



# OPTICAL-RADAR SYNERGY FOR WETLAND CHARACTERIZATION AND SURFACE WATER MONITORING

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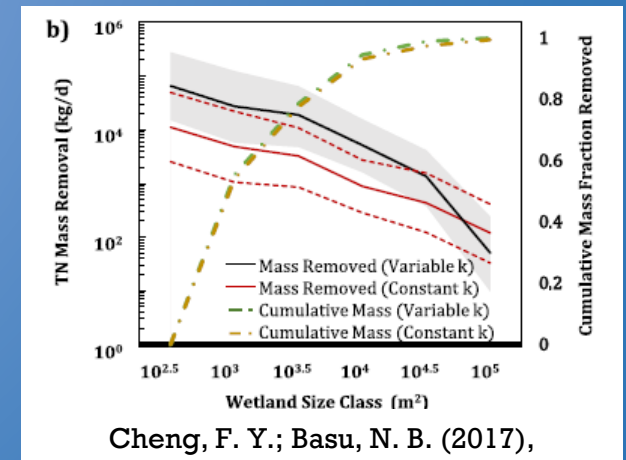
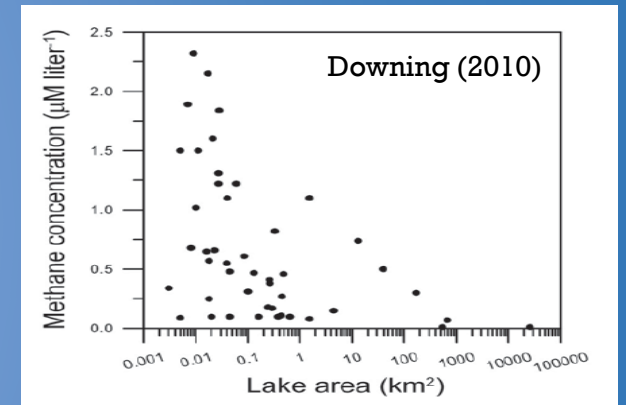
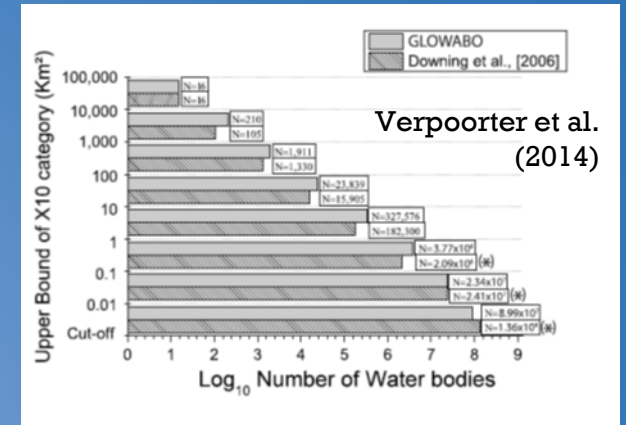
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<sup>8</sup> Department of Geography, Environment and Geomatics, University of Guelph

# WETLANDS HIGHLY VALUABLE BUT EXTREMELY COMPLEX

- Provide a wide range of ecosystem services
  - Biodiversity, nutrient/pollutants removal, .....
- Small ones disproportionately more important
  - Out-number larger ones
  - Higher methane concentration
  - More effective in removing nutrients
- ..... but have complex and highly variable water regimes and vegetation communities



# HIGH REQUIREMENTS FOR PRODUCING US NATIONAL WETLAND INVENTORY (NWI) DATA



## Final Draft Wetlands Mapping Standard

FGDC Wetlands Subcommittee

January 2009



U.S. Fish & Wildlife Service

Data Collection Requirements and Procedures for Mapping Wetland,  
Deepwater, and Related Habitats of the United States (version 2)

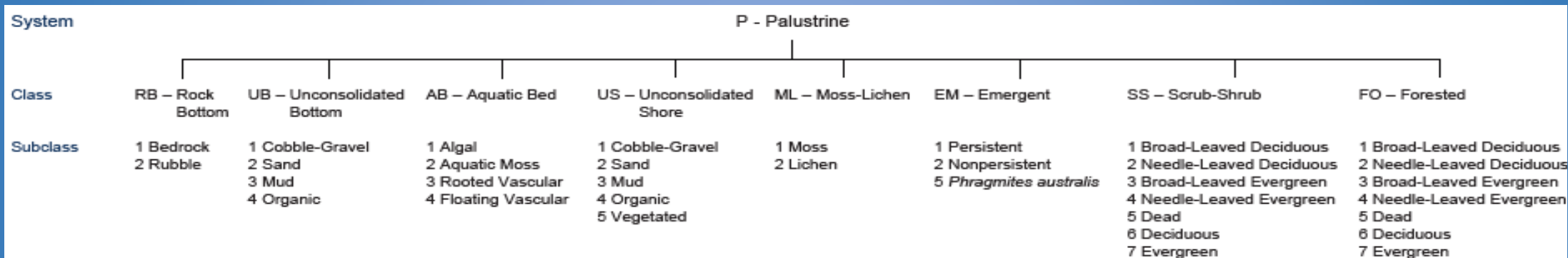
U.S. FISH & WILDLIFE SERVICE - ECOLOGICAL SERVICES  
DIVISION OF BUDGET AND TECHNICAL SUPPORT  
BRANCH OF GEOSPATIAL MAPPING AND TECHNICAL SUPPORT  
FALLS CHURCH, VA 2204

*REVISED AUGUST 2015*

# KEY NWI MAPPING REQUIREMENTS

**Table 2. Spatial Resolution Requirements of Base Imagery**

	Lower 48 States, Hawaii, & Territories *	Estuarine & Lacustrine Deepwater **	Alaska (Including Deepwaters)
Resolution	1m	3m	5m
Scale	1:12,000	1:24,000	1:63,360



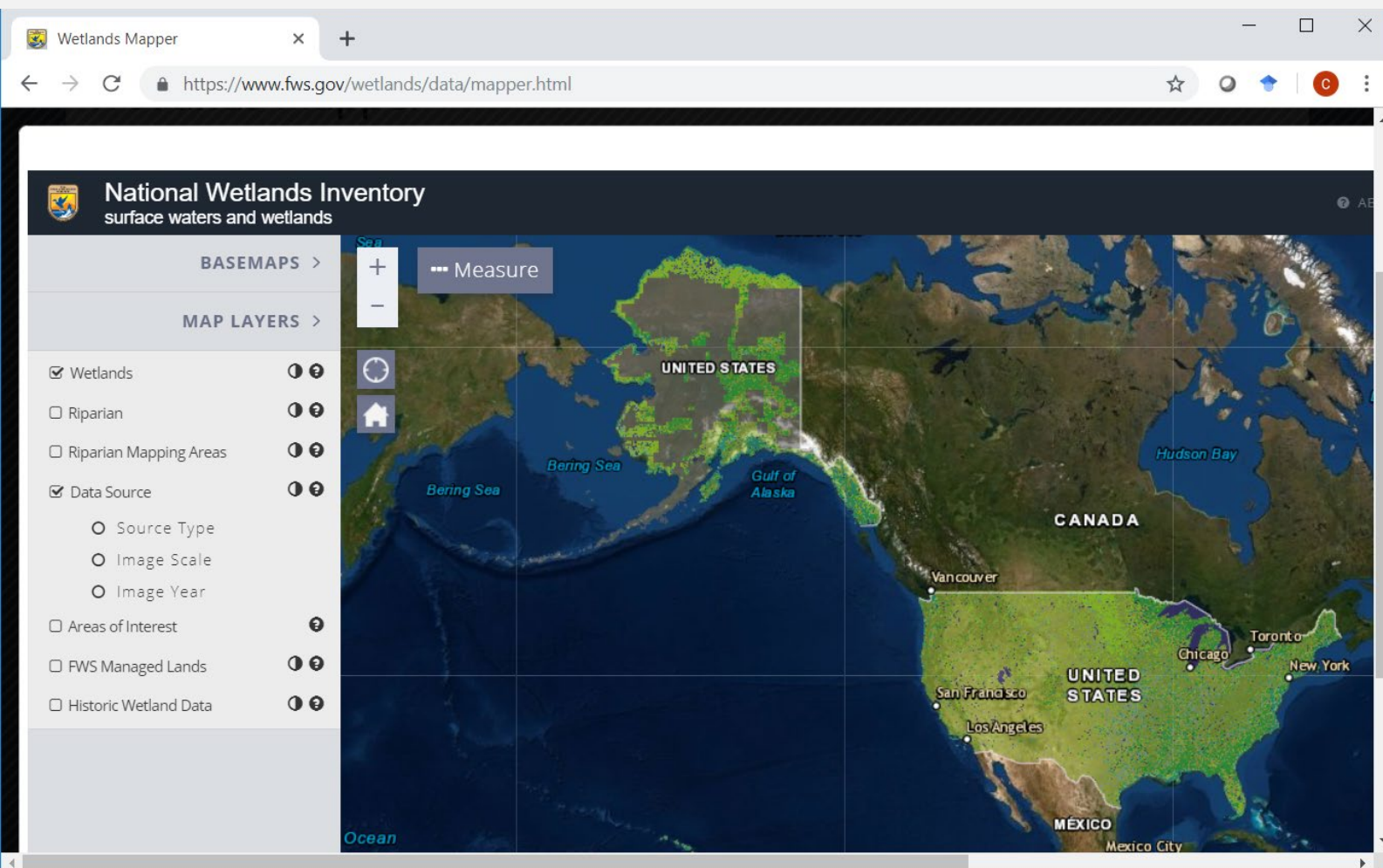
## MODIFIERS

In order to more adequately describe the wetland and deepwater habitats, one or more of the water regime, water chemistry, soil, or special modifiers may be applied at the class or lower level in the hierarchy.

Water Regime			Special Modifiers	Water Chemistry		Soil
Nontidal	Saltwater Tidal	Freshwater Tidal		Halinity/Salinity	pH Modifiers for Fresh Water	
A Temporarily Flooded	L Subtidal	S Temporarily Flooded-Fresh Tidal	b Beaver	1 Hyperhaline / Hypersaline	a Acid	g Organic
B Seasonally Saturated	M Irregularly Exposed	Q Regularly Flooded-Fresh Tidal	d Partly Drained/Ditched	2 Euhaline / Eusaline	t Circumneutral	n Mineral
C Seasonally Flooded	N Regularly Flooded	R Seasonally Flooded-Fresh Tidal	f Farmed	3 Mixohaline / Mixosaline (Brackish)	i Alkaline	
D Continuously Saturated	P Irregularly Flooded	T Semipermanently Flooded-Fresh Tidal	m Managed	4 Polyhaline		
E Seasonally Flooded/ Saturated		V Permanently Flooded-Fresh Tidal	h Diked/Impounded	5 Mesohaline		
F Semipermanently Flooded			r Artificial Substrate	6 Oligohaline		
G Intermittently Exposed			s Spoil	0 Fresh		
H Permanently Flooded			x Excavated			
J Intermittently Flooded						
K Artificially Flooded						

# CURRENT STATUS OF NWI DATA

<https://www.fws.gov/wetlands/data/mapper.html>



## Challenges in Alaska

- Base imagery
- Field data collection difficult
- Mapping procedure largely driven by image analysts

# EXPLORE USE OF AVAILABLE OPTICAL AND RADAR ASSETS FOR WETLAND MAPPING

## Optical



Landsat-8: 16-day revisit



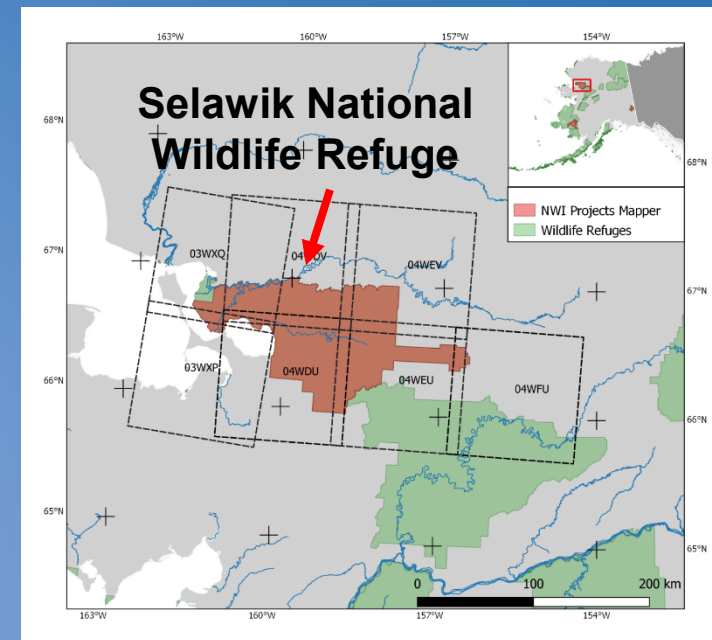
Sentinel-2: 10-day one satellite, 5-day with two satellites



## RADAR



Sentinel-1: 12-day revisit one satellite, but current acquisition scenario for two satellites provides 12-day revisit in most areas



# METHODS TARGET VEGETATION AND WATER REGIME OF THE CLASSIFICATION SYSTEM

System

P - Palustrine

Class

RB – Rock Bottom    UB – Unconsolidated Bottom    AB – Aquatic Bed    US – Unconsolidated Shore    ML – Moss-Lichen    EM – Emergent    SS – Scrub-Shrub    FO – Forested

Subclass

1 Bedrock 2 Rubble	1 Cobble-Gravel 2 Sand 3 Mud 4 Organic	1 Algal 2 Aquatic Moss 3 Rooted Vascular 4 Floating Vascular	1 Cobble-Gravel 2 Sand 3 Mud 4 Organic 5 Vegetated	1 Moss 2 Lichen	1 Persistent 2 Nonpersistent 5 <i>Phragmites australis</i>	1 Broad-Leaved Deciduous 2 Needle-Leaved Deciduous 3 Broad-Leaved Evergreen 4 Needle-Leaved Evergreen 5 Dead 6 Deciduous 7 Evergreen	1 Broad-Leaved Deciduous 2 Needle-Leaved Deciduous 3 Broad-Leaved Evergreen 4 Needle-Leaved Evergreen 5 Dead 6 Deciduous 7 Evergreen
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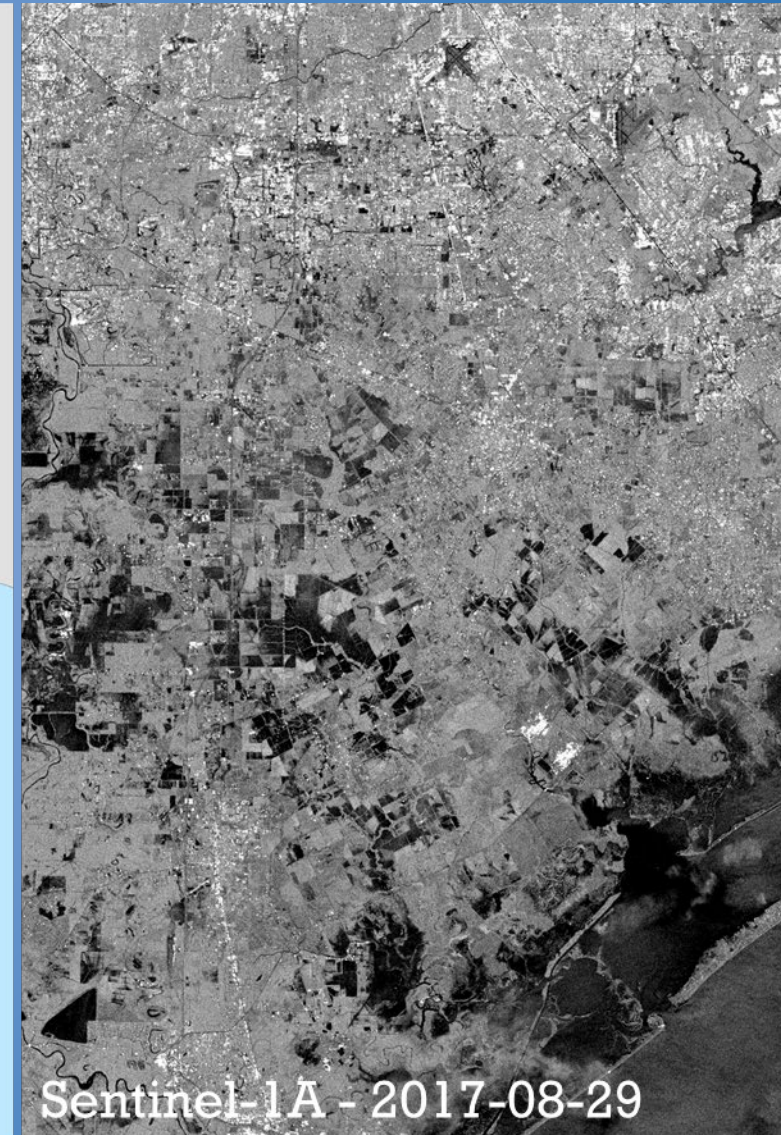
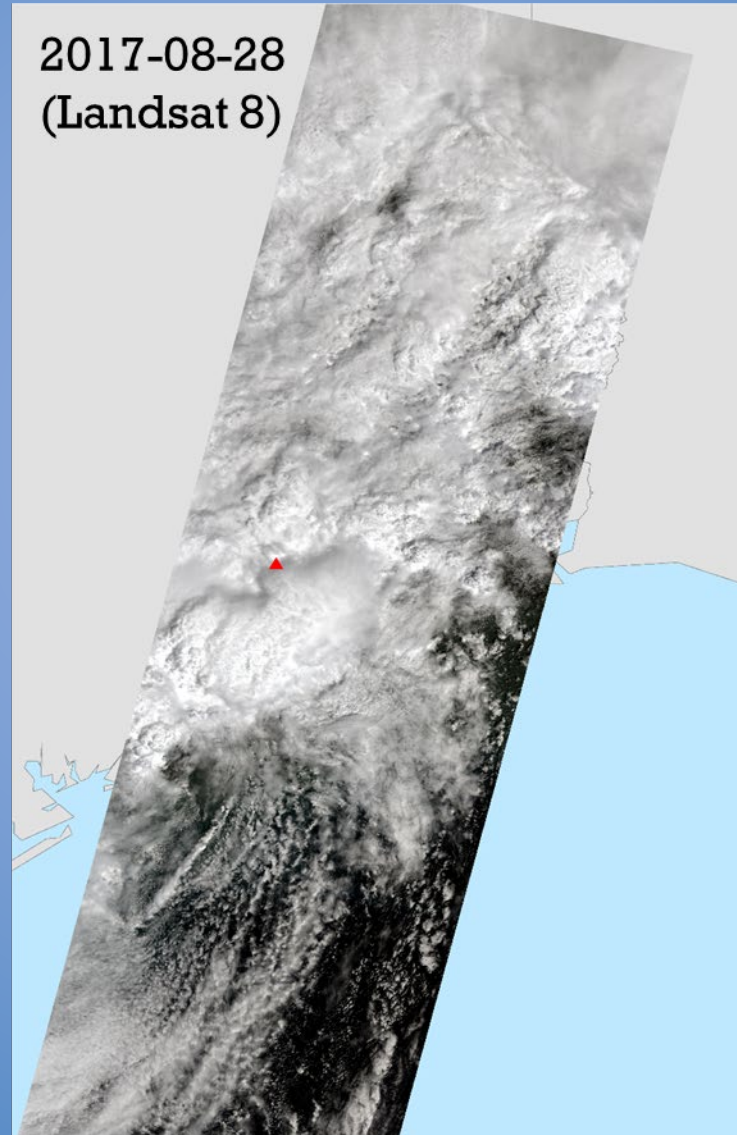
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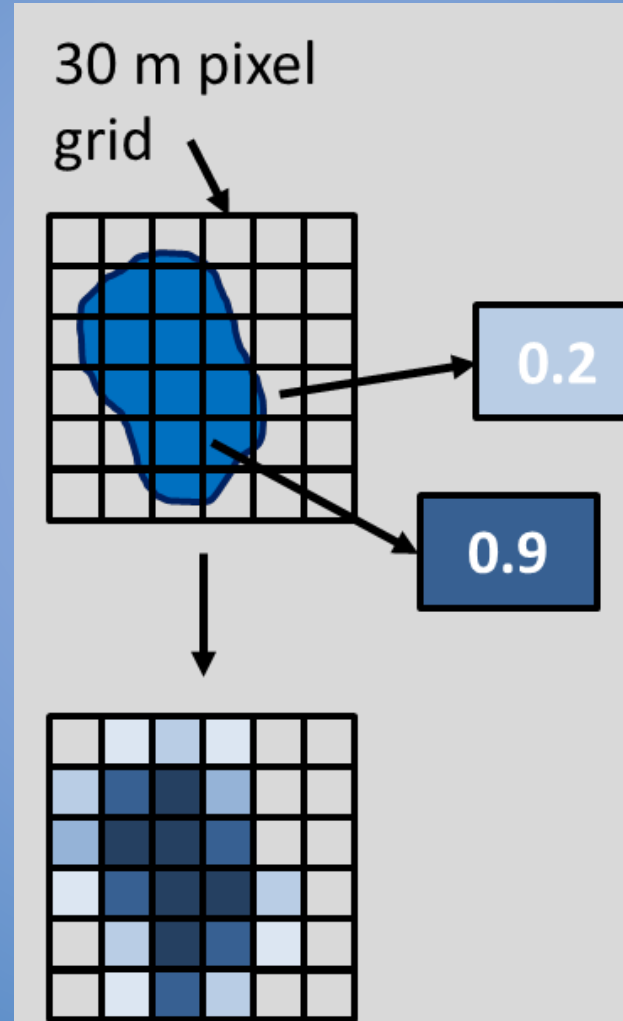
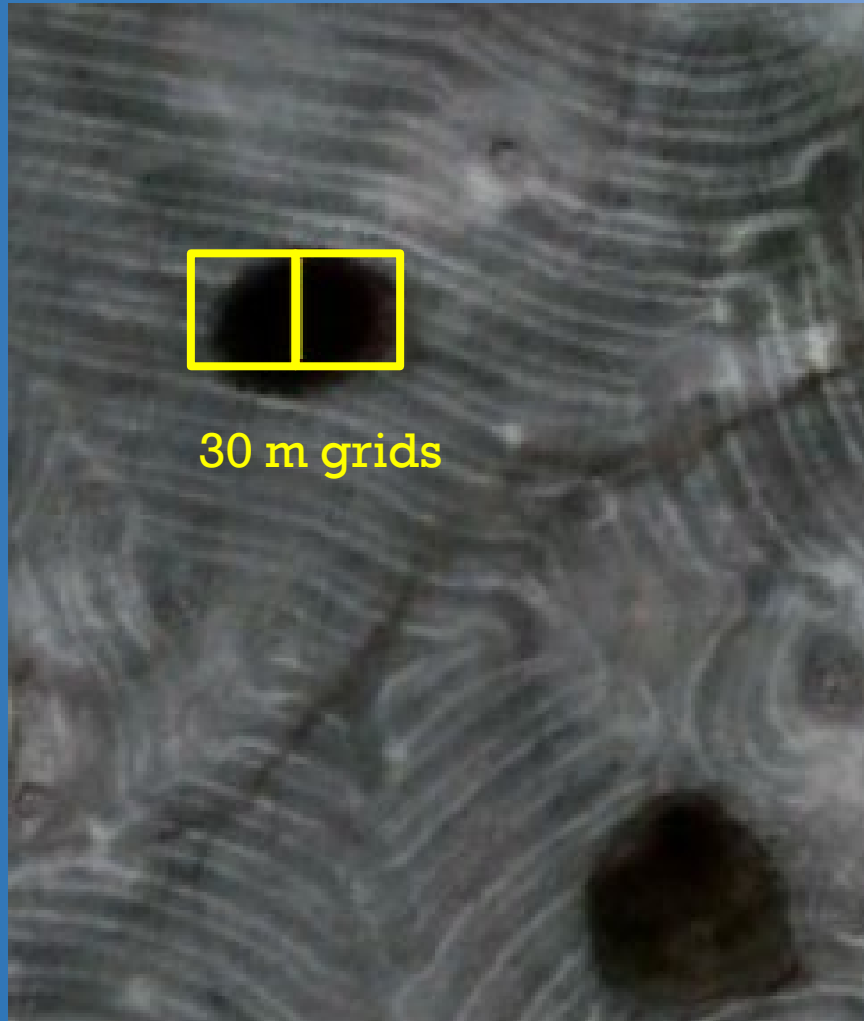
# WATER REGIME MAPPING APPROACH

- Automated surface water mapping algorithms
  - Optical algorithm
  - Radar algorithm
- Temporal aggregation to reduce noises





# AUTOMATED ESTIMATION OF SUBPIXEL WATER FRACTION (SWF)



waffls:

“water fraction from Landsat and Sentinel-2”

python package for automated SWF estimation

<https://github.com/bendv/waffls>

remote sensing

Automated Quantification of Surface Water Inundation in Wetlands Using Optical Satellite Imagery

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Abstract

We present a fully automated and scalable algorithm for quantifying surface water inundation in wetlands. Requiring no external training data, our algorithm estimates sub-pixel water fraction (SWF) over large areas and long time periods using Landsat data. We tested our SWF algorithm over three wetland sites across North America, including the Humber Inflow Region, the Delaware Peninsula and the Chesapeake, representing a gradient of inundation and vegetation conditions. We validated SWF at 30-m resolution with accuracies ranging from a normalized root mean square error of 0.1 to 0.19 when compared with various high-resolution ground and airborne datasets. SWF estimates were more accurate to define structural features compared to previously published surface water datasets, accurately depicting water bodies, large heterogeneously inundated surfaces, narrow water channels and canopy-covered water features. Despite the enhanced accuracy, several sources of errors affected SWF estimates, including emergence of floating vegetation and forest canopy shadows from topographic features, urban structures and unmasked clouds. The automated algorithm described in this article allows for the production of high temporal resolution wetland inundation data.

remote sensing

Automated Quantification of Surface Water Inundation in Wetlands Using Optical Satellite Imagery

MDPI

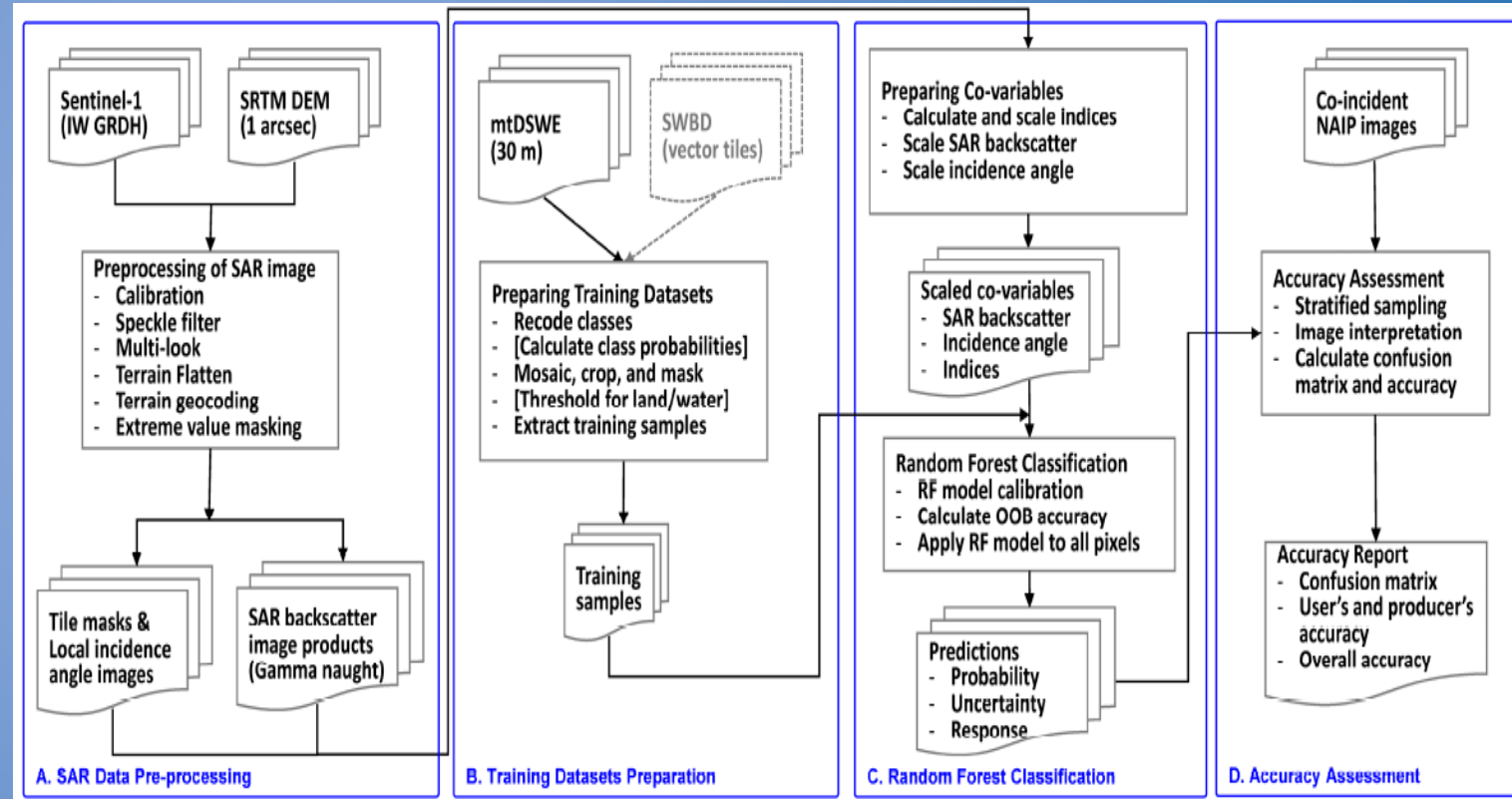
remote sensing

1099-4450

# USE OPTICAL RESULTS TO AUTOMATE THE RADAR MAPPING ALGORITHM

## Key steps:

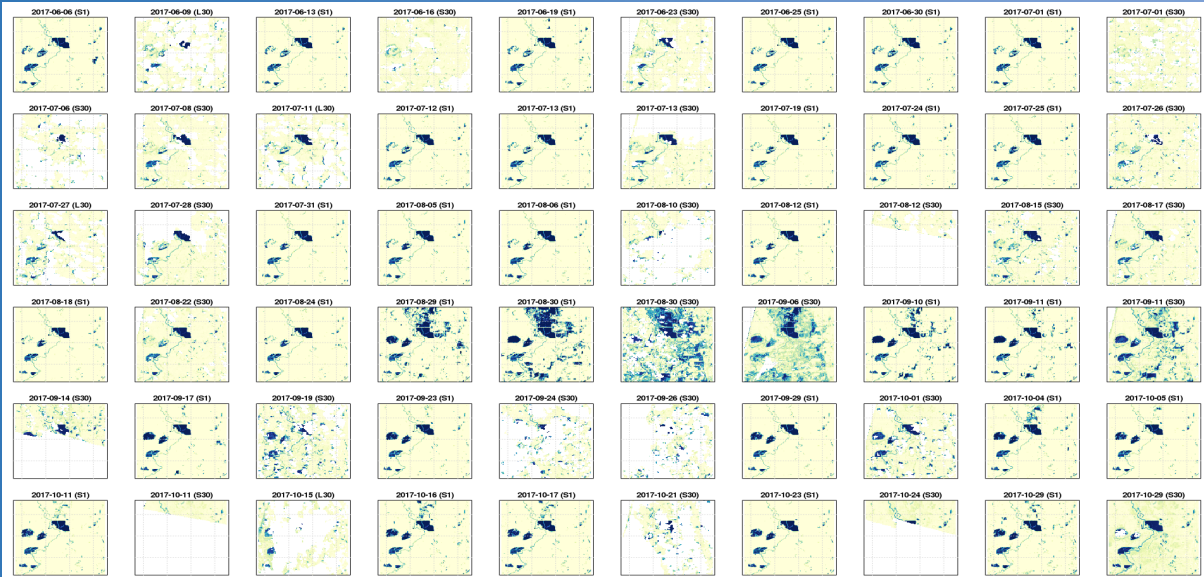
- Use historical optical results to identify
  - Permanent water bodies
  - Upland never inundated
- Select water and non-water training samples from the above
- Train machine learning algorithm and use it to produce classification map



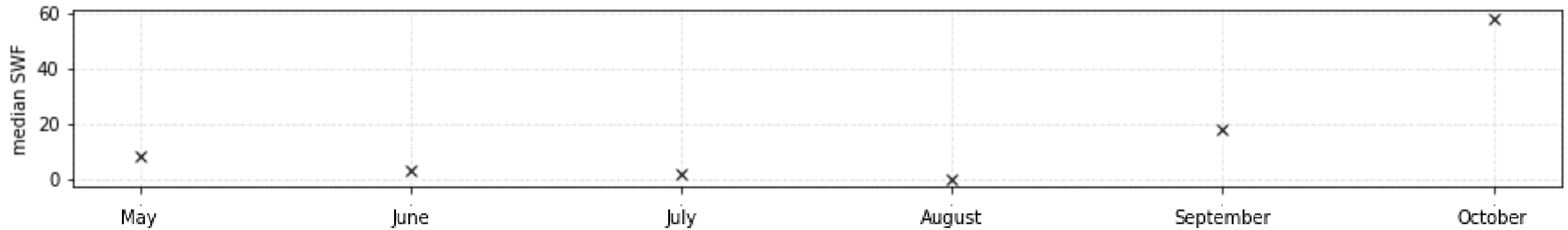
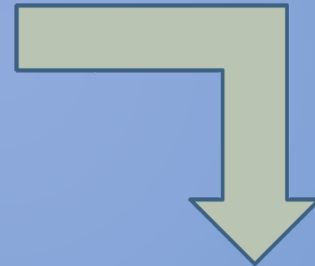
# RESULTS FROM INDIVIDUAL OPTICAL-SAR OBSERVATIONS TEMPORALLY DETAILED, BUT VERY NOISY



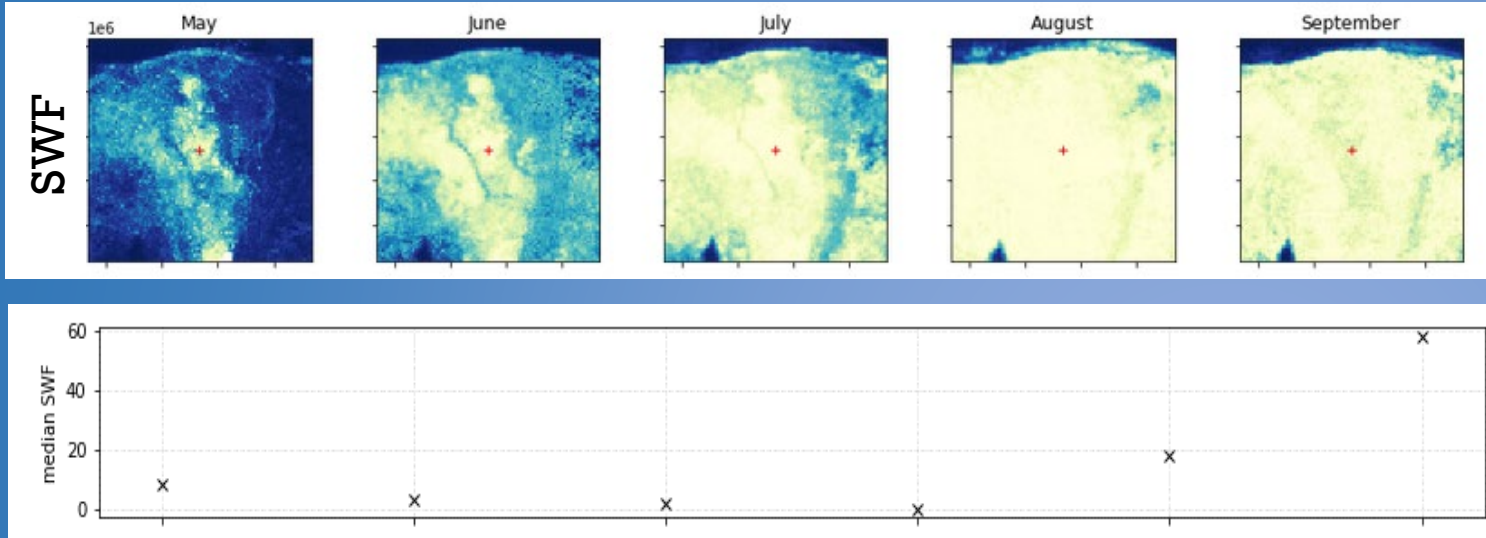
# TEMPORAL AGGREGATION TO REDUCE NOISES



Monthly Median



# DEFINE WATER REGIME CLASSES BASED ON MONTHLY COMPOSITES



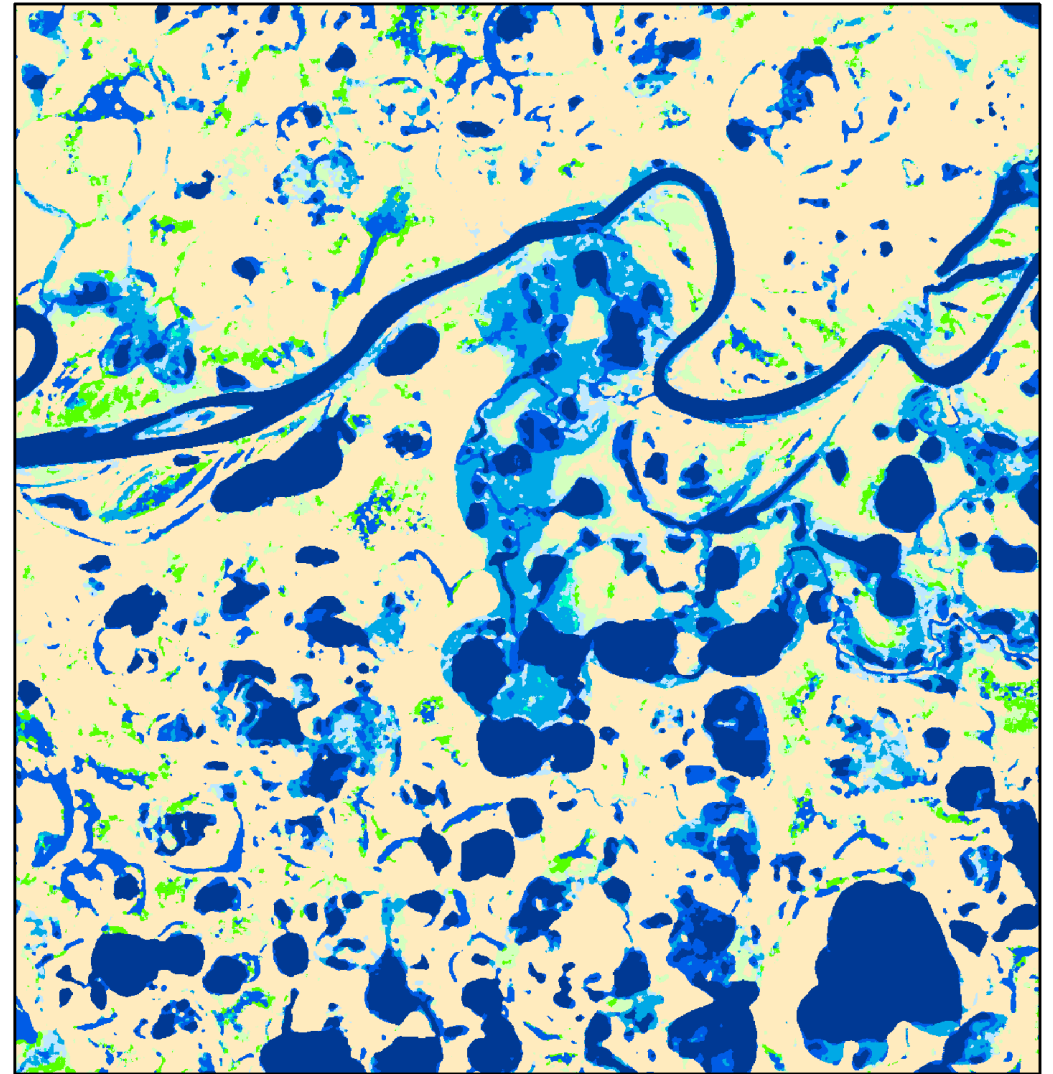
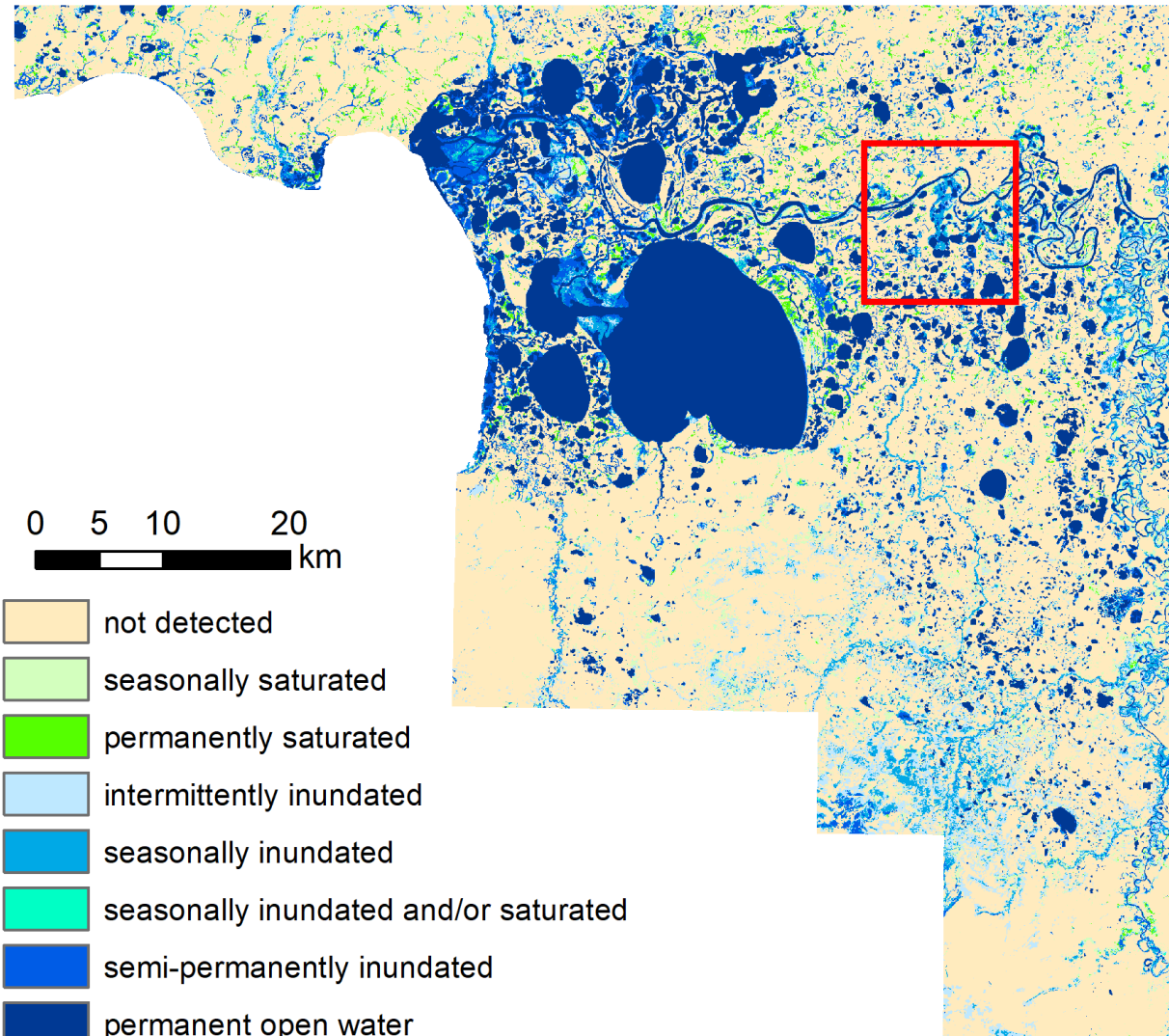
Quantitative  
Rules to Define  
Surface Water  
Regime Classes

- 0 – not detected
- 1 – seasonally saturated
- 2 – permanently saturated
- 3 – intermittently inundated
- 4 – seasonally inundated
- 5 – seasonally inundated or saturated
- 6 – semi-permanently inundated
- 7 – permanent open water

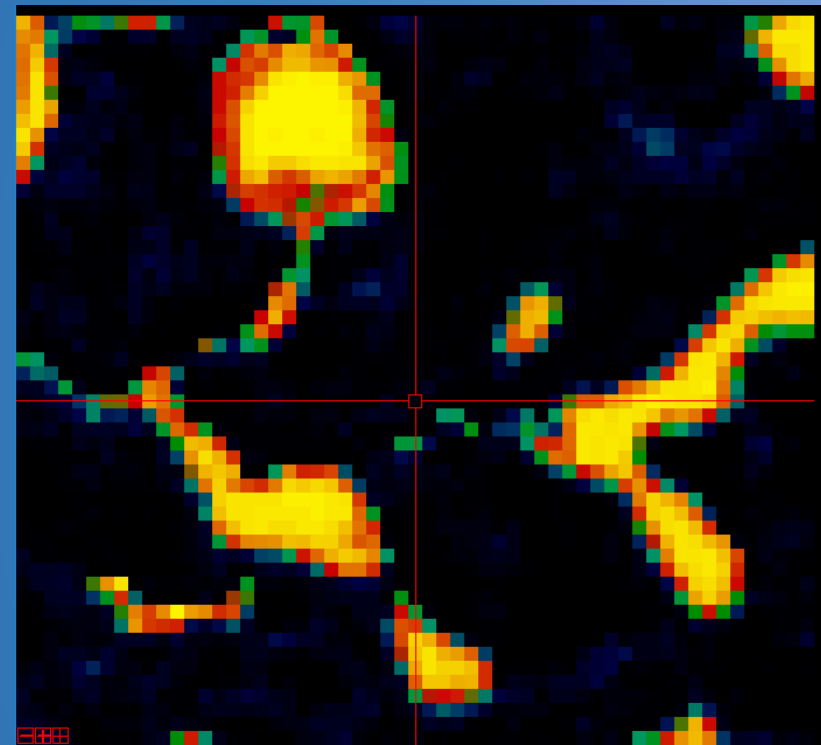
**Seasonally Flooded (C)** Surface water is present for extended periods (generally for more than a month) during the growing season, but is absent by the end of the season in most years. When surface water is absent, the depth to substrate saturation may vary considerably among sites and among years.

**Intermittent (R4)** This Subsystem includes channels that contain flowing water only part of the year. When the water is not flowing, it may remain in isolated pools or surface water may be absent.

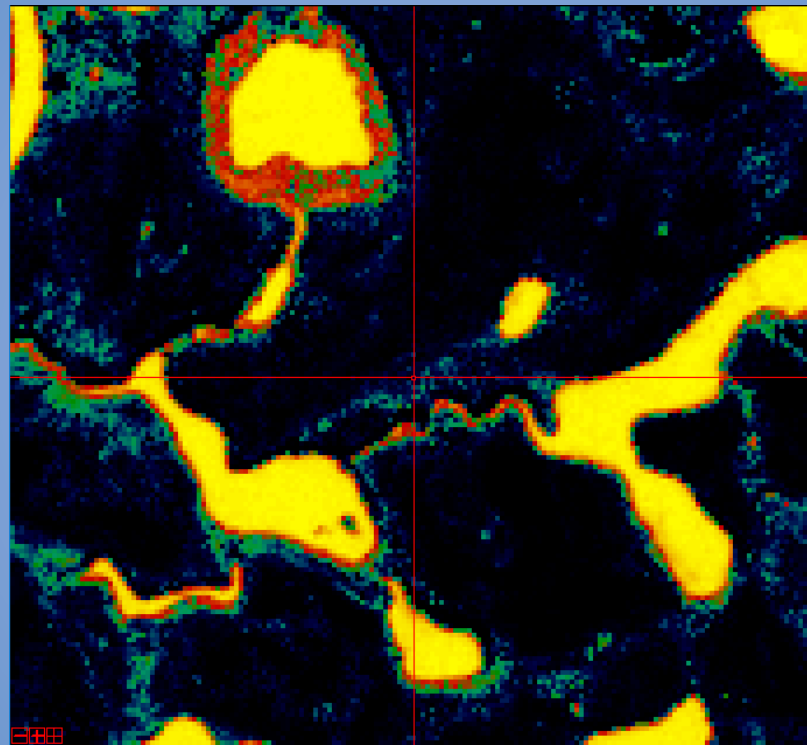
# Rule-Based Surface Water Regime Map



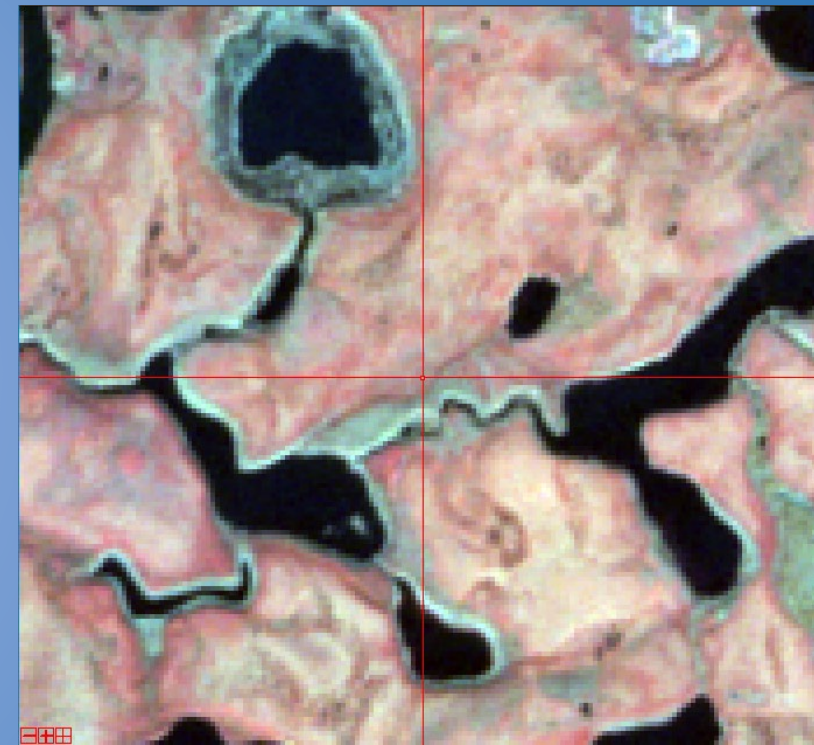
# 10-M DATA MUCH BETTER THAN 30-M DATA FOR SMALL WETLANDS



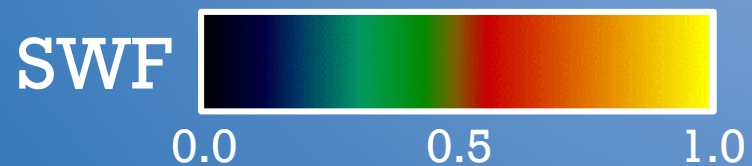
SWF (30m)



SWF (10m)



NIR-R-G (10m)

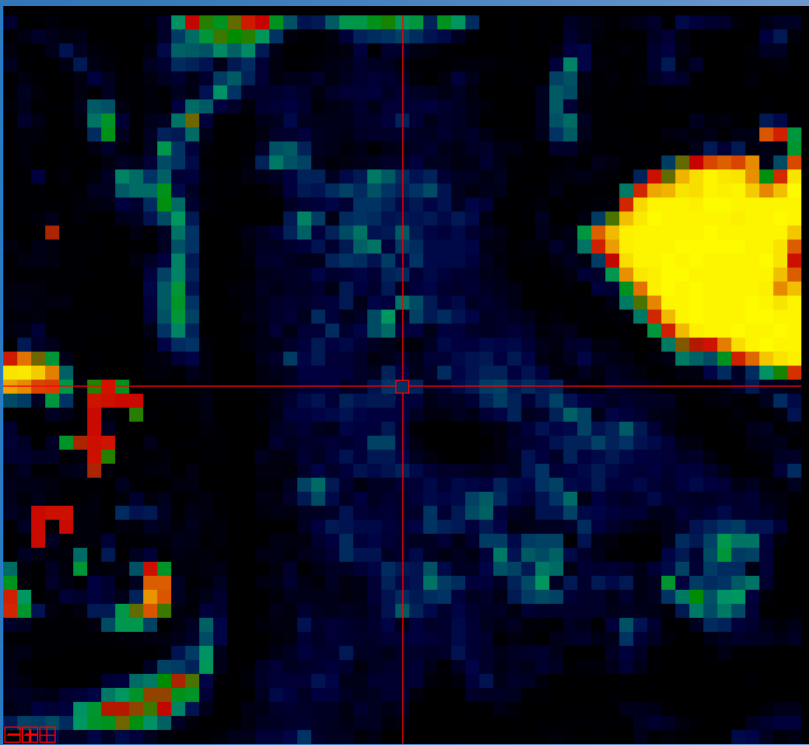


0.0

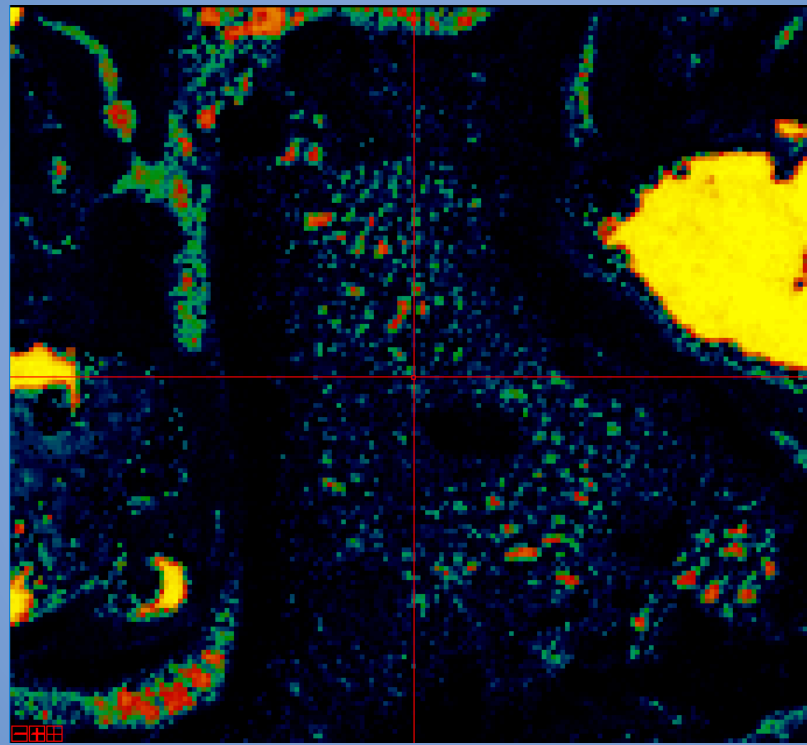
0.5

1.0

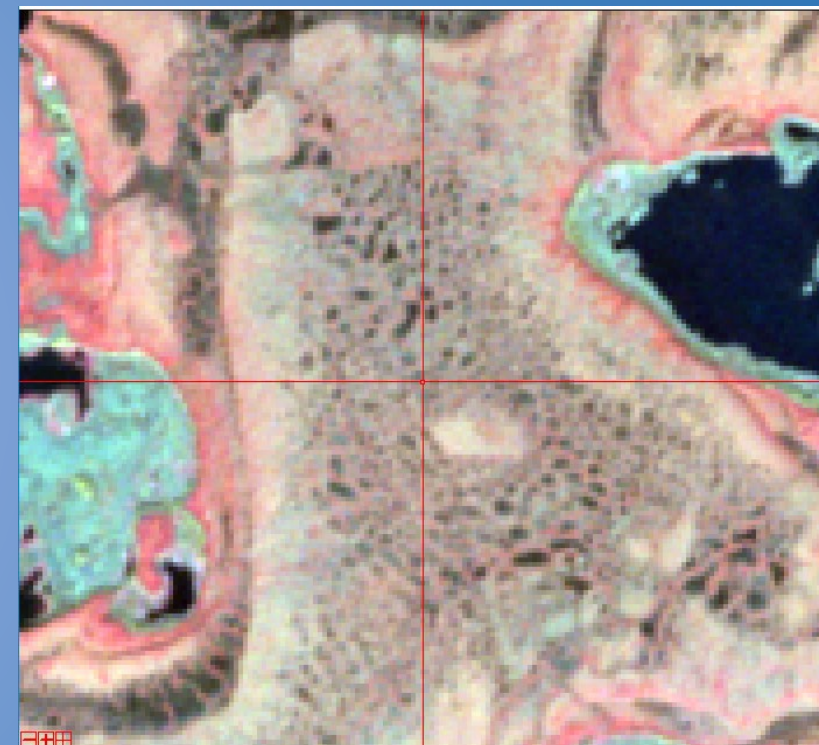
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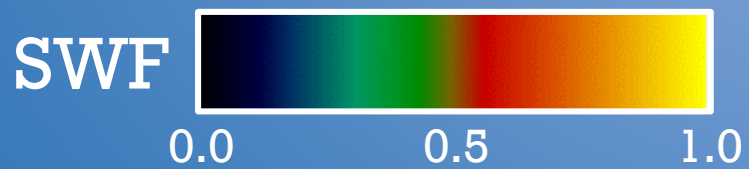
SWF (30m)



SWF (10m)



NIR-R-G (10m)





# VEGETATION CLASSIFICATION: USE STANDARD MACHINE LEARNING APPROACH

System	P - Palustrine							
Class	RB - Rock Bottom	UB - Unconsolidated Bottom	AB - Aquatic Bed	US - Unconsolidated Shore	ML - Moss-Lichen	EM - Emergent	SS - Scrub-Shrub	FO - Forested
Subclass	1 Bedrock 2 Rubble	1 Cobble-Gravel 2 Sand 3 Mud 4 Organic	1 Algal 2 Aquatic Moss 3 Rooted Vascular 4 Floating Vascular	1 Cobble-Gravel 2 Sand 3 Mud 4 Organic 5 Vegetated	1 Moss 2 Lichen	1 Persistent 2 Nonpersistent 5 <i>Phragmites australis</i>	1 Broad-Leaved Deciduous 2 Needle-Leaved Deciduous 3 Broad-Leaved Evergreen 4 Needle-Leaved Evergreen 5 Dead 6 Deciduous 7 Evergreen	1 Broad-Leaved Deciduous 2 Needle-Leaved Deciduous 3 Broad-Leaved Evergreen 4 Needle-Leaved Evergreen 5 Dead 6 Deciduous 7 Evergreen

Simplify to create classes derivable from remote sensing data



# MONTHLY MEDIAN COMPOSITES USED IN VEGETATION CLASSIFICATION

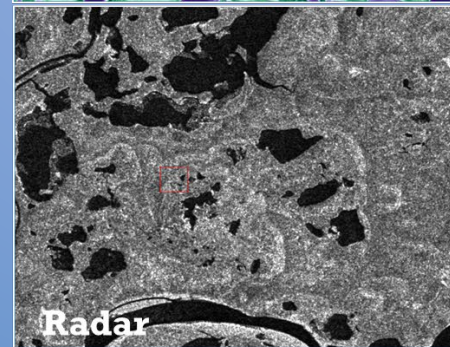
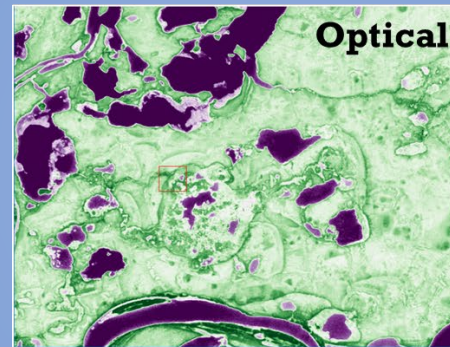
## Optical:

- spectral bands
- indices, e.g.:
  - NDVI, NDWI, NDSI

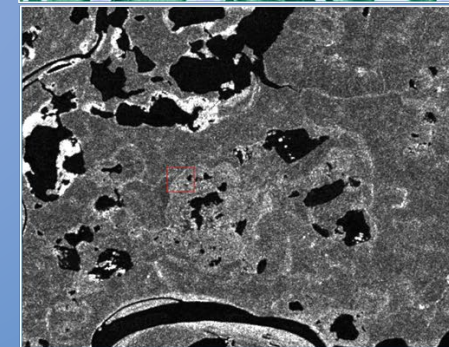
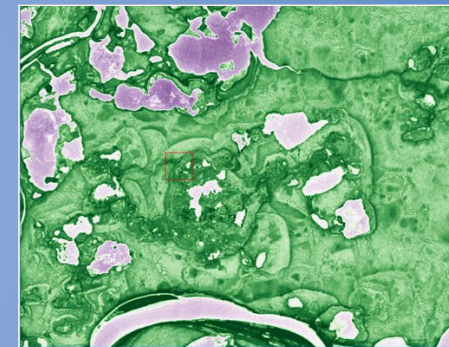
## Radar:

- vv, vh backscatter
- indices, e.g.:
  - $vh/(vv+vh)$

June



July



August .....

# VEGETATION CLASSIFICATION ACCURACY STATISTICS

## Reference

Predicted		NV	SS	SS/FO	EM	EM/SS	Sum	OA
	NV	3543	363	107	729	265	5007	<b>0.65</b>
	SS	383	2772	28	691	727	4601	
	SS/FO	230	185	4620	161	278	5474	
	EM	552	626	63	2221	578	4040	
	EM/SS	292	1054	182	1198	3152	5878	
	Sum	5000	5000	5000	5000	5000	25000	

Cell values = # of 10-m pixels

NV: not vegetated

SS: scrub-shrub

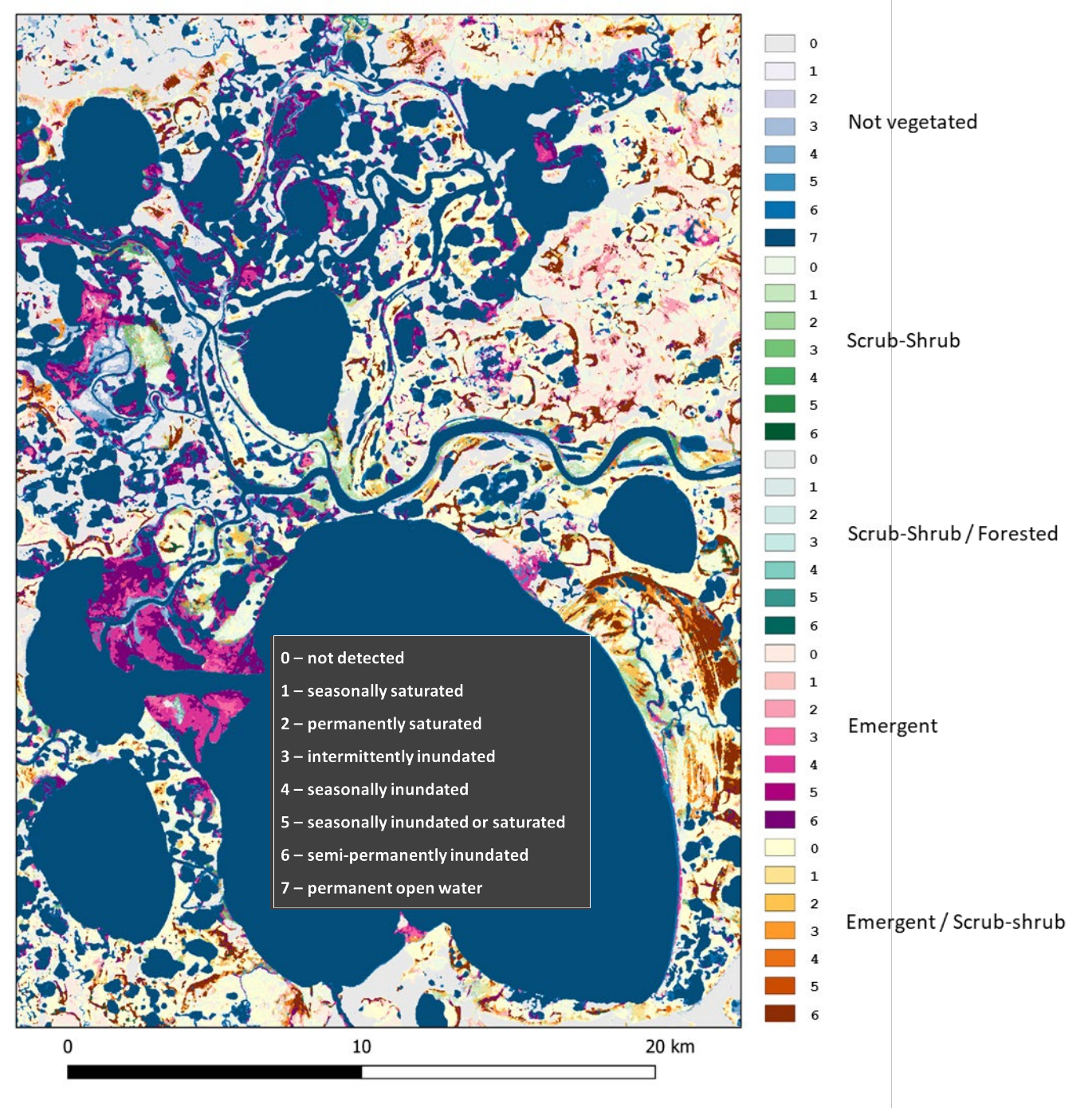
FO: forested

EM: emergent

OA: overall accuracy

# WETLAND CLASSIFICATION BY VEGETATION TYPE AND WATER REGIMES

- 36 classes in total
  - 5 veg x 7 water + open water)
- Visually appealing
- Assessment
  - Vegetation: ~65%
  - Water regime: no reference data to assess, but method objective and repeatable
- Products being evaluated by NWI towards developing standard compliant products over Alaska



# SUMMARY AND FUTURE OUTLOOK

- Detailed wetland classification possible with synergistic use of optical and radar time series observations
  - Quantitative measure of water regimes
  - Detailed wetland classification that considers both water regime and vegetation type
  - Great improvements in mapping small wetlands using 10 m data instead of 30 m data
- Significantly improved radar capabilities forthcoming
  - RADARSAT constellation - sub-weekly, C band
  - NISAR - L band, better mapping of woody wetlands
  - Optical-SAR synergy -> much needed spatial-temporal details for highly dynamic aquatic systems
    - Strong technical skills on cloud platforms needed to realize the full potential

# THANK YOU!

NASA LCLUC Program

US FWS NWI Program

Free Data Policy

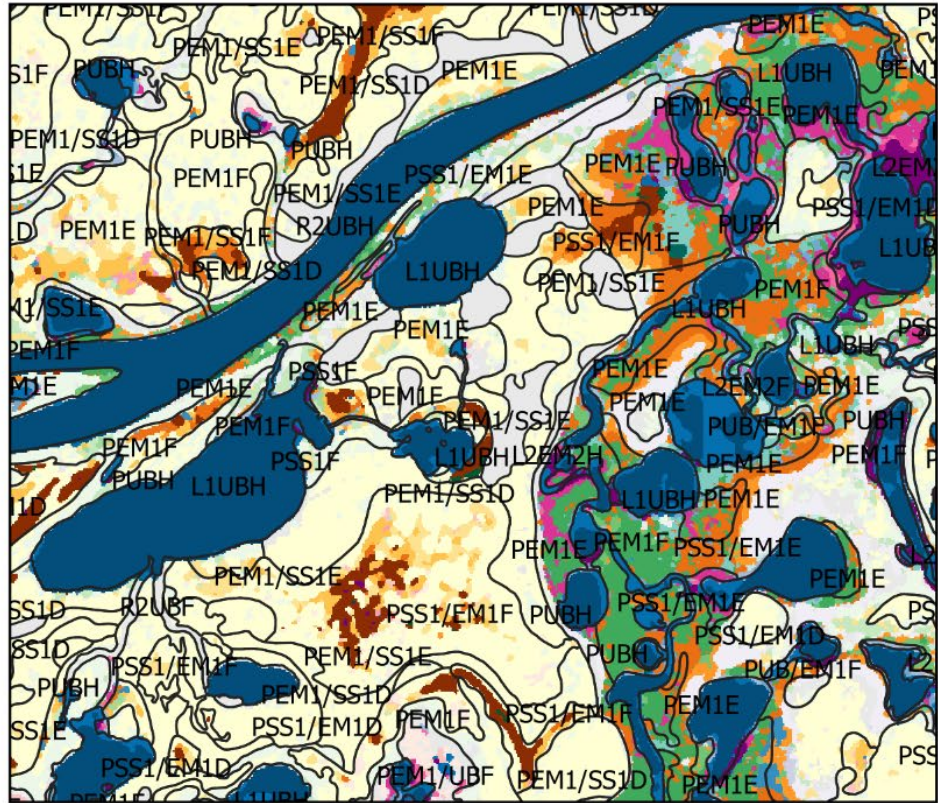
NASA HLS Project

Google Earth Engine

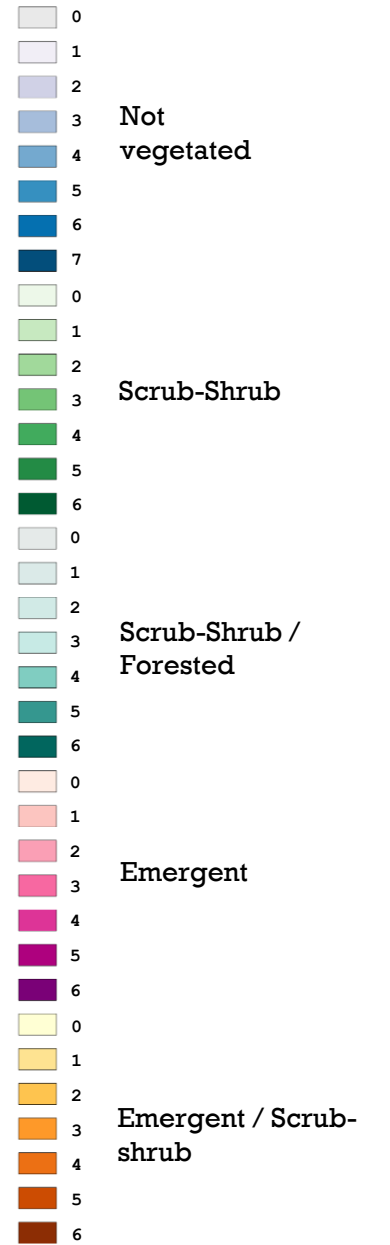
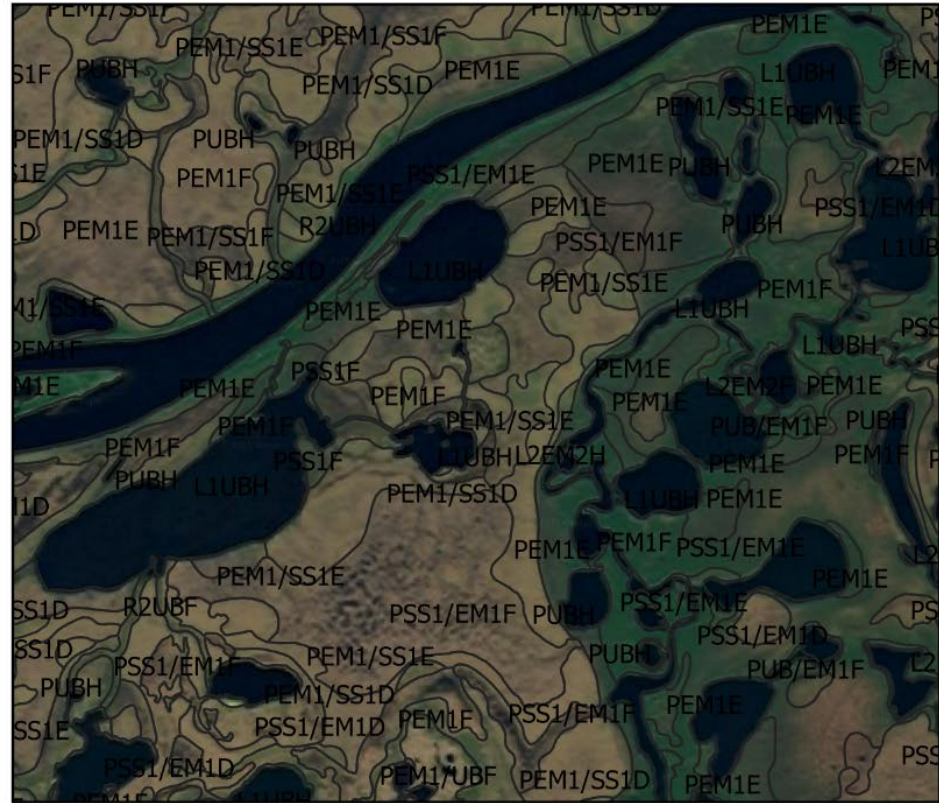
NASA NEX

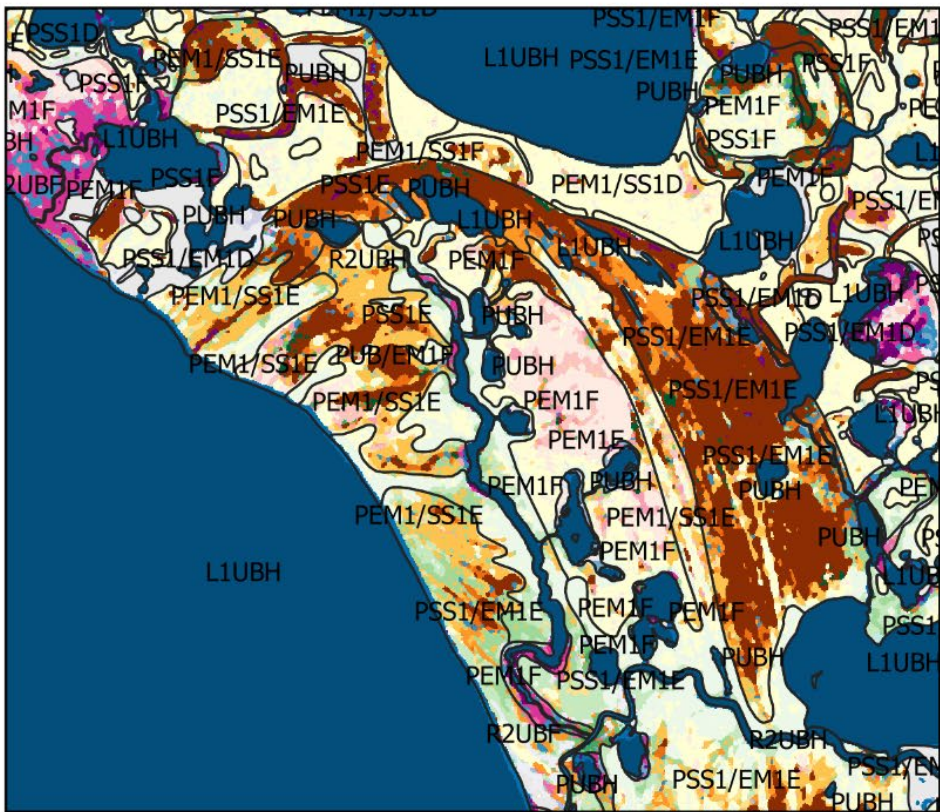
Contact for follow-up discussions

Chengquan (Cheng) Huang  
cqhuang@umd.edu



0 1.5 3 km





0 1.5 3 km

