

# FROM FREQUENCIES TO FLORA

FOURIER ANALYSIS AS A KEY  
TO SPECIES IDENTIFICATION



# OUR TEAM



ALEX

INTRO



CAITLIN

METHOD



DEAN

RESULTS



JANE

DISCUSSION

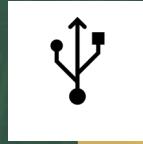


MORGAN

CONCLUSION



# OUR VISION



## Distribute

Make this technology accessible to the public



## Identify Materials

Using Fourier analysis, identify a wide range of materials



## Analyse Condition

Check the condition of a material using NDT



# OUR GOALS

## Distinguish

Distinguish between Ash, Oak and Beech blocks based on frequencies produced when struck



## Determine

Determine relationship between volume of block and peak frequency

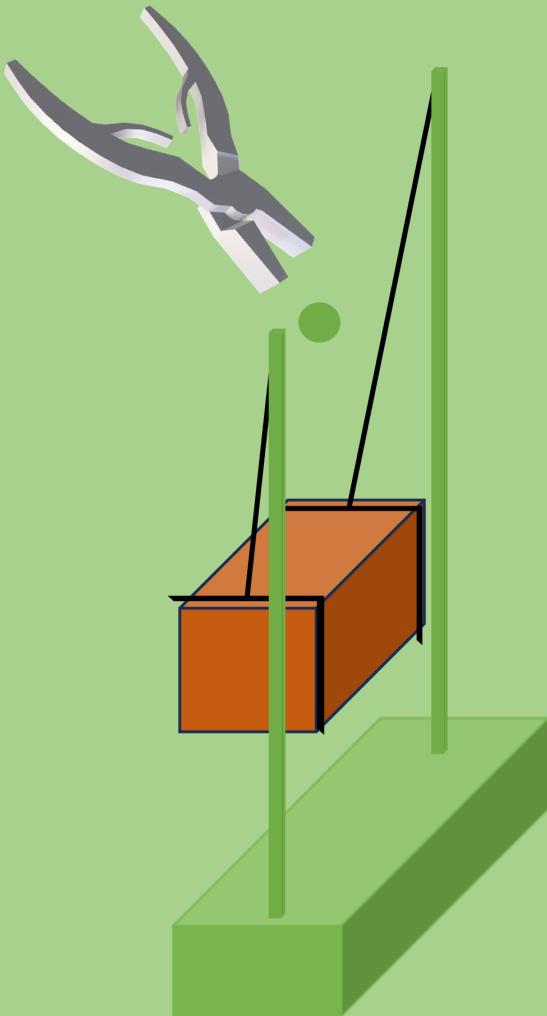


## Predict

Use graphic analysis to predict frequencies for any volume of block



# METHOD



## Block of wood suspended with dimensions:

Largest = ~60mm $\pm$ 1mm x ~60mm $\pm$ 1mm x ~105mm $\pm$ 1mm  
Medium= ~60mm $\pm$ 1mm x ~60mm $\pm$ 1mm x ~52.5mm $\pm$ 1mm  
Smallest= ~30mm $\pm$ 1mm x ~60mm $\pm$ 1mm x ~52.5mm $\pm$ 1mm

## Steel ball dropped

The steel ball weighed 60g  $\pm$  0.5g and had a 25mm  $\pm$  0.1mm diameter. Dropped from a height of 80cm  $\pm$  3cm.

## Sound recorded

Phone kept a distance 24cm  $\pm$  0.1cm from the block and 21cm  $\pm$  0.1cm below the block.



# METHODS



## Graphing

A graph of amplitude in decibels vs time was recorded in the DecibelX app, with an uncertainty of  $\pm 0.05$  db in amplitude and  $\pm 0.05$ s in time.



## Fourier Transform

Python was used to perform a Fourier transform

# RESULTS- Python

butter\_highpass

highpass\_filter

remove\_background

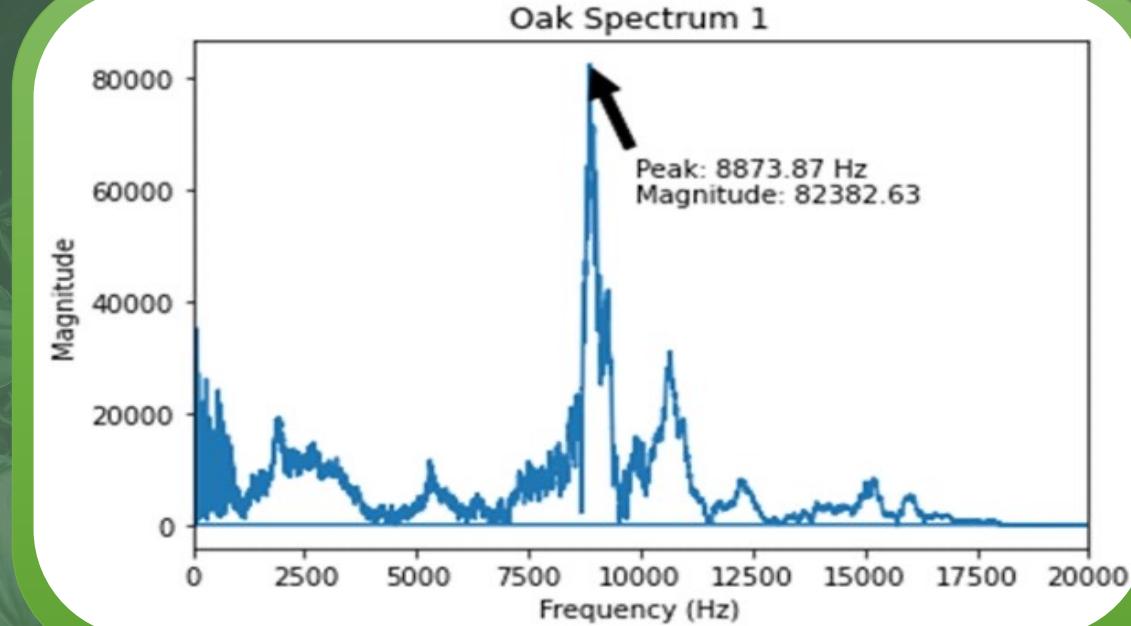
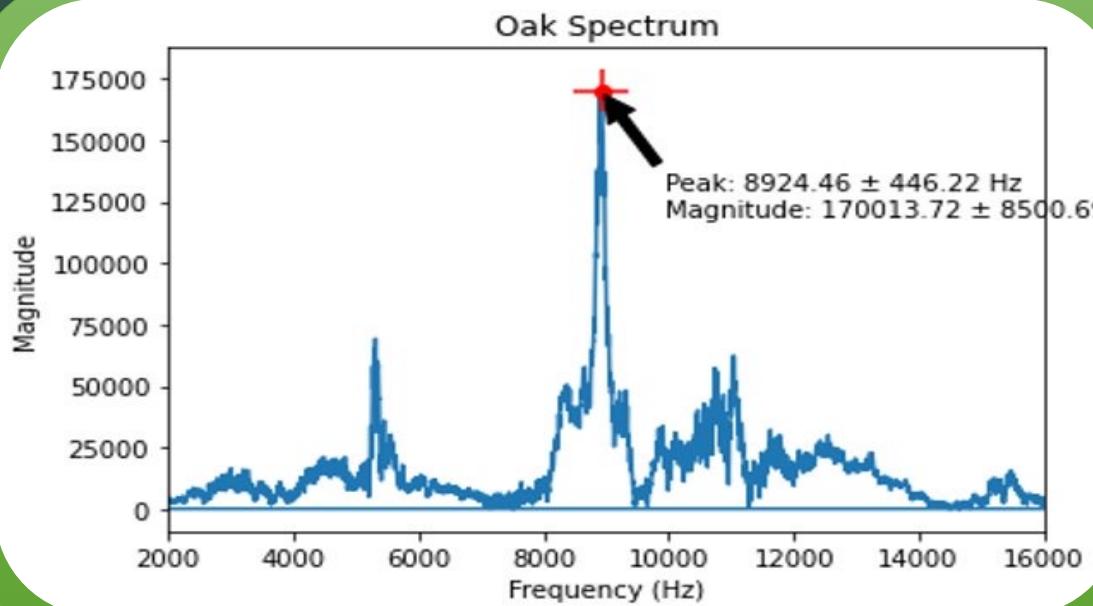
compute\_fourier\_transform

Plotting

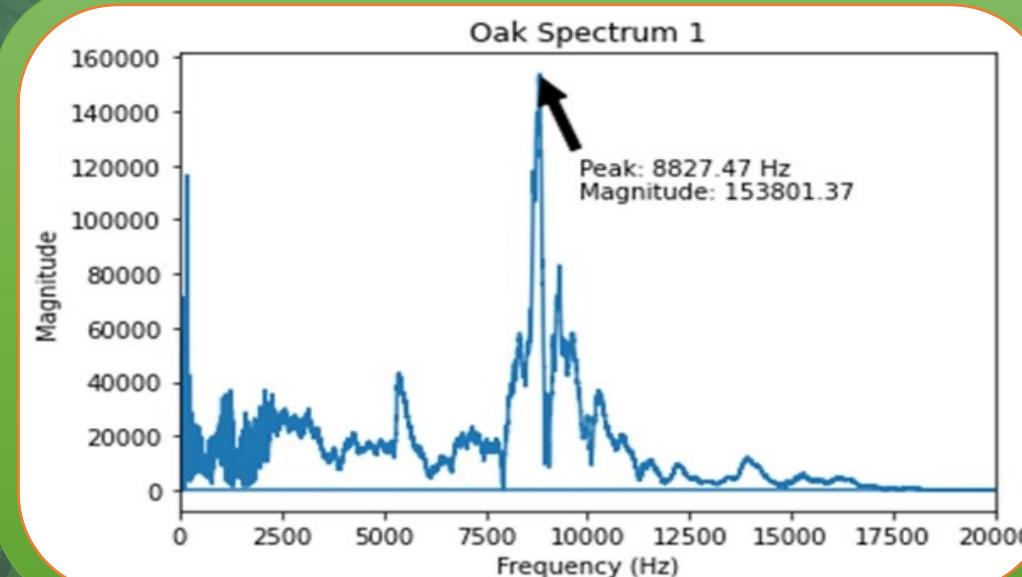
```
1 import numpy as np
2 from scipy.io import wavfile
3 import matplotlib.pyplot as plt
4 from scipy.signal import butter, lfilter
```

```
6 def butter_highpass(cutoff, fs, order=5):
7     nyq = 0.5 * fs
8     normal_cutoff = cutoff / nyq
9     b, a = butter(order, normal_cutoff, btype='high', analog=False)
10    return b, a
11
12 def highpass_filter(data, cutoff, fs, order=5):
13     b, a = butter_highpass(cutoff, fs, order=order)
14     y = lfilter(b, a, data)
15     return y
16
17 def remove_background(data, background_data):
18     if len(data) > len(background_data):
19         data = data[:len(background_data)]
20     else:
21         background_data = background_data[:len(data)]
22     return data - background_data
23
24 def compute_fourier_transform(filename, background_filename=None, trim_start_sec=0):
25     rate, data = wavfile.read(filename)
26
27     if background_filename:
28         _, background_data = wavfile.read(background_filename)
29         data = remove_background(data, background_data)
30
31     if trim_start_sec > 0:
32         trim_start_samples = int(trim_start_sec * rate)
33         data = data[trim_start_samples:]
34
35     if len(data.shape) == 2:
36         data = np.mean(data, axis=1)
37
38     # Remove DC offset
39     data = data - np.mean(data)
40
41     # Apply Hamming window
42     data = data * np.hamming(len(data))
43
44     # High-pass filtering to remove frequencies
45     data = highpass_filter(data, 4000, rate)
46
47     # Compute the Fourier Transform of the audio
48     spectrum = np.fft.fft(data)
49
50     freq = np.fft.fftfreq(len(spectrum), d=1/rate)
51
52     return freq, spectrum
```

# RESULTS



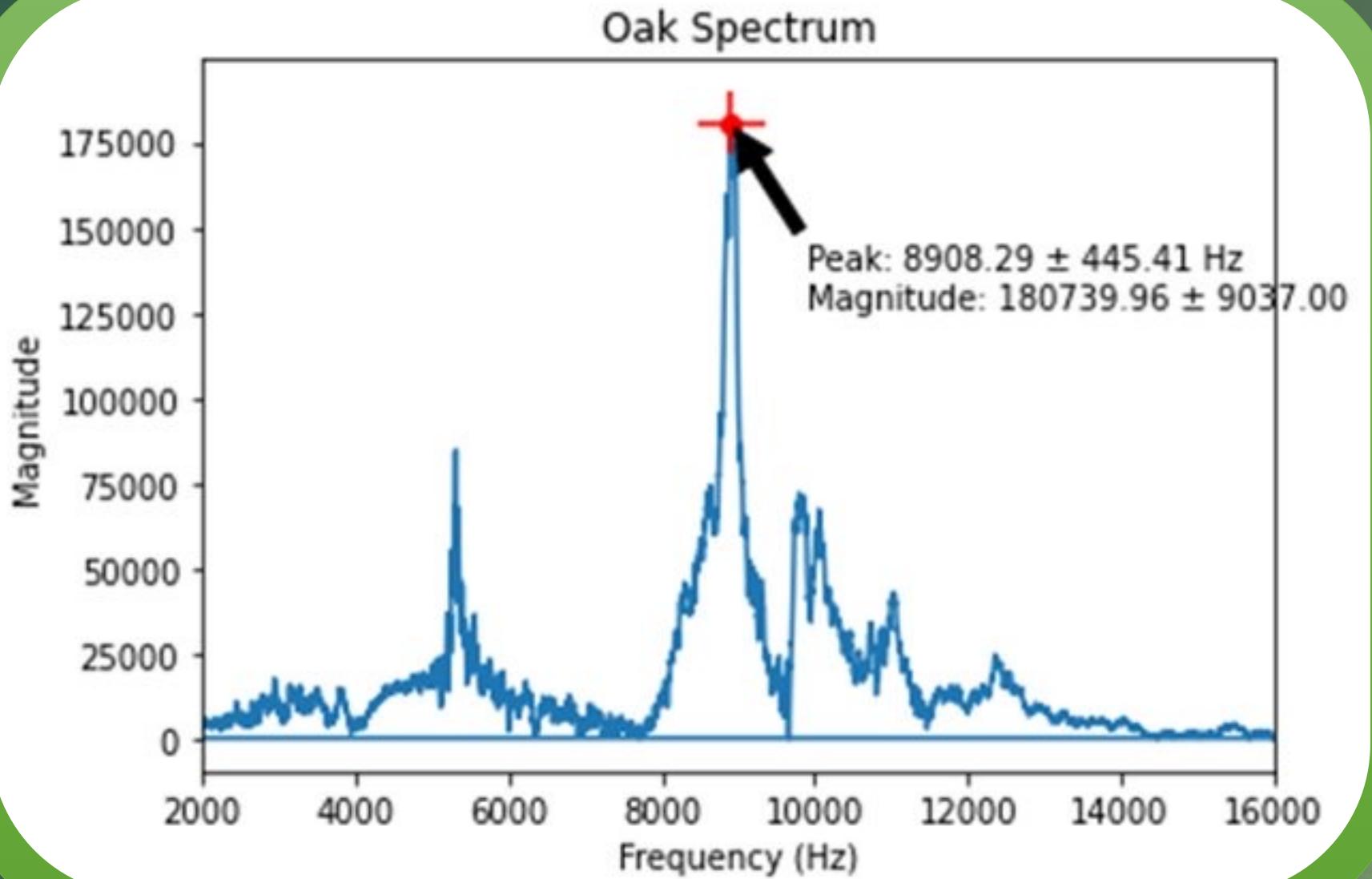
Day 1



Day 2

Day 3

# RESULTS

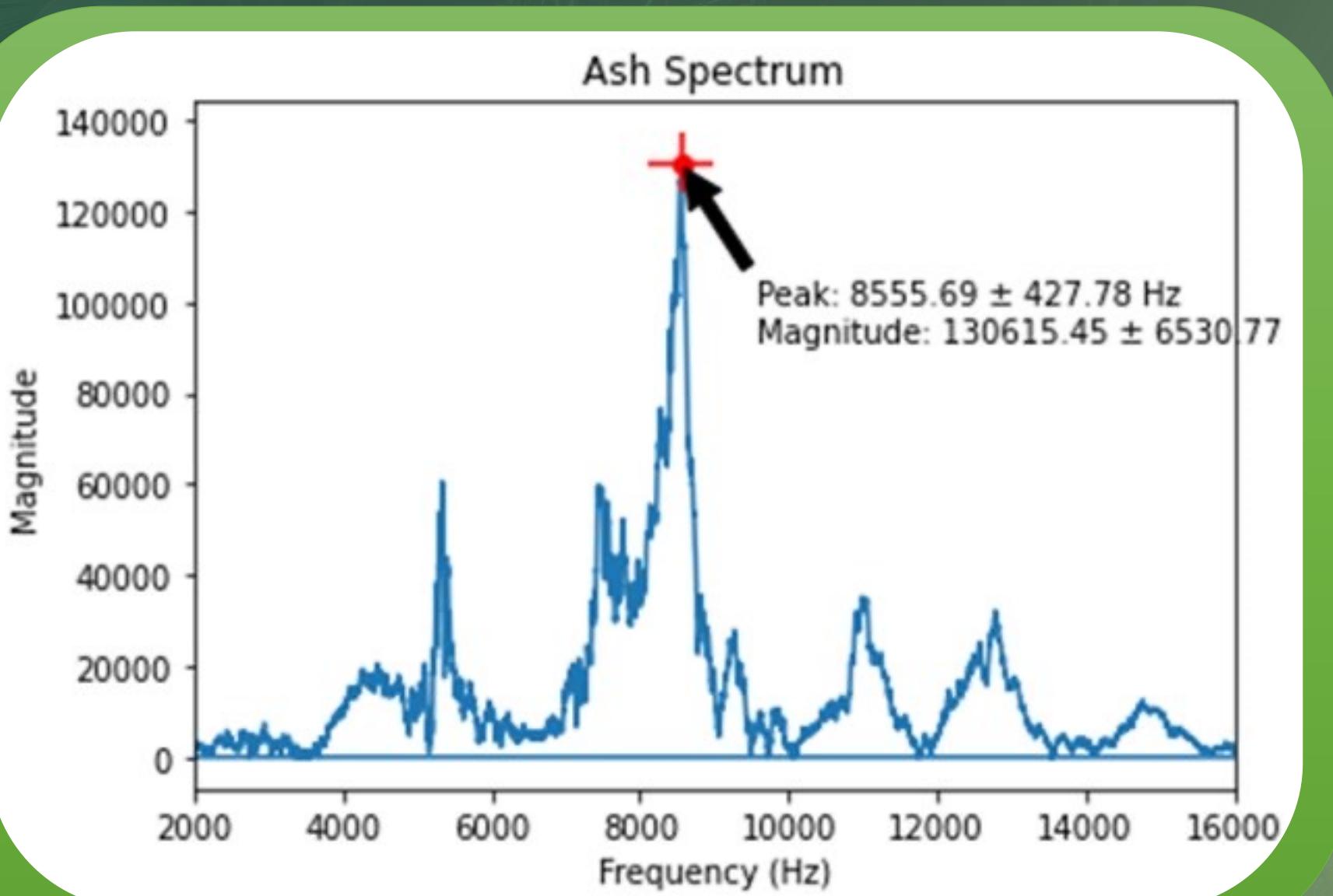


- Nyquist frequency, which was consistently 22050.0 Hz, half the usual sampling rate for WAV files.

Oak:  $8908.29 \pm 445.41$  Hz -  
Frequency Resolution:  
 $0.4306640625$  Hz

Full Blocks

# RESULTS

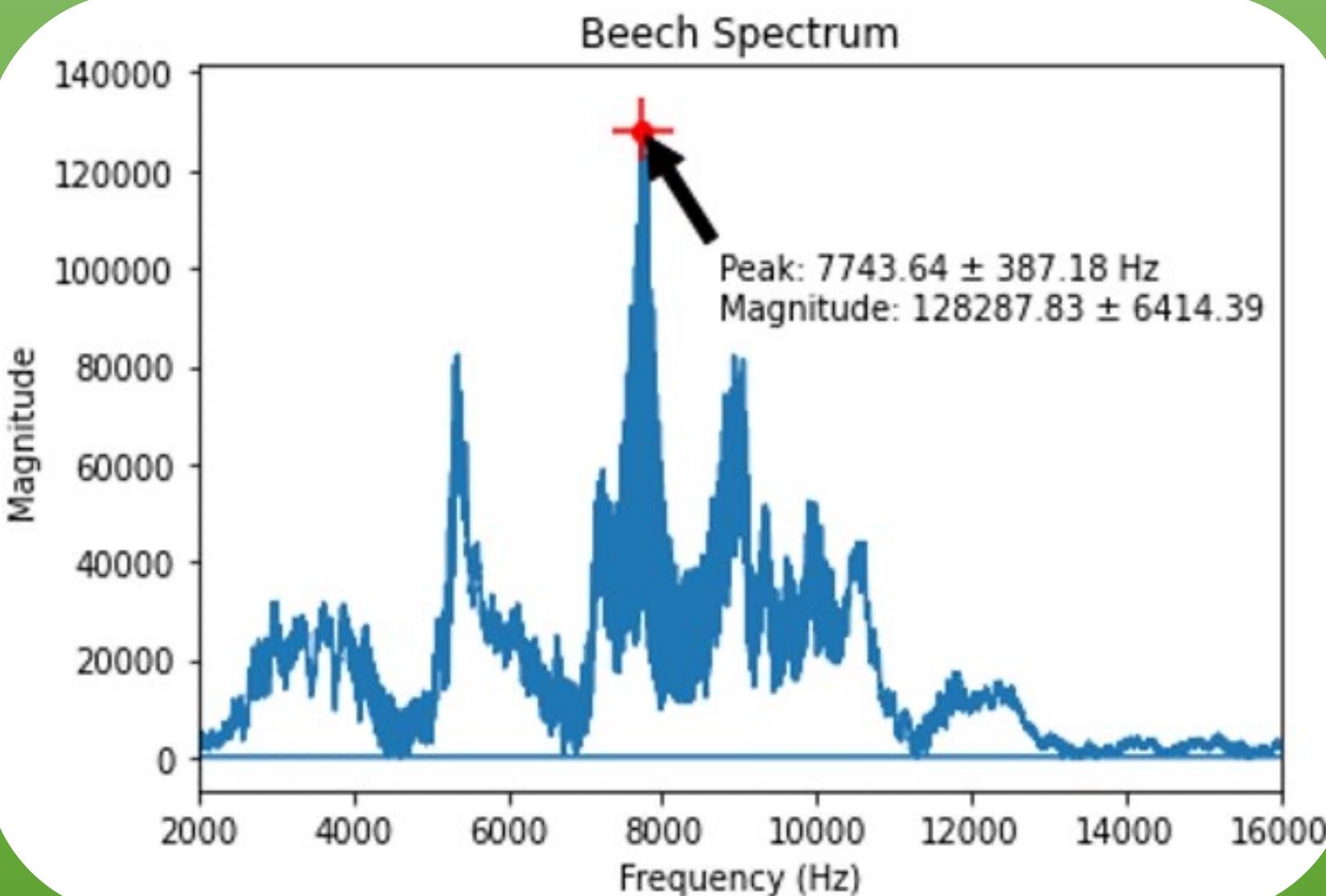


Nyquist frequency, which was consistently 22050.0 Hz, half the usual sampling rate for WAV files.

Ash:  $8555.69 \pm 427.78$  Hz  
- Frequency Resolution:  
 $0.566663240131579$  Hz

Full Blocks

# RESULTS

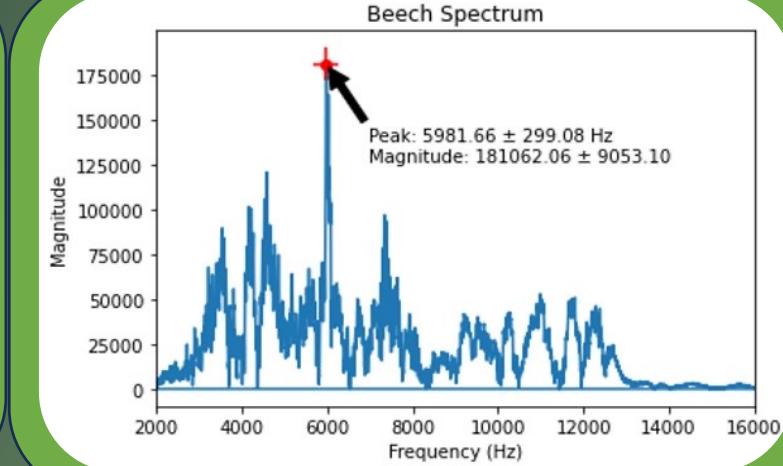
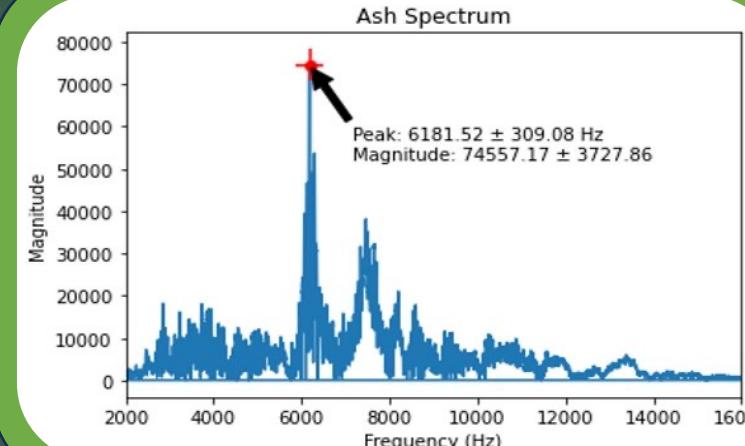
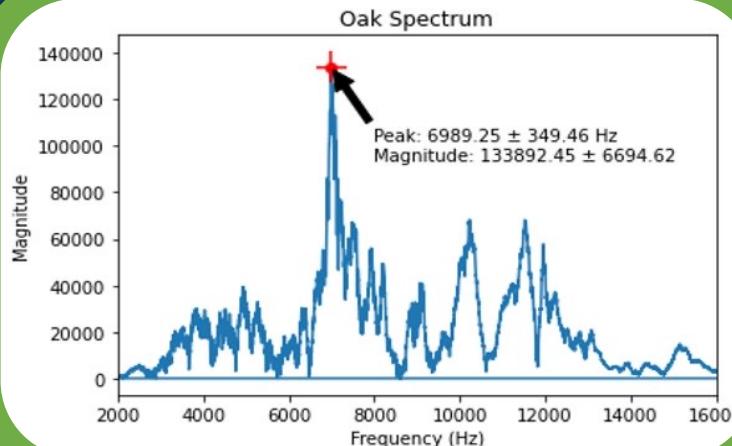


- Nyquist frequency, which was consistently 22050.0 Hz, half the usual sampling rate for WAV files.

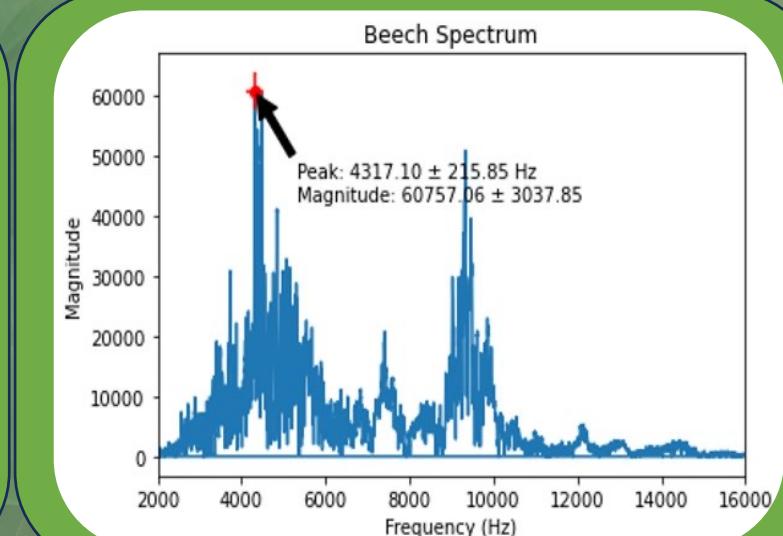
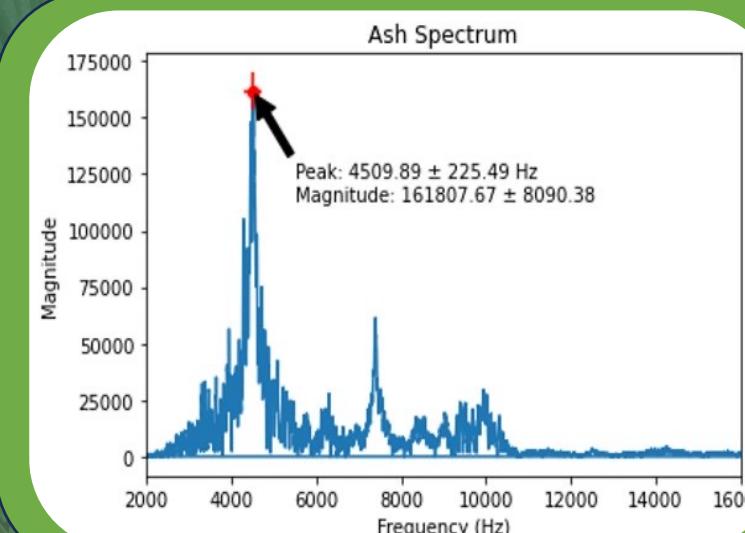
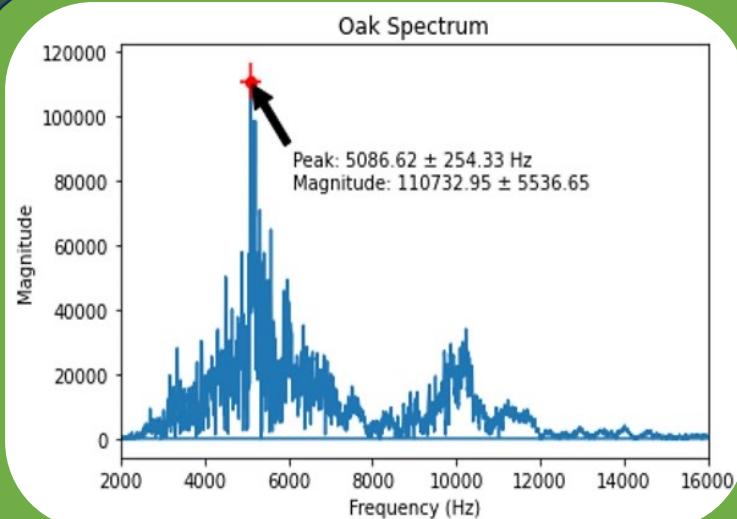
Beech:  $7743.64 \pm 387.18$  Hz  
- Frequency Resolution:  
 $0.4024897780373832$  Hz

Full Blocks

# RESULTS



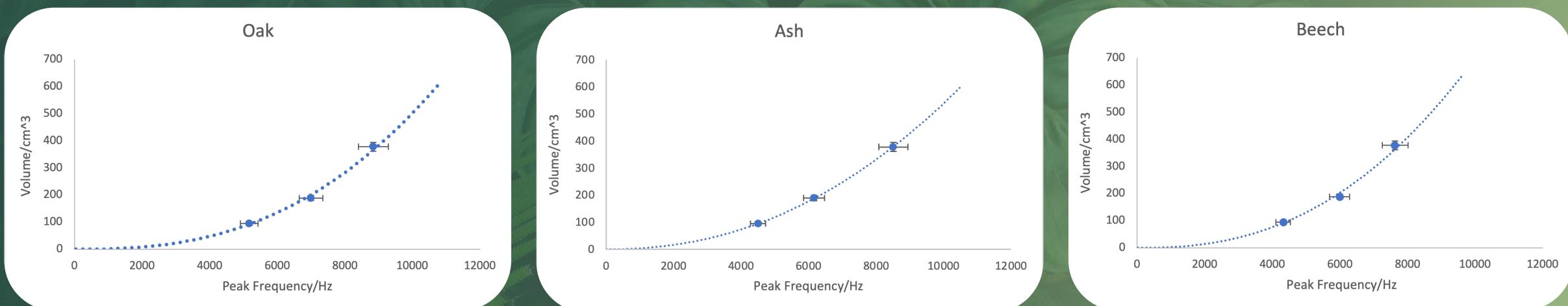
$\frac{1}{2}$  Blocks  $\uparrow$   $\downarrow$   $\frac{1}{4}$  Blocks



# RESULTS



Difference between sizes (Hz)	Type	Full	1/2	1/4
1845.817 1823.62	Oak	8835.215 +- 441.76 Hz	6989.40 +- 349.47 Hz	5165.78 +- 258.289 Hz
2345.747 1662.43	Ash	8510.347 +- 425.517 Hz	6164.60 +- 308.23 Hz	4502.17 +- 225.1085 Hz
1640.903 1668.23	Beech	7626.233 +- 381.411 Hz	5985.33 +- 299.27 Hz	4317.10 +- 215.855 Hz



# DISCUSSION-RESULTS



**Consistency across experiments**



**Smaller volume, smaller frequency**



**Differences in materials ascertained**



**Resonance occurring**

# FUTURE ASPIRATIONS



Grading Materials Using NDT



Identify Wider Range of Materials  
Using NDT



FTIR Used to Identify Disease



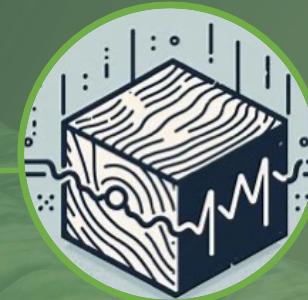
Ecosystem Protection

# CONCLUSION



## Material Identification

Demonstrated the ability to use NDT for distinguishing samples.



## Frequency Relationship

Exhibited the relationship between block size and frequency.



## Material Grading

Demonstrated the ability to use NDT for identifying defaults in samples.

# REFERENCES



<https://www.bing.com/images/create/infographic-image-of-a-clipboard-with-the-word-con/655128568a544ba5b95618d6e1c7f684?FORM=GNCRE>

Images sourced from bing image generator

Information obtained from several studies such as

<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9147555/>

<https://bioresources.cnr.ncsu.edu/resources/vibrational-characteristics-of-four-wood-species-commonly-used-in-wood-products/>