Ross Sea

The Last Ocean

Presentation by Group 4

URL to code repo: https://github.com/Aufuray/ross-sea-project

Group 4 Members:

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The last ocean?

The Ross Sea ecosystem is the **last** <u>intact</u> marine ecosystem left on Earth!

Unlike many other areas of the world's oceans, the Ross Sea's top predators are still abundant.





The last ocean?



The Ross sea is home to species found **nowhere** else in the world - **uninterrupted** by humans.

Commercial fishing now threatens survival in the Ross Sea.

You can help save the Last Ocean by donating and working with the New Zealand government. lastocean.org/Take-Action

Yes. We care about the planet.

But that's not our assignment today 😃



Our Dataset:

Date: December 2nd, 2015

(Julian Date: 336)

Time: 00:41 am to 20:25 pm

Sensors: VIRS and MODIS

(8 VIRS, 12 MODIS snaps)

Our Stack:

Python: Scripting Language

NumPy: To work with np arrays

SciPy: Load matfiles

To Compute Empirical CDF

Histogram matching

K-means clustering

Pandas: To create Time series

Matplot: Plot Graphs and charts

Game Plan

- Work with Sea Ice Concentration (mw_sic) data
 - Determine where the sea ice is located
 - Try to use histogram matching to normalize images from different sensors
- Visualising independent components
 - Determine where land, water, cloud, ice components overlap
 - See if any element(s) encroached more into the other over time
- Use K-means Algorithm
 - Separate the individual components into clusters
 - See if the clusters validate the analysis of the ICA in step 2



Challenge:

Analysing the data from the [mw_sic] in our images, we found some irregularities between the VIIRS and MODIS sensors. 60% of our data files are from MODIS instrument, so we decided not to use VIIRS images alone.

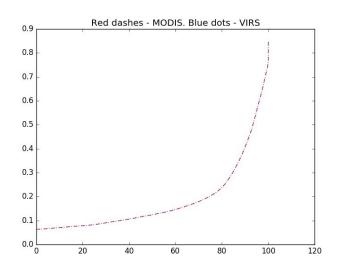
Solution: Histogram Matching and Linear Interpolation.

We used the VIIRS images as reference and MODIS images as source of our histogram match computation.

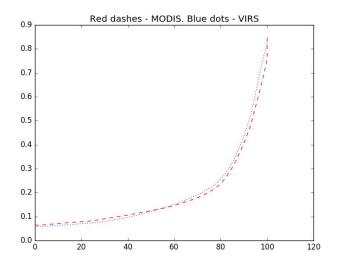
We used images that were taken approx. 20mins apart from each other.

This is because images with large time produced very different histograms.

Before Histogram MatchingSample Sea Ice discrepancy between VIIRS and MODIS sensors



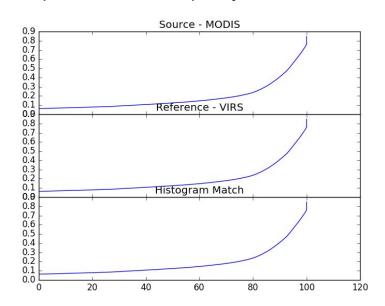
Images from MODIS and VIIRS sensors taken **4 mins** apart



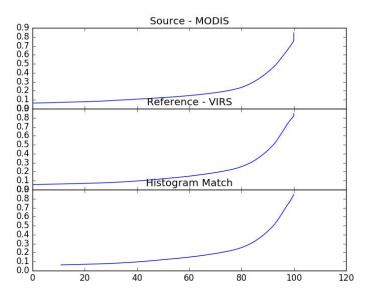
Images from MODIS and VIIRS sensors taken **16 hours** apart

After Histogram Matching

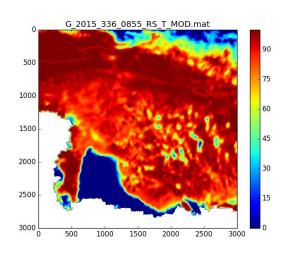
Sample Sea Ice discrepancy between VIIRS and MODIS sensors

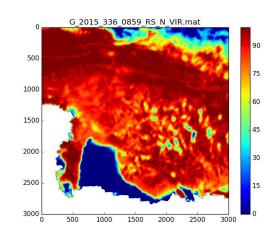


Using images from MODIS and VIIRS sensors taken **4 mins** apart



Using images from MODIS and VIIRS sensors taken **16 hours** apart





Histogram Match of images taken approx. 20 mins apart

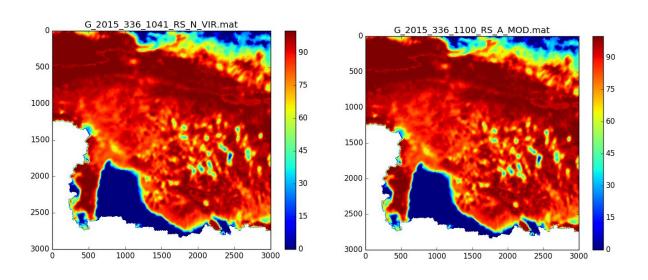
VIIRS image used as reference.
MODIS used as source.

Conclusion:

The closer the images were taken to each other, the more similar the resulting histogram matched image was.

MODIS image taken at time: 0855

VIIRS image taken at time: 0859



VIIRS image taken at time: 1041

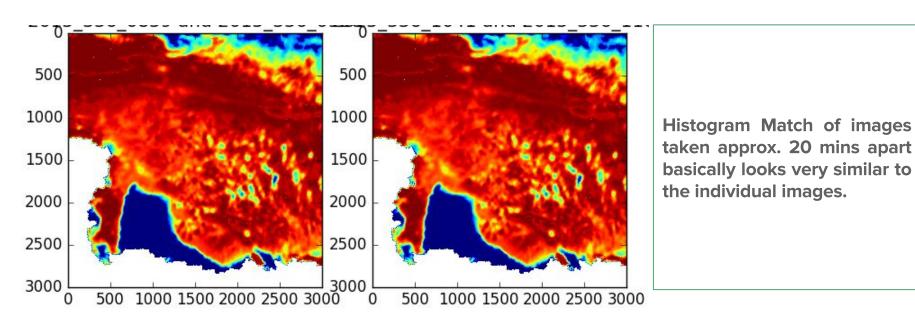
MODIS image taken at time: 1100

Histogram Match of images taken approx. 20 mins apart

VIIRS image used as reference.
MODIS used as source.

Conclusion:

The closer the images were taken to each other, the more similar the resulting histogram matched image was.

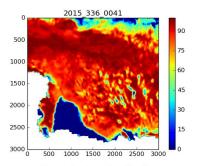


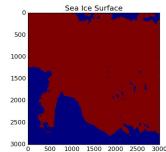
Hist Match between Images taken at time: **0859** and **0900**

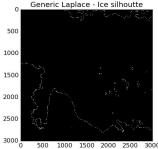
Hist Match between Images taken at time: 1041 and 1100

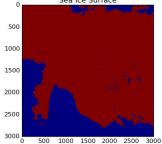


Sea Ice concentration and Surface for G 2015 336 0041 RS N VIR.mat









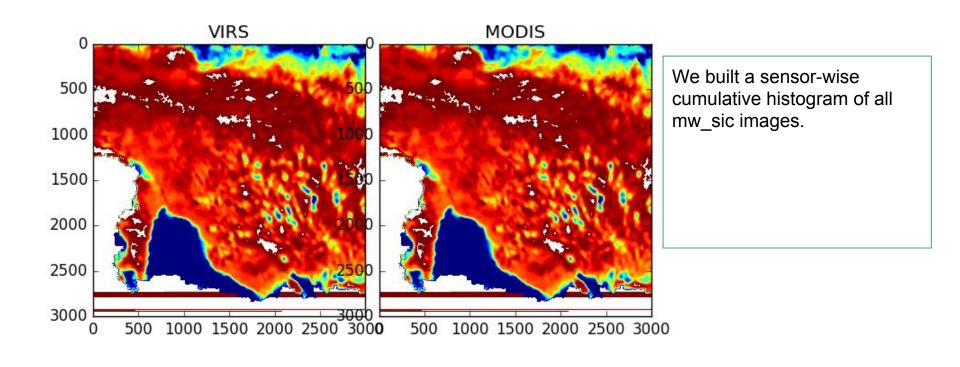
In the microwave sea ice readings, we set a threshold:

- <= 40 for other components,
- **40** for sea ice.

Hence we were able to classify and separate the image into two portions.

We also applied a **Generic** Laplace filter on the image to create a silhouette that highlights the edges.

Cumulative histogram match of sea ice images by sensors



Visualising independent components over time



Visualising independent components over time

Challenge:

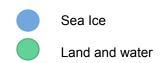
Plot a component-wise analysis of each image and build a Time Series. See how the Time Series changes over time.

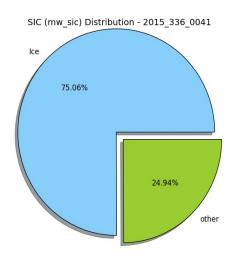
Solution:

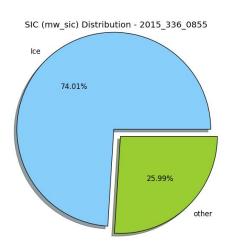
Use the Land mask and Sea ice array values.

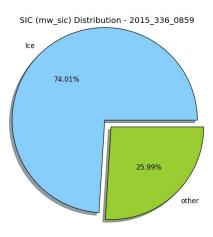
Plot percentage of components in a pie chart and translate that to bar graph of time series.

Visualising components over time





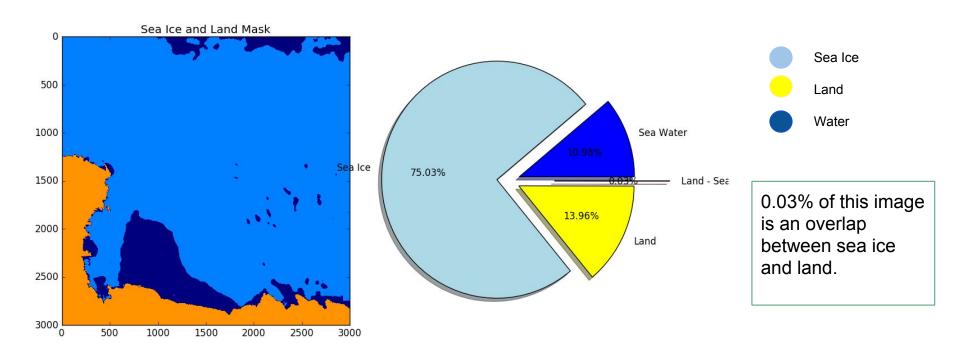




From MODIS snap taken at **0855**

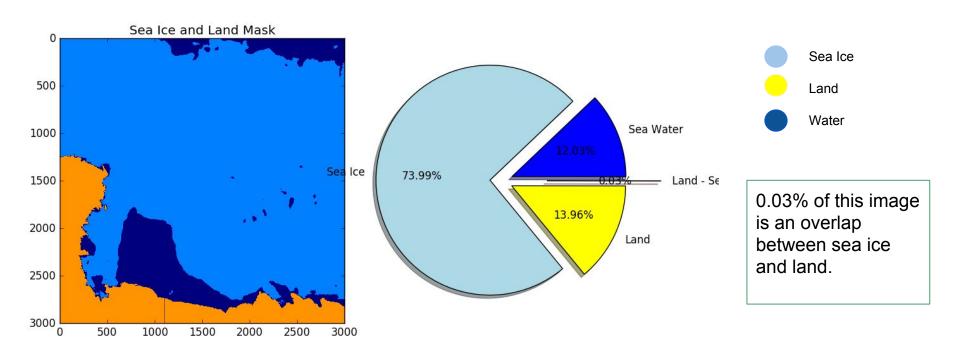
From VIIRS snap taken at 0859

Visualising components in single image



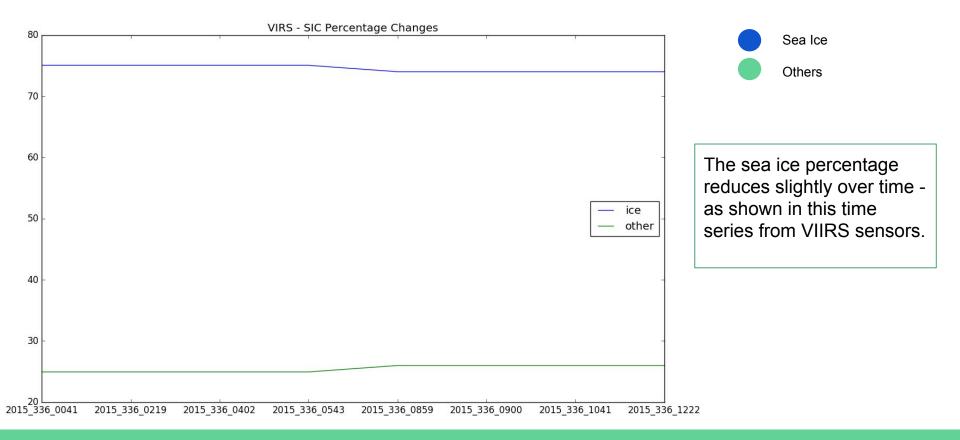
From VIIRS snap taken at **0041**

Visualising components in single image

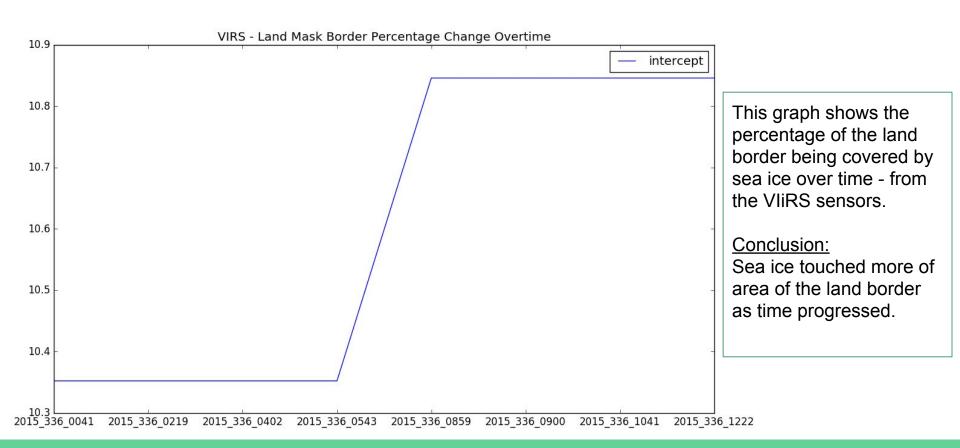


From MODIS snap taken at 0855

Time series of Sea Ice percentage (VIIRS Sensor)



Time series of Sea Ice at land border (VIIRS Sensor)



Using K-means

Solving a classification problem



Using K-Means

Challenge:

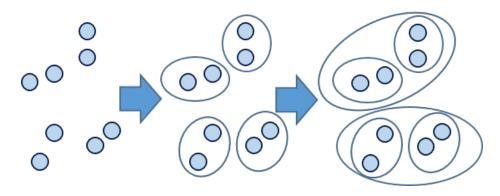
Group the data into clusters of individual components (cloud, ice, water, land). By far the toughest challenge we had.

Solution:

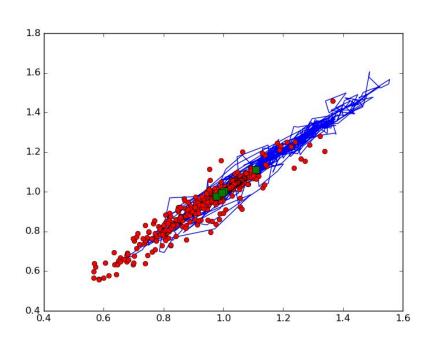
K-means clustering algorithm from Scipy.

How K-Means works?

- K-Means clustering is a technique in which the algorithm groups the data entities into a specified number of clusters (k).
- K-Means algorithm selects k random centroid points, and assigns each data entity to a cluster based on the shortest distance to a centroid.



Using K-Means - wrong implementation





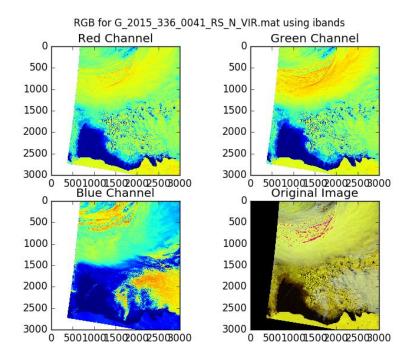
Result was highly skewed and irregular, Because the data <u>insufficient - too small</u> to tell components apart.

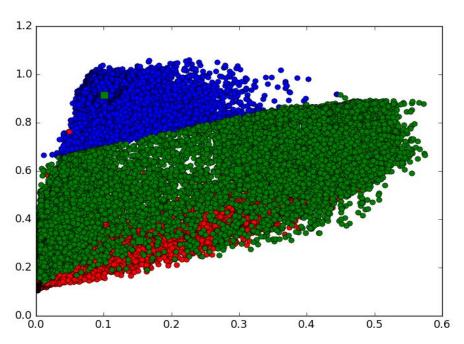
Basically, we were using one dimension. Also, were using wrong RGB slices from the data set.

Using VIIRS image taken at time: 0041

Using K-Means - 2nd attempt



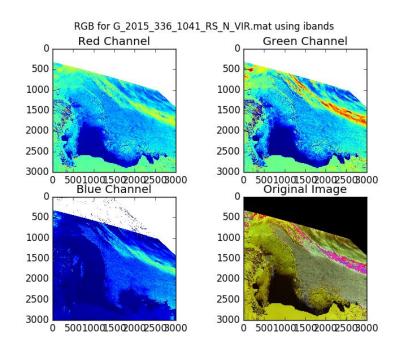


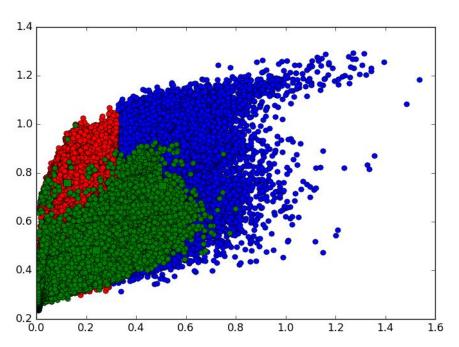


Using VIIRS image taken at time: 0041

Using K-Means - 2nd attempt

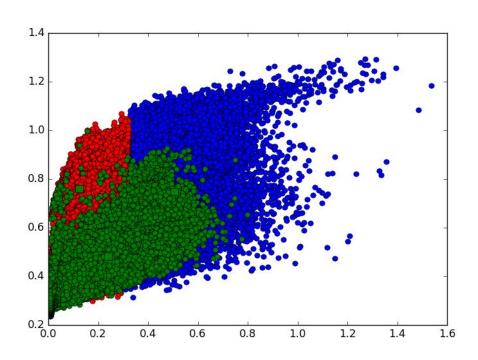






Using VIIRS image taken at time: 1041

Using K-Means

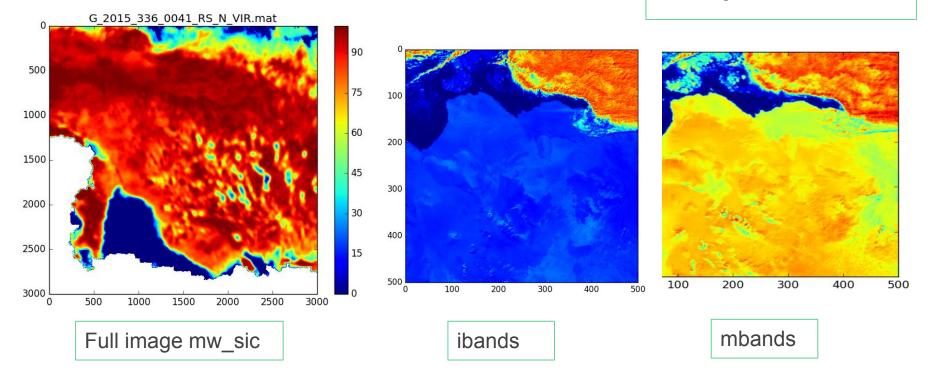


We basically wanted our k-means output to validate the analysis of our time series and graph visualisations.

To verify that the sea ice clusters truly encroach more into land over time.

Using K-Means with 2 centroids

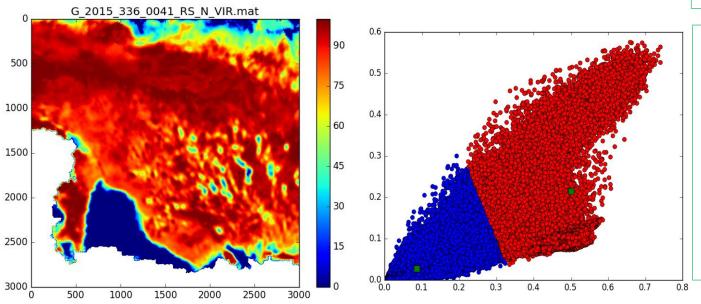
Using ibands and mbands on Lower right 500 X 500 slice



Using K-Means with 2 centroids



Using ibands and mbands

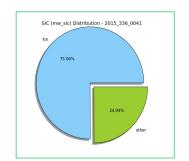


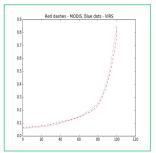
From clustering the data into sea ice and other components.

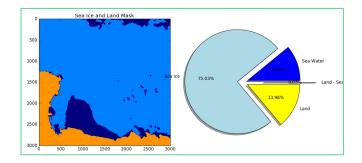
The clusters are clearly divided as our original mw_sic image plot.

In conclusion ...

- Going forward, it is possible to build a prediction model using the movement of ice and see observing how it spreads in and out of the land over time.
- Before applying histogram matching, difference in time between images should be taken into consideration because of the moving components.
- K-means clustering can be used to validate inferences from the ICA.
 We worked with 2 centroids for this project, we can improve to work with 3 centroids.









Random Fact: The Last Ocean campaign runs on donations ... You should consider donating.

Thank You!

URL to Code Repo:

https://github.com/Aufuray/ross-sea-project

References:

Images: Group 4, Wikipedia Random Facts: Lastocean.org

scikit-learn.org

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