Speech Signal Processing — Exercise 2 — Frequency Domain Speech Analysis

Timo Gerkmann, Robert Rehr

In this exercise session the topic of frequency domain signal analysis is tackled. Here, you will write another function which takes the segmented blocks from the previous exercise as an input argument and uses them to compute a short-time Fourier transform (STFT). Further, you will write a function which reconstructs the time domain signal from an STFT.

Download the file Exercise2.zip from STiNE which contains the signals speech1.wav and phone.wav.

1 Short-time fourier transform

Write a Matlab function which computes the short-time Fourier transform (STFT). Use your my_windowing function from the previous session as a starting point, i.e., copy it an rename it to my_stft.m. Extend the function header of the copied version as follows:

```
[m_stft, v_freq, v_time] = my_stft(v_signal, fs, frame_length, frame_shift,
    v_analysis_window)
```

The input and output parameters are defined as follows:

v_signal	vector containing the time domain signal
fs	sampling rate in Hz
frame_length	frame length in seconds
frame_shift	frame shift in seconds
v_analysis_window	vector containing that contains the spectral analysis window (This vector should have the same length as the frames, i.e., frame_length in samples)
m_stft	a matrix which stores the complex short-time spectra in each column
v_freq	a vector which contains the frequency axis (in units of $Hertz$) corresponding to the computed spectra
v_time	time steps around which a frame is centered (as in previous exercise)

The function needs to be extended by the following steps. They need to be performed for each frame:

- 1. After you extracted a frame from the signal, the analysis window needs to be applied.
- 2. Use the built-in Matlab function fft to calculate the DFT.
- 3. Only keep the lower half of the spectrum and remove the upper half. Make sure that the frequency bin at the Nyquist frequency is still included.
 - Why are the computed spectra complex conjugate symmetric?
 - What may be the advantage of only considering one half of the spectrum?

- How can you compute the frequency for each spectral bin? How many sampling points does the spectrum have after you removed the mirrored part while including the Nyquist frequency bin?
- 4. Store the transformed frames in the columns of the output matrix m_stft.

2 Spectral analysis

If not stated otherwise, the following exercises should be performed for both signals.

- a) Use your own function to compute the STFT and plot the logarithmic magnitude spectrogram in dB using the following parameters.
 - frame length: 32 ms
 - frame shift: 8 ms
 - window function: periodic Hann window

You can create the Hann window with v_analysis_window = hann(frame_length_samples, 'periodic'), where frame_length_samples is the frame length in samples. The spectrogram can be plotted using the following Matlab command

```
imagesc(v_time, v_freq, 10*log10(max(abs(m_stft).^2, 10^(-15)))
```

It tells Matlab to use the vector v_time for the x-axis and v_freq for the y-axis. Here, the vector v_time contains the time instants for each block / each spectrum. The imagesc command uses by default a reversed y-axis. The axis can be inverted by using axis xy after the imagesc call.

- Why is the magnitude plotted in dB? Why is it reasonable to introduce a lower limit? What is the lower limit in the command given above in dB?
- b) Identify the voiced, unvoiced and silence segments in the spectrogram by eye.
 - Describe their appearance and what distinguishes them. Is it possible to identify the different voicing types more easily in comparison to the time domain representation?
- c) Produce the same plot as in a) but this time using a frame length corresponding to 8 ms and a frame shift of 2 ms. Further, create a plot for a frame length of 128 ms and a frame shift of 32 ms.
 - How well can you distinguish single sinusoidal components? Short impulses? Explain the influence of the different parameter settings.
- d) Only for the speech signal estimate the fundamental frequency using the auto-correlation-based method of the last exercise session. Plot the estimated fundamental frequency onto the spectrogram. The parameter setting should be the one used in a). This can be achieved by issuing the following commands after the imagesc command:

```
hold on
plot(v_time, v_fundamental_frequency, 'k')
hold off
```

• Do the estimated fundamental frequencies follow the harmonic structures in the spectrogram? You may also want to plot higher harmonics by multiplying your estimated fundamental frequencies with a positive integer value. This way, you can see the precision of the estimated frequencies more precisely.

3 Synthesis from the STFT domain (Inverse STFT)

Use the provided Matlab function my_inverse_stft to synthesize a time-domain signal using overlap-add. The function has the following header:

```
v_signal = my_inverse_stft(m_stft, fs, frame_length, frame_shift, v_synthesis_window)
```

The input and output parameters have the following meaning:

m_stft	matrix containing the STFT spectra which have been generated using the function from Section 1.
fs	the sampling rate in Hz
frame_length	frame length in seconds that was used for separating the signal into blocks
frame_shift	frame shift in seconds used for the frames in ${\tt m_stft}$
v_synthesis_window	vector containing a synthesis window function
v_signal	vector which contains the synthesized time domain signal

Check if the provided function works correctly with the following signal v_test_signal = ones(2048, 1). Assume the sampling rate is 16 kHz.

- 1. Generate the STFT of v_test_signal using your function from Section 1 with a frame length of 32 ms and a frame shift of 16 ms. Employ a √Hann-window as analysis window. This window can be generated via sqrt(hann(frame_length_samples, 'periodic')).
- 2. Resynthesize the signal using your new function my_inverse_stft. Use the periodic √Hann-window also as synthesis window. For frame_length and frame_shift, use the same values that you employed for generating the STFT. Finally, plot the synthesized signal.
 - Is it possible to perfectly reconstruct the input signal? Are there parts where a perfect reconstruction is not possible when a $\sqrt{\text{Hann}}$ -window is used as analysis and synthesis window?
 - What happens, when you omit the parameter 'periodic' in the window generation? Which error can you observe in the reconstructed signal? Explain the difference in the window function which causes this behavior.