

Comparing Spectral Bands for Object Detection at Sea using Convolutional Neural Networks

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Motivation – Autonomous navigation at sea

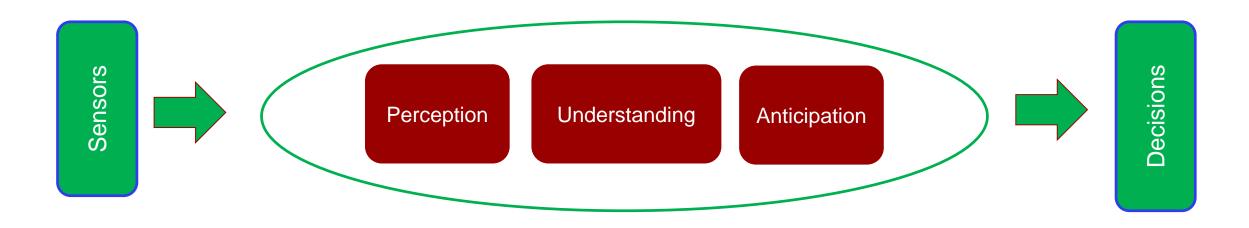
Overall goal:

To realize autonomous or assisted navigation at sea.

Steps to goal:



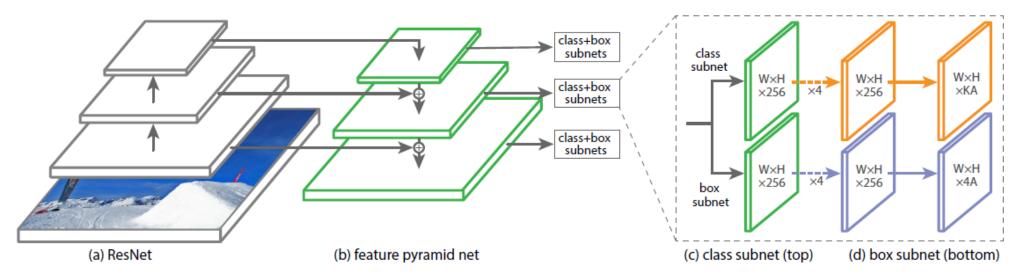
- Detect and classify nearby objects.
- Determine position and anticipate future behavior.
- Decide upon maneuvering based on object behavior and COLREGs.





Convolutional Neural Nets for object detection: RetinaNet

- One-stage detector
 - Backbone feature extractor
 - Feature map extracted using feature pyramid net
 - Predetermined anchor sizes assigned at each layer using stride
 - Anchors sent to classifier



Lin et al: Focal Loss for Dense Object Detection 2017



Challenges with vision system

Challenges:

- Variation in visibility due to changing weather
- Reflections from water
- Day and night operation

Objectives:

- A vision system effective in multiple spectral bands for image based object detection.
- Test platform outputting the same image in different spectral ranges.



Implemented vision system

Camera Model	Wavelength	Resolution	Bit depth	HFOV
JAI GO-5000C	Visible range	2560x2048px	12 bit	55°
JAI GO-5000M	NIR 800-1000nm	2560x2048px	12 bit	55°
Teledyne Dalsa Calibir 640	LWIR 8000-14000nm	640x480px	14bit	42.5°





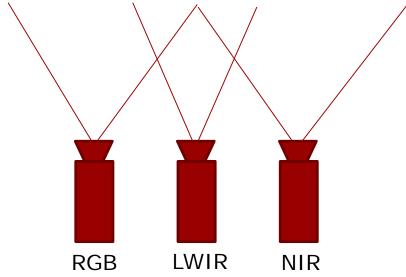




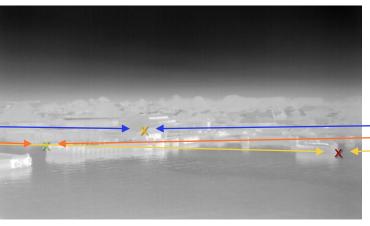




Image Alignment















Dataset

	Images	Buoy	Ship
Visible range	9229	13894	6256
NIR	9229	13889	6256
LWIR	9229	13845	6219
Validation set	923	1431	624



Example image from the Dataset.



Results - performance metrics

$$precision = \frac{TP}{TP + FP}$$

$$recall = \frac{TP}{TP + FN}$$

$$AP(q) = \int_0^1 p(r|q)dr, \ mAP = \frac{1}{Q} \sum_{q=1}^Q AP(q)$$

	Visible range	NIR	LWIR
Precision	0.96	0.96	0.95
Recall	0.9	0.94	0.86
mAP @0.25 loU	0.98	0.98	0.96
mAP @0.50 loU	0.95	0.96	0.89
mAP @0.75 loU	0.45	0.42	0.37

What is there



What CNN says



True Positive



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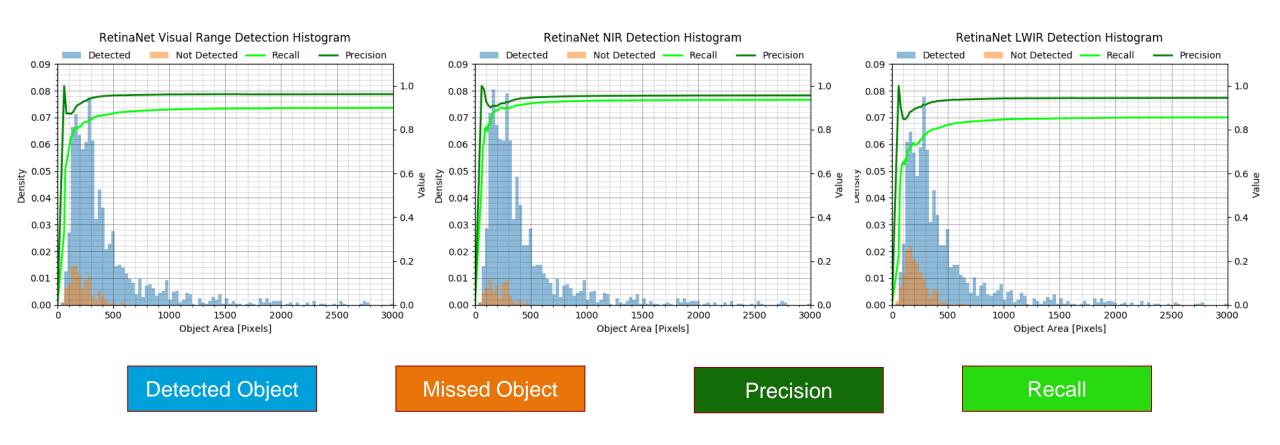
What CNN says



False Negative



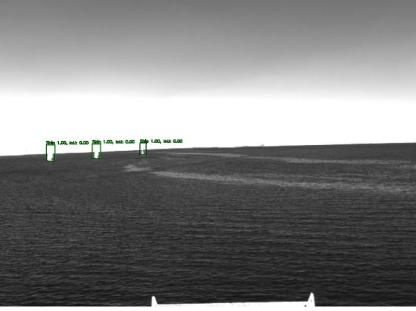
Pixel size of objects needed for detection: Detection Histograms

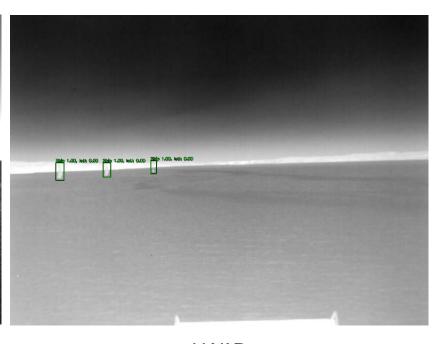




Examples – Daytime



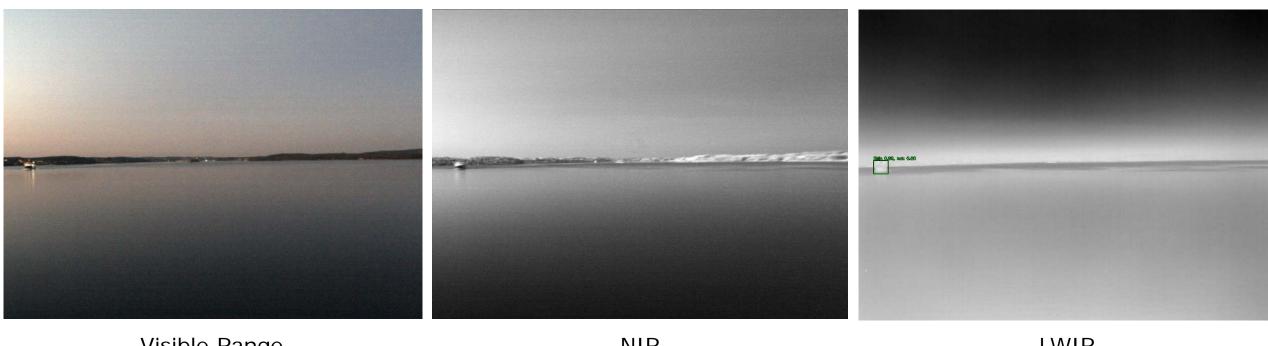




Visible Range NIR LWIR



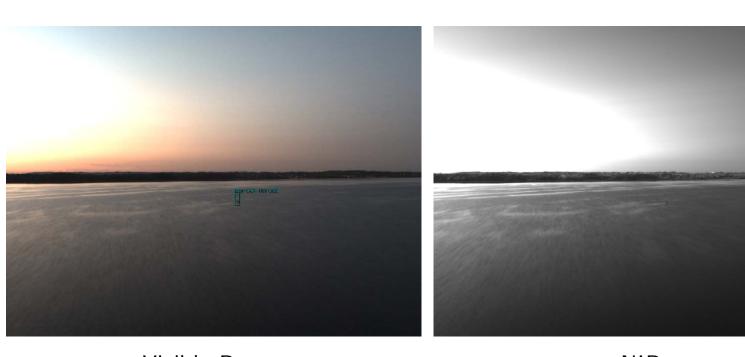
Examples – Dusk

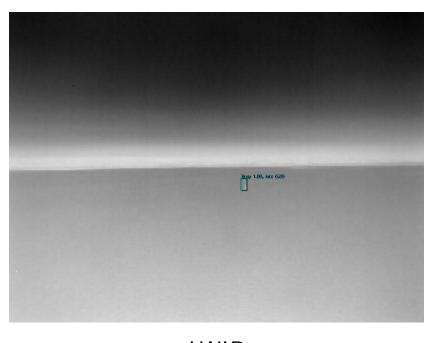


Visible Range NIR **LWIR**



Examples – Dusk





Visible Range NIR LWIR



Navigation perspective

M/F Marstal Ferry: 5.45 nmi away



	Visible	NIR	LWIR
Pixels Area (recall 0.5)	70	60	70
Minimum Detection Distance (recall 0.5)	1.06 nmi	1.27 nmi	1.06 nmi



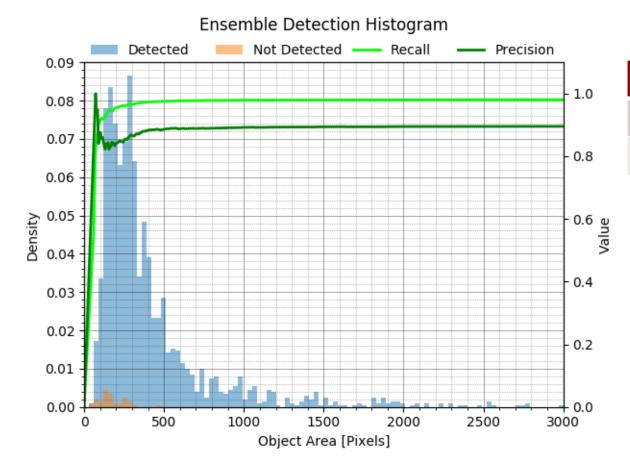
Conclusions

- 3 Spectral ranges assessed for object detection in maritime images.
- Possible to detect with high probability, detecting up to 94% of objects in single image.
 - Missed objects are far away and will be detected as they approach.
 - Object tracking will improve detection
- Each spectral range has its advantages.
 - Visible range includes the color of navigational lights.
 - NIR best for detection in day time.
 - LWIR best for detection during low light.



Recent and Future work

Ensemble modeling



	Visible range	NIR	LWIR	Ensemble
Precision	0.94	0.942	0.938	0.904
Recall	0.92	0.949	0.86	0.984

Manuscript submitted: Frederik E. T. Schöller, Martin K. Plenge-Feidenhans'l, Jonathan D. Stets, Mogens Blanke: Object Detection Performance for Marine Autonomous Crafts using Ensemble Models.



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