Wetland Inventory System for NL using Satellite Data (What has been Done and What Is the Next Step)

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Wetland Service

- Flood control
- Erosion control
- Water purification
- Shoreline protection
- Soil and water conservation
- Carbon storage
- Recreation and tourist activities

Wetlands in NL



Wetland Classification Methods

- Traditional (e.g., field work) Remote Sensing
 - Expensive
 - Time-consuming
 - Not practical for large areas
 - No practical for wetland change detection and monitoring
 - Accessibility issues
 - Necessary for remote sensing methods



- Cost effective
- Real-time data
- Large coverage
- Repetitive observation
- No limitation regarding the accessibility



Remote Sensing

- Remote sensing is defined as the measurement of object properties on the earth's surface using data acquired from aircrafts and satellites
- Remote sensing satellites for wetland classification:













Hyperspectral



SmallSats



NL WETLAND INVENTORY SYSTEM

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Importance of Wetland Classification in NL

- ~18% of NL is covered by wetlands
- Over the last decades, industrialization, urbanization, and agricultural activities have posed a serious threat to wetlands in the province
- Newfoundland has no comprehensive wetland inventory



Wetland Classification in NL (3 years ago)

- Only two small areas were classified
- Using basic methods
- Some sporadic field works



Wetland Classification in NL (up to now) Study areas





Wetland Classification in NL (up to now) Field data

- Conducted in summer 2015, 2016, 2017
- Ancillary information, including GPS points, on-site photographs, field notes on dominant vegetation, and hydrology were collected at each wetland site
- The GPS points were inserted into ArcMap and, then, the boundary delineation was conducted using high resolution images
- Based on Canadian Wetland Classification System (CWCS)
- Five wetland classes: Bog, Fen, Marsh, Swamp, and Shallow Water
- Three non-wetland classes: Deep Water, Urban, and Upland



Study area	Class	Number of polygons	Area (ha)
	Bog	83	269
	Fen	39	80
Avalon	Marsh	50	62
	Swamp	45	47
	Shallow Water	40	110
	Bog	30	357
Grand Falls	Fen	61	194
Windsor	Marsh	45	102
windsor	Swamp	30	47
	Shallow Water	21	52
	Bog	31	236
	Fen	54	121
Deer Lake	Marsh	24	19
	Swamp	40	56
	Shallow Water	23	68
	Bog	38	779
	Fen	31	98
Gros Morne	Marsh	31	50
	Swamp	42	48
	Shallow Water	27	64
	Bog	28	395
	Fen	29	139
Goose Bay	Marsh	21	78
	Swamp	23	35
	Shallow Water	11	19

Satellite data

- Optical
 - RapidEye
 - Landsat-8 (free)
 - Sentinel-2 (free)
 - ASTER (free)
- SAR
 - RADARSAT-2 (free)
 - ALOS-1 (free)
 - ALOS-2
 - Sentinel-1 (free)
 - TerraSAR-X
- Aerial
 - Canadian Digital Elevation Model (CDEM)



Preliminary analyses

Comparison between classification algorithms



Evaluation of confusion matrix

On	nission ((in %).	: User	Accurac	y, 1100	lucci A	accuracy.	, chois	or com	unssion ,	and e	1101 01
					R	eferenc	e Data					
		В	F	M	SW	S	Up	DW	Ur	Total	C	UA
	В	20279	4241	1476	0	590	0	0	83	26669	24	76
	F	1672	7368	841	0	987	0	0	0	10868	32	68
	М	541	70	3941	92	0	1295	0	0	5939	34	66
Data	SW	0	0	0	1445	0	0	60	0	1505	4	96
fied	S	78	740	859	0	3470	2316	0	32	7495	54	46
assi	Up	0	682	0	0	178	24304	0	0	25164	3	97
C	DW	0	0	0	0	0	0	69312	0	69312	0	100
	Ur	0	0	0	0	0	0	0	19340	19340	0	100
	Total	22570	13101	7117	1537	5225	27915	69372	19455	166292		-
	0	10	44	45	6	34	13	0	1		OA= 91	
	PA	90	56	66	94	66	87	100	99		K= 0.87	
OA: Overall Accuracy B: Bog			S: Swamp		C: Commission							
K: Kappa Coefficient F:		F: Fen		Up: Upland		O: Omission						
PA	: Produ	cer Accu	uracy	M: Ma	rsh		DW: D	eep Wat	er			
UA	A: User .	Accuracy	y I	SW: Sł	allow V	Water	Ur: Ur	ban			_	



Pixel-based vs. object-based method

Evaluation of multi-temporal satellite data

Accuracy	June (2015/06/19)	August (2015/08/15)	November (2015/11/26)	Combination of multi-date images
OA (%)	86	88	81	88
APA (%)	60	64	56	68
AUA (%)	57	60	53	63



Wetland Classification in NL (up to now) General method



Wetland classified maps





Wetland classified maps





Classification accuracies



Other analysis (spectral analysis of wetlands)

Evaluated optical features

The spectral bands of wavelength range for	f RapidEye, Sentinel 2A, each band is provided in th	ASTER, and Landsat 8 ne parentheses, and is in r	used in this study (Th nicrometers).		
RapidEye Sentinel 2A		ASTER	Landsat 8		
B1_Blue (0.44-0.51)	B2_Blue (0.46-0.52)	B1_Green (0.52-0.6)	B2_Blue (0.45-0.51)		
B2_Green (0.52-0.6)	B3_Green (0.54-0.58)	B2_Red (0.63-0.69)	B3_Green (0.53-0.59)		
B3_Red (0.63-0.69)	B4_Red (0.65-0.68)	B3N_NIR (0.76-0.86)	B4_Red (0.64-0.67)		
B4_RE (0.69-0.73)	B5_RE (0.698-0.712)	B10_TIR (8.13-8.48)	B5_NIR (0.85-0.88)		
B5_NIR (0.76-0.85)	B6_RE (0.733-0.747)	B11_TIR (8.48-8.83)	B6_SWIR (1.57-1.67)		
	B7_RE (0.773-0.793)	B12_TIR (8.93-9.28)	B7_SWIR (2.11-2.29)		
	B8_NIR (0.784-0.9)	B13_TIR (10.25-10.95)	B10_TIR (10.6-11.19)		
	B8A_RE (0.855-0.875)	B14_TIR (10.95-11.65)			
	B11_SWIR (1.565-1.655))			
	B12_SWIR (2.1-2.28)				
The spectral indices of wetland types.	, textural and ratio features	evaluated in this study for	r separability analyses		
Spectral indices	$Brightness, NDWI = \frac{Gree}{Gree}$ $RE-NDVI = \frac{NIR-RE}{NIR+RE}, NI$ 0.5	$\frac{en-NIR}{en+NIR}, DVI=NIR-RedDSI=\frac{SWIR-NIR}{SWIR+NIR}, SAVI=0$	$\frac{d}{NIR+Red}, \frac{NIR-Red}{NIR+Red}, \frac{(1+L)(NIR-Red)}{NIR+Red+L}; L =$		
Ratio features	Blue Brightness' SWIR Brightness	' Red RE ' Brightness ' Brightness	' NIR Brightness'		
Texture features	Standard deviation of the the satellites.	polygons obtained from a	ll spectral bands of		

Visual analysis (spectral signature)



Statistical analysis



Best optical spectral bands (summary)

The most useful spectral bands for discriminating between each pairs of wetland classes in late Spring (June) using two distance measures. The spectral bands are ordered based on their separability measures (the spectral bands indicated in the upper right half of the table and the lower left half of the table were obtained by the T-statistics and U-statistics, respectively).

	Bog	Fen	Marsh	Swamp	Shallow Water
Bog	×	Landsat8 NIR RapidEye RE ASTER NIR RapidEye NIR	RapidEye RE Landsat8 NIR	RapidEye Red	RapidEye NIR
Fen	Sentinel2 RE5 Landsat8 NIR	×	Landsat8 NIR RapidEye RE	RapidEye Red	RapidEye NIR
Marsh	Landsat8 NIR RapidEye RE	RapidEye RE	×	Landsat8 NIR RapidEye RE	RapidEye RE
Swamp	Landsat8 Red	Sentinel2 RE7 Sentinel2 RE6 Sentinel2 NIR ASTER NIR Sentinel2 RE8 RapidEye Red	Landsat8 NIR RapidEye RE	×	RapidEye NIR
Shallow Water	Landsat8 NIR Landsat8 SWIR6	Landsat8 NIR Landsat8 SWIR6 Landsat8 SWIR7	Sentinel2 RE6	Sentinel2 RE6	×



Other analysis (backscattering analysis of wetlands)

Evaluated SAR features

SAR features investigated for se	parability analysis of wetland classes (80 features in total).							
Sentinel-1								
(5 features): horizontal transmit and horizontal receive polarization of Sentinel-1 (S1_HH), horizontal transmit and vertical receive polarization of Sentinel-1 (S1_HV),								
vertical transmit and vertical reco	eive polarization of Sentinel-1 (S1_VV), S1_HH/HV= $\frac{ S_{HH} ^2}{ S_{HV} ^2}$, S1_VV/HV= $\frac{ S_{VV} ^2}{ S_{HV} ^2}$							
	ALOS-2							
(3 features): horizontal transmit	and horizontal receive polarization of ALOS-2 (A2_HH), horizontal transmit and vertical receive polarization of ALOS-2 (A2_HV),							
$A2_HH/HV = \frac{ S_{HH} ^*}{ S_{HV} ^2}$								
	RADARSAT-2							
Covariance matrix (16 features)	Diagonal elements (C11= S _{HH} ² , C22= S _{HV} ² , C33= S _{VV} ²), Total Power (TP = S _{HH} ² + 2 S _{HV} ² + S _{VV} ²), Off-diagonal elements,							
	Polarization ratios (R2_HH/HV= $\frac{ S_{HH} ^2}{ S_{HV} ^2}$, R2_VV/HV= $\frac{ S_{VV} ^2}{ S_{HV} ^2}$, R2_HH/VV= $\frac{ S_{HH} ^2}{ S_{VV} ^2}$, R2_HH/TP= $\frac{ S_{HH} ^2}{TP}$, R2_HV/TP= $\frac{ S_{HV} ^2}{TP}$,							
	$R2_VV/TP = \frac{ 5y_U ^2}{m_P}$							
Coherency matrix (9 features)	Diagonal elements (111, 122, 133), Off-diagonal elements							
Freeman-Durden (3 features)	Surface scattering (odd_F), Double-bounce scattering (dbl_F), Volume scattering (vol_F)							
Van Zyl (Van Zyl, 1989), (3 features)	3 Surface scattering (odd_V), Double-bounce scattering (dbl_V), Volume scattering (vol_V)							
Yamaguchi (Yamaguchi et al., 2005), (4 features)	Surface scattering (odd_Y), Double-bounce scattering (dbl_Y), Volume scattering (vol_Y), Helix scattering (hlx_Y)							
Krogager (Krogager, 1990), (3 features)	Sphere scattering (Ks), Diplane scattering (Kd), Helix scattering (Kh)							
Cloude-Pottier (7 parameters)	Cloude-Pottier (7 parameters) Anisotropy (A), Entropy (H), Alpha, Beta, Delta, Gamma, Lambda							
Touzi (4 features)	Symmetric scattering type magnitude (<u>Alpha_s</u>), Symmetric scattering type phase (<u>P_Alpha_s</u>), Orientation angle (ψ), Helicity (τ)							
Eigen value parameters (23 features)	First pseudo probability (P1 = $\frac{\lambda_1}{\lambda_1 + \lambda_2 + \lambda_3}$), Second pseudo probability (P2 = $\frac{\lambda_2}{\lambda_1 + \lambda_2 + \lambda_3}$), Third pseudo probability (P3 = $\frac{\lambda_3}{\lambda_1 + \lambda_2 + \lambda_3}$), First Eigen value (Lambda1), Second Eigen value (Lambda2), Third Eigen value (Lambda3), Shannon entropy (SH), Normalized Shannon entropy (NSH), Polarimetry Shannon entropy (PSH), Normalized polarimetry Shannon entropy (NSH), Polarimetry Shannon entropy (NSH), Normalized intensity Shannon entropy (NSH), Normalized intensity Shannon entropy (NSH), Anisotropy (AE), Luenburg Anisotropy (LA), Anisotropy 12 (A12), Single-bounce eigenvalues relative difference (Merd), Normalized ouble-bounce eigenvalues relative difference (Merd), Normalized Neurone eigenvalues relative difference (Merd), N							



Best SAR features and decomposition methods(summary)

The most useful SAR features (provided in the upper right half of the table) and the decomposition methods (provided in the lower left half of the table) for discriminating between each pair of wetland classes. The features and decomposition methods are ordered based on their separability measures (refer to Table 5.3 for the acronyms of the selected SAR features).

	Bog	Fen	Marsh	Swamp	Shallow Water
Bog	×	Alpha	Alpha_s	R2_HH/TP	A
		Alpha_s	Alpha	R2_HH/HV	N_derd
		dbl_F	vol_F	A12	R2_HH/VV
		Н	А	A2_HH/HV	
		S1_HH/HV	R2_HV/TP	PA	
		odd_Y	P2	N_serd	
		PA	S1_HH/HV	S1_HH/HV	
Fen	H/A/Alpha	×	S1_HH/HV	A2_HH/HV	N_derd
	Freeman-Durden		N_derd	serd	A
			А	R2_HH/HV	R2_HH/VV
			R2_HH/HV	R2_HH/TP	R2_HH/HV
			R2_HH/TP	N_serd	N_serd
			R2_HV/TP	A12	serd
			dbl_F	P1	R2_VV/HV
Marsh	H/A/Alpha	Freeman-Durden	×	R2_HH/HV	A
	Freeman-Durden	H/A/Alpha		R2_HV/TP	S1_VV/HV
				R2_HH/TP	N_derd
				A2_HV	R2_VV/TP
				serd	R2_VV/HV
				N_derd	N_serd
				N_serd	PF
					serd
Swamp	H/A/Alpha	Freeman-Durden	H/A/Alpha	×	R2_HH/HV
			Freeman-		N_derd
			Durden		A
					serd
					N_serd
					R2_VV/HV
					PF
					R2_HH/TP
Shallow Water	H/A/Alpha	-	H/A/Alpha	-	×
	Freeman-Durden				

Other analysis (Advanced classification algorithms to improve the accuracy)



Selected Journal Publications

- Amani, M., Salehi, B., Mahdavi, S., Granger, J., & Brisco, B. (2017). "Wetland classification in Newfoundland and Labrador using multi-source SAR and optical data integration". GIScience & Remote Sensing, 54(6), 779-796.
- Amani, M., Salehi, B., Mahdavi, S., Granger, J. E., Brisco, B., & Hanson, A. (2017). "Wetland Classification Using Multi-Source and Multi-Temporal Optical Remote Sensing Data in Newfoundland and Labrador, Canada". Canadian Journal of Remote Sensing, 43(4), 360-373.
- Mahdavi, S., Salehi, B., Amani, M., Granger, J. E., Brisco, B., Huang, W., & Hanson, A. (2017).
 "Object-based classification of wetlands in Newfoundland and Labrador using multi-temporal PolSAR data". Canadian Journal of Remote Sensing, 43(5), 432-450.
- Mahdavi, S., Salehi, B., Granger, J., Amani, M., & Brisco, B. (2017). "Remote sensing for wetland classification: a comprehensive review". GIScience & Remote Sensing, 1-36.
- Amani, M., Salehi, B., Mahdavi, S., Brisco, B., & Shehata, M. (2018). "A Multiple Classifier System to improve mapping complex land covers: a case study of wetland classification using SAR data in Newfoundland, Canada". International Journal of Remote Sensing, 1-14.
- Amani, M., Salehi, B., Mahdavi, S., & Brisco, B. (2018). "Spectral analysis of wetlands using multi-source optical satellite imagery". ISPRS Journal of Photogrammetry and Remote Sensing, 144, 119-136.

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Wetland Classification in NL (Next Step)

Limitation of the previous works

- Wetland maps from only 5 pilot sites (1% of the province in total)
- No wetland map from the entire province
- Did not evaluate the changes over years and did not estimate the amount of loss/gain in wetland areas
- Did not consider most of the non-wetland classes
- Conducted solely based on remote sensing
- Some of the satellite data were not free

Wetland Classification in NL (Next Step) Preliminary map

• NL-wide wetland map in 2018 using multi-date landsat-8 imagery



Wetland Classification in NL (Next Step)

How to improve the accuracy of maps

- Use a combination of optical (Landsat-8+Sentinel-2) and SAR (Sentinel-1) data
- Include more field data
- Include more non-wetland classes
- Remove small classified areas to make the maps less noisy
- Produce wetland maps each year and evaluate the changes, gain, and loss in wetland areas over time
- Relate the results to the other variables (carbon storage, ducks migration, etc.)

Wetland Classification in NL (Next Step)

Support / Collaboration





Thanks

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