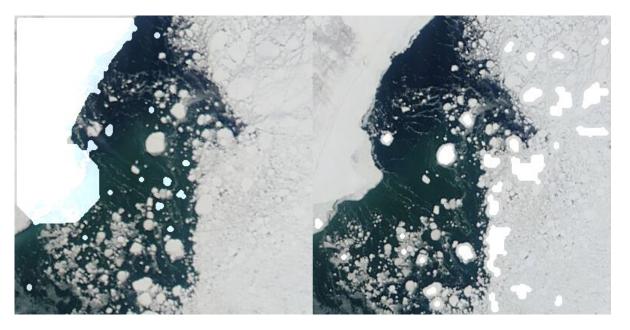
### Identifying causes of error in Segmentation F

Daniel Watkins, Oct. 9, 2024

*Problem:* A persistent error in which the large portions of land-masked regions are labeled as sea ice floes in the final segmentation

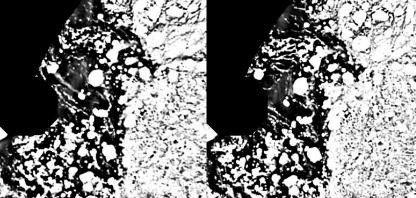


Example of error: test\_images/298-hudson\_bay. Two images from the same day with similar properties; in one, the land mask is labeled as a floe, and no true floes are detected.

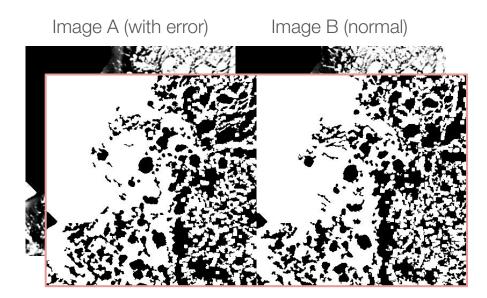
After running the preprocessing and segmentation up to segmentation B, I stepped through the processing steps in segmentation F one-by-one.

1. Input from segmentation B

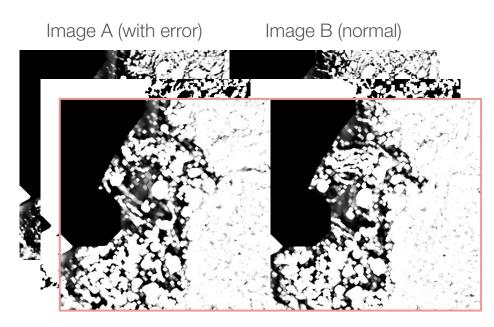
 Image A (with error)
 Image B (normal)



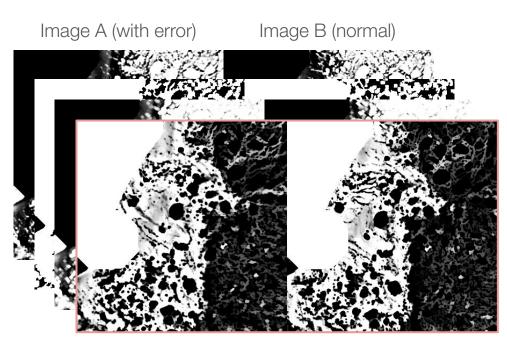
- 1. Input from segmentation B
- 2. Identification of leads (intersection of watershed output and application of area opening)



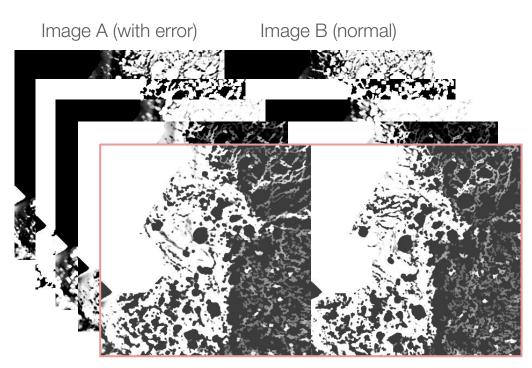
- 1. Input from segmentation B
- 2. Identification of leads (intersection of watershed output and application of area opening)
- 3. Dilation of "not ice" object



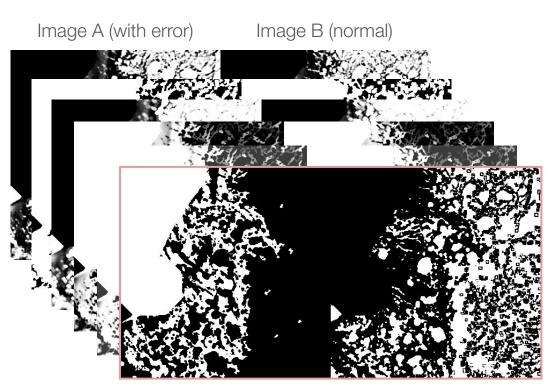
- 1. Input from segmentation B
- 2. Identification of leads (intersection of watershed output and application of area opening)
- 3. Dilation of "not ice" object
- 4. Reconstruction of "not ice"



- 1. Input from segmentation B
- 2. Identification of leads (intersection of watershed output and application of area opening)
- 3. Dilation of "not ice" object
- 4. Reconstruction of "not ice"
- 5. Product of "not ice", "ice lead"



- 1. Input from segmentation B
- 2. Identification of leads (intersection of watershed output and application of area opening)
- 3. Dilation of "not ice" object
- 4. Reconstruction of "not ice"
- 5. Product of "not ice", "ice lead"
- 6. Application of k-means



### Diagnosing issue in the *k*-means step

The *k*-means algorithm divides a grayscale image into k clusters. Currently k is set to 4. The function does the following:

- 1. gray\_image  $\leftarrow$  reconstructed\_leads
- 2. classes  $\leftarrow$  kmeans(gray\_image, k=4)
- 3. ice\_class  $\leftarrow$  mode( classes[ice\_labels] )
- 4. segmented\_image  $\leftarrow$  reshape(classes)
- 5. return segmented\_image == ice\_class

# Diagnosing issue in the k-means step

The *k*-means algorithm divides a grayscale image into k clusters. Currently k is set to 4. The function does the following:

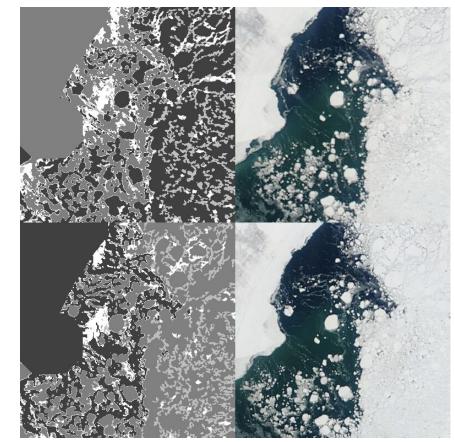
- $1. \quad gray\_image \gets reconstructed\_leads$
- 2. classes  $\leftarrow$  kmeans(gray\_image, k=4)
- 3. ice\_class  $\leftarrow$  mode( classes[ice\_labels] )
- $4. \quad segmented\_image \leftarrow reshape(classes)$
- 5. return segmented\_image == ice\_class

Image B (normal)

Image A (with error)

#### k-means output

True color image



# Diagnosing issue in the k-means step

The *k*-means algorithm divides a grayscale image into *k* clusters. Currently k is set to 4. The function does the following:

- gray\_image ← reconstructed\_leads 1.
- 2. classes  $\leftarrow$  kmeans(gray\_image, k=4)
- ice\_class  $\leftarrow$  mode( classes[ice\_labels] ) 3.
- segmented\_image  $\leftarrow$  reshape(classes) 4.
- 5. return segmented\_image == ice\_class

The result of the *k*-means algorithm looks perfectly reasonable in both cases. My suspicion: we need to improve the accuracy of the ice label step. From what I can see, the error is actually in ice labels – it labels some sections of the landmask as ice there.

Next steps

- Validate the find ice labels function
- Verify that framework is robust to chance that no ice\_labels are found

(normal) ന Image

#### *k*-means output

#### True color image

