## goertzel\_test

November 13, 2023

```
[]: import numpy as np
import matplotlib.pyplot as plt
from scipy.signal import bessel, lfilter, welch
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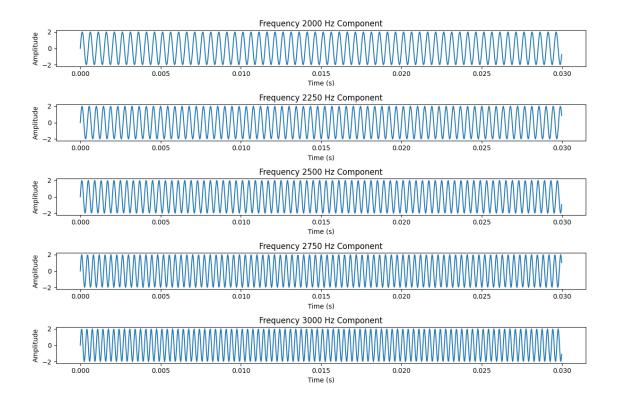
Goertzel reference: https://www.embedded.com/the-goertzel-algorithm/

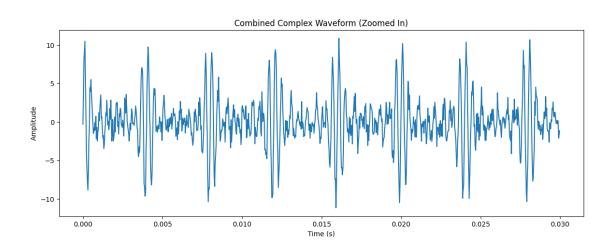
```
[]: # Goertzel algorithm function
     def goertzel(samples, sample_rate, target_freq, n):
         The Goertzel algorithm is used to calculate the magnitude of a specific \Box
      ⇔frequency component within a signal.
         This is particularly efficient for computing the spectral content at_{\sqcup}
      ⇔specific frequencies of interest.
         k = int(0.5 + n * target freq / sample rate)
         omega = (2 * np.pi / n) * k
         coeff = 2 * np.cos(omega)
         q0, q1, q2 = 0, 0, 0
         for sample in samples:
             q0 = coeff * q1 - q2 + sample
             q2 = q1
             q1 = q0
         real = (q1 - q2 * np.cos(omega))
         imag = (q2 * np.sin(omega))
         return np.sqrt(real**2 + imag**2)
     # Function to calculate magnitude response over a range of frequencies
     def magnitude_response(signal, sample_rate, freq_range, n):
         This function calculates the magnitude response of a signal over a range of [1]
      \hookrightarrow frequencies.
         It is useful for analyzing how different frequencies are represented in the \sqcup
      \hookrightarrow signal.
         magnitudes = []
         for freq in freq_range:
             magnitude = goertzel(signal, sample_rate, freq, n)
             magnitudes.append(magnitude)
```

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return magnitudes
# Function to apply a 4th order Bessel filter
def apply_bessel_filter(signal, sample_rate, cutoff_freq):
    11 11 11
    This function applies a 4th order Bessel filter to the signal.
    The Bessel filter is known for its linear phase response but has a gentler
 ⇔roll-off compared to other filters.
    The cutoff frequency determines the frequency point at which the filter 
 ⇔starts attenuating the signal.
    11 11 11
    b, a = bessel(N=4, Wn=cutoff freq/(0.5*sample rate), btype='low',
 →analog=False)
    return lfilter(b, a, signal)
def calculate_psd rfu(signal, sample_rate, center_freq, bandwidth):
    11 11 11
    This function calculates the Power Spectral Density (PSD) and Relative \Box
 \hookrightarrow Field Unit (RFU) of the signal.
   PSD provides a measure of the signal's power content versus frequency.
    RFU is calculated as the sum of PSD values within a specified frequency \Box
 ⇒range, indicating the signal's strength in that range.
    11 11 11
    f, Pxx = welch(signal, fs=sample_rate, nperseg=1024)
    # Find the frequency range of interest
    freq_range = (f >= center_freq - bandwidth/2) & (f <= center_freq +
 ⇒bandwidth/2)
    # Isolate the PSD values within the frequency range of interest
   Pxx_interest = Pxx[freq_range]
    f interest = f[freq_range]
    # Calculate RFU as the sum of PSD in the frequency range of interest
   rfu = np.sum(Pxx_interest)
    return f_interest, Pxx_interest, rfu
# Function to add white Gaussian noise to a signal
def add_white_noise(signal, noise_level):
    HHHH
    This function adds white Gaussian noise to the signal.
    The noise level is a fraction of the signal's standard deviation, allowing \Box
 ⇔control over the signal-to-noise ratio.
    Adding noise can be useful for testing the robustness of signal processing \Box
 \hookrightarrow algorithms.
    11 11 11
    mean_noise = 0
    std_noise = noise_level * np.std(signal)
    noise = np.random.normal(mean_noise, std_noise, len(signal))
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return signal + noise
```

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[]: # Noise level as a fraction of the signal's standard deviation
     noise_level = 0.3 # Adjust this to increase or decrease noise level
     # Signal parameters
     sampling rate = 32000 # Hz
     duration = 1
                            # seconds
     frequencies = [2000, 2250, 2500, 2750, 3000] # Hz
     amplitude = 2.0 # Amplitude factor
     # Time array
     t = np.arange(0, duration, 1 / sampling_rate)
     signals = [amplitude * np.sin(2 * np.pi * f * t) for f in frequencies]
     # Plotting parameters
     plot_duration = 0.03 # seconds to display in the plot
     samples_to_plot = int(sampling_rate * plot_duration)
     plt.figure(figsize=(12, 8))
     # Plot each frequency component within a smaller time window
     for i, signal in enumerate(signals):
        plt.subplot(len(frequencies), 1, i + 1)
        plt.plot(t[:samples_to_plot], signal[:samples_to_plot])
        plt.title(f"Frequency {frequencies[i]} Hz Component")
        plt.xlabel("Time (s)")
        plt.ylabel("Amplitude")
        plt.tight_layout()
     plt.show()
     # Combining signals
     combined_signal = np.sum(signals, axis=0)
     combined signal = add white noise(combined signal, noise level) # un-comment_
     ⇔this line to inject noise
     # Plotting the combined signal within the same smaller time window
     plt.figure(figsize=(14, 5))
     plt.plot(t[:samples_to_plot], combined_signal[:samples_to_plot])
     plt.title("Combined Complex Waveform (Zoomed In)")
     plt.xlabel("Time (s)")
     plt.ylabel("Amplitude")
     plt.show()
```

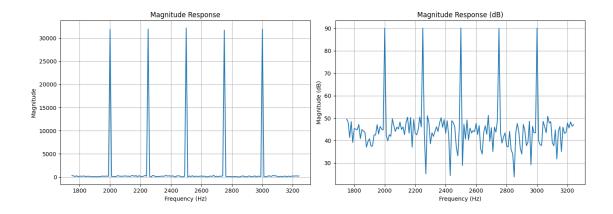




```
[]: # Applying the Goertzel algorithm
n = len(combined_signal)
results = {f: goertzel(combined_signal, sampling_rate, f, n) for f in_u
frequencies}

# Displaying the results
for f, magnitude in results.items():
```

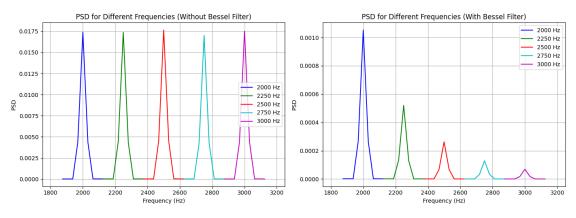
```
print(f"Frequency {f} Hz, Magnitude: {magnitude}")
# Define frequency range for the magnitude response
freq_start = 1750 # Start frequency in Hz
freq_end = 3250  # End frequency in Hz
freq_step = 10
                  # Frequency step in Hz
frequency_range = np.arange(freq_start, freq_end, freq_step)
# Calculate magnitude response with and without windowing
n = len(combined_signal)
magnitudes = magnitude response(combined signal, sampling rate,
 →frequency_range, n)
magnitudes_windowed = magnitude response_windowed(combined_signal,_
 sampling_rate, frequency_range, n)
# Convert magnitudes to decibels
magnitudes_db = 20 * np.log10(magnitudes)
magnitudes_windowed_db = 20 * np.log10(magnitudes_windowed)
# Plotting setup
plt.figure(figsize=(14, 5))
# Linear scale
plt.subplot(1, 2, 1)
plt.plot(frequency_range, magnitudes)
plt.title("Magnitude Response")
plt.xlabel("Frequency (Hz)")
plt.ylabel("Magnitude")
plt.grid(True)
# dB scale
plt.subplot(1, 2, 2)
plt.plot(frequency_range, magnitudes_db)
plt.title("Magnitude Response (dB)")
plt.xlabel("Frequency (Hz)")
plt.ylabel("Magnitude (dB)")
plt.grid(True)
plt.tight_layout()
plt.show()
Frequency 2000 Hz, Magnitude: 31899.737316131977
Frequency 2250 Hz, Magnitude: 31968.882585711617
Frequency 2500 Hz, Magnitude: 32175.856921205406
Frequency 2750 Hz, Magnitude: 31703.625554230355
Frequency 3000 Hz, Magnitude: 31950.3091766614
```



```
[]: # Apply Bessel filter
     cutoff frequency = 0.05 * sampling rate # Normalized cutoff frequency
     filtered_signal = apply_bessel_filter(combined_signal, sampling_rate,__
      ⇔cutoff_frequency)
     # Apply Hamming window to both filtered and original signals
     windowed_signal_filtered = filtered_signal * np.hamming(len(filtered_signal))
     windowed signal original = combined signal * np.hamming(len(combined signal))
     # Frequencies of interest
     frequencies_of_interest = frequencies
     paddinging = 250 # Hz
     # Plotting setup
     plt.figure(figsize=(14, 5))
     # Plot PSD without Bessel filter
     plt.subplot(1, 2, 1)
     for i, center_frequency in enumerate(frequencies_of_interest):
         freqs, psd, rfu = calculate_psd_rfu(windowed_signal_original,_
      ⇒sampling_rate, center_frequency, bandwidth)
         plt.plot(freqs, psd, label=f'{center_frequency} Hz', color=colors[i %__
     →len(colors)])
     plt.title("PSD for Different Frequencies (Without Bessel Filter)")
     plt.xlabel("Frequency (Hz)")
     plt.ylabel("PSD")
     plt.xlim(min(frequencies_of_interest) - paddinging,__

¬max(frequencies_of_interest) + paddinging)
     plt.legend()
     plt.grid(True)
     # Plot PSD with Bessel filter
```

```
plt.subplot(1, 2, 2)
for i, center_frequency in enumerate(frequencies_of_interest):
   freqs, psd, rfu = calculate psd_rfu(windowed_signal_filtered,_
 sampling_rate, center_frequency, bandwidth)
   plt.plot(freqs, psd, label=f'{center_frequency} Hz', color=colors[i %__
 →len(colors)])
plt.title("PSD for Different Frequencies (With Bessel Filter)")
plt.xlabel("Frequency (Hz)")
plt.ylabel("PSD")
plt.xlim(min(frequencies_of_interest) - paddinging,__
 plt.legend()
plt.grid(True)
# Show the plots
plt.tight_layout()
plt.show()
```



```
[]: from scipy.signal import freqz

"""

The plot shows how the filter attenuates frequencies above the cutoff frequency.

The Bessel filter is characterized by a smooth transition from the passband to—

the stopband,

and the plot illustrates this with a gradual roll-off in gain beyond the cutoff—

frequency.

The gain at the cutoff frequency is typically 0.707 (or -3dB), which is a—

standard reference point for filter design.

"""

# Define the Bessel filter parameters
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b, a = bessel(N=4, Wn=cutoff_frequency/(0.5*sampling_rate), btype='low',_
 →analog=False)
# Compute the frequency response
w, h = freqz(b, a, worN=8000)
# Plot the frequency response
plt.figure(figsize=(10, 6))
plt.plot(0.5 * sampling_rate * w / np.pi, np.abs(h), 'b')
plt.plot(cutoff_frequency, 0.5*np.sqrt(2), 'ko')
print(cutoff_frequency)
plt.axvline(cutoff_frequency, color='k')
plt.xlim(0, 0.5 * sampling_rate)
plt.title("Frequency Response of 4th Order Bessel Filter")
plt.xlabel('Frequency [Hz]')
plt.ylabel('Gain')
plt.grid()
plt.show()
```

## 1600.0

