

# Creating training data sets for the automatic identification of wildlife sounds: a narwhal example



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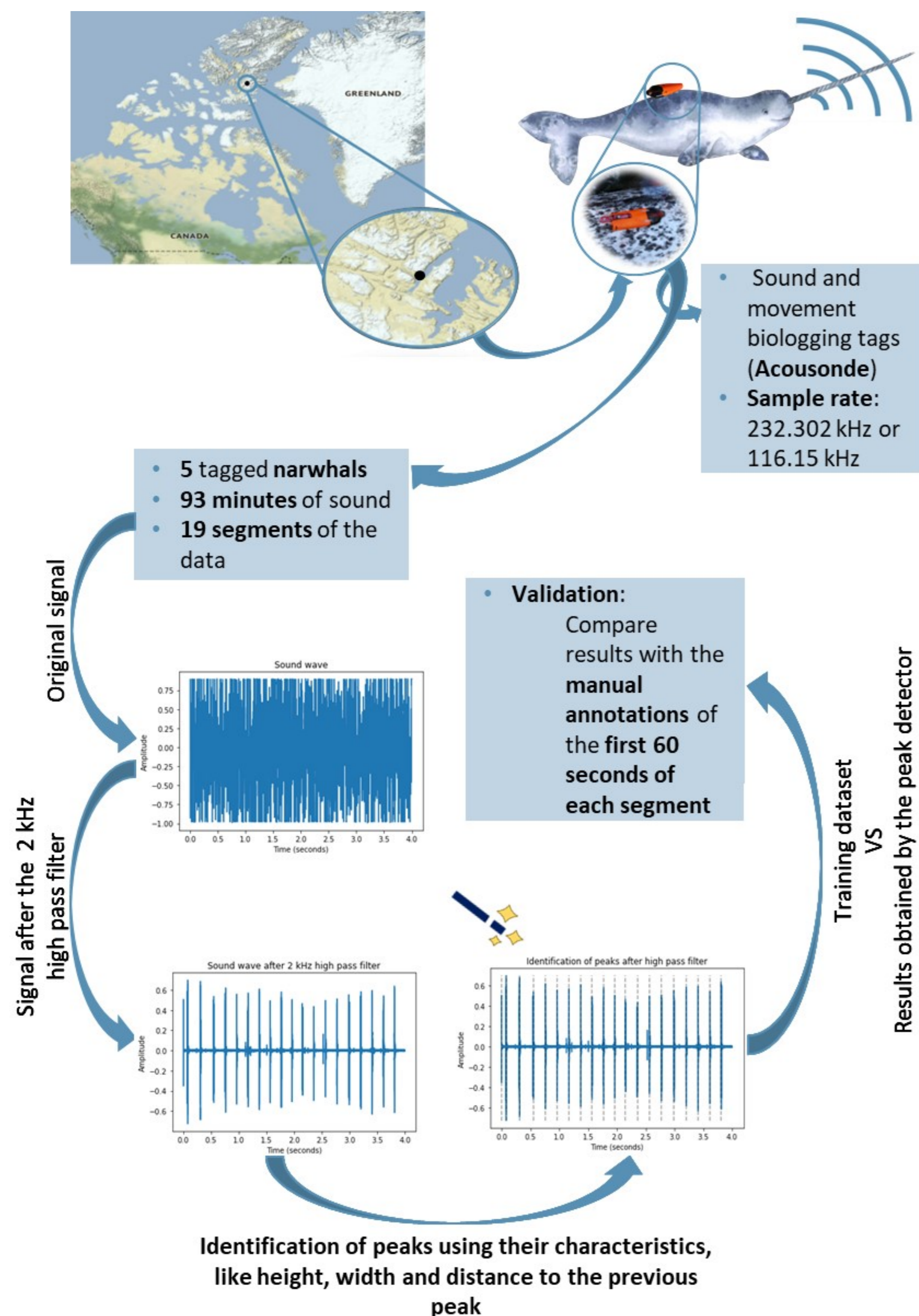
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## Introduction

- ↑ Capacity for **automated data collection** → **Large amount of data** → Search for **automatic methods** for data processing;
- **For example:** in bioacoustics it is possible to obtain **acoustic data for long periods** leading to **terabytes of data**;
- **Animal vocalizations** are often **embedded in background noise** that can make their **detection challenging** (Huynh et al. (1998));
- **High pass filters remove** low-frequency noise that is often associated with **environmental sounds** while **preserving** the higher frequency sounds that are typically associated with **animal vocalizations** (Xie et al. (2021));
- As a result, **high pass filters** have become an **essential tool** in acoustic ecology and conservation biology, **helping** researchers to **monitor and understand the vocal behavior** of a wide range of animal species.

## Methodology



## Results

- **31899** clicks identified **automatically**
  - **3401** clicks identified **manually** from the **first 60 seconds** of each segment
- **Signal-to-noise ratio:** -11.124 → -0.1691
- **Mean Squared Error:** 0.0531 → 3.350e-03
- **Precision** ( $\frac{TP}{FP+TP}$ ): 0.339
- **Sensitivity** ( $\frac{TP}{FN+TP}$ ): 0.671
- **False negative rate** ( $\frac{FN}{FN+TP}$ ): 0.329
- **Balanced accuracy** ( $\frac{TP+FN+TN+FP}{2}$ ): 0.835
- **F-1 score** ( $\frac{2TP}{2TP+FP+FN}$ ): 0.451
- Where TP= true positives, FP = false positives, FN= false negatives, TN= true negatives

## Discussion and Conclusions

The **high pass filter** shows a significant **increases** in the **signal-to-noise ratio**. This leads to an **easier identification of the click sounds** produced by the narwhals. The **automatic detection** was done by **finding the peaks** in the denoised signal, with certain characteristics, like height, width and distance to the previous peak. The comparison between the automatic detection and the manually annotated datasets shows that the **high pass filter generated peaks that are easily identified as vocalizations**. With the use of the high pass filter method we get a **precision of 0.656**, a **sensitivity of 0.707** and an **F-1 score of 0.680**. Therefore, these results show that the **high pass filter** is an easy and **straightforward way to denoise the sound**. This will **allow easier identification of vocalizations** through other methods that are more robust like neural networks.

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## References

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