

**Question #1:** (1 pts) How many hours did you spend on this homework?

**Question #2:** (7 pts) *The Discrete Fourier Transform over Time*

The DFT or DTFT alone is not particularly useful for describing many complex signals. For example, music contains many frequencies that all occur at different times. The DFT / DTFT of an entire piece of music shows all of those frequencies. As an example, consider the signal

$$x[n] = \cos\left(\frac{2\pi n^2}{10000}\right).$$

This is known as a linear frequency-modulated “chirp” signal whose instantaneous frequency  $\omega[n]$  changes linearly with time. Chirp signals are commonly used in RADAR processing to extract time-varying time-shifts (distance from a target) and frequency/doppler-shifts (velocity of a target). In this context, the instantaneous frequency  $\omega[n]$  is defined by

$$x[n] = \cos((\omega[n]/2)n).$$

More specifically, the instantaneous frequency  $\omega[n]$  is the time-derivative of the expression inside the cosine, also known as the instantaneous phase.

- (a) Determine the instantaneous frequency of  $x[n]$  for  $n = 0$ ,  $n = 1000$ , and  $n = 2500$ . Express the frequencies as a function of  $\pi$  (e.g.,  $\pi/100$ ).
- (b) Plot  $x[n]$  with a length of 2501. Label the horizontal axis “Samples” and vertical axis “Amplitude.” Also, use the `fft` function to compute the DFT of  $x[n]$  (i.e.,  $X[k]$ ). Use the `abs` function in MATLAB to plot the magnitude response  $|X[k]|$ . Label your horizontal axis “Normalized frequency [rad / s]” with values  $2\pi k/N$ . Label your vertical axis “Magnitude.” After plotting, apply the command `axis([0 2*pi 0 max(abs(X))])` to see the figure from 0 to  $2\pi$  in frequency and 0 to the maximum value in magnitude.

Since the DFT / DTFT is poor for analyzing such signals, we may instead analyze signal **segments**. Yet, this can affect the precision of our results. To demonstrate this, consider the cosine

$$y[n] = \cos((\pi/2)n)$$

and a rectangular window defined by

$$w[n] = u[n] - u[n - N]$$

- (c) Compute and sketch the magnitude and phase of the DTFT of  $y[n]$ .
- (d) Compute and sketch the magnitude and phase of the DTFT  $Z(\omega)$  of a segment of the cosine signal, defined by  $z[n] = y[n]w[n]$  for  $N = 10$ .

- (e) **(EEE 5502 Only)** Show that the “bandwidth” of the  $W(\omega)$  (defined as the length between the two locations nearest to  $\omega = 0$  where  $W(\omega) = 0$ ) is inversely proportional to the width of the rectangular window  $N$ .
- (f) **(EEE 5502 Only)** This result implies that we cannot simultaneously analyze the frequency content of a signal with perfect temporal resolution and perfect frequency resolution. We often refer to this as a time-frequency uncertainty principle. Explain how the previous result shows this.

**Question #3:** (6 pts) *The Discrete Fourier Transform over Time*

As we illustrated in the previous problem, the DFT is not particularly useful for describing chirp signals. In this problem, we will compute the DFT over time (i.e., a short-time Fourier transform) to better describe the chirp’s behavior.

- (a) Code the following short-time Fourier Transform process for your chirp  $x[n]$ :
- 1) Compute  $M = \text{floor}(N/W)$ , the number of length- $W$  segments in  $x[n]$  (where  $N = 2501$  is the length of the chirp).
  - 2) Initialize a matrix  $\text{STFT} = \text{zeros}(W, M);$ .
  - 3) Extract the first  $W$  samples (samples 0 to  $W - 1$ ) of the signal
  - 4) Compute the DFT (using the `fft` function) of these  $W$  samples (note: this is equivalent to computing the windowed signal segment)
  - 5) Store the result of the DFT in  $\text{STFT}(:, m)$  where  $m = 1$ .
  - 6) Iteratively repeat steps # 3 to # 5 for the next  $W$  samples (i.e., samples  $W$  to  $2W - 1$  and then samples  $2W$  to  $3W - 1$  ... until you reach samples  $(M - 1)W$  to  $MW - 1$ ). Increase  $m$  with each iteration.

Use the MATLAB code

```
imagesc(0:(M-1), 2*pi*(0:(W-1))/W, abs(STFT))
xlabel('Time [samples]');
ylabel('Normalized Frequency [rad/s]');
```

to plot the short-time Fourier transform magnitude values for a  $W$  of your choosing.

- (b) Plot the short-time Fourier transform values for 8 different values of  $W$ . Specifically, consider  $W$  equal to 10, 20, 40, 80, 160, 320, 640, and 1280.
- (c) Your plots should illustrate the time-frequency uncertainty principle. For which value of  $W$  do you think best illustrates the chirp signal?

**Question #4:** (7 pts) *Audio DFT over Time*

Load the audio .mp4 file `rudenko_01.mp4` into MATLAB using

```
[x, Fs] = audioread('rudenko_01.mp4');
```

- (a) Compute the short-time Fourier transform of your Rudenko music `x`. Use the MATLAB code

```
imagesc((0:(M-1))*W/Fs, 2*pi/W*(0:(W-1)), abs(STFT))  
xlabel('Time [seconds]')  
ylabel('Normalized Frequency [rad/s]');
```

to plot the short-time Fourier transform magnitude values for  $W = 10,000$ . Use the `axis` command from Question #2 to focus only on the lower frequencies from 0 to  $\pi/20$ . Note that the code above is plotting time in seconds since we know the sampling rate.

- (b) Use slightly different MATLAB code

```
imagesc((0:(M-1))*W/Fs, 2*pi/W*(0:(W-1)), 10*log10(abs(STFT)./  
    max(max(abs(STFT))), [-20 -5])  
xlabel('Time [seconds]')  
ylabel('Normalized Frequency [rad/s]');
```

to plot the short-time Fourier transform magnitude values for the same  $W$ . This normalizes the data so that the maximum value is 1 (0 dB) and plots the values in the decibel range of -20 dB to -5 dB to make the data is easier to visualize. Again, use the `axis` command from Question #2 to focus only on the lower frequencies from 0 to  $\pi/20$ .

- (c) Just as in coding assignment #3, create a vector corresponding to the impulse response of a 10000-point running average filter

$$h[n] = \frac{1}{10000} \sum_{k=0}^{10000-1} \delta[n-k].$$

Use `conv` or `fft` to filter  $x[n]$  and get an output  $y[n]$ . Then compute the short-time Fourier transform of the filtered signal. Plot the short-time Fourier transform magnitude values for the same  $W$  you chose in part (b).

- (d) How does the filter affect the short-time Fourier transform? How does the filter affect the music?