

Automated Classification of Operational SAR Sea Ice Images

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Why Processing Sea Ice Images

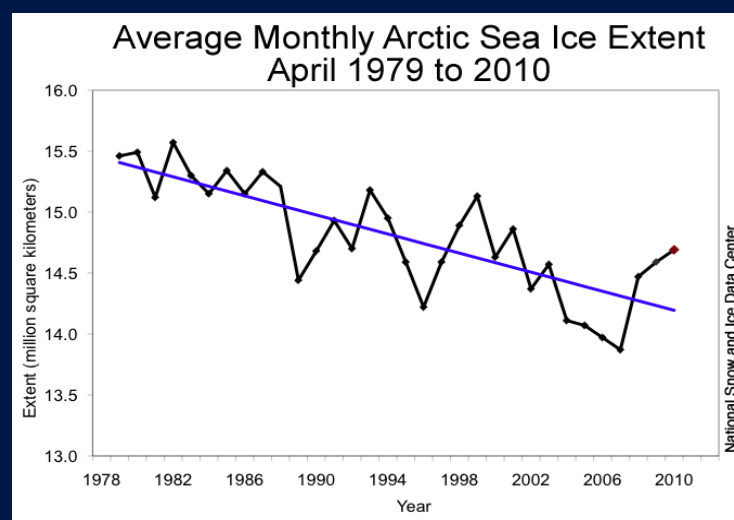
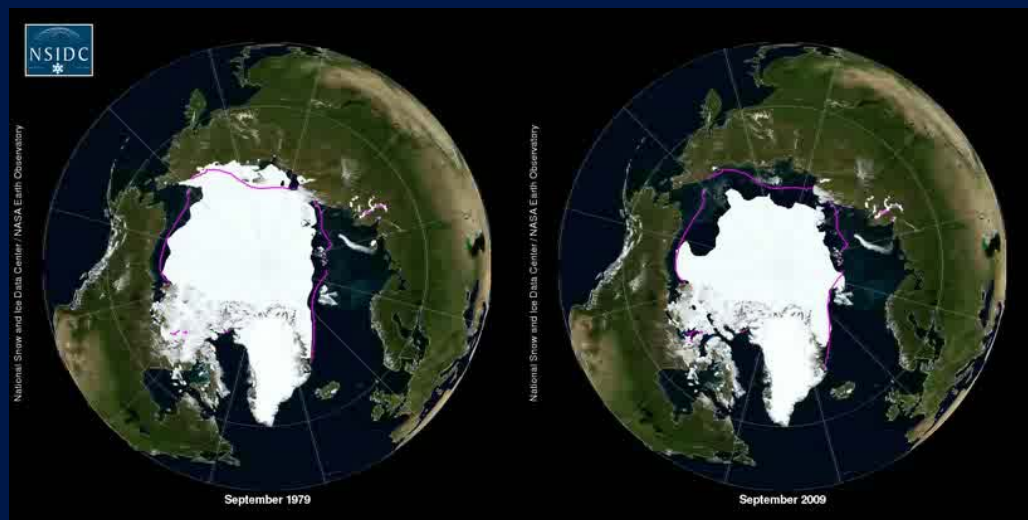
•Ship Navigation

- CIS manually processing approximately 4000, 10Kx10K SAR sea-ice images, annually

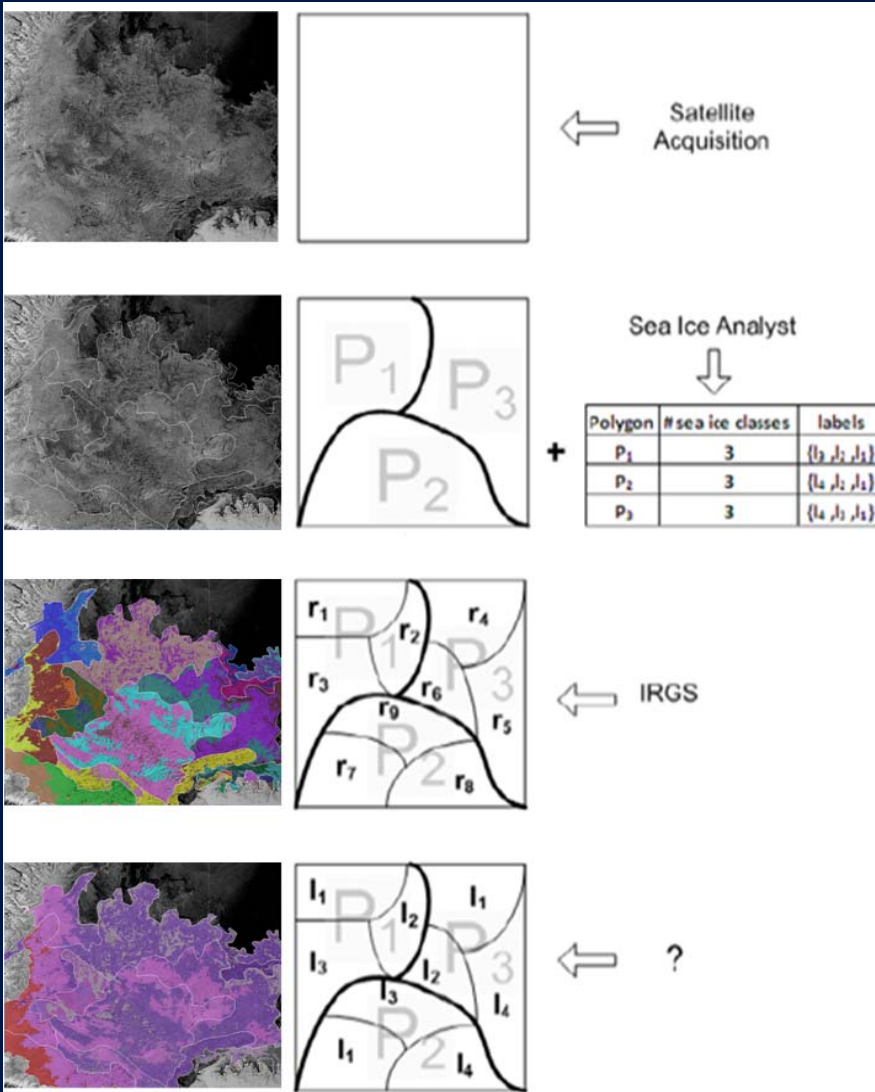
•Climate Research

- Importance of implications of climate change

The ships are trapped in ice in the Baltic Sea of Sweden



Problem Definition



Large synthetic aperture radar (SAR) scenes

Trained ice analysts divide SAR images into "polygon" areas and then identify the number and type of ice classes per polygon

Full scene classification can be performed by first segmenting each polygon into distinct regions algorithmically.

How to perform labeling if there is insufficient information to assign a sea ice label for each region within an individual polygon?

Sea Ice Labeling Model

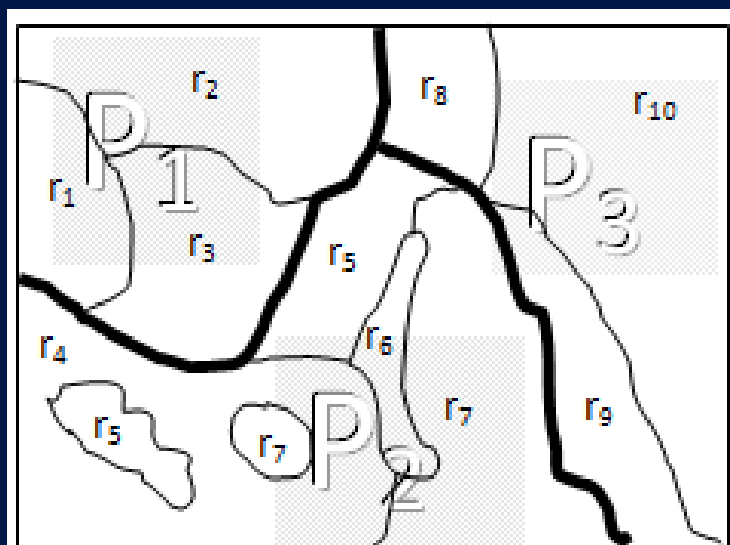
- *Markov random field formulation using joint information to label each region in a full SAR scene has been developed to solve sea ice labelling problem*
- *The model effectively defines the spatial and statistical relationship between the regions across the full SAR sea ice scene*

Implemented Method (highlights)

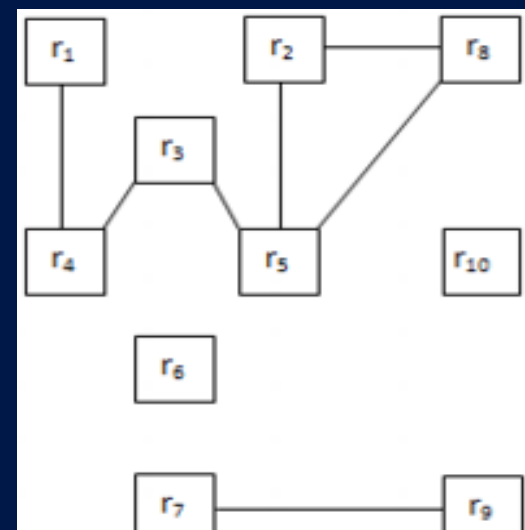
- ***MRF model for sea ice labeling***
 - *is compatible with any segmentation technique*
 - *utilize new neighbourhood for polygon interactions*
 - *utilize edge penalty as a spatial term*
 - *includes domain knowledge as a constraint*
 - *incorporates information from all polygons in the definition of energy*
 - *utilize simulated annealing and Metropolis sampling for global optimization*

Polygonal Neighbourhood

- *Two regions are polygonal neighbours if the boundary exist between them and they do not belong to the same polygon*

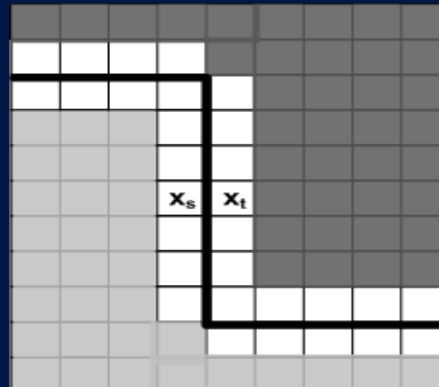


The example of segmentation result



Polygonal neighbourhood

Edge Strength



White are boundary sites

- *The boundary is the set of all sites connecting two regions based on first-order neighbourhood*

$$\partial r_{ij} = \{s, t | s \in r_i, t \in r_j, t \in N_s\}$$

- *Edge strength is the absolute difference between the boundary sites*

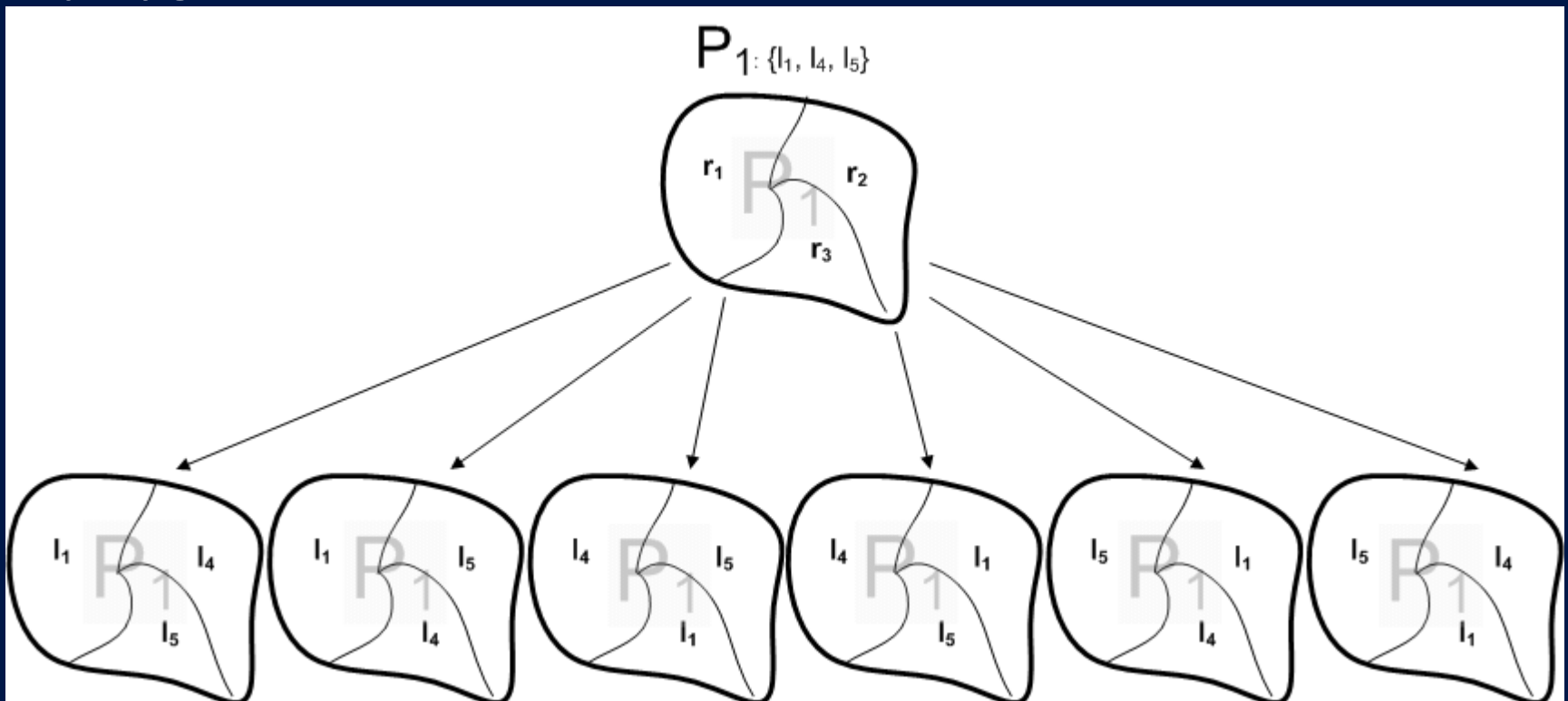
$$\nabla_{r_i r_j} = \sum_{s, t \in \partial r_{i, j}} |x_s - x_t|$$

- *Edge penalty is high if the edge strength is small*

$$g(\nabla_{r_i r_j}) = 1 - \nabla_{r_i r_j}$$

Incorporating Domain Knowledge

- The labelling realization of regions in any polygon is constrained to permutation of sea-ice classes given for that polygon*



MAP estimation as Energy Minimization

Feature energy model Gaussian mixture

$$E(r_i) = \sum_{s \in r_i} \left\{ \frac{1}{2} \ln(2\pi)^K |\Sigma_{z_i}| + \frac{1}{2} (x_s - u_{z_i})^T \Sigma_{z_i}^{-1} (x_s - u_{z_i}) \right\}$$

Feature energy of a single region

$$E_f = \sum_{i=1}^{n_r} E(r_i)$$

Clique Energy Model

$$V_2(r_i, r_j)_{(r_i, r_j) \in N_r} = \begin{cases} \beta g(\nabla_{r_i, r_j}) & z_i \neq z_j \\ 0 & z_i = z_j \end{cases}$$

Pair wise clique potential

$$E_r = \sum_{r_i, r_j \in N_p} V_2(r_i, r_j)$$

MAP estimation

$$\arg \min_{\{z_s | s \in S\}} (\alpha E_f + \beta E_r)$$

Optimization

- *The goal is optimizing objective function:*

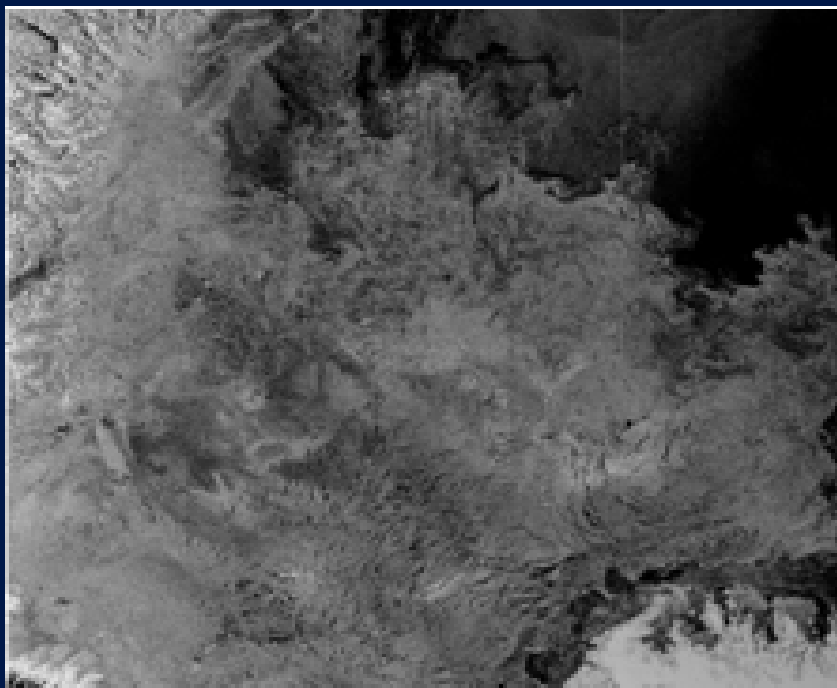
$$\arg \min_{\{z_s | s \in S\}} \alpha \sum_{i=1}^{N_r} \sum_{s \in r_i} \left\{ \frac{1}{2} \ln(2\pi)^K |\Sigma_{z_i}| + \frac{1}{2} (x_s - u_{z_i})^T \Sigma_{z_i}^{-1} (x_s - u_{z_i}) \right\} \\ + \beta \sum_{r_i, r_j \in N_p} \left\{ [1 - \delta(z_i, z_j)] \left[1 - \sum_{s, t \in \partial r_{i,j}} |x_s - x_t| \right] \right\}$$

- *Using*
 - *Simulated annealing and Metropolis Sampling*

Iteration	region	polygon	size M	size N	time
100	55	22	5008	5387	79s

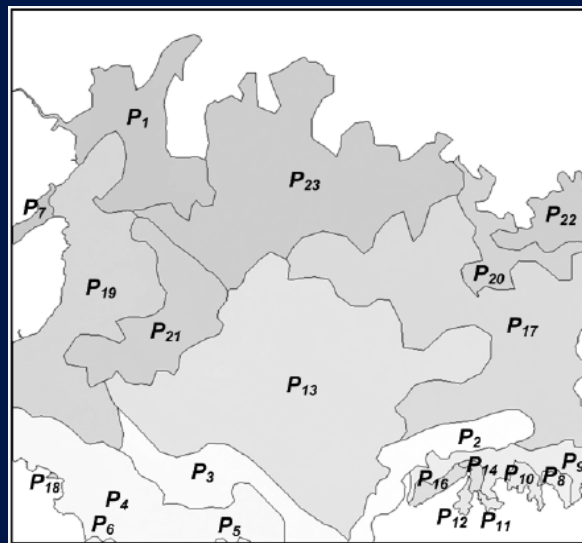
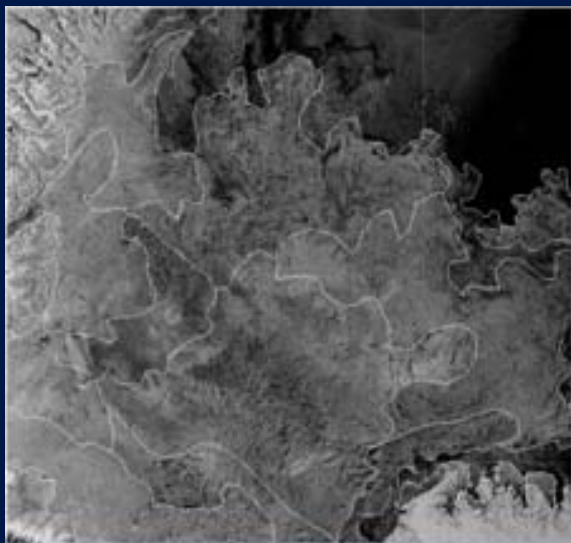
Operational SAR Sea Ice Image

- **Operational SAR Sea Ice Image of Arctic region (Baffin Bay)**
 $\{69.7270^\circ, 76.9649^\circ\} /$
 $\{-80.5216^\circ, -66.0010^\circ\}$



- **Acquired by RADARSAT-1 satellite in Scan SAR Wide mode on October 30, 2005**

CIS Metadata



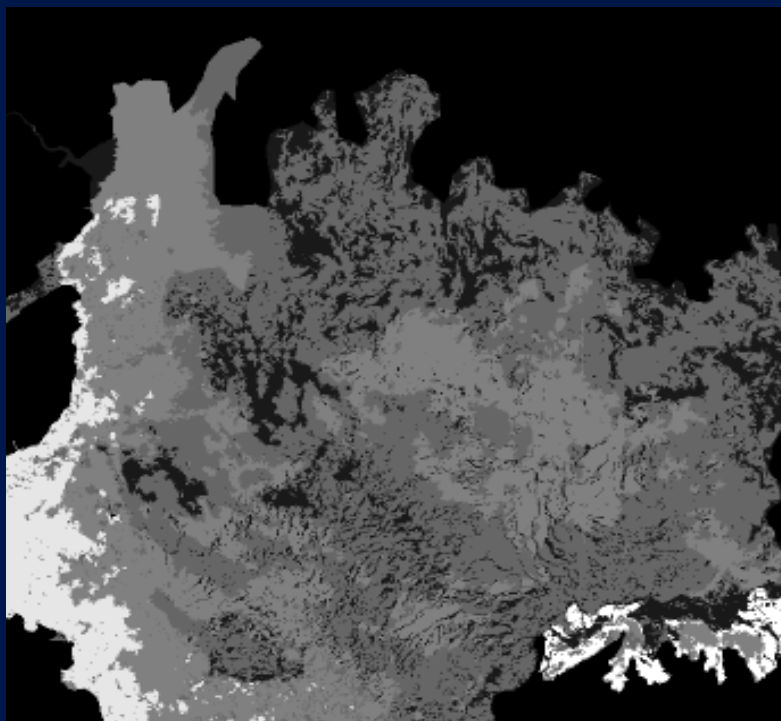
Polygon	#iceclasses	labels
P_1	3	{5, 4, 1}
P_2	2	{4, 1}
P_3	3	{5, 4, 1}
P_4	3	{9, 5, 4}
P_5	2	{4, 1}
P_6	3	{9, 5, 4}
P_7	2	{4, 1}
P_8	2	{L, 5}
P_9	3	{L, 4, 1}
P_{10}	2	{L, 5}
P_{11}	2	{L, 5}
P_{12}	2	{L, 5}
P_{13}	3	{5, 4, 1}
P_{14}	3	{L, 5, 4}
P_{15}	3	{L, 5, 4}
P_{16}	3	{5, 4, 1}
P_{17}	2	{4, 1}
P_{18}	3	{9, 5, 4}
P_{19}	2	{4, 1}
P_{20}	3	{5, 4, 1}
P_{21}	2	{4, 1}
P_{22}	2	{4, 1}

- *2x2 block averaged to a pixel resolution 100m*
- *Separated into closely located 22 polygons*
- *World Meteorological Organization (WMO) standard, labels refer to new ice (1), grey ice (4), grey white ice (5), multiyear ice (9), and fast ice (L)*

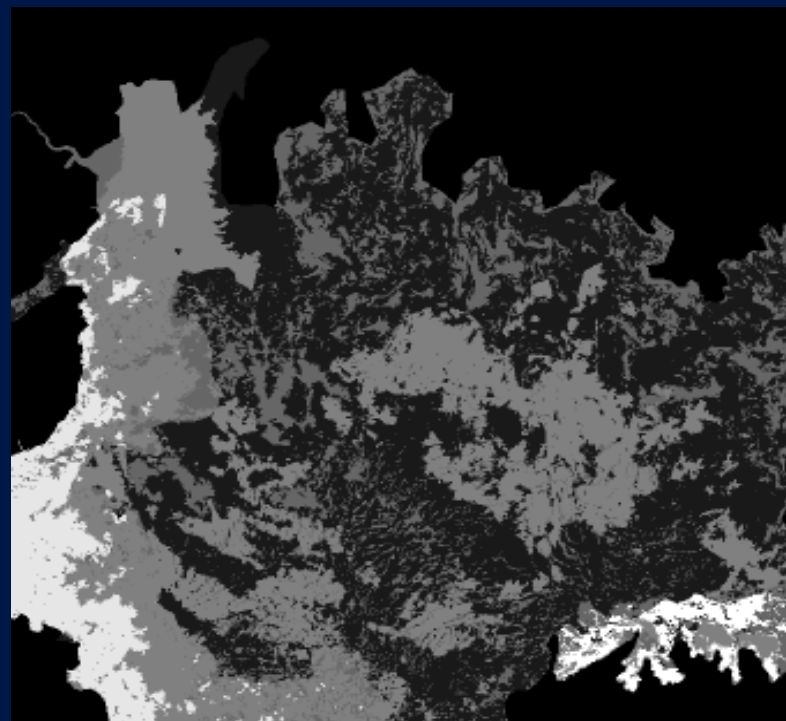
Validation

- *Due to the logistical impossibility field ground truth for the operational SAR sea ice image is not available*
- *The segmentation and labeling results presented here have been validated by a trained SAR sea ice expert at CIS*

Labeling Results

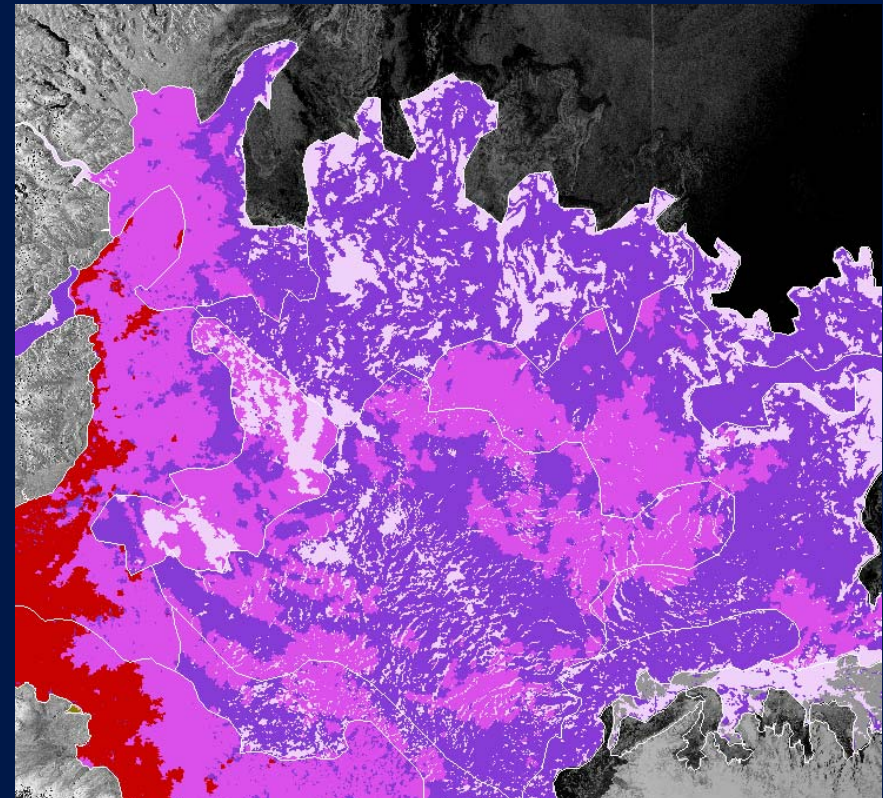
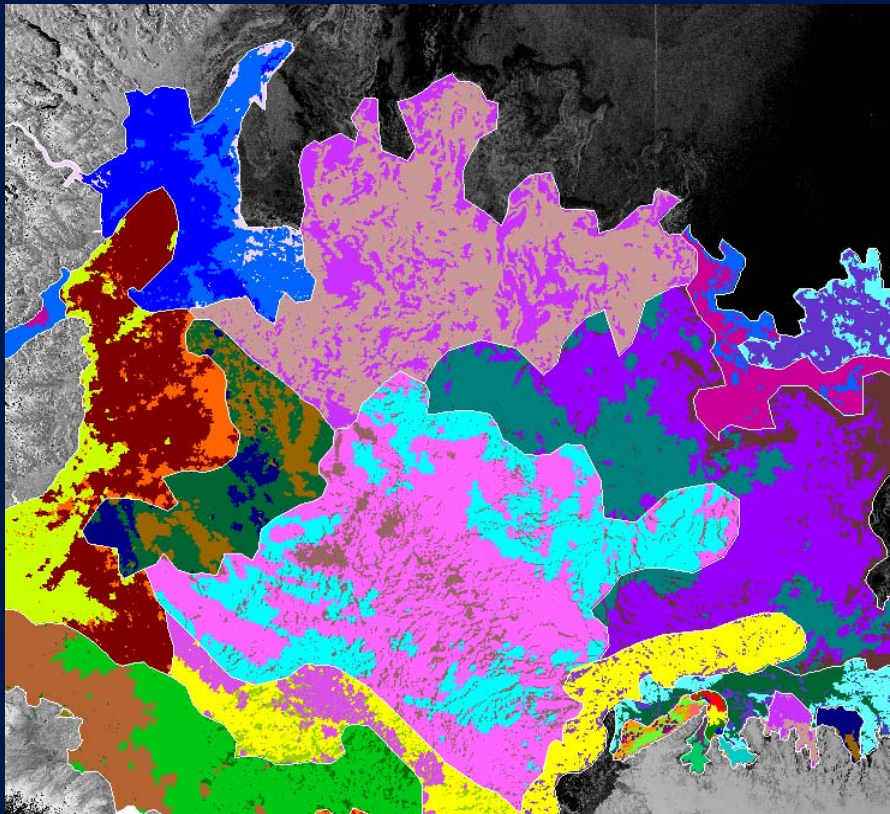



Proposed method ~80%+ accuracy



GMM ~40% accuracy

Labeling Result WMO Color Code



						
Ice Free	Open Water	New Ice	Grey Ice	Grey-White Ice	First-Year Ice	Thin First-Year Ice
						
Medium First-Year Ice	Thick First-Year Ice	Old Ice	Second-Year Ice	Multi-Year Ice	Fast Ice	Icebergs

Limitations

- *The number of classes provided might be determined incorrectly by the ice analyst*
- *Ice analyst might be biased toward assigning thicker ice type in some polygons.*
- *Polygons are drawn manually so the boundaries are often imprecise and accidentally include ice types not recorded for a particular polygon*
- *The segmentation might fail over some complex scenarios resulting in incorrectly formed regions*

Conclusion

- *If the operator provided information is accurate then the segmentation has a stronger accuracy and the labeling process produces a more accurate pixel classified map*
- *This approach has been successfully applied to operational CIS data*
- *Is the first known successful end-to-end process for automatically classifying operational SAR sea ice images*



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