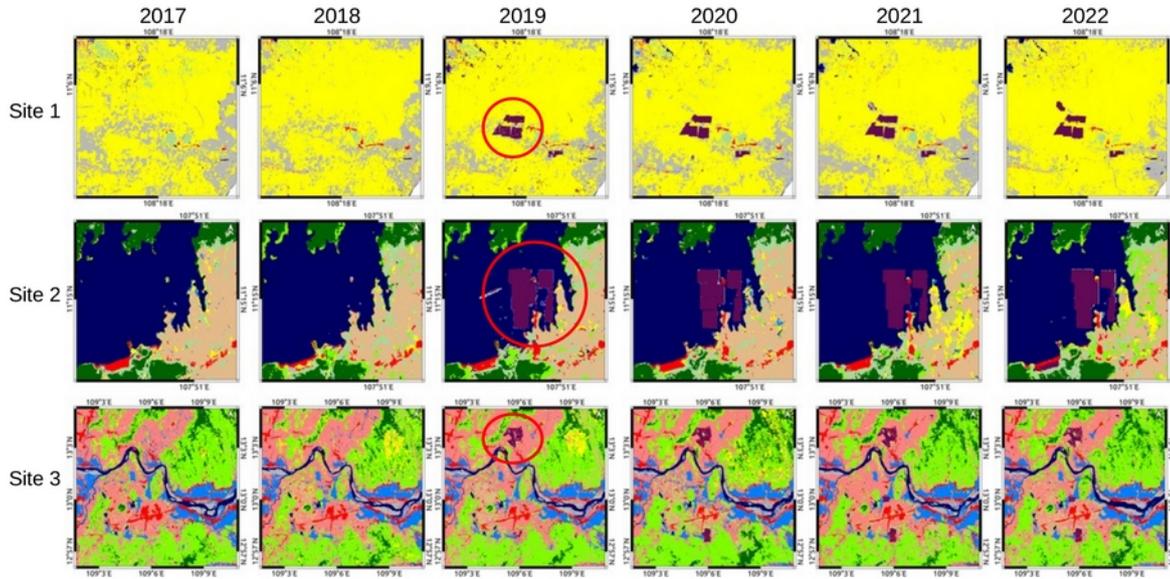
## Frequency of change



# The rapid increase of solar panel facilities

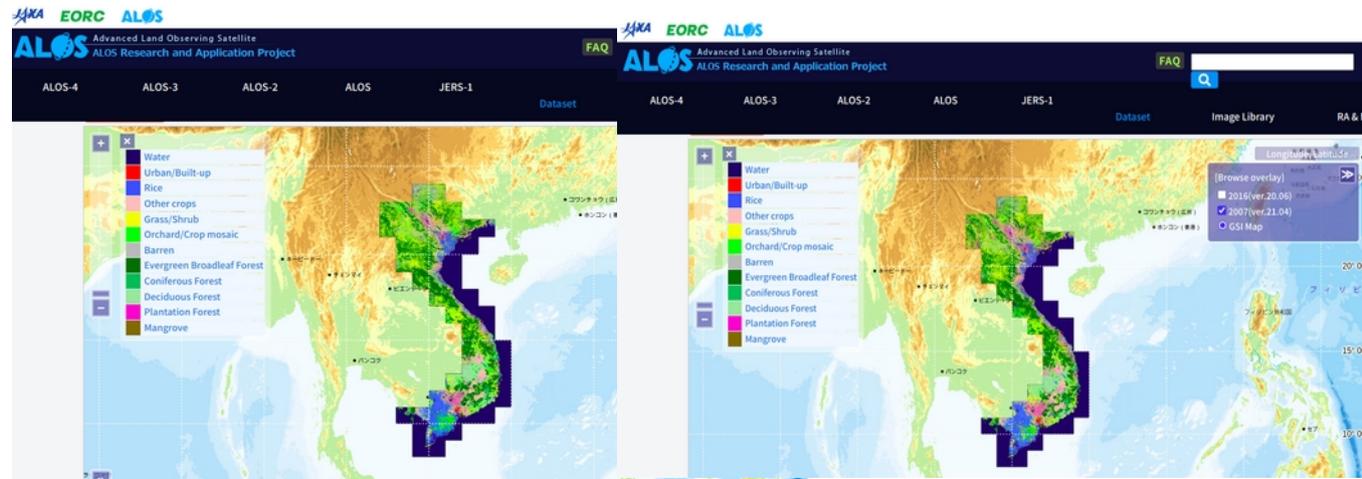
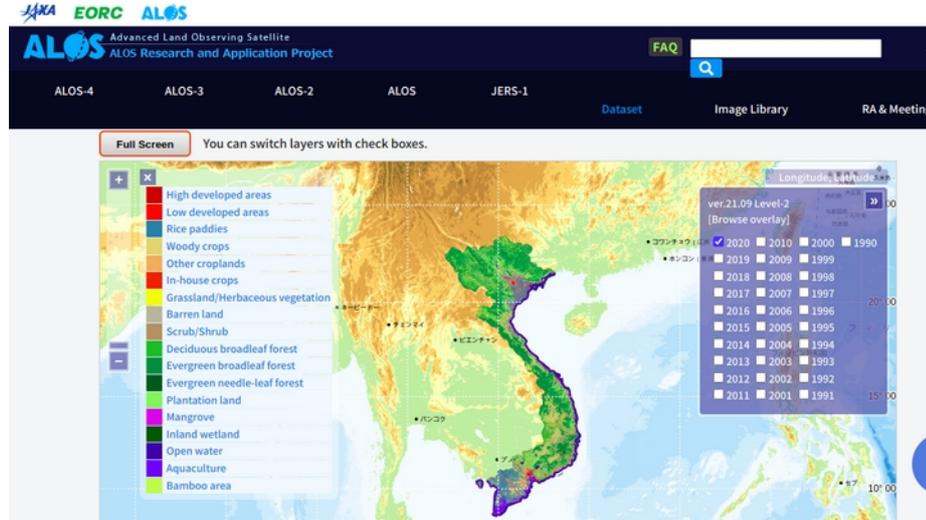
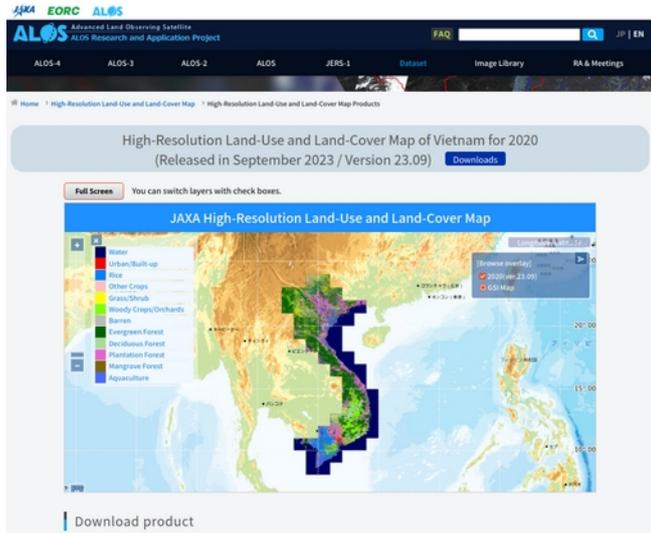


Reported by Viet Nam Energy Partnership Group, 2020



Water	EBF
Urban/Built-up	ENF
Rice	Deciduous Forest
Other Crops	Plantation Forest
Grass/Shrub	Mangrove Forest
Woody Crops/Orchard	Aquaculture
Barren	Solar Panel

# Published Data on JAXA's website



# Conclusions and Recommendations

- The new CNN model showed good capability to produce a high accuracy LULC maps, especially, in preserving the shape of objects
- Accuracy assessment using ~40,000 points showed high accuracy of more than 90% for all maps from 2017-2022.
- Deforestation was detected in some places, corresponding with the fact that deforestation has been occurring in Vietnam.
- Solar panel facilities has been rapidly increasing since 2019

## **Suggestions for future studies:**

- More investigation on the contributed factors to accuracy of LULC maps, such as temporal profile of satellite data, arrangement of feature (bands, indices), number of features, combination of different satellite images.
- The contribution of number of training data to accuracy of LULC maps
- Improving the classification algorithm such as: model structure, model's-parameters optimization



*Thank you for listening*

HOPE