



# Automated passive acoustic detection and aural classification of blue and fin whale calls

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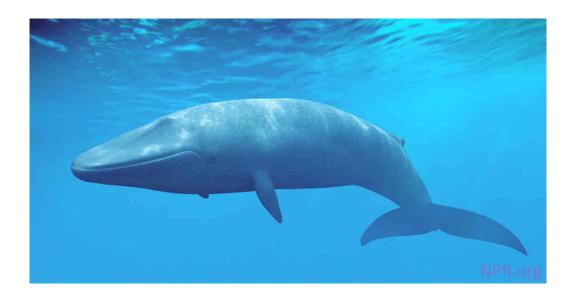
DCLDE Workshop 2015 La Jolla, CA 13-16 July 2015





### Introduction

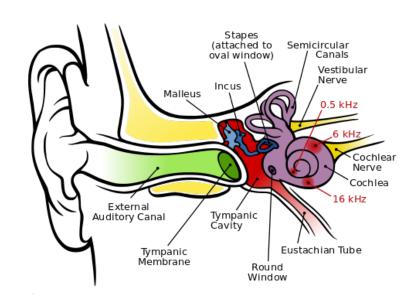
- Want to develop robust classifier capable of identifying different species and removing false detections
- Implement a general automatic detector with a high detection rate
  - Accept a high false positive rate
- Detections passed to automatic classifier to reduce number of false detections and classify marine mammal species

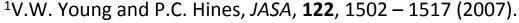




### **Introduction – Aural Classifier**

- Aural classifier previously developed at Defence R&D Canada (DRDC)
  - Uses perceptual signal features that model how humans perceive sound
  - Designed for broadband signals that may be more complicated than calls considered here
  - Use simple Bayesian classifier (Gaussian statistics)
  - Previously been successfully used for inter-species discrimination of cetacean vocalizations





<sup>&</sup>lt;sup>2</sup>S.M. Murphy and P.C. Hines, *JASA*, **135**, 626 – 636 (2014).

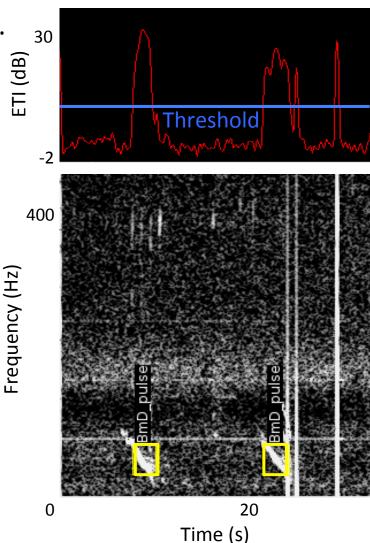


<sup>&</sup>lt;sup>3</sup>C.M. Binder and P.C. Hines, *JASA*, **135**, 2113 – 2125 (2014).

### **Detections**



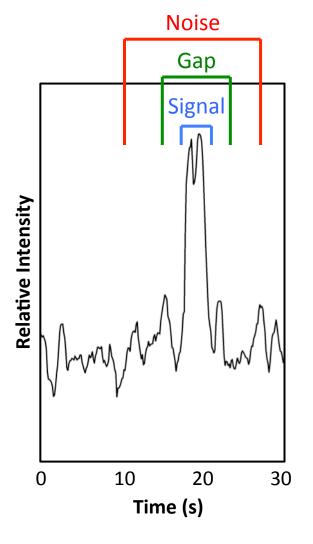
- Detections performed by Geospectrum Inc. using band-limited energy detector developed for DRDC under previous contracts
  - Detection function calculated by estimating short-term energy average in signal band and dividing by longer average of noise energy
- True detections are considered to be detections that overlap with analysts' annotated calls
  - Only D and 40 Hz calls
- Detections are placed in center of .WAV
   file to be input to aural classifier algorithm





### **Detection Results**

Detector	Blue whale D	Fin whale 40 Hz
Frequency Band (Hz)	30 – 70	50 – 65
SNR threshold (dB)	4.0	6.0
Signal window length (s)	1.0	0.2
Number of true detections [Detection rate]	4 268 [89.4 %]	1 450 [88.1 %]
Number of false detections [False positive rate *]	210 975 [27.3 s/(FP)]	311 875 [18.5 s/(FP)]

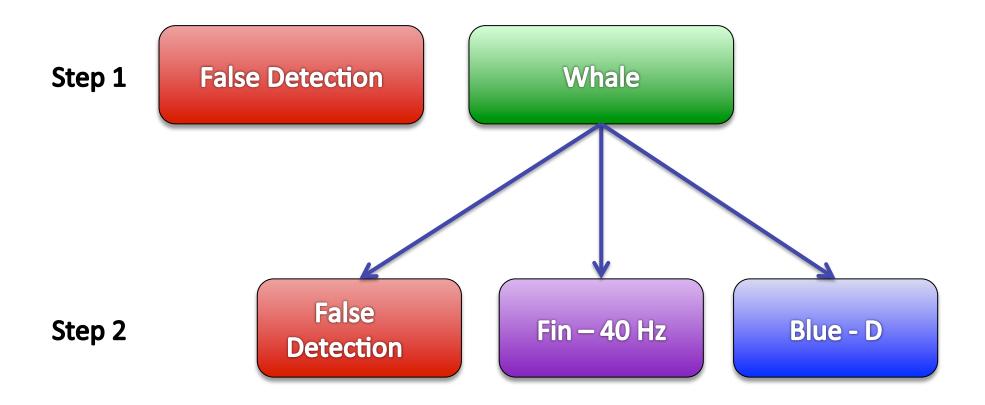


<sup>\*</sup> The time between false detections; considers anything other than blue or fin calls a false detection



### **Classification method**

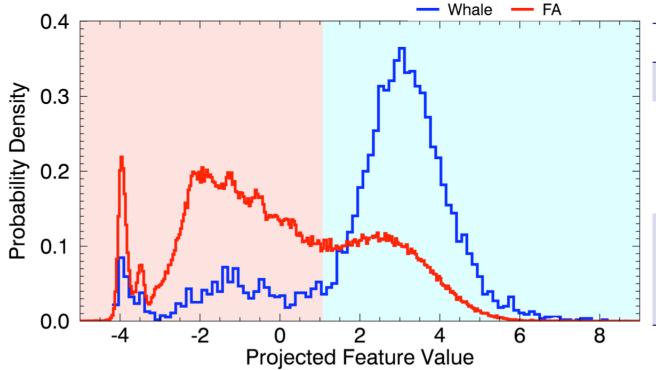
Performed a two step classification with data from each of the detectors





# **Blue Whale D Call Detector Results**

- Stage 1
  - Use all calls that were classified as whale as inputs to the next processing stage



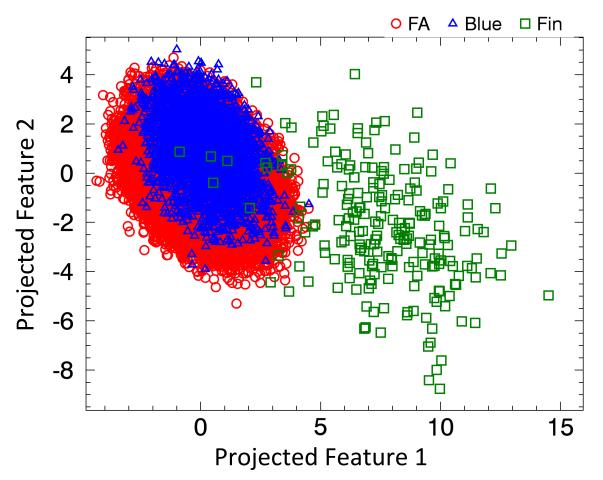
Overall accuracy	67.8%
AUC	0.785
Number of correctly classified whale calls	3 464 (81.2%)
Number of incorrectly classified false detections	68 397 (32.4%)



### **Blue Whale D Call Detector Results**

Stage 2

■ This is the final classification decision for data from blue whale detector.



Overall accuracy	65.4%
Blue accuracy	63.2%
Fin accuracy	91.8%
False detection accuracy	65.4%
AUC	0.785
Precision	0.169
Recall	0.635



# Fin Whale 40 Hz Detector Results

Performed same two stage method for fin whale 40 Hz detector

Stage 1		
Overall accuracy	82.7%	
AUC	0.694	
Number of correctly classified whale calls	1 762 (42%)	
Number of incorrectly classified false detections	36 284 (11.6%)	

Stuge 1	
Overall accuracy	75.8%
Blue accuracy	88.1%
Fin accuracy	79.0%
False detection accuracy	75.2%
AUC	0.950
Precision	0.022
Recall	0.681

Stage 2



### **Combined Results**

Combined results from both the blue whale D and fin whale 40 Hz call detectors

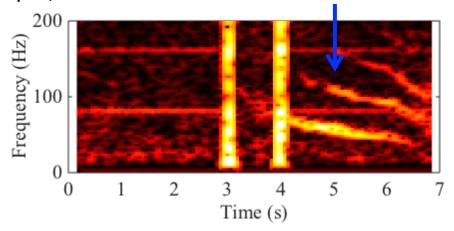
Precision	0.120
Recall	0.638

- Possible explanations for poor results
  - Detector generated too many false detections
  - A lot of variability in false detections, particularly between the sites
  - Other whale calls not annotated, so they were classified as false detections



### **DCPP** A site

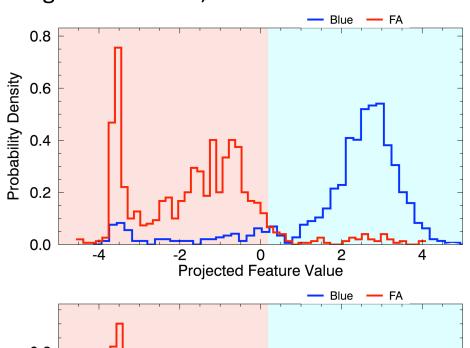
- Look at reduced set to get a better idea of strengths/challenges
  - Calls from July 2013; 955 signals for each of blue whale calls and false detections in training set; same number of calls in testing set
  - Classification results ok, ~70% of blue whale calls and ~90% of false detections correctly classified
- Further reduced set by removing detections that had been labelled as true detections but were contaminated with other loud signals
  - For example,





### **DCPP** A site

- Further reduced set with "bad" detections removed
  - Training set: 764 blue, 955 false detections
  - Testing set: 775 blue, 955 false detections



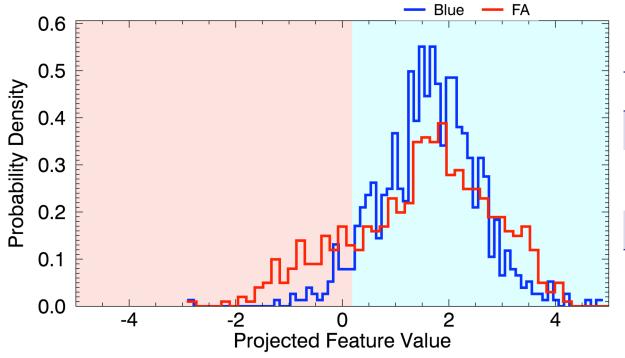
		Projected Feature Value	
		— Blue — FA	
Probability Density	0.6 0.4 0.2 0.0	-4 -2 0 2 4 Projected Feature Value	

# Training Overall accuracy 91% Blue accuracy 88% False detection 93% accuracy AUC 0.92

Testing	
Overall accuracy	91%
Blue accuracy	87%
False detection accuracy	94%
AUC	0.92

# **Between-site Variability**

- Apply classifier trained with data from DCPP A site to calls recorded at CINMS B during June 2012
  - 757 blue whale calls and 642 false detections

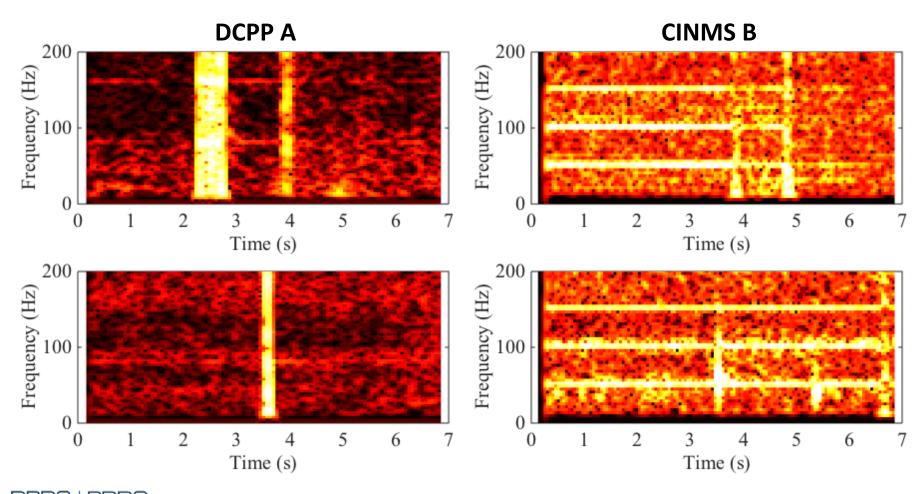


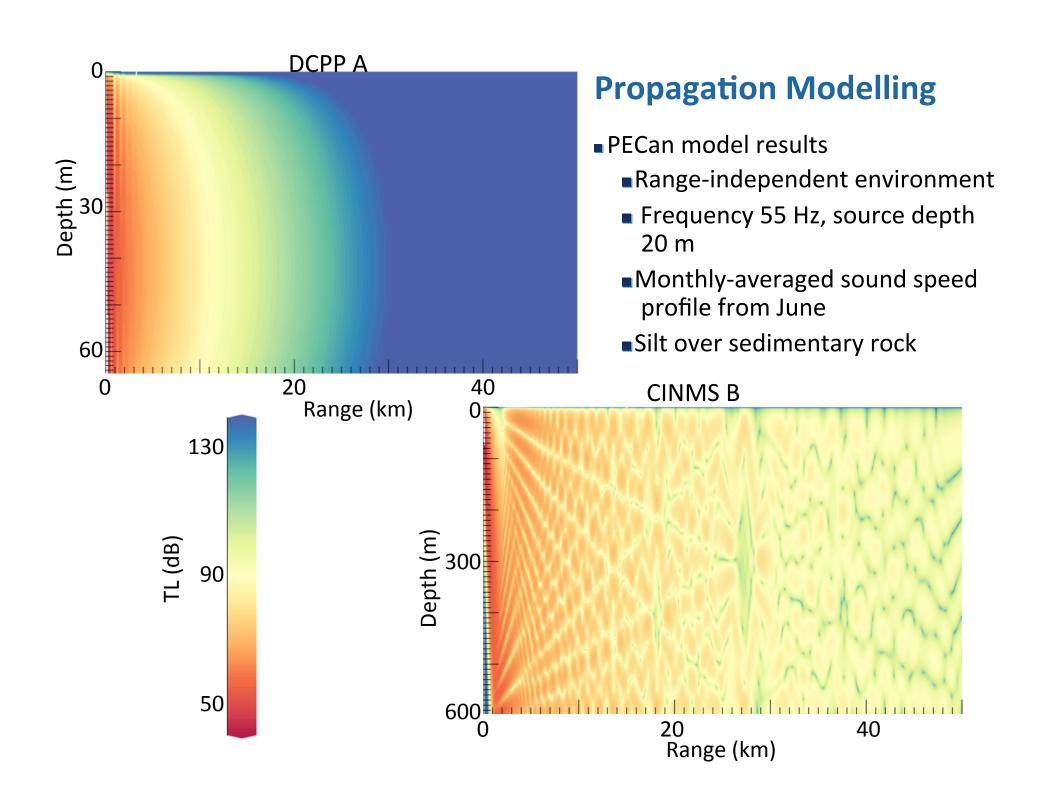
Testing		
Overall accuracy	60%	
Blue accuracy	94%	
False detection accuracy	19%	
AUC	0.52	



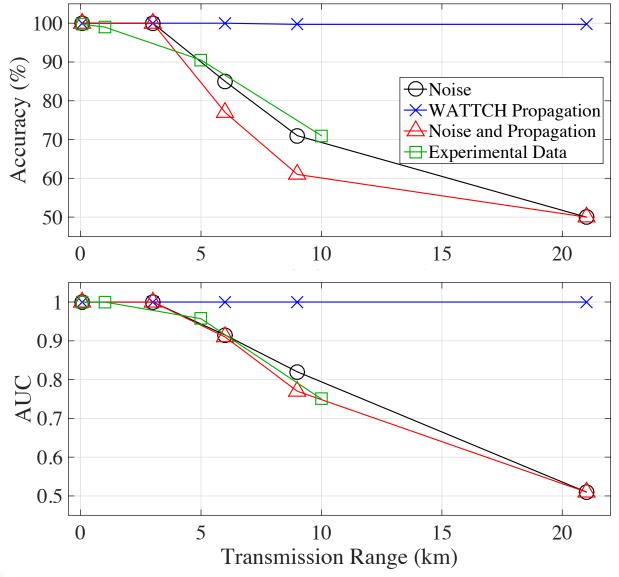
# **Between-site Variability**

 Majority of false detections from CINMS B site different from those at DCCP site A





# **Performance Dependence on SNR and Propagation**





# **Concluding Remarks**

- Aural classifier did not perform particularly well
  - Detector settings need to be refined to limit the number of false detections
  - Large variability in types of noise in false detection class
  - Features were originally designed for transients with more complicated time-frequency features
- When developing training set all whale calls in band of interest should be annotated not just a few call types
- Need to keep in mind impact of propagation and SNR on classifier performance



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