

National Cropland Classification with Agriculture Census Information and EO Datasets

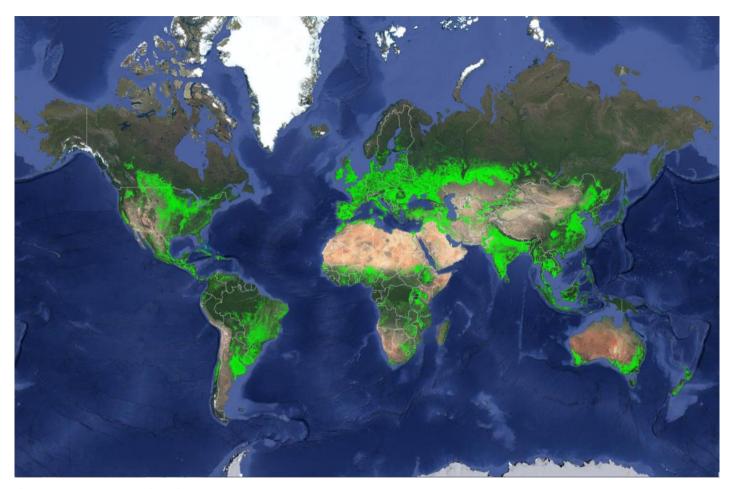
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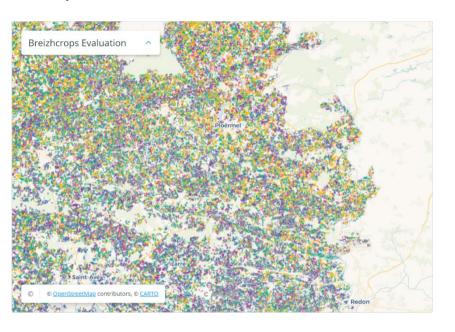


Global Food Security -> Global cropland mapping

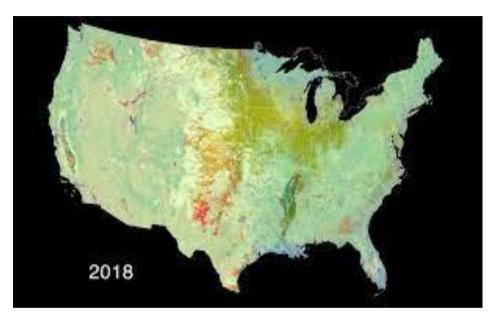


https://www.croplands.org/app/map.

Cropland datasets



BreizhCrops https://breizhcrops.org/

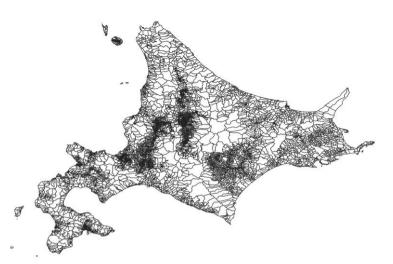


Cropland Data Layer (CDL) of USA https://www.nass.usda.gov/Research_and_Science/Cropla_nd/Release/index.php

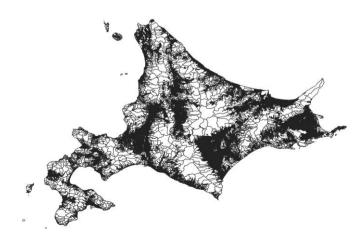
Objective: Japan Cropland with Census information and EO datasets



Source from Wiki



Agricultural rural communities (7295)



Crop field boundary (~900K) from MAFF

	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
01Hokkaido												
02Aomori												
03Iwate												
04Miyagi												
05Akita												
06Yamagata												
07Fukushima												
08Ibaraki												
09Tochigi												
10Gunma												
11Saitama												
12Chiba												
13Tokyo												
14Kanagawa												
15Niigata												
16Toyama												
17Ishikawa												
18Fukui												
19Yamanashi												
20Nagano												
20Nagano 21Gifu												
22Shizuoka												
225nizuoka												
23Achi												
24Mie												
25Shiga												
26Kyoto												
27Osaka												
28Hyogo												
29Nara												
30Wakayama												
31Tottori												
32Shimane												
33Okayama												
34Hiroshima												
35Yamaguchi												
36Tokushima												
37Kagawa												
38Ehime												
39Kochi												
40Fukuoka												
41Saga												
42Nagasaki												
43Kumamoto												
44Oita												
45Miyazaki												
45Miyazaki 46Kagoshima												
46Kagosnima 47Okinawa												
4/Okinawa												

Hokkaido





2015: Apr-May (L8)

2015: Aug-Sep (L8)

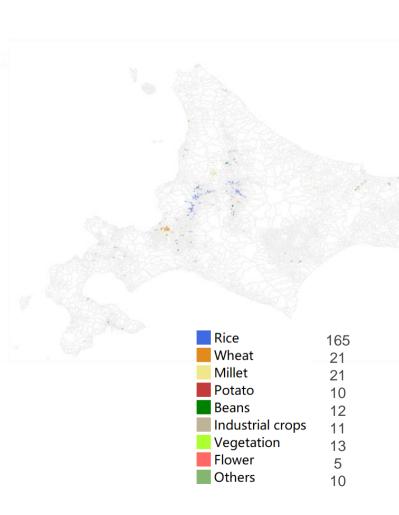
Hokkaido



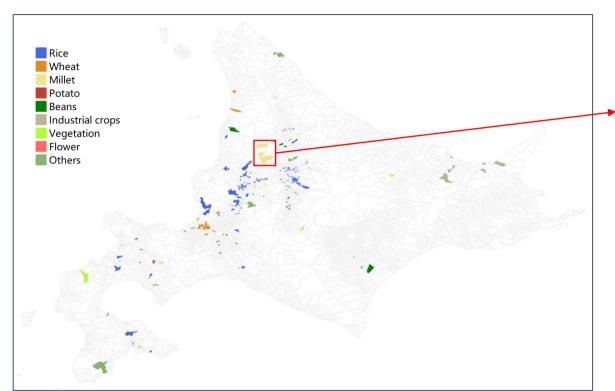


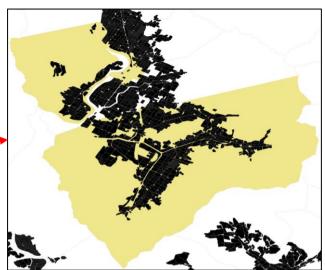
2015: Apr-May (S1)

Pure Labels from Census Data



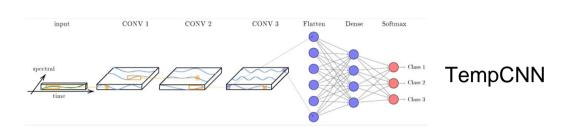
- Pure ratio = (sale class area) 📫 (crop area)
- Ratio threshold definition
 - Rice > 0.8 or 0.9 (except Tokyo and Okinawa)
 - Others > 0.2-0.9 based on the numbers
 - top 10 or top 5 according to the small numbers
 - Only few (2-5) numbers that all selected
 - Selected the largest numbers

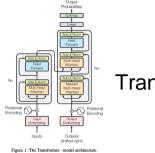




	Rice	Wheat	Millet	Potato	Beans	Industrial	Vegetation	Flower	Others
01Hokkaido	>0.8	>0.5	>0.5	Top 10	>0.5	>0.5	>0.5	Top 5	Top 10
02Aomori	>0.8	>0.4	Top 5	Top 5	Top 5	>0.1	Top 5	>0.1	>0.3
03Iwate	>0.8	Only 2	>0.4	Top 5	Top 5	Top 5	>0.1	>0.3	Top 5
04Miyagi	>0.9	Only 4	>0.4	>0.2	Top 5	>0.1	Top 5	>0.2	>0.3
05Akita	>0.8	Only 2	Top 5	Top 5	>0.1	Top 5	>0.1	Top 5	>0.3
06Yamagata	>0.9	Top 5	>0.4	Top 5	Top 5	>0.2	>0.1	>0.1	Top 5
07Fukushima	>0.9	Only 2	>0.4	Top 5	Top 5	>0.1	>0.1	>0.1	>0.3
08Ibaraki	>0.9	Top 5	>0.4	Top 5	Top 5	>0.1	>0.1	>0.1	>0.3
09Tochigi	>0.8	>0.67	>0.45	Top 5	Top 5	>0.1	>0.1	>0.1	>0.3
10Gunma	>0.8	>0.6	Only 5	Top 5	Top 5	>0.6	Top 10	Top 5	Top 5
11Saitama	>0.9	>0.6	Only 7	Top 5	>0.1	>0.6	>0.5	>0.6	Top 10
12Chiba	>0.9	Top 5	Only 1	>0.5	Top 10	Only 2	Top 10	>0.5	>0.5
13Tokyo	>0.1	Only 4	None	>0.1	Top 5	>0.2	>0.1	>0.2	Top 5
14Kanagawa	>0.6	Only 2	Only 1	Top 5	Top 5	Top 10	Only 4	>0.1	Only 4
15Niigata	>0.9	Only 2	Top 5	Only 4	Top 5	Top 5	Top 5	Top 8	>0.2
16Toyama	>0.9	Top 10	Top 5	Top 5	Top 10	None	Only 4	Top 5	Top 5
17Ishikawa	>0.9	Top 10	Top 5	Top 6	Top 10	Only 1	Only 4	Top 5	Top 4
18Fukui	>0.9	Top 10	Top 5	Top 5	Top 9	None	Top 5	Only 2	Only 4
19Yamanashi	>0.7	Only 4	Top 10	Top 5	Top 5	> 0.1	Top 5	Only 2	None
20Nagano	>0.8	Top 10	>0.5	Top 5	Top 5	Top 5	Top 10	Top 10	Top 5
21Gifu	>0.8	>0.25	Top 5	Top 5	Top 5	>0.6	Top 5	Top 5	>0.1
22Shizuoka	>0.8	Top 5	Only 2	>0.3	Only 3	>=1	>0.6	>0.5	Top 10
23Achi	>0.9	>0.4	Only 2	Top 5	>0.3	>0.4	>0.3	>0.4	Top 5
24Mie	>0.9	>0.4	None	Top 5	>0.3	>0.6	Only 4	>0.3	Top 5
25Shiga	>0.9	>0.3	Top 5	Only 5	Top 5	>0.3	Only 3	Only 2	>0.1
26Kyoto	>0.8	Only 5	Only 5	Top 5	Top 5	>0.6	Top 5	Top 5	Top 5
27Osaka	>0.8	None	None	Top 5	Top 5	None	Only 1	>0.4	Only 2
28Hyogo	>0.9	>0.2	Top 5	Top 5	>0.3	Top 5	>0.1	>0.4	Top 5
29Nara	>0.8	Only 4	Only 2	Top 5	Only 2	>0.4	Only 1	>0.4	None
30Wakayama	>0.8	None	None	>0.4	Only 5	>0.4	>0.6	>0.3	Only 2
31Tottori	>0.8	Only 2	>0.2	Top 5	Top 5	Top 5	>0.6	Top 2	>0.6
32Shimane	>0.9	Only 2	>0.2	Top 5	Top 5	Top 5	Top 5	Top 5	>0.0
33Okayama	>0.9	>0.4	>0.2	Top 5	>0.3	>0.1	>0.1	Top 5	>0.1
34Hiroshima	>0.9	Only 1	>0.2	>0.3	>0.1	>0.1	>0.1	Only 4	>0.1
35Yamaguchi	>0.9	>0.1	Only 3	Top 5	>0.1	>0.3	>0.1	>0.1	Top 5
36Tokushima	>0.9	Only 3	Only 3	>0.9	>0.1	>0.2	>0.6	>0.1	>0.1
37Kagawa	>0.9	>0.4	None	>0.2	Only 5	>0.3	>0.4	>0.3	>0.1
38Ehime	>0.8	>0.4	Only 3	Top 5	Top 5	Top 5	>0.2	>0.3	Top 5
39Kochi	>0.9	None	None	>0.1	Only 4	>0.6	>0.2	>0.2	>0.2
40Fukuoka	>0.9	>0.9	Only 2	Top 5	>0.5	>0.6	>0.4	>0.4	>0.2
41Saga	>0.8	>0.9	None	Top 5	>0.4	>0.6	>0.4	Top 5	>0.3
42Nagasaki	>0.8	>0.2	Top 5	>0.9	>0.1	>0.6	>0.0	>0.3	>0.1
43Kumamoto	>0.8	>0.6	>0.2	>0.4	>0.1	>0.6	>0.6	>0.6	>0.1
44Oita	>0.8	>0.6	>0.1	>0.4	>0.3	>0.4	>0.4	Top 5	>0.4
45Miyazaki	>0.8	Only 3	>0.1	>0.2	Top 5	>0.4	>0.4	>0.6	>0.4
46Kagoshima	>0.8	Only 1	>0.2	>0.9	>0.1	>0.9	>0.4	>0.4	>0.3
47Okinawa	>0.8	None	>0.2	>0.2	Only 1	>0.9	>0.0	>0.4	Only 2

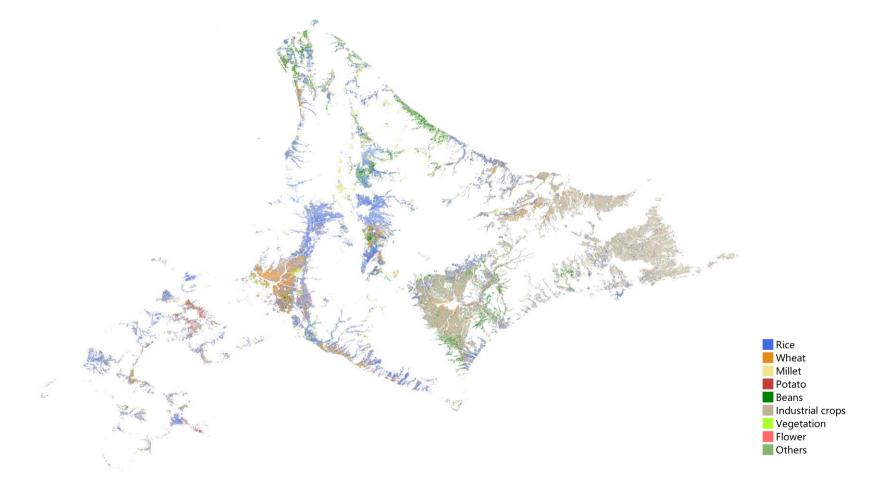
Classification methods comparisons (Ibaraki)



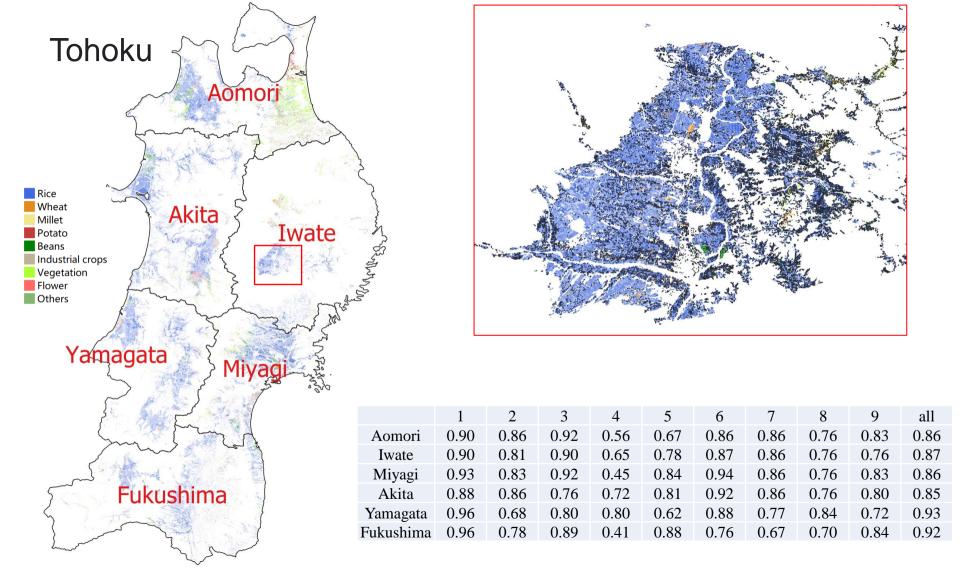


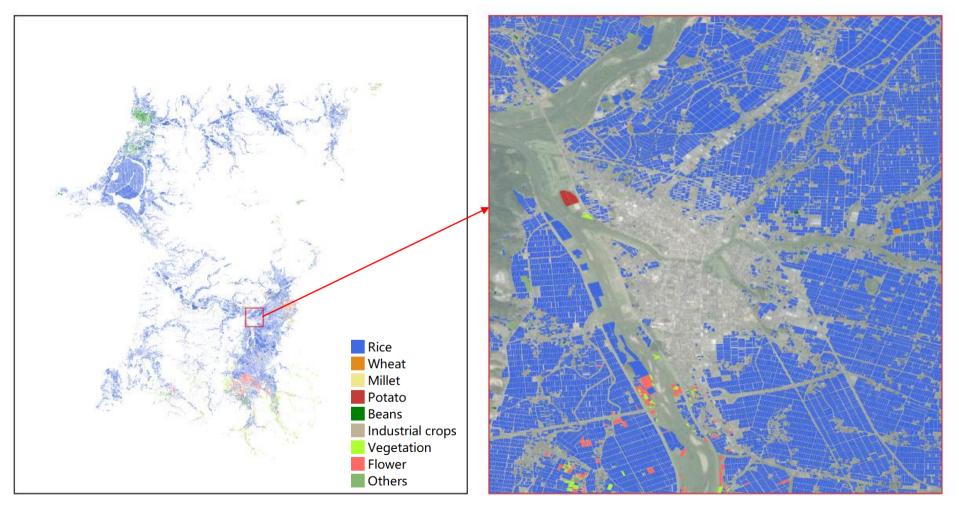
Transformer

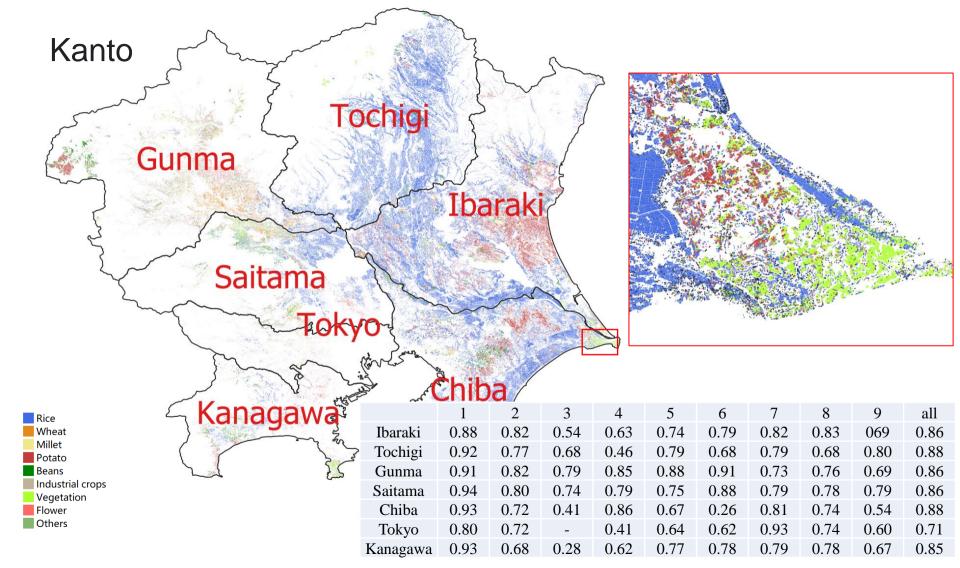
Method	Rice	Wheat	Millet	Potato	Beans	Industrial	Vegetation	Flower	Others	OA	mean F1
RF (pixel-based)	0.86	0.41	0.72	0.57	0.46	0.62	0.58	0.54	0.76	0.86	0.61
RF (parcel-based)	0.87	0.40	0.71	0.58	0.45	0.63	0.57	0.55	0.75	0.87	0.62
TempCNN	0.85	0.40	0.77	0.68	0.47	0.67	0.59	0.55	0.77	0.86	0.63
LSTM	0.82	0.38	0.74	0.66	0.43	0.63	0.51	0.47	0.63	0.83	0.61
Transformer	0.86	0.41	0.71	0.63	0.46	0.68	0.57	0.55	0.84	0.88	0.63

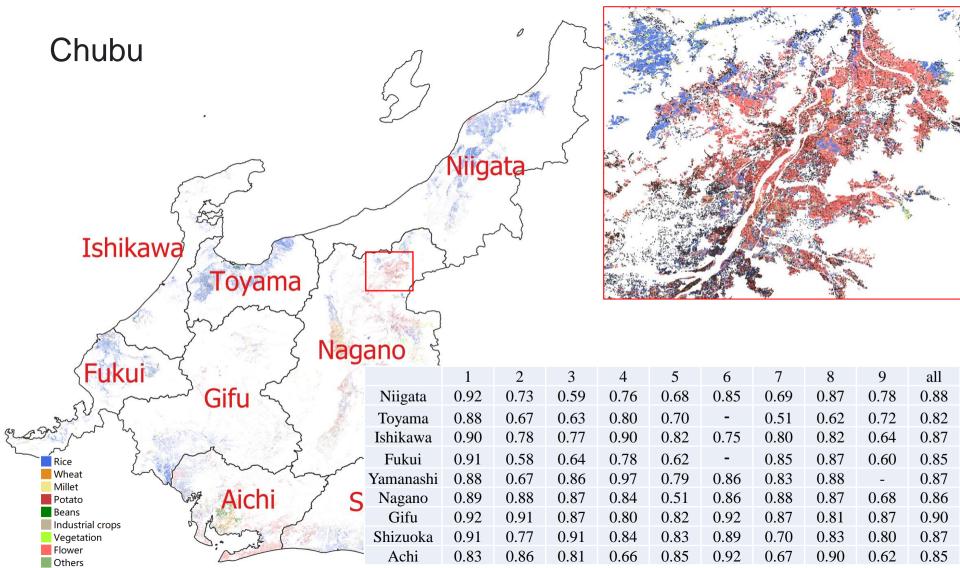


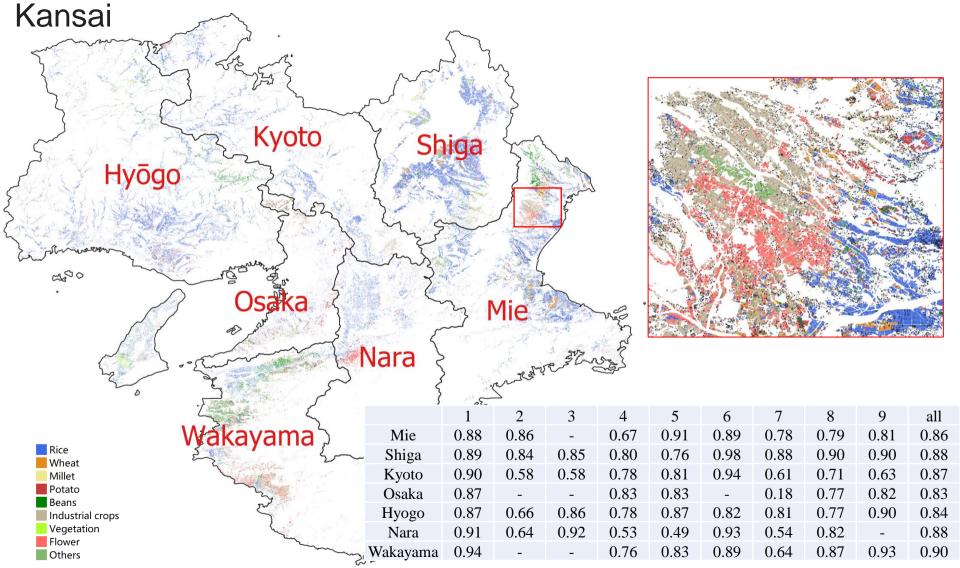
Classification result (F1 score: 0.85)



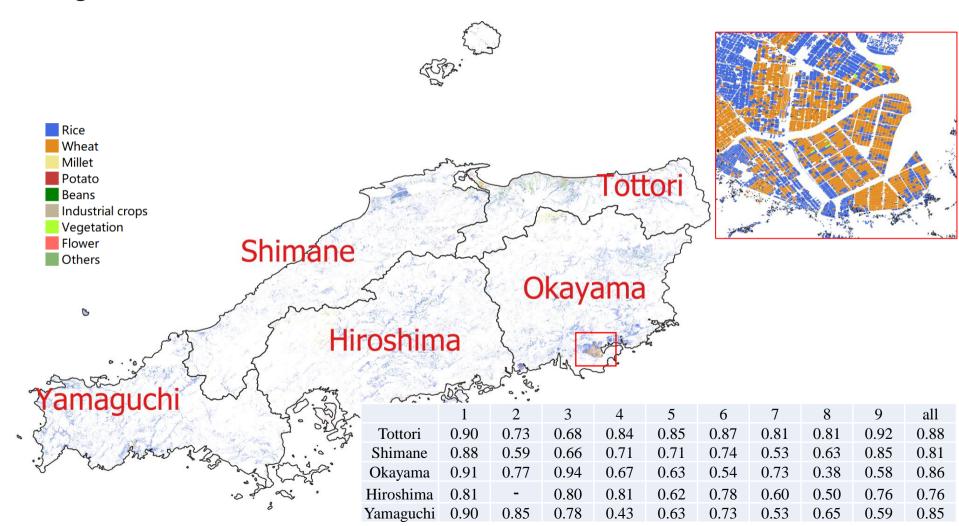




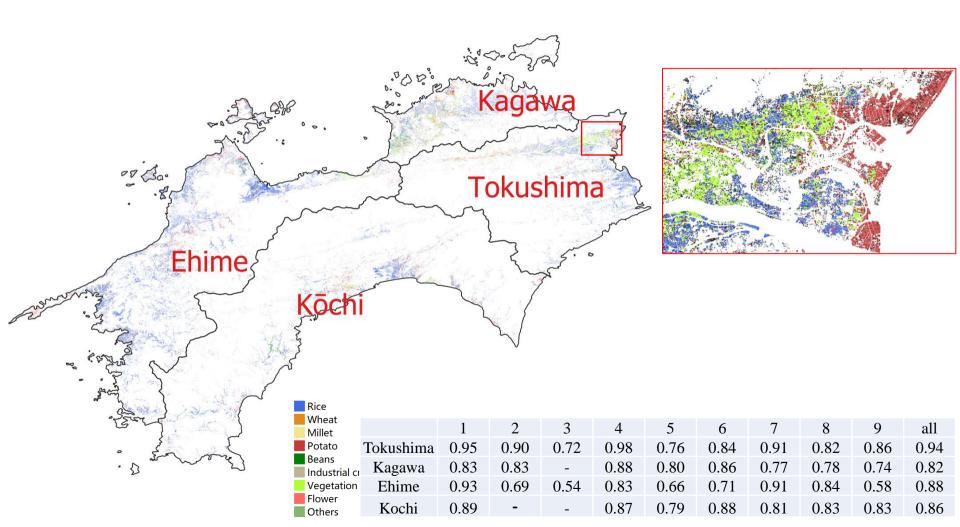


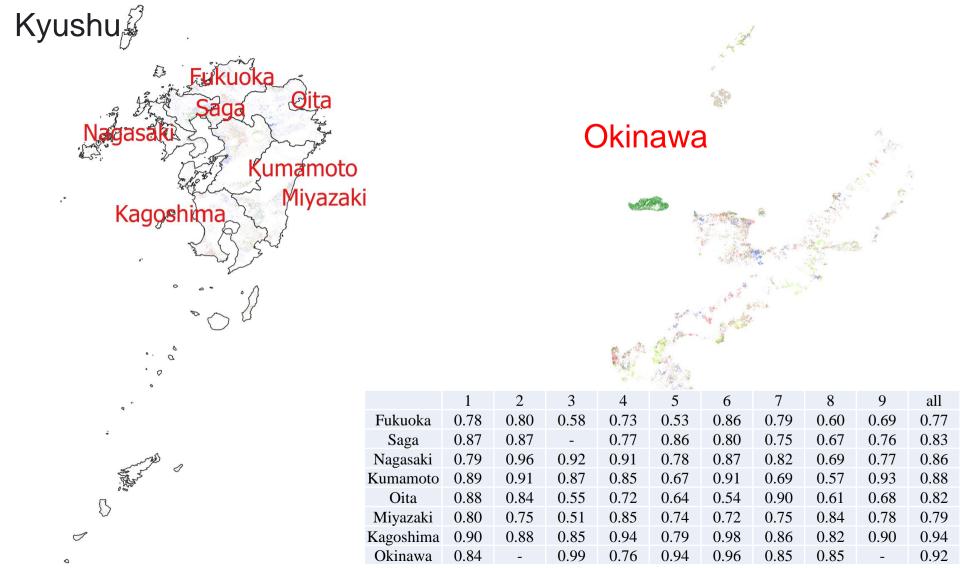


Chugoku



Shikoku





Conclusions

- ✓ Accuracies of rice > 85% over 45 prefectures
- ✓ Accuracies of other crop types differ in different prefectures
- ✓ First high-resolution cropland map of Japan

- ✓ Limited labels (except rice)
- ✓ Very poor ratio for other classes
- ✓ Unbalanced class ratio
- ✓ Sparse labels



- Noisy label learning
 Imbalanced label learning
- Real ground truth labels
- High-resolution datasets