NAWM's Wetland Mapping Consortium

Government of Alberta Inventory Pilot Project

## Next Generation Wetland Mapping in Alberta using Al

Lyle Boychuk & Becky Edwards Ducks Unlimited Canada

Alberta

Government







### Acknowledgements

#### EPA

Angela Burkinshaw Amanda Cooper Danielle Cobbaert Courtney Kelly Craig Mahoney Josh Montgomery Kristyn Mayner Mina Nasr Nicole Skakun Vanessa Swarbrick Jonathan Thompson Marsha Trites-Russell

#### DUC

Thorsten Hebben Lyle Boychuk Lindsay McBlane Becky Edwards Josh Evans Mark Kornder Andrea Dechene Kevin Smith Al Richard

Thank you!

#### ABMI

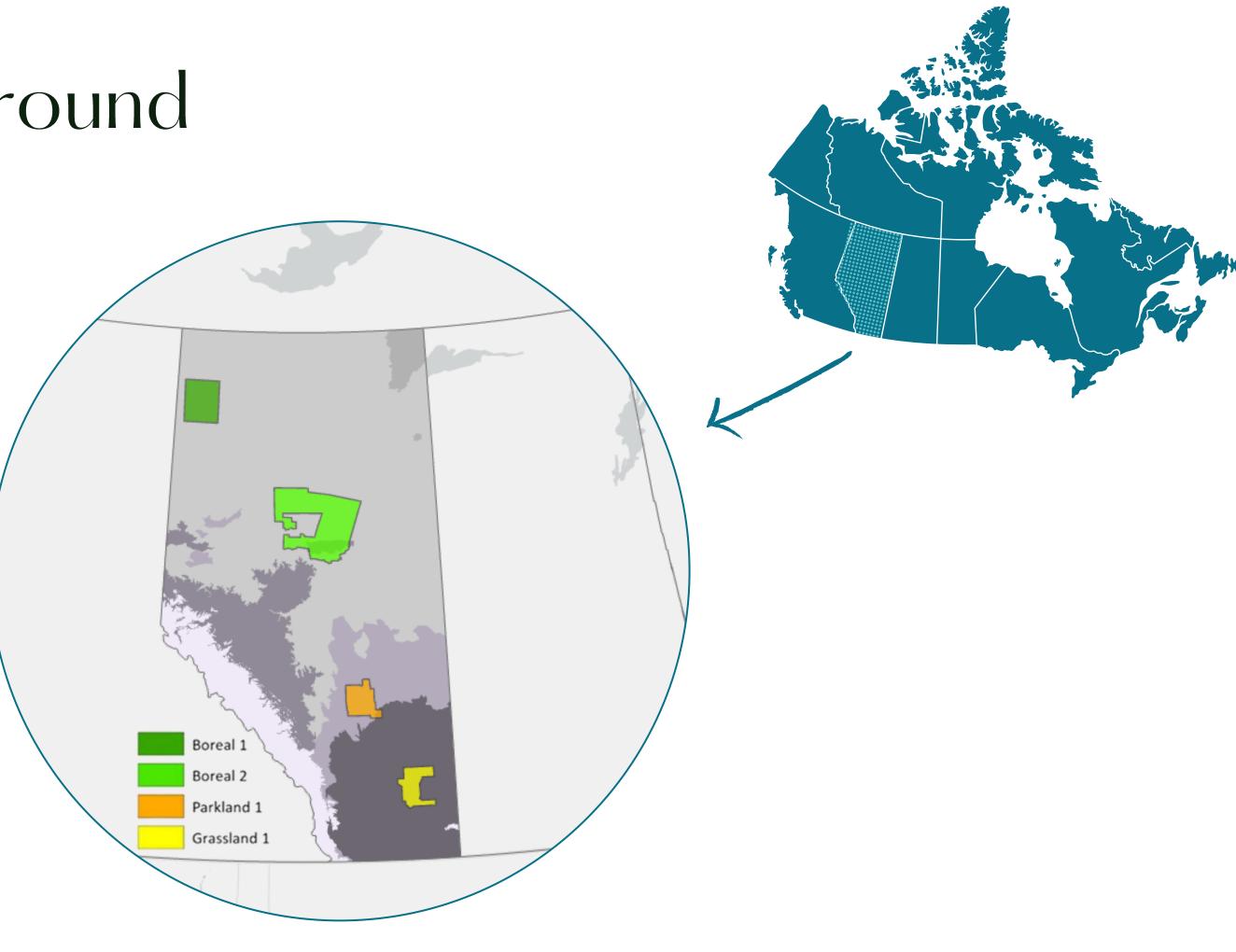
Cynthia McClain Jim Herbers Jen Hird John Simms Jenet Dooley Michael Merchant Syd Toni Thuy Doan Fiona Gregory Jacqueline Dennett

## Project Background

#### **Pilot study areas**

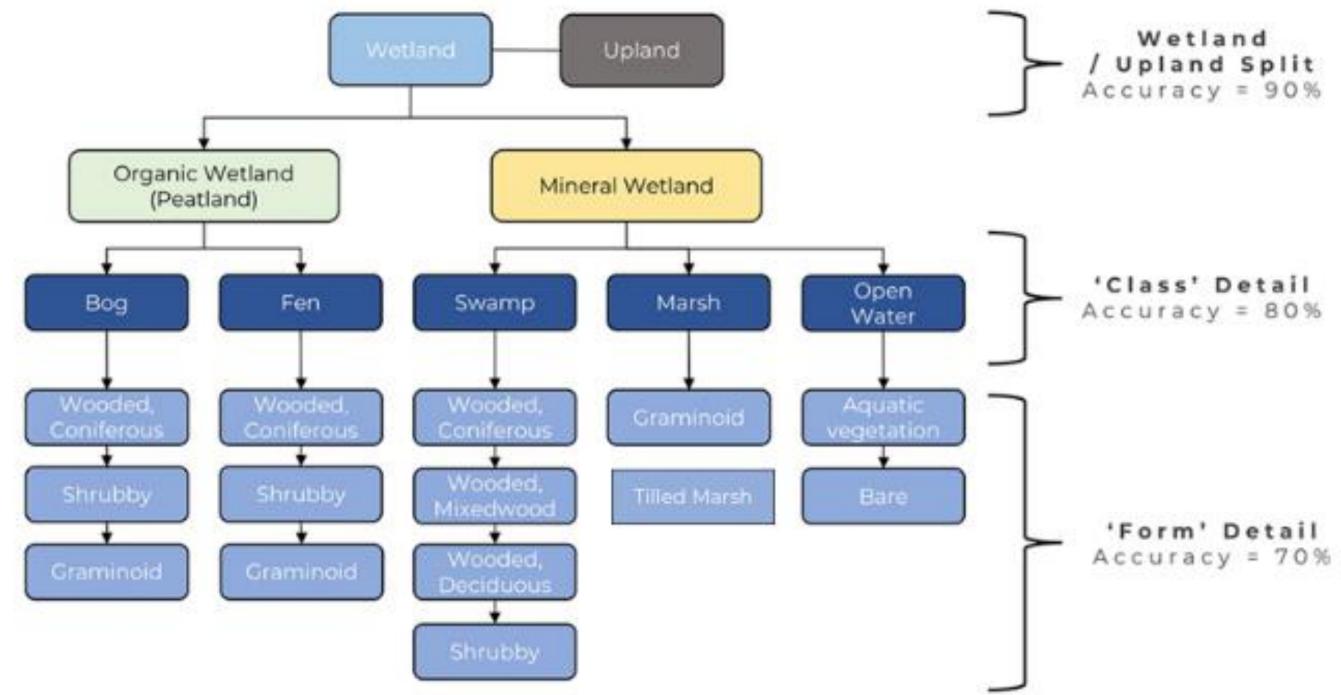
Four pilot areas within the Boreal, Parkland and Grassland Natural Regions.

These areas represent the broad diversity of wetland and land uses in the province



### Project Background

#### **Alberta Wetland Classification System**

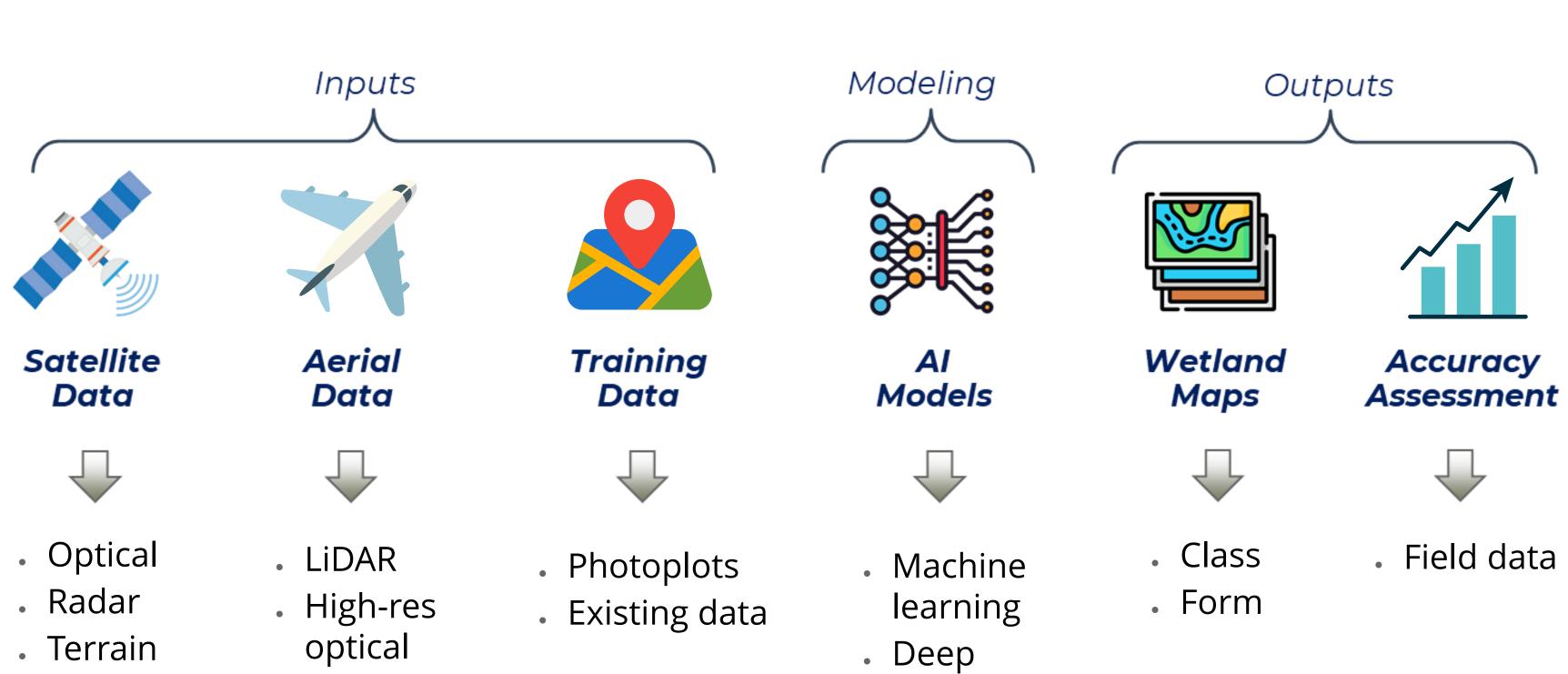




# Methods



## Methodology

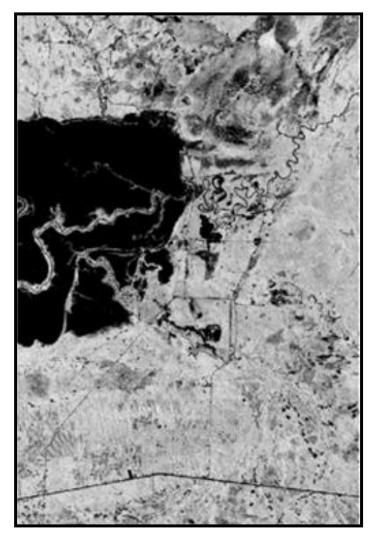


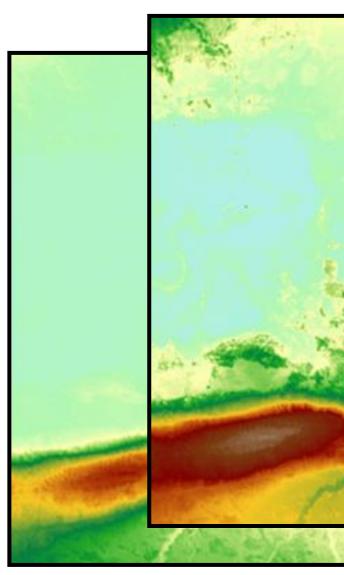
learning

### Satellite Data

#### **Boreal Metrics**







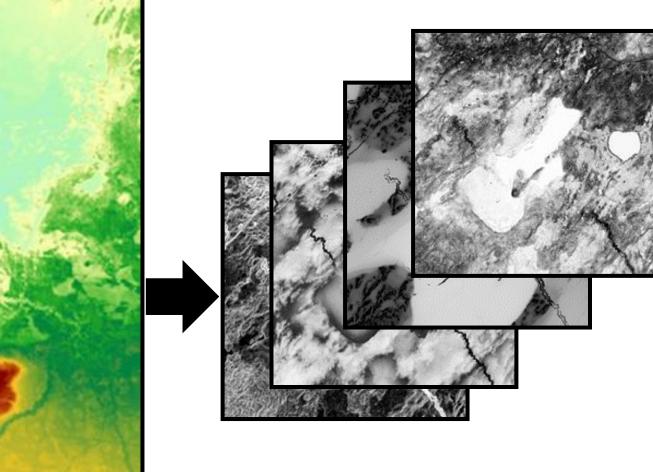
**Multispectral** Sentinel-2 (10m)

Radar **Sentinel-1** (10m)

Topography **DEM/LiDAR** (1m)



### Topographic Derivatives





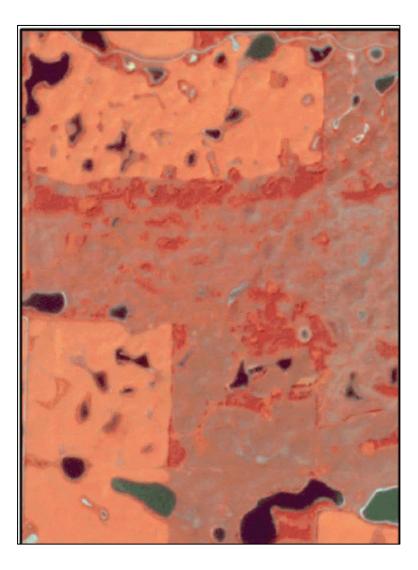


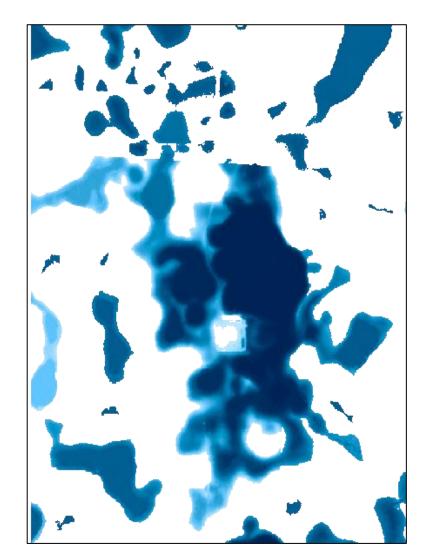


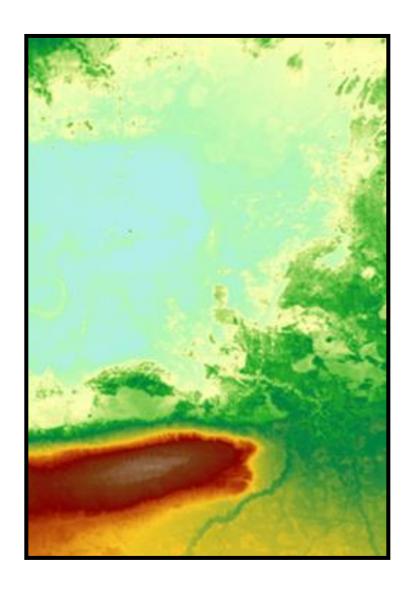


### Satellite Data

#### **Prairie Metrics**







**Multispectral** Planet (3m)

Terrain **High Res. DEM** 

**Topography LiDAR** 

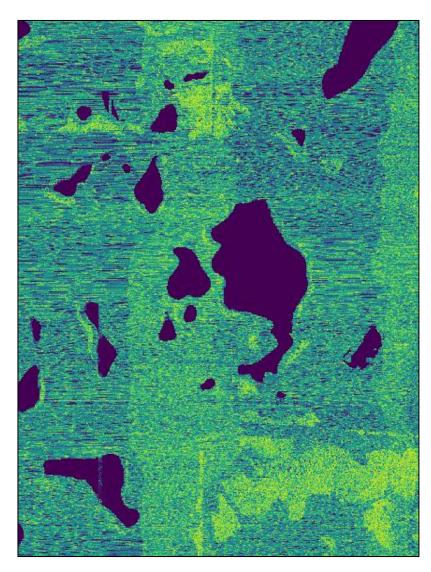
Alberta











### **Metrics derived** from Point Cloud

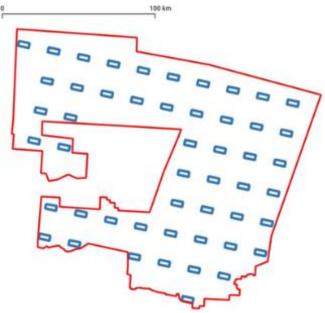
## Training Data

#### **Boreal 3x7 plots**

- ABMI 3x7 km photoplots were used for training AI models.
- Photoplots were created in 2016 and needed to be updated following standards.
- They were enhanced by DUC using the vegetation attributes, hydrological cues, canopy and tree species code etc.







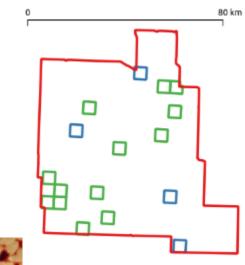
## Training Data

#### **Prairie 5x5 plots**

- DUC 5x5 km photoplots were used for training AI models.
- Photoplots were created using a combination of 2018, 2021, and 2023 stereo imagery.
- They were collected with existing CWI mapping protocols which were adapted to include open water and swamp forms.

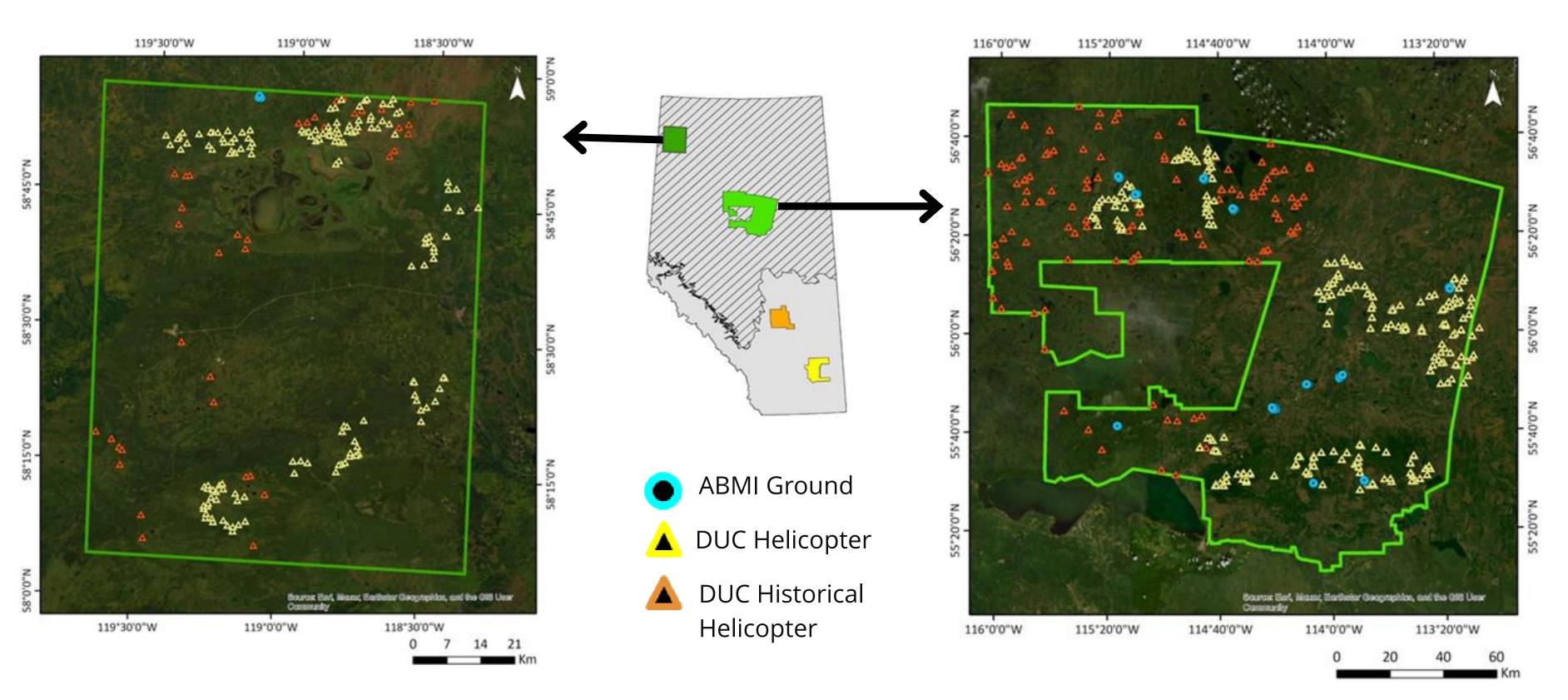


Marsh

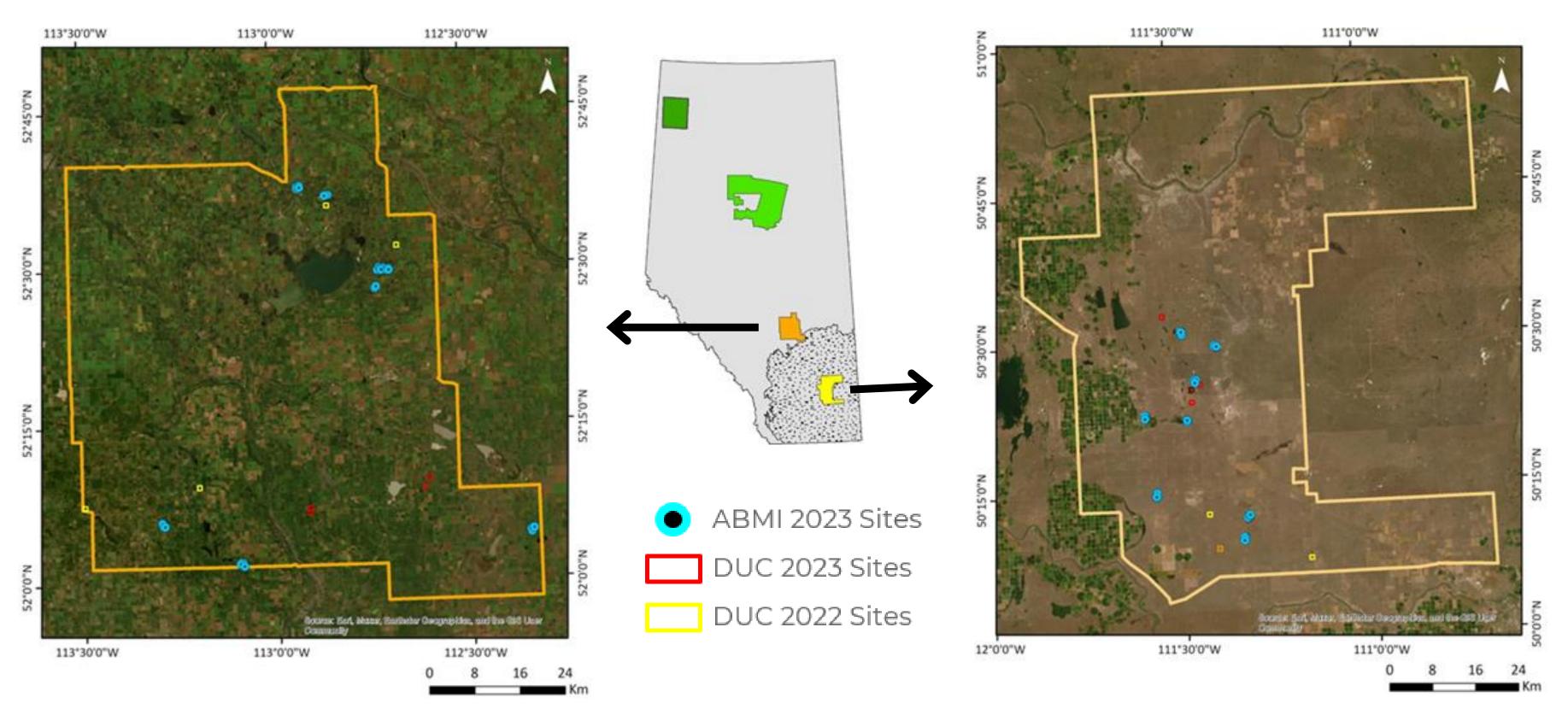




#### **Boreal Field Data Collection**

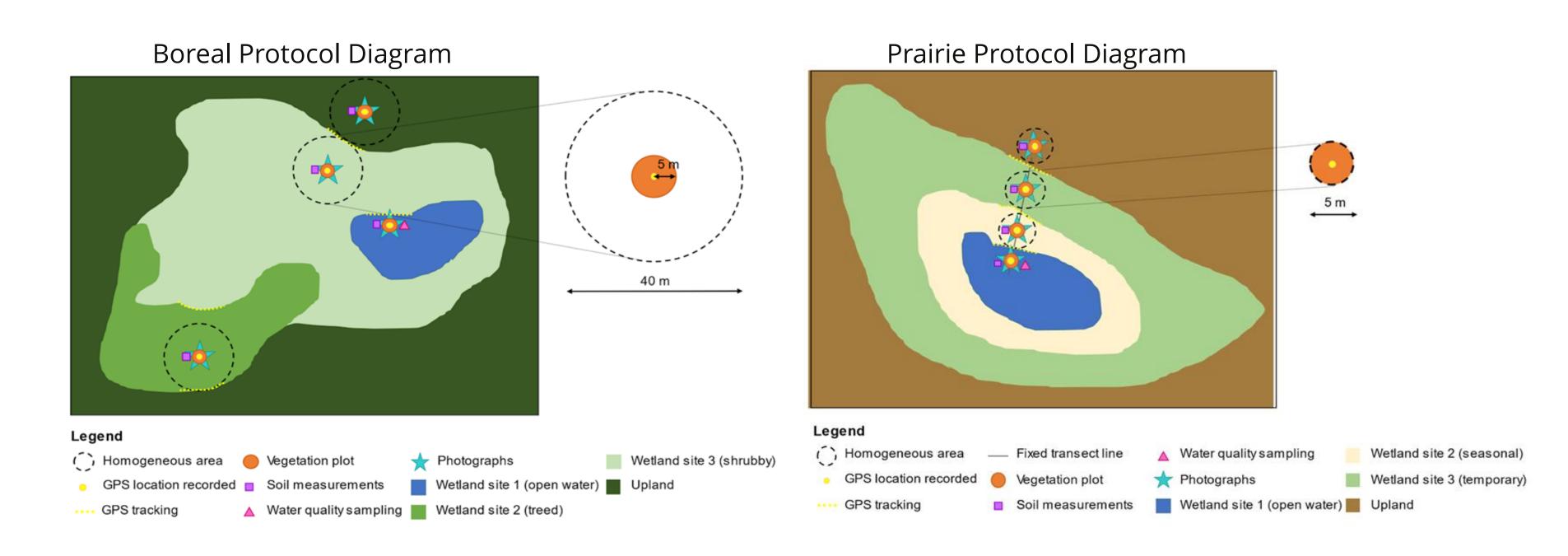


#### **Prairie Field Data Collection**



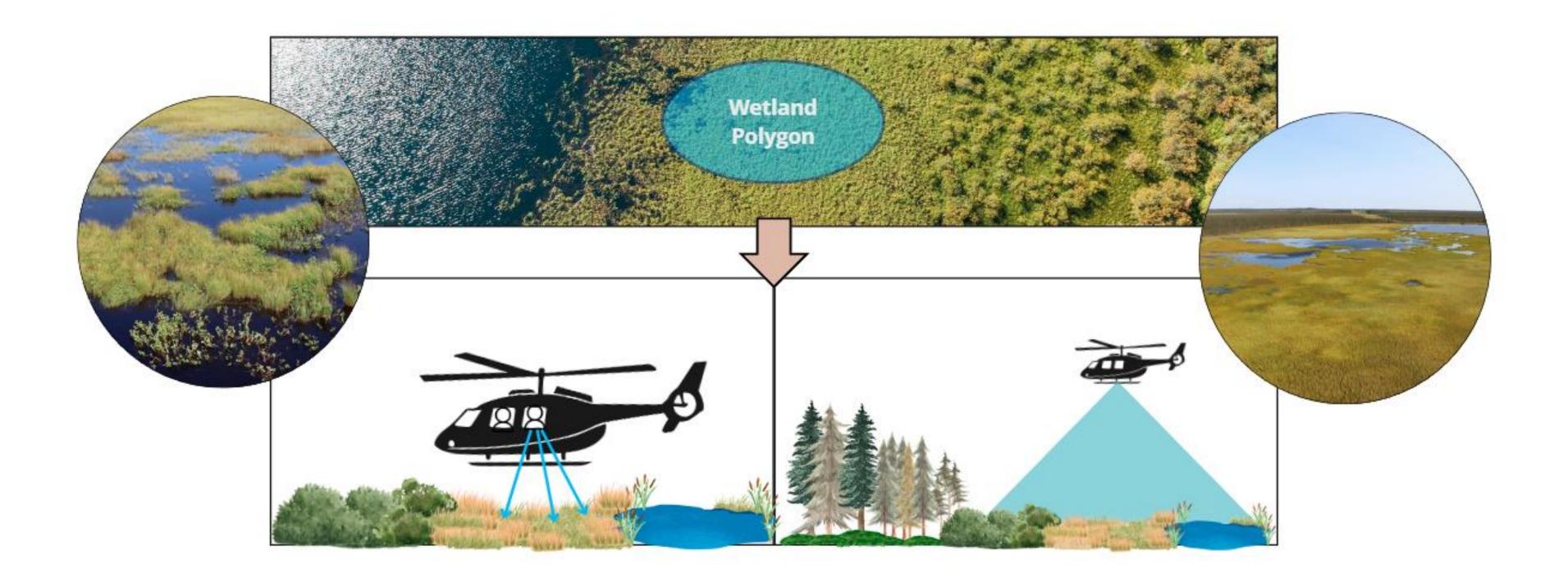
#### **Field Protocols**

Developed new protocols for field collection and classification of AWCS wetlands



#### **Helicopter-based Protocols**

Species type | Percent coverage (%) | Species heights (m) | Class/Form | Site photographs



### Validation Data **Prairie Field Protocols**



#### **Field Survey**

- validation polygons.
- Information collected:

  - Class/Form/Type

  - Disturbance
  - Site photographs

• Targeted sampling of individual wetlands (points/edge) to collect wetland representative features and minority classes. • Full parcel sampling conjunction with UAV

acquisition and photo interpretation to compile

 Vegetation species types • Vegetation species heights (m) • Vegetation percent cover (%) • Wetland edge / Soil observation • Hydrological indicators



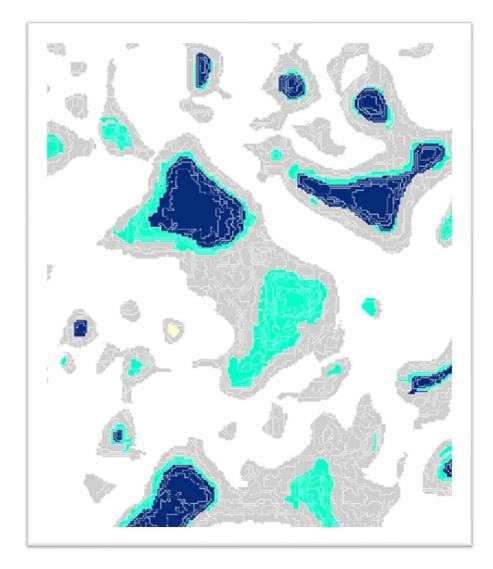
#### **Machine Learning for Boreal and Prairie**

Machine Learning relies on the algorithm to identify patterns and make predictions based on structure data



#### Boreal Point Based

- **Balanced Sampling**
- Pixels are independent
- **Ensemble learning** • methods XGBoost





Point samples

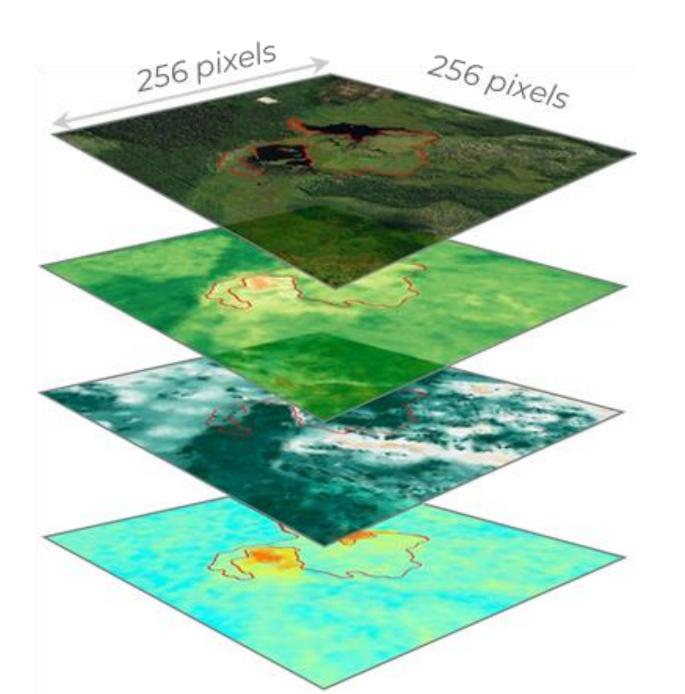
### Prairie Segment based

- **Balanced Sampling** •
- XGBoost and Random Forest evaluated

## Al Modeling

#### **Deep Learning for Boreal and Prairie**

Deep Learning uses neural networks to automatically extract features and learn from large volumes of unstructured data



#### **Deep Learning**

- **Convolutional Neural Network** ullet
- **U-Net Architecture**
- Segmentation model that trains using image patches



#### **Boreal Validation**

Overall Accuracy (%)						
Pilot	Al Model	Inputs	Wetland vs. Upland	Class	Form	
			(Standard 90%)	(Standard 80%)	(Standard 70%)	
Boreal-1	Deep Learning	Important features	98	82	58	
		Important features + lidar	96	68	71	
	Machine Learning	Important features	95	72	-	
		Important features + lidar	97	78	74	
Accuracy Benchmark						
Boreal-2	Deep Learning	Important features	95	81	63	
		Important features + lidar	95	81	65	
	Machine Learning	Important features	96	84	-	
		Important features + lidar	98	87	70	

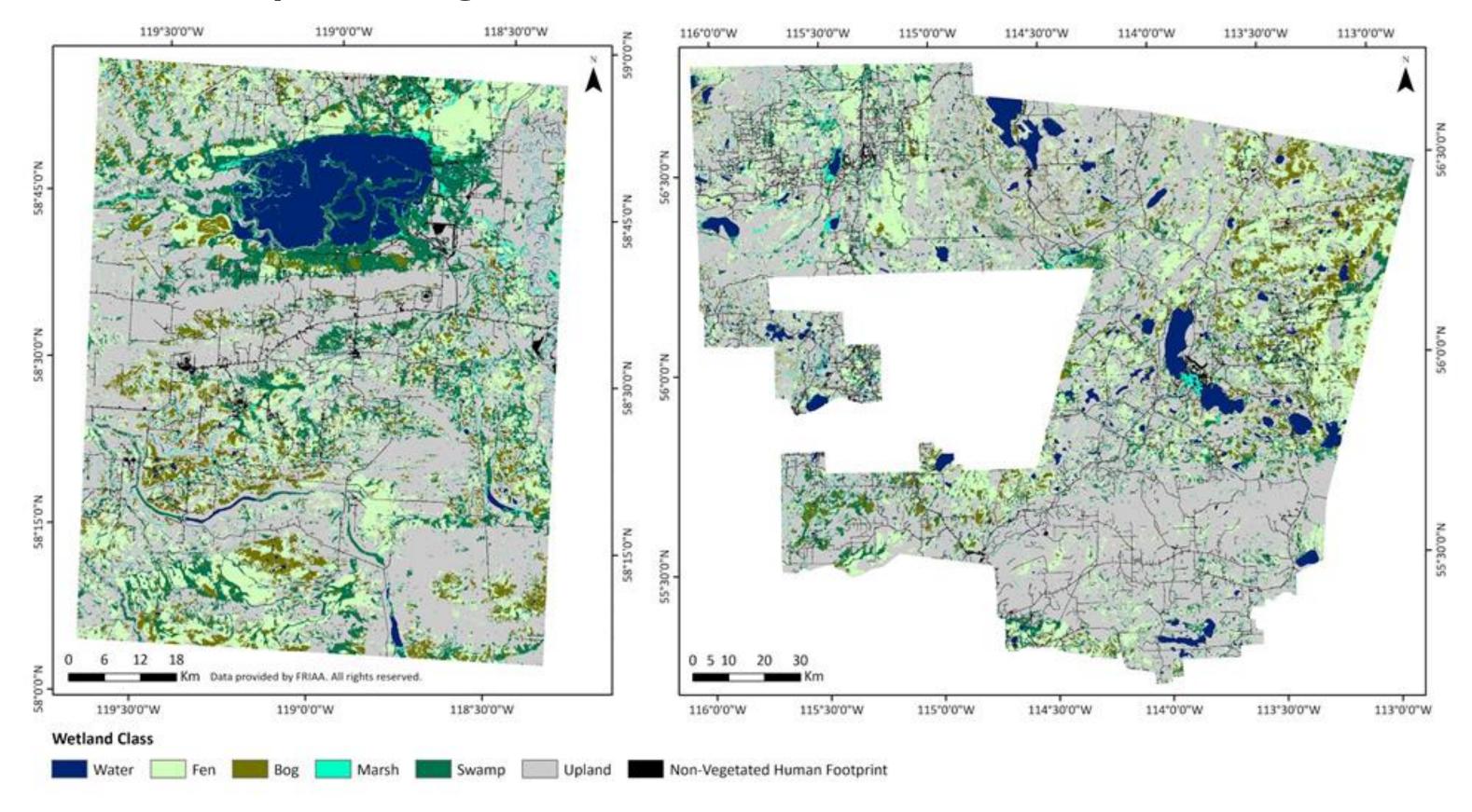
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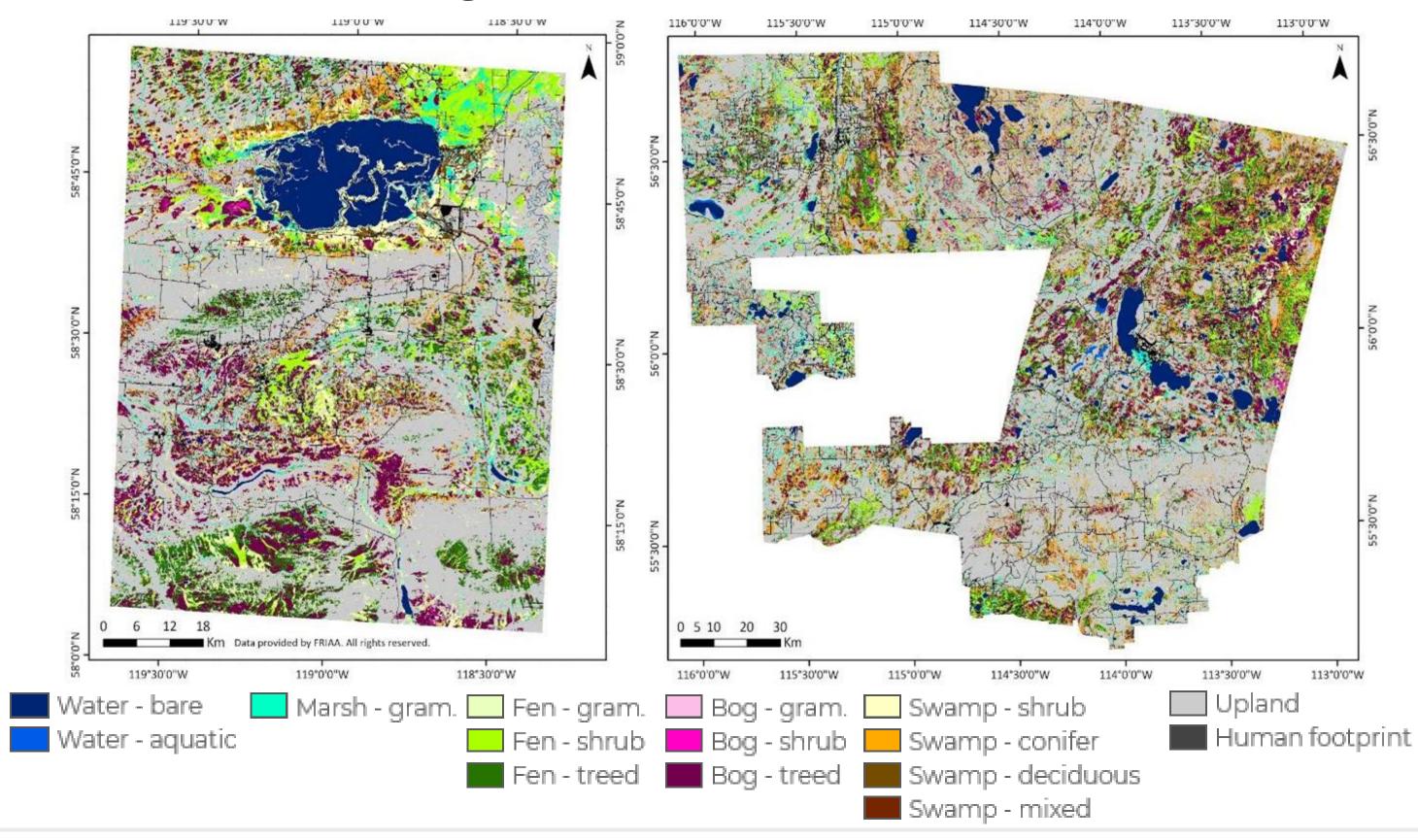




#### **Boreal Class Deep Learning**



#### **Boreal Form Machine Learning**

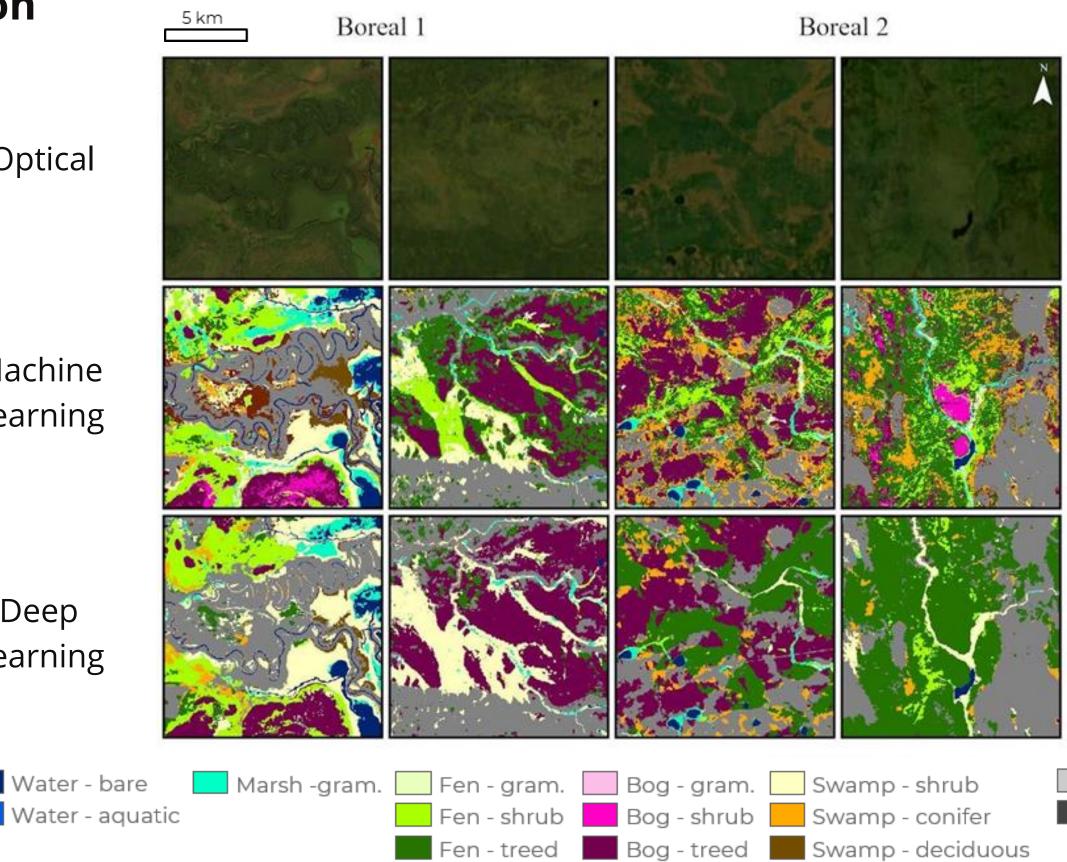


#### Comparison

Optical

Machine Learning

Deep Learning



- Follows the ecological gradient at the form level (i.e., transitional boundaries between treed, shrub and graminoid)
- More granular
- Enhanced detection of rare wetlands

- Less noise
- Enhanced boundaries

Swamp - mixed

Upland Human footprint

#### **Prairie Validation**

Overall Accuracy (%)					
Pilot	Al Model	Inputs	Wetland vs. Upland (Standard 90%)	<b>Class</b> (Standard 80%)	Form
Parkland-1	Deep Learning	Important features + LiDAR	85	64	64
	Machine Learning	Important features + LiDAR	87	66	66

#### Accuracy Benchmark

Grassland-1	Deep Learning	Important features + LiDAR	93	95	95
	Machine Learning	Important features + LiDAR	83	82	82

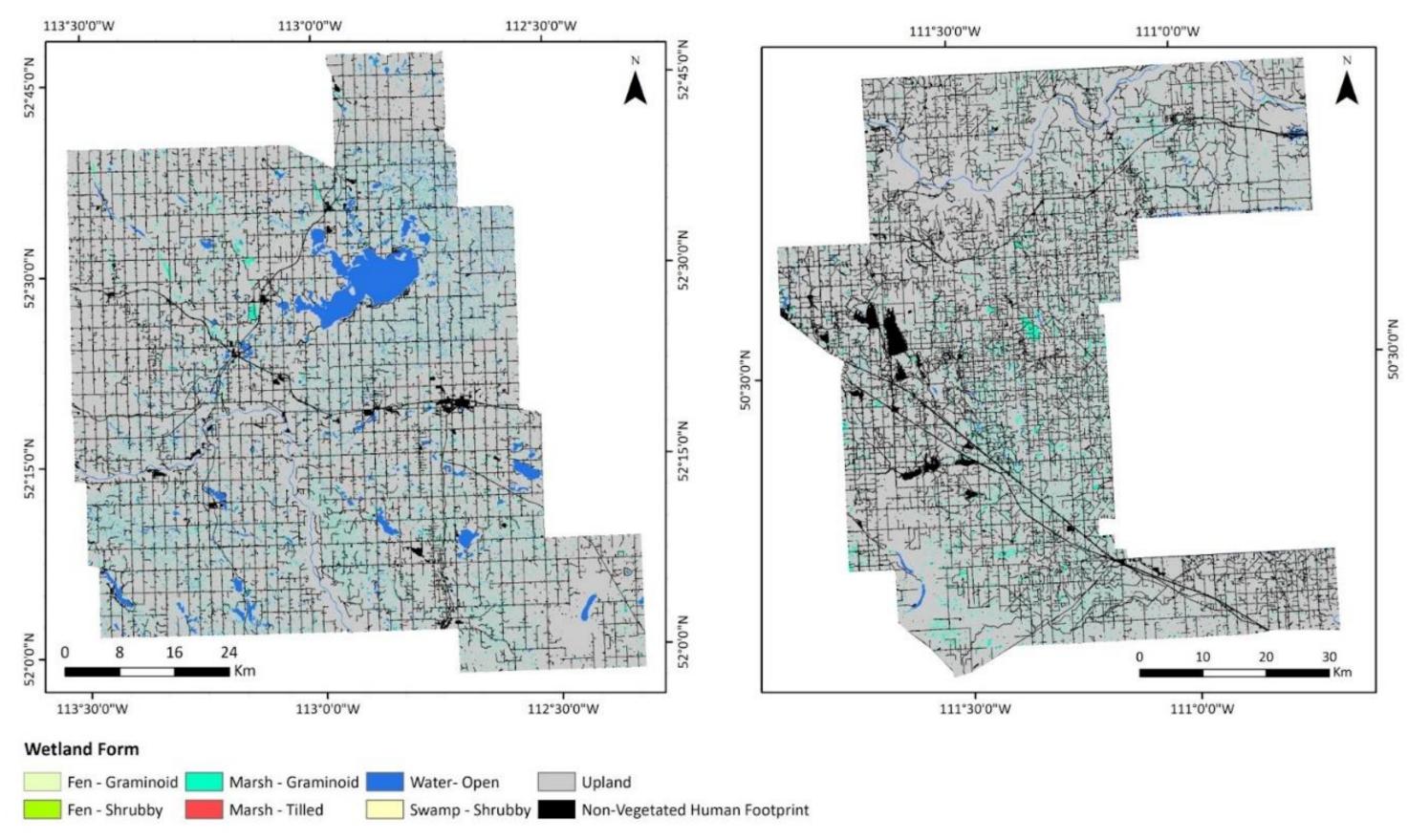
#### Accuracy Benchmark





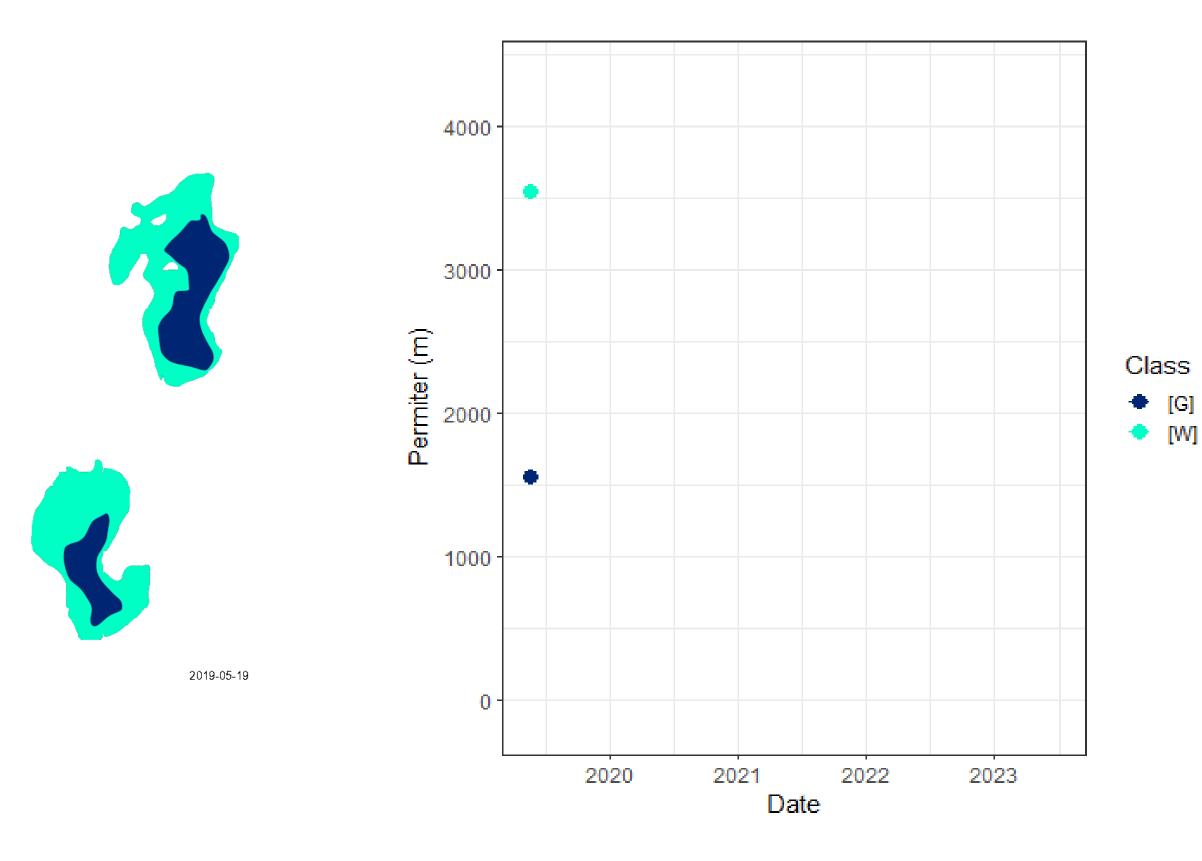


#### **Prairie Validation**



#### **Parkland vs Grassland**

- Parkland
  - More complex veg.
  - High seasonal variability
  - Most wetlands <1ha</li>
  - Challenging to align satellite imagery with ground validation



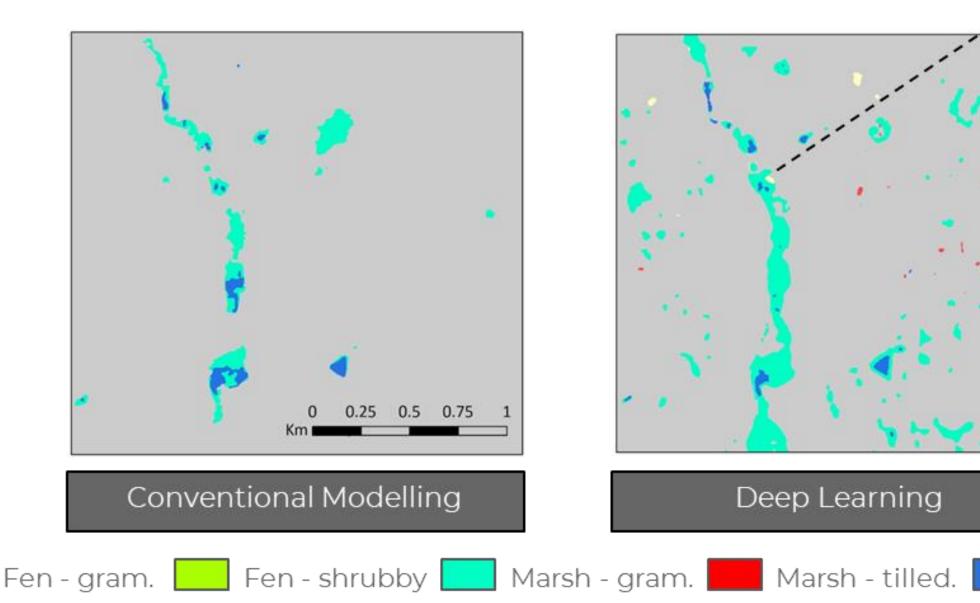


#### Open water

#### **Model Comparison**

 Only Machine Learning was capable of mapping parkland fen





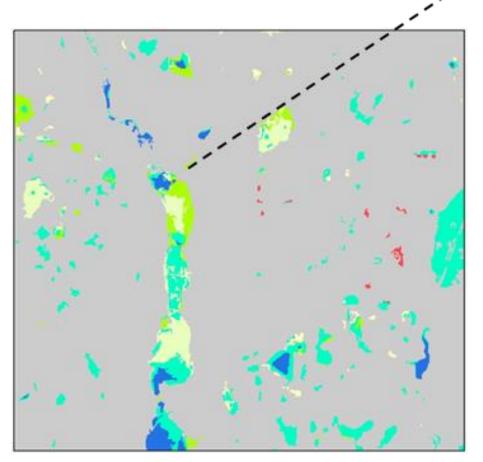


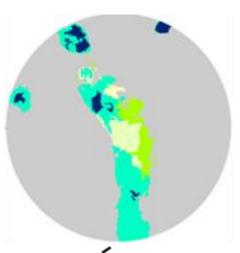
S

Swamp - shrubby



#### Machine Learning





# Take away Message



# Key takeaways

## Successfully Met Standards

Pilot project successfully tested and advanced AI methods. The top DL or ML methods in 3 pilot areas met the GOA's wetland mapping standards for accuracies >70%.

#### Resolution matters

### DL vs ML

Machine learning based methods require higher accuracy in the training data, whereas DL methods require more training data. We recommend a hybrid ML/DL approach for the model pipeline.

# LiDAR

In the Boreal, LiDAR had a significant impact on the form level. In the Prairie, bare earth is critical for defining isolated depressions, but point cloud derivatives (i.e., CHM) were negligible.

#### Field Data

Dedicate appropriate resources to collecting a large, detailed, and representative field reference dataset. Focus on rare or unique forms.

### NAWM's Wetland Mapping Consortium

# Thank you!

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