

## University of Washington – ECE Department

### EE235 Lab 5 – Time Domain to Frequency Domain

### Background Material

In this lab, we will learn how to transform signals from the time domain to the frequency domain and identify the frequencies of  $e^{j\omega t}$  that comprise a periodic signal. The concepts that we'll focus on are finding the Fourier Series coefficients  $a_k$  of  $x(t)$ , identifying frequency components of  $x(t)$ , and understanding the relationship between Fourier Series coefficient index  $k$  and frequency  $\omega$ .

Concepts TO REVIEW	Concepts You Will Learn
<ul style="list-style-type: none"> <li>• Generating a signal in time domain</li> <li>• Playing audio signal</li> <li>• Subplots – usage, labeling, titles</li> <li>• Extracting a list of elements in a vector</li> <li>• Vector concatenation</li> <li>• Loops, functions</li> </ul>	<ul style="list-style-type: none"> <li>• Using <code>fft</code> and <code>fftshift</code> functions</li> <li>• Finding magnitude and phase</li> <li>• A few other useful functions</li> </ul>

## 1.0 Frequency Analysis on a Computer

In this lab, because we are working on a computer, the signals are discrete-time signals with a finite length. As a result, the Fourier transform of those signals are discrete and finite in length. That is, we actually use a discrete Fourier transform (DFT), which differs from the continuous-time Fourier transform that you in class in two ways: i) the frequency domain is discrete, and ii) the frequency domain is over the window  $[-fs/2, fs/2)$  where  $fs$  is the sampling frequency (in Hz).

In order to compute the DFT, we use the Fast Fourier Transform (FFT) using `numpy` in Python (i.e., `import numpy as np`).

```

xf = np.fft.fft(x,nfft)      # FFT of time signal x nfft frequency samples
xfs = np.fft.fftshift(xf)    # shift FFT of xf to be centered around 0
yt = np.fft.ifft(yf,nt)      # inverse FFT of yf, optional nt specifies length of yt

```

Function `numpy.fft.fft()` returns a vector whose frequency samples correspond to the range  $[0, fs)$ . In order to make the frequency range be  $[fs/2, fs)$ , we use function `numpy.fft.fftshift`. If your DFT is length  $N$  and the range is  $[0, fs)$ , then the interval between frequency samples corresponds to  $\Delta f = fs/N$ . So, if you are interested in  $X(f)$ , then you need to use the index that is the nearest integer to  $f/\Delta f$ . In this lab, we plot the spectrum of  $x(t)$ , which is positive. Hence, you take the absolute value of the Fourier transformed signal using function `numpy.abs()`.

## 2.0 Other Needed Functions

(1) Generating sinusoidal signals

<code>numpy.pi</code>	<code># the value of <math>\pi</math></code>
<code>numpy.sin(x)</code>	<code># sin(x), where x could be a single time or an array of time samples</code>
<code>numpy.cos(x)</code>	<code># cos(x), where x could be a single time or an array of time samples</code>

Here is a sample code to generate a sinusoidal signal with frequencies 941Hz and 1336Hz. 'numpy' is imported as 'np'.

```
fs=8000
t=np.arange(0,0.25,1/fs)
d0=np.sin(2*np.pi*941*t)+np.sin(2*np.pi*1336*t)
```

(2) Finding the index whose element is larger than a threshold

Consider an array `a = [3, 4, 1, -6, 6]`. We want to make an array 'b' that contains indices of the elements of the array `a`, which is larger than 3. Then, we expect that `b = [1, 4]` because '4' and '6' are larger than 3 and their indices are 1 and 4, respectively. Here is a sample code.

```
threshold = 3
a = np.array([3, 4, 1, -6, 6])
ind = np.arange(len(a))
b = ind[a[ind] > threshold]
```

When finding the frequency component of a signal from the spectrum, we find the indices of elements in an array that represents the spectrum, which has the value larger than a predetermined threshold. (i.e., arrays 'a', 'b', and threshold denote the array representing a spectrum, array containing the indices, and predetermined threshold.)