

QUANTIFYING ICE-SHIP INTERACTIONS FROM AN OPTICAL PERSPECTIVE

Fall 2023 Semester Closeout and Roundup

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Agenda

- 1) Mask-RCNN
- 2) Local Region Ice-Statistics Extraction
- 3) Force Prediction (WIP)
 - a) RNN: LSTM
- 4) Future Works Proposals

Mask-RCNN

- Mask-RCNN Def: Mask - Region Convolution Neural Network
- Built to extend capabilities of Faster-RCNN, adding only small computational overhead in comparison [1]. The ‘extension’ specifically is a branch to predict an object’s mask in parallel to the bounding box prediction.
- 2016 COCO challenge winners (pre-trained weights available @matterport Github)

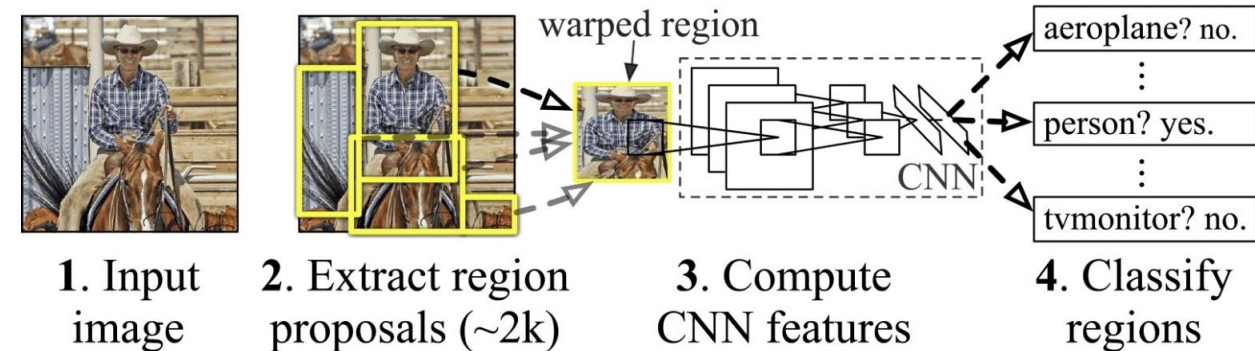
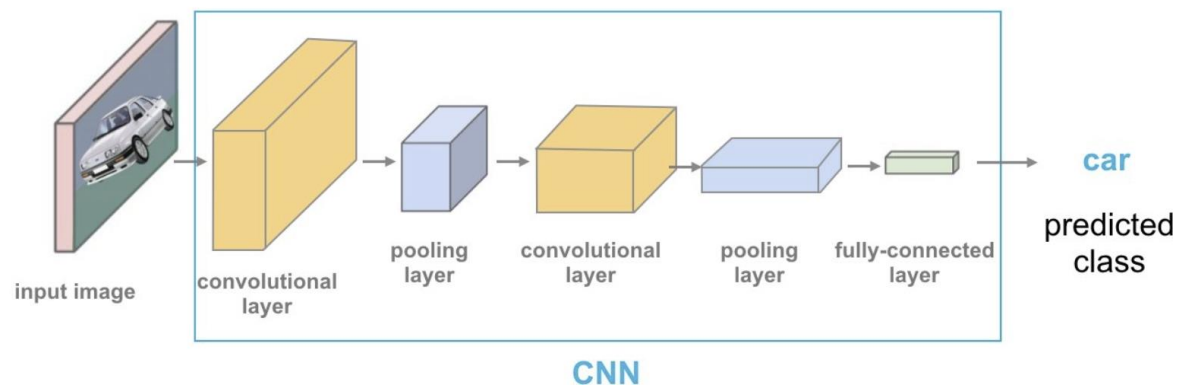


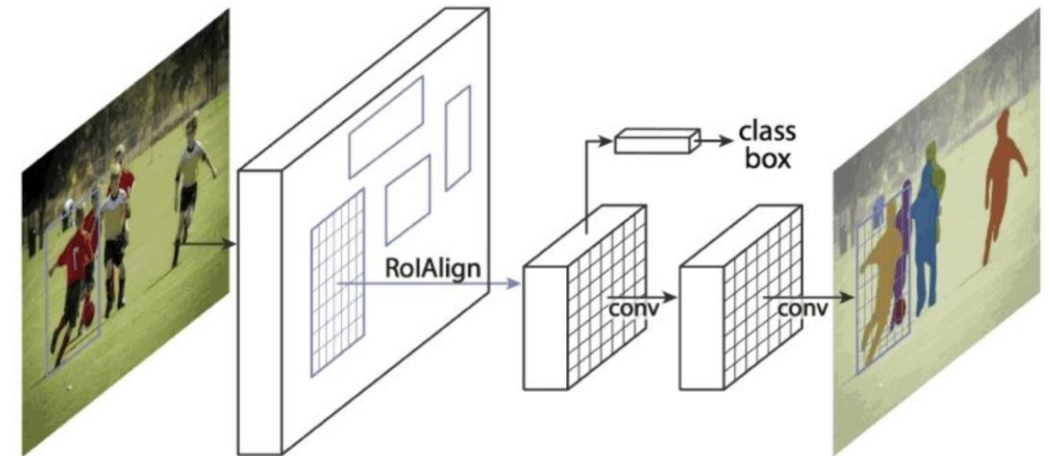
Figure: CNN & RNN Architectures [2]

[1] K. He, G. Gkioxari, P. Dollár, and R. Girshick, Mask R-CNN, <https://arxiv.org/abs/1703.06870> (accessed Dec. 4, 2023).

[2] E. Odemakinde, “Everything about mask R-CNN: A beginner’s guide,” viso.ai, <https://viso.ai/deep-learning/mask-r-cnn/> (accessed Dec. 4, 2023).

Mask-RCNN

Fast-RCNN opposed to faster RCNN (shown here), uses selective search rather than RPN layer



Mask R-CNN – The Mask R-CNN Framework for Instance Segmentation

Figure: Faster-RCNN & MRNN Architectures [2]

[2] E. Odemakinde, “Everything about mask R-CNN: A beginner’s guide,” viso.ai, <https://viso.ai/deep-learning/mask-r-cnn/> (accessed Dec. 4, 2023).

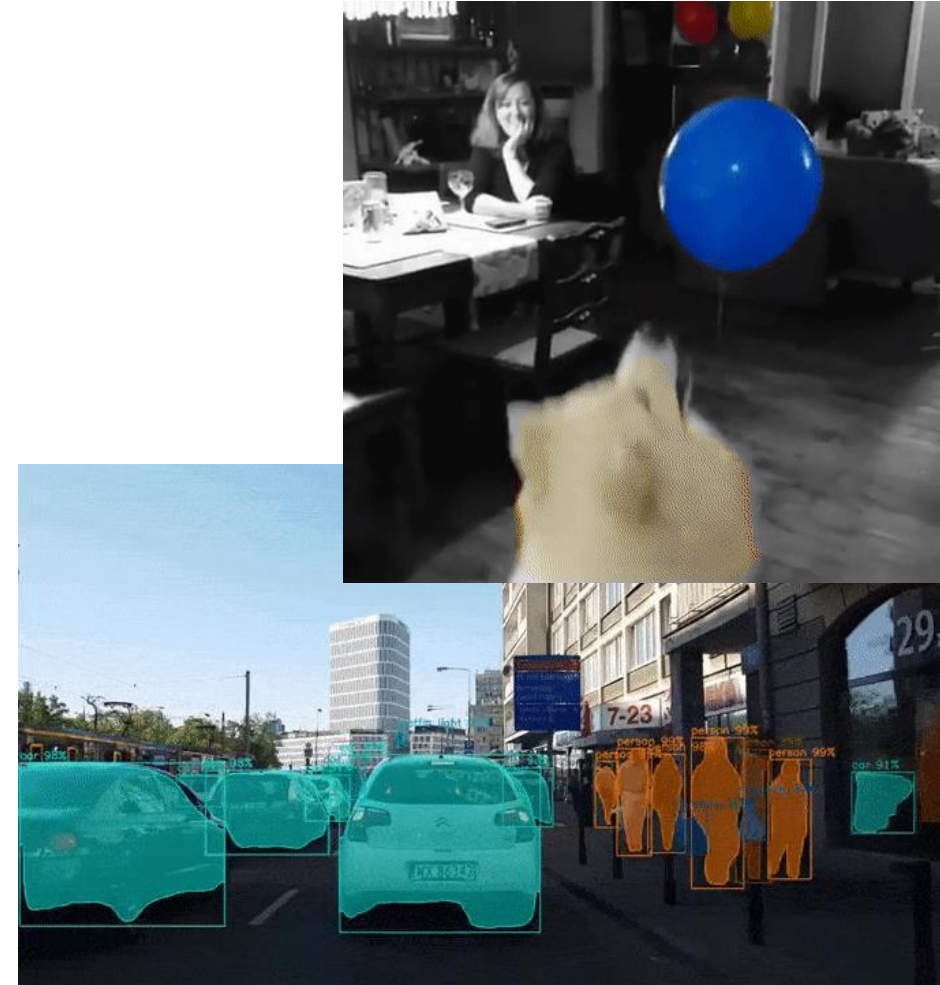
Mask-RCNN

[Matterport Mask-RCNN](#) implementation using Python [3]

Lot of good samples & pre-trained weights to get started from. Majority of samples are built in Jupyter notebooks, so it makes it easy to understand.

Some examples of implementations;

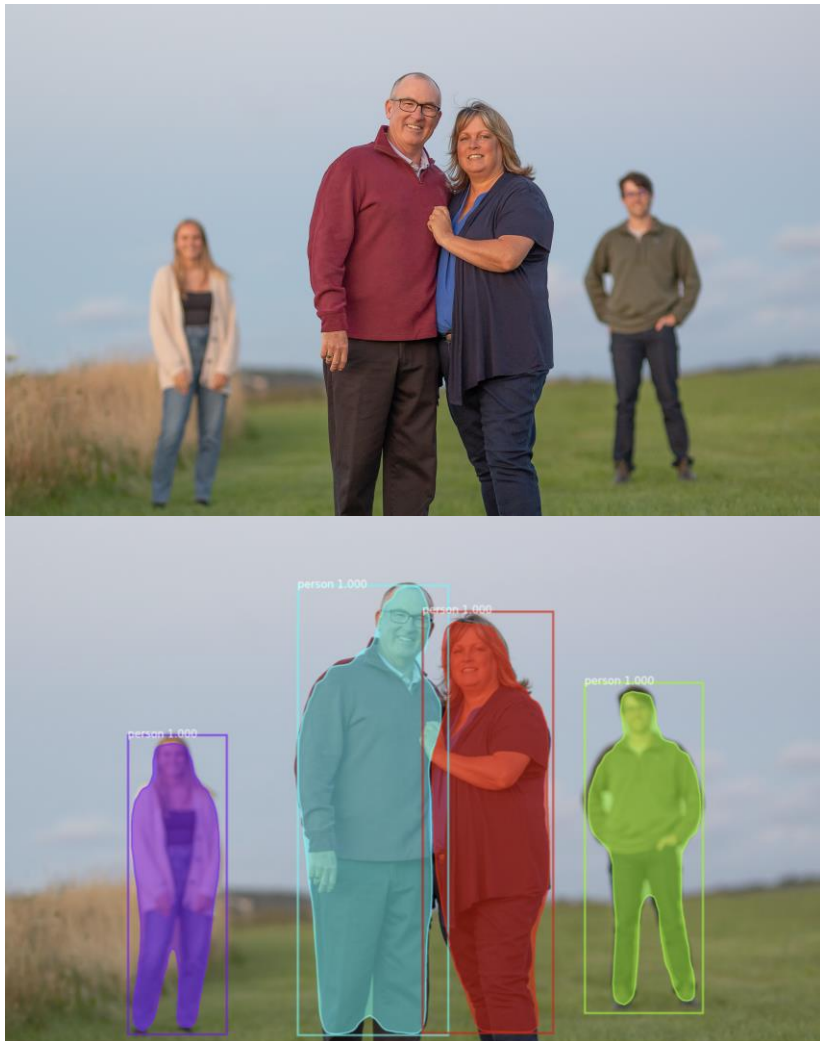
- a) Splash of color; merges predicted mask in colour with greyscale background (Top RHS)
- b) Generic full colour video implementation from dash cam footage on busy street (Bottom RHS).



MRCNN video examples [3].

[3] W. Abdulla, "Mask R-CNN for Object Detection and Instance Segmentation on Keras and TensorFlow," GitHub, 2017. [Online]. Available: https://github.com/matterport/Mask_RCNN

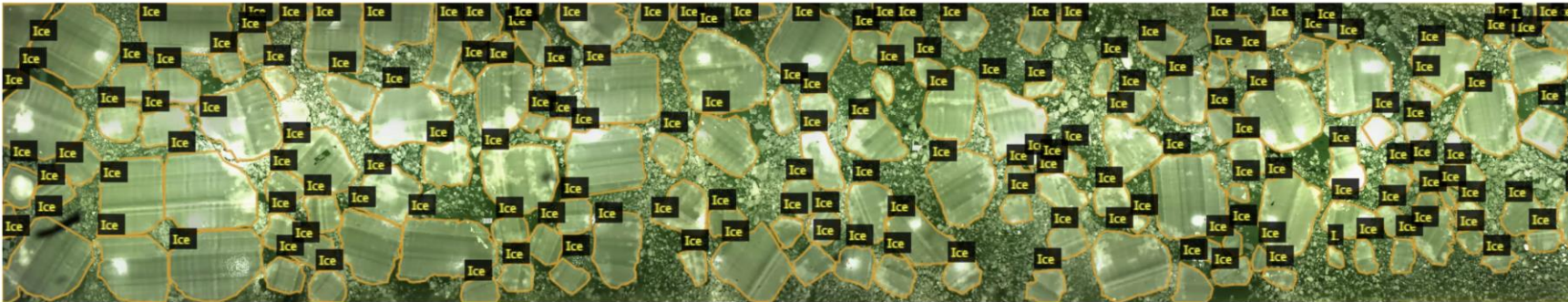
Mask-RCNN



Figures: MRCNN pre-trained on COCO capability examples.

Mask-RCNN

- Applying this to Ice-Ship Force Prediction
 - Transfer learning from pretrained weights based on COCO 2016 dataset, to keep training time under control [3].
- Annotation Style
 - COCO – need to input the ground truth mask



```

    },
    "region_attributes": {
      "Object": "Ice"
    }
  },
  "file_attributes": {}
},
"100m_dist_9ths_1p2kts_0p4m_0deg_001_c_overhead_frame384.png5167382": {
  "filename": "100m_dist_9ths_1p2kts_0p4m_0deg_001_c_overhead_frame384.png",
  "size": 5167382,
  "regions": [
    {
      "shape_attributes": {
        "name": "polygon",
        "all_points_x": [
          4,
          267,
          354,
          370,
          285,
          0
        ],
        "all_points_y": [
          534,
          534,
          534,
          534,
          534,
          534
        ]
      }
    }
  ]
}

```

Figure: VIA Image Annotation Tool Showing Mask Annotations and COCO Annotation Format

[3] W. Abdulla, "Mask R-CNN for Object Detection and Instance Segmentation on Keras and TensorFlow," GitHub, 2017. [Online]. Available: https://github.com/matterport/Mask_RCNN

Mask-RCNN

- Annotation overhead
 - Non-annotated regions are assigned region attribute: “Background” (BG)
 - Everything in the image must be annotated.
 - We chose to allow brash ice to be considered “background” **Future works**

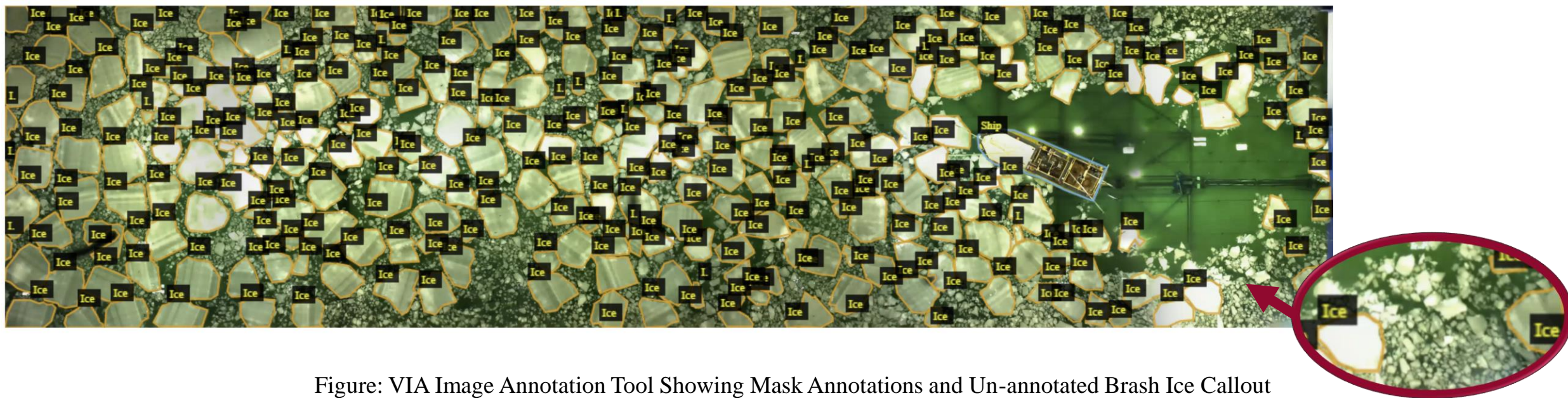


Figure: VIA Image Annotation Tool Showing Mask Annotations and Un-annotated Brash Ice Callout

Ice-Statistics Extraction

What kind of statistics?

- Total Pixel Area **Future Works** – convert to *real* measurement
- Mean floe size
- Standard Deviation of floe size

Ice beyond 2SL forward, and 0.5SL behind have no effect on the ship per Dr. Shameem [4]

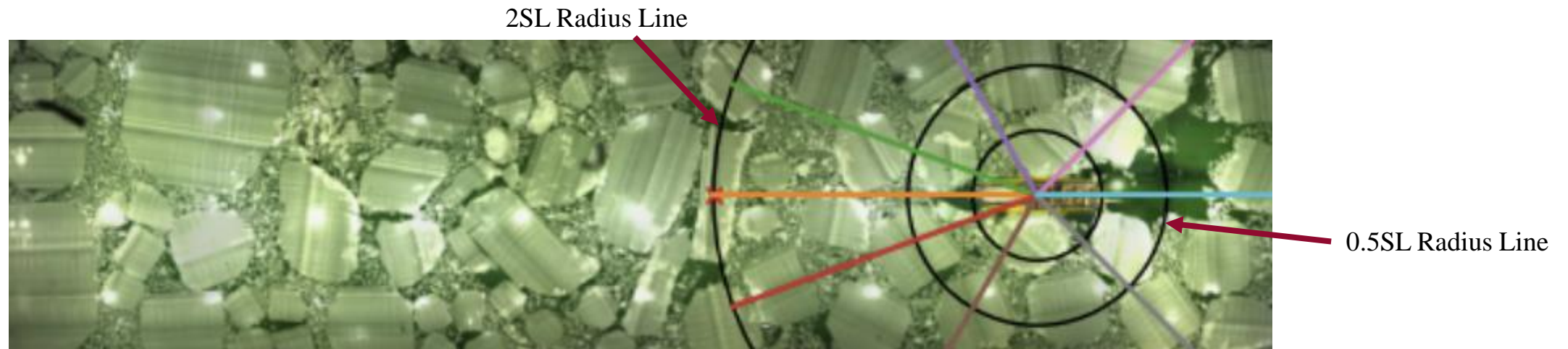
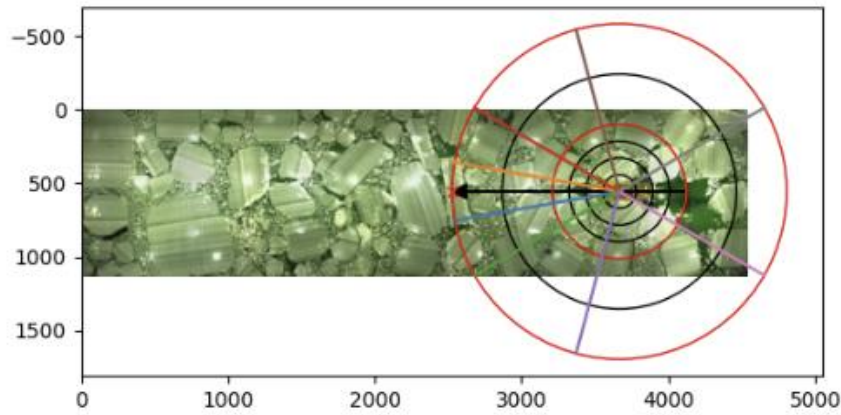


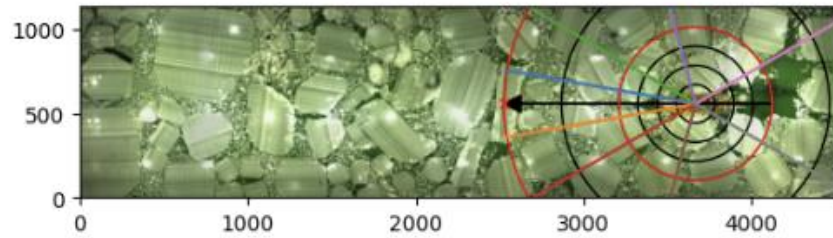
Figure: 2SL (2.5SL) and 0.5SL (1SL) Region Proposal Callout

[4] S. Islam et al., “Physical Model Testing for Supporting Ice Force Model Development of DP Vessels in Managed Ice,” in OTC Arctic Technology Conference, 2018.

Ice-Statistics Extraction



(a) Uncropped



(b) Cropped

Figure: Initial Region Proposals

Divisions

- Angles:
 - $\pm [10, 30, 75, 150]$
- Distance (Radius from ship center)
 - $\pm [0.25, 0.5, 0.75, 1, 1.75, 2.5]$

Divisions

- Angles:
 - $\pm [0, 20, 60, 135, 180]$
- Distance (Radius from ship center)
 - $\pm [0.5, 1, 2.5]$

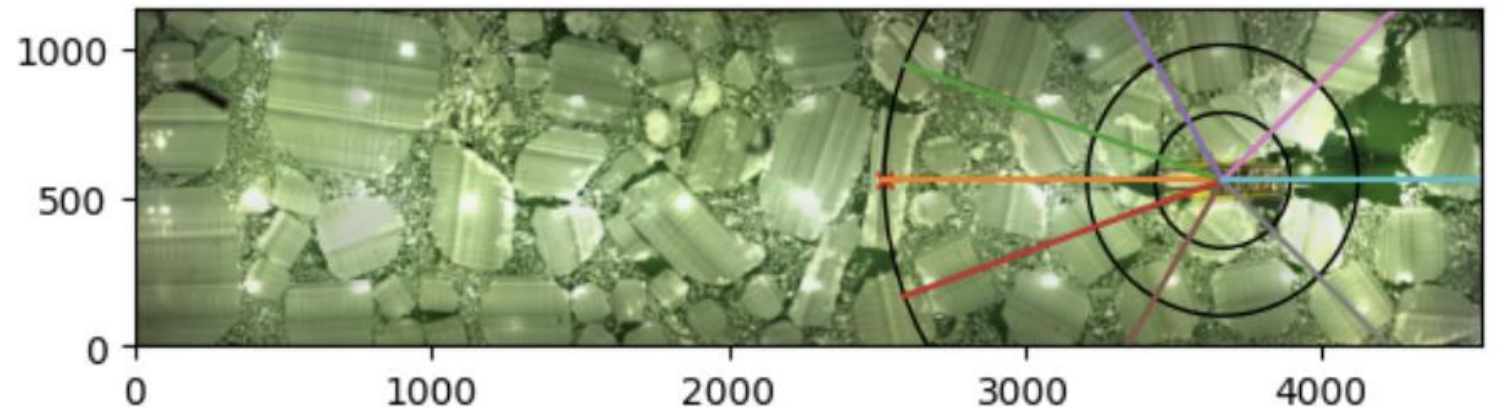
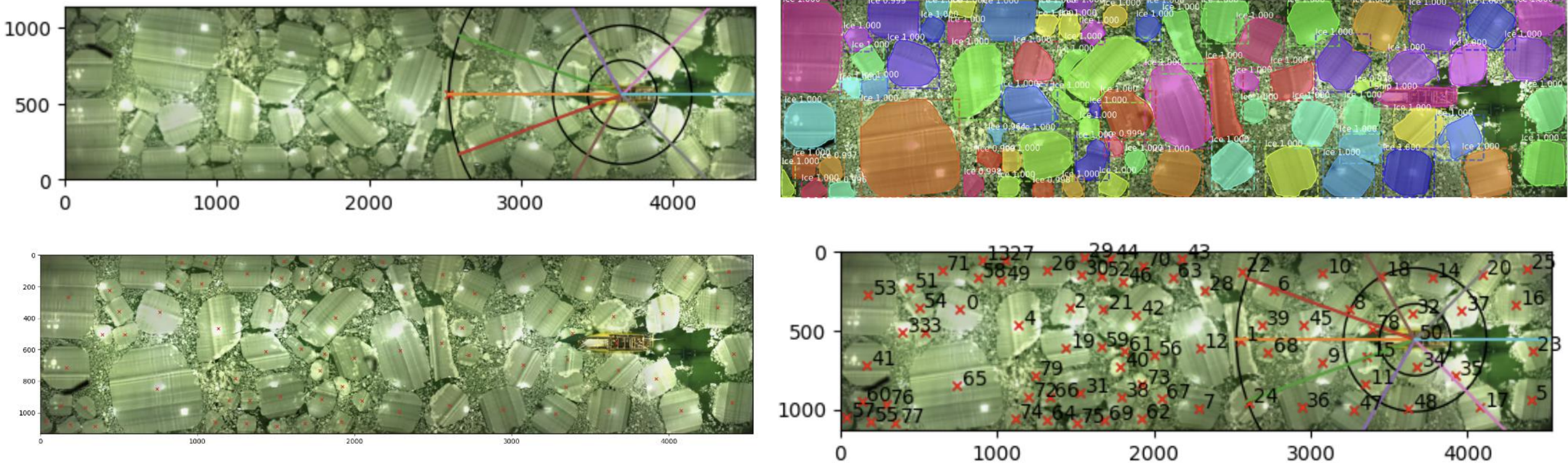
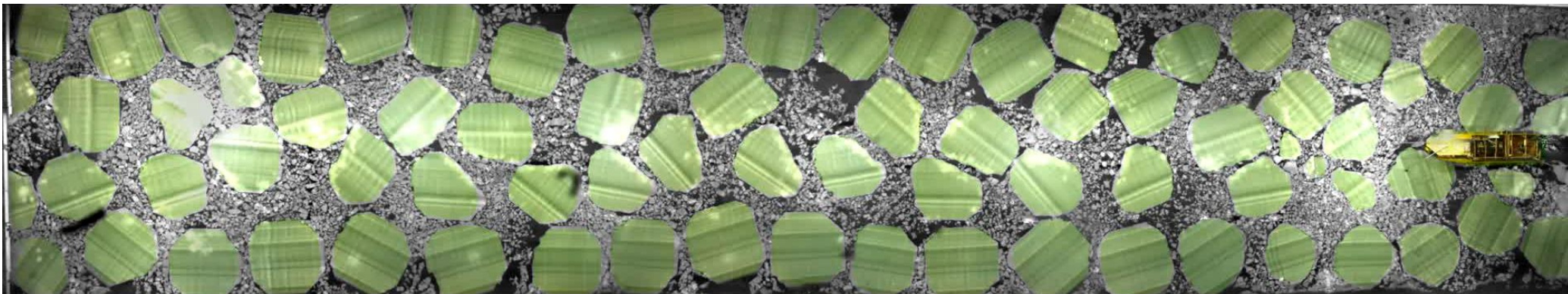


Figure: Current Region Proposals

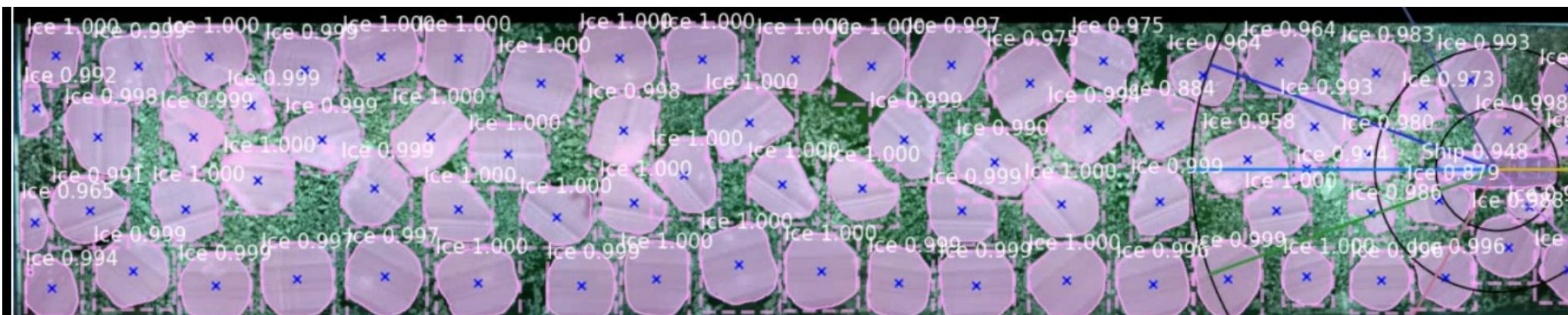
Ice-Statistics Extraction



Ice-Statistics Extraction



Video: Masked Detections (splash of color)



Video: Masked Detections (overlaid masks, scores and regions)

Ice-Statistics Extraction

Current Assumptions (limitations)

- 0° yaw
 - Can maintain this with selective datasets from NRC testing. **Future works: Difficult challenge mathematically to robustly determine heading within 5 degrees.**
- Consistent pixel density
 - OK for this implementation. Other cameras/datasets will need to re-define distance **Future works: based on known distance in the image.**
- No processing of NRC provided data
 - Ship speed data is measured as carriage speed not instantaneous ship speed **Future works**

Ice-Statistics Extraction

Output to CSV (one row per frame, one file per video)

- | | | |
|-----------------------------------|---|--|
| 1.FrameN (int) | } | Directly from NRC DAQ (CSV input) |
| 2.Carriage_Speed_DP (float) | | |
| 3.Global_FX (float) | | |
| 4.Global_FY (float) | | |
| 5.Global_FT (float) | | |
| 6.Global_MZ (float) | | |
| 7.Floe_Size (string) | | |
| 8.Drift_Speed (float) | | |
| 9.Ice_Thick (float) | | |
| 10.Drift_Angle (float) | | |
| 11.Region_0:228.0_0:20 - Ice Conc | } | Repeated for each region (pre-defined, easily changed) |
| 12.Region_0:228.0_0:20 - Mean | | |
| 13.Region_0:228.0_0:20 - StdDev | | |

Ice-Statistics Extraction

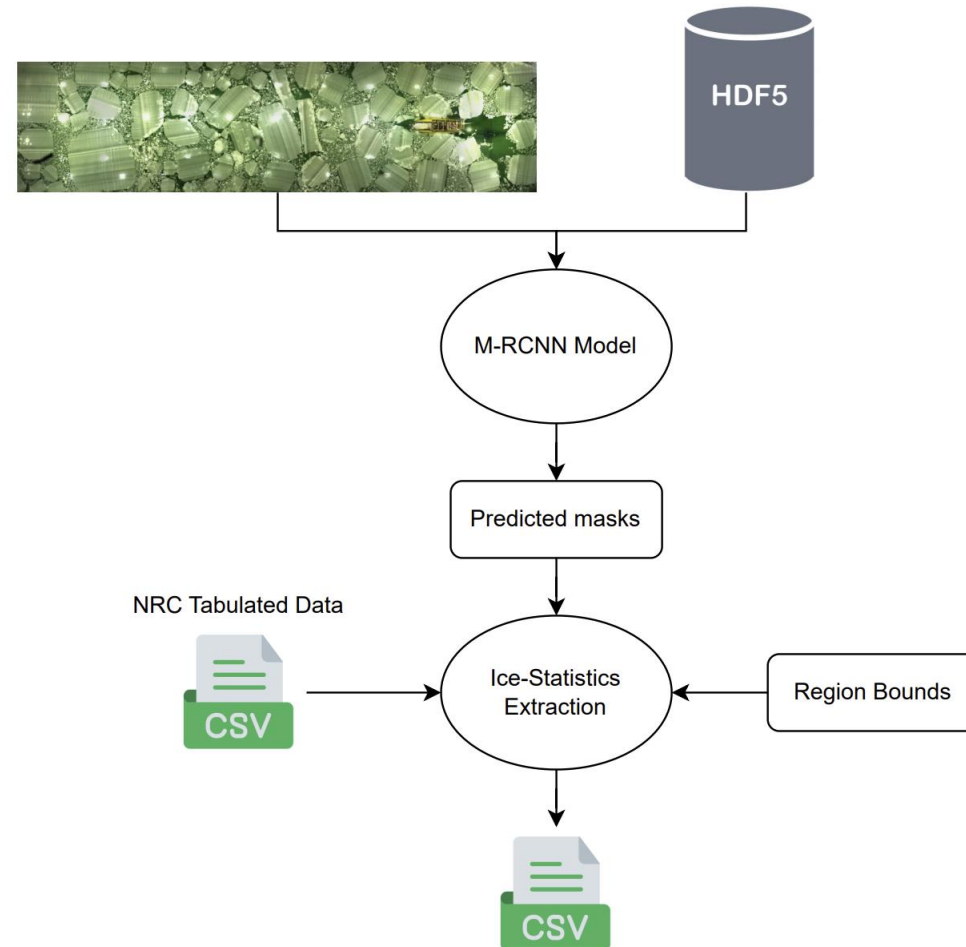


Figure: Summary Flowchart of Ice-Statistics Extraction

Force Prediction

Using output of ice-extraction (CSV), train LSTM RNN model using several of the videos available from NRC Ice testing in 2022.

- I want to try to show the correlation between a high concentration of ice in front of the vessel, indicates a likely collision a few time steps later.
 - Would be nice to be able to make some correlations here between size of the ice and which region it is in relative to the ship.

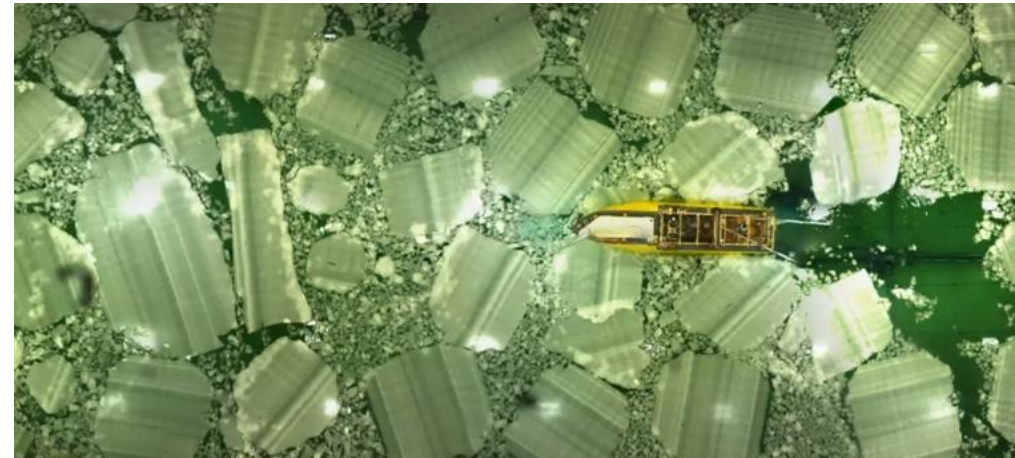
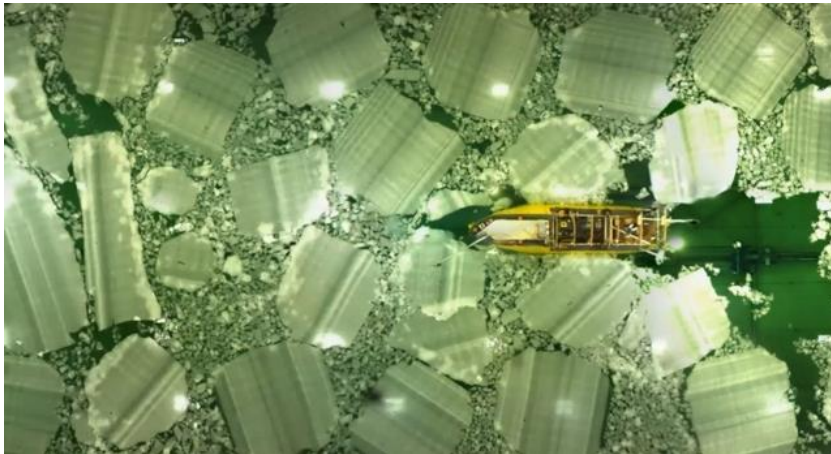


Figure: Two Different Collision Type Instances (Rubbing/Oblique Collision)

Future Works Proposals

Proposed future works in terms of experimental setup for the next 5 semesters, in order of importance (to me).

1. Direct form of collision force measurement
 - a) Tactile sensor sheet [5]
 - b) Strain gauges on hull
2. Sensor Mounting Location
 - a) Ship mounted
3. CCG Live Webcam footage
 - a) Outfitting live hull with strain gauges, tactile sensor sheet?
4. Using Wayne Pearson's Ice Simulator Model
 - a) Infinite data source
5. Alternative sensors
 - a) SAR, LiDAR, etc.

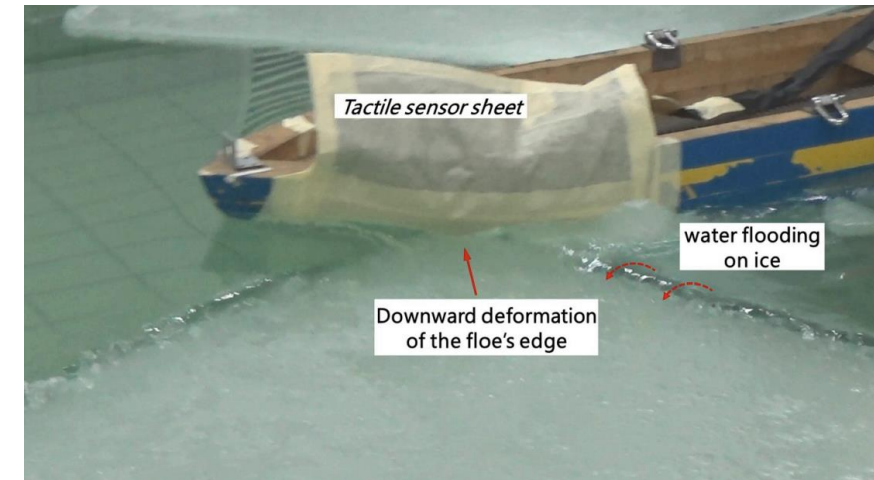


Figure: Tactile Sensor Sheet [5]

[5] Sun and Y. Huang, "Investigations on the ship-ice impact: Part 1. experimental methodologies," Marine Structures, vol. 72, p. 102772, 2020

Future Works Proposals

Proposed future contribution direction

1. Add brash/slush ice considerations, similar to B.Dowden [4]
 - a) Generally expanding ice type capabilities
2. Decrease reliance on ice types provided by NRC; consideration of ice thickness/density/age/inertia without this being an external input.
3. Considering/identifying/tracking individual ice collision events.
4. Expanding prediction time horizon to all ice pieces in a decent region around the ship
 - a) POC that this could be used for trajectory planning/tracking operations
5. Reinforcement learning focusing on generalizing architecture across different vessels

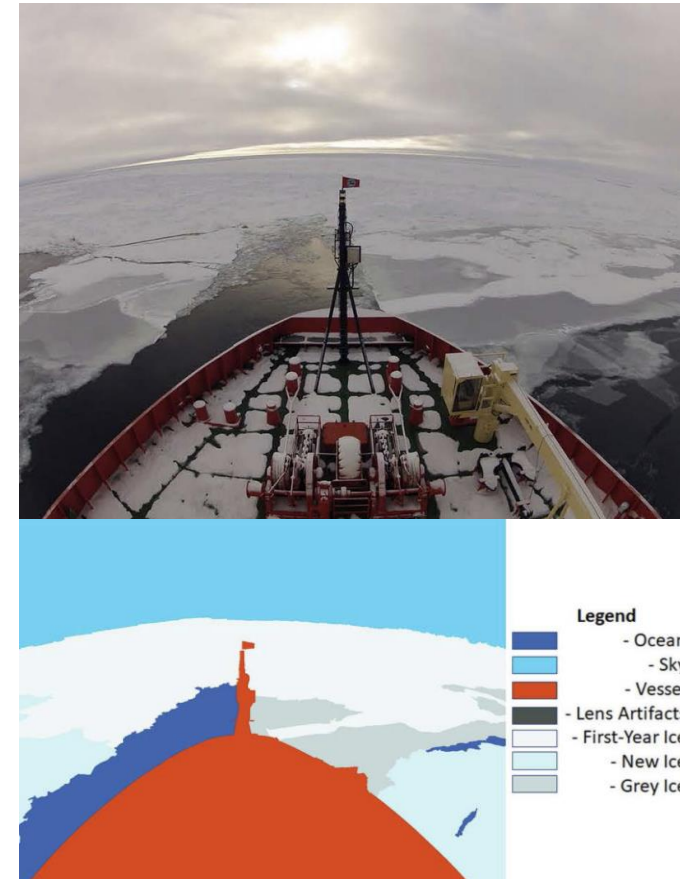


Figure: Semantic Segmentation including brash ice [6]

[6] B. Dowden, O. De Silva, W. Huang, and D. Oldford, "Sea ice classification via deep neural network semantic segmentation," IEEE Sensors Journal

Thank-you all for the help & support this semester!