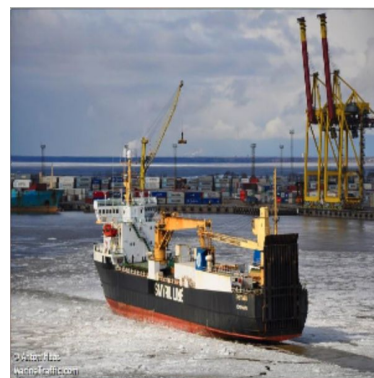




Background

How can we police the Open Seas?

- **Challenges:** The scale of the world's ocean makes it difficult to know what and where people are fishing.
- **Opportunities:** Autonomous drones can fly long distances and potentially identify and report malicious fishing behavior in the open ocean.
- We have unprecedented access to what types of gear that ships are using.
- A critical task for a potential is getting images of these ships.



Methods

Transfer Learning

Approach 1

- **Vanilla:** Use pretrained Resnet weights as a feature extractor for images, train last layer
- **Accuracy:**
- **Pitfalls:** (Empirically) Training does not converge ~50% of time

Approach 2

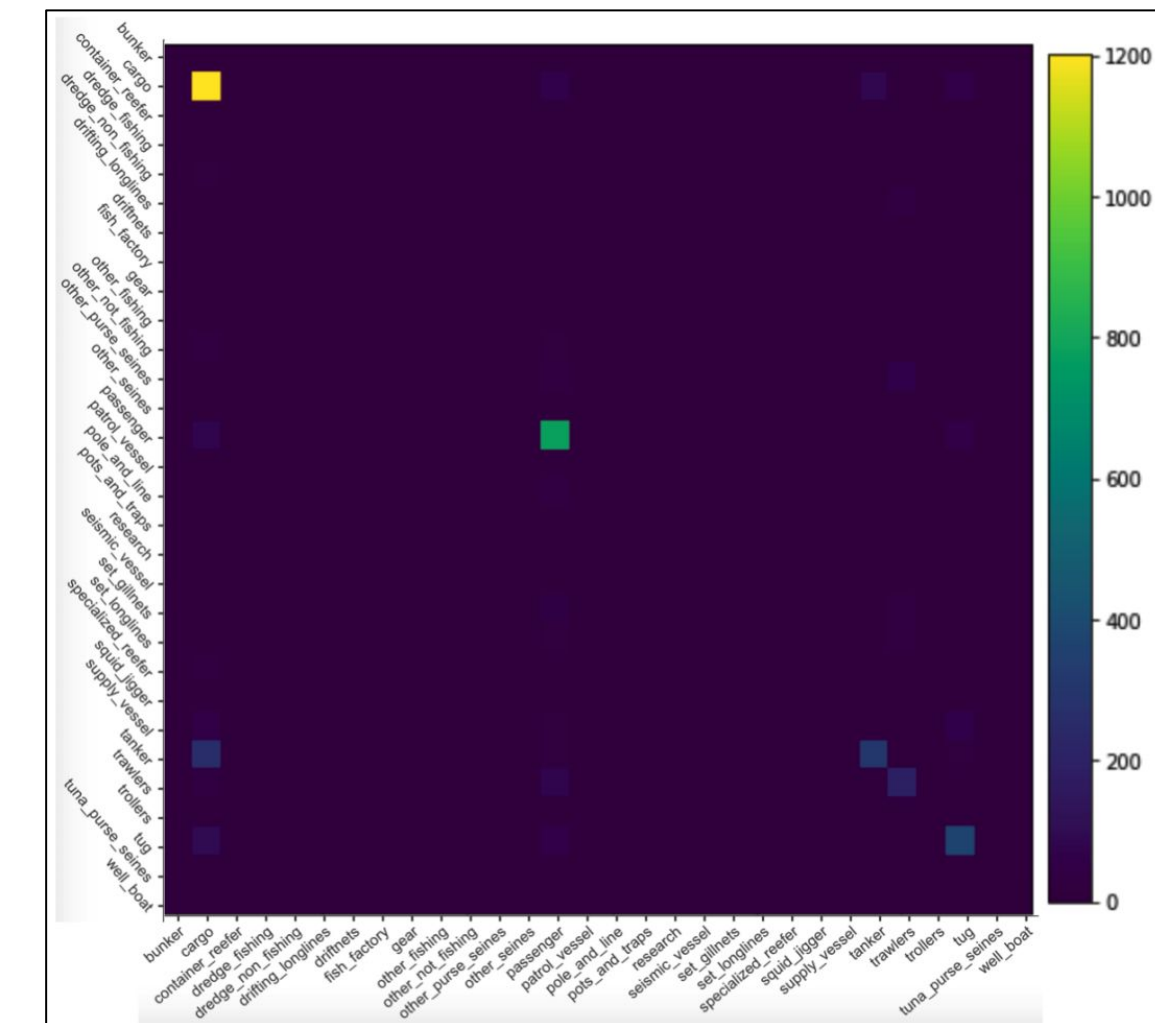
- **Pretrain:** Use ship samples from Marvel dataset, which has grainy classes of ships
- **Accuracy:** silhouette score.
- **Pitfalls:** Class imbalance has a tougher time getting less-seen classes using C.E.

Results

(3) Network is biased towards overrepresented classes

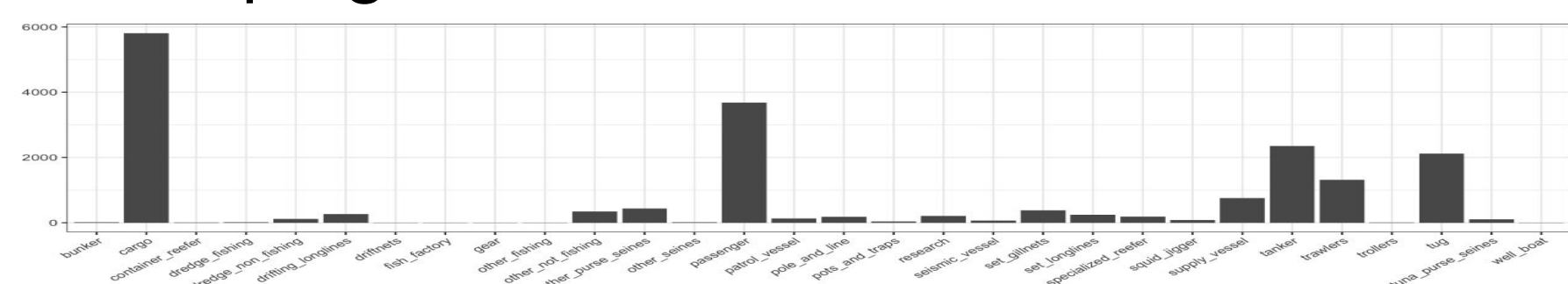
$$w_y = \frac{1 - \beta}{1 - \beta^{n_y}}$$

Since cargo ships make up the majority of ships on the ocean, the data set is biased towards these vessels



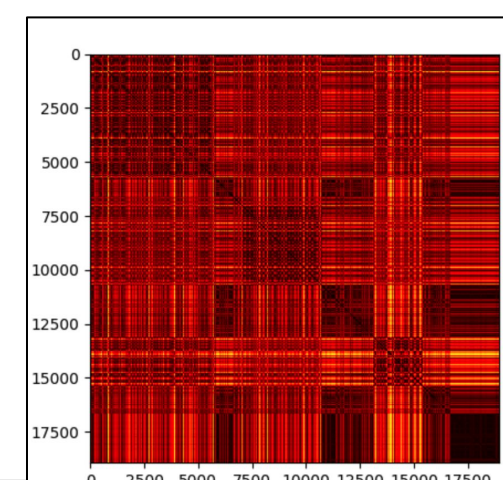
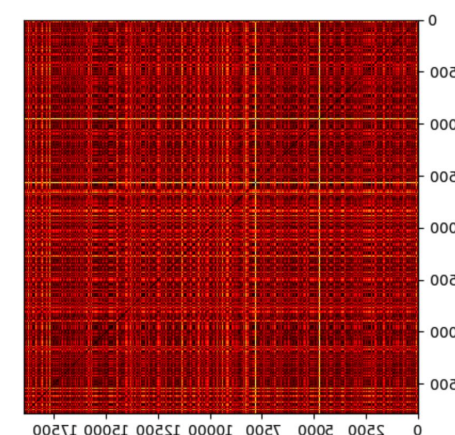
Methods

- **Initial data:** n = 1.5 million ship ids from satellite-intercepted signals.
- **Scraping:** Create web scraper that captures ship images from MarineTraffic.com
- **Labels:** Use labels from database, courtesy of Global Fish Watch, alongside images to create labeled data set for sup. learning
- **Final dataset:** ~25k images surfaced after scraping, 30 classes.

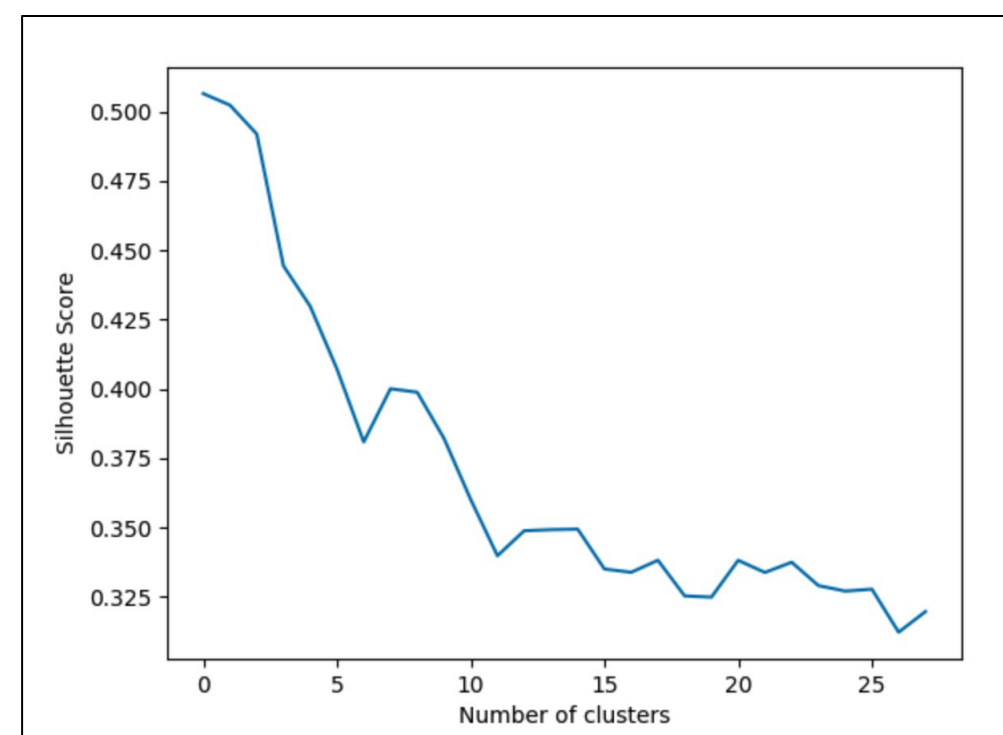


Results - Geometry

(1) Class Labels provide structure to data set - Euclidean



(2) Number of ships that the network is picking out is less than 30



Conclusions/Acknowledgements

- **Error Analysis:** Use weighted cross entropy for loss function
- **Future approach (1):** Use t-sne to understand the distribution of classes in latent space
- **Future approach (2):** utilize results from various clustering methods to better inform what features of ships this neural network is extracting from the images
- Networks also need more gpu time, it is possible that accuracy could be as high as .85

Acknowledgments: Class distribution and confusion matrix courtesy of Emma Gee.

