# EECS 16A Designing Information Devices and Systems I Homework 6A

# This homework is due Wednesday, August 5, 2020, at 23:59. Self-grades are due Sunday August 9, 2020, at 23:59.

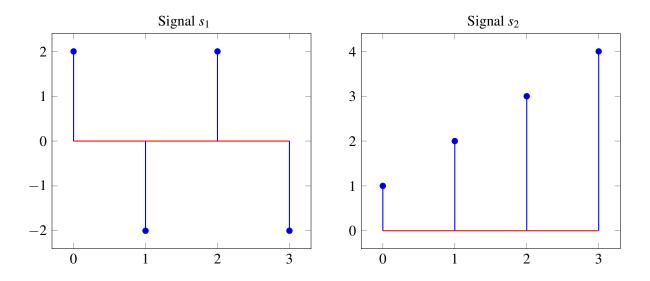
#### **Submission Format**

Your homework submission should consist of **one** file.

• hw6A.pdf: A single PDF file that contains all of your answers (any handwritten answers should be scanned) as well as your IPython notebook (if any) saved as a PDF.

Homework Learning Goals: The objectives of this homework is to familiarize you with the concept of cross-correlation.

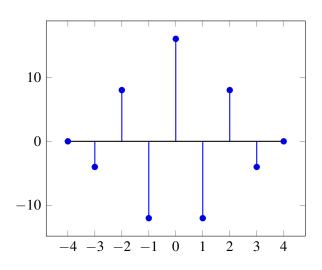
#### 1. Mechanical Linear Correlation



Assume that both of the above signals extend to  $\pm\infty$ , and are 0 everywhere outside of the region shown in the above graphs. First, we will demonstrate the procedure for linear correlation by computing the linear correlation between signal  $s_1$  with itself (i.e.  $\operatorname{corr}_{\vec{s}_1}(\vec{s}_1)[k]$ ). This is referred to as the linear autocorrelation. This can be computed by evaluating the inner product between the signal and the shifted version of the signal (outlined in the below tables). Here, we compute this quantity for shifts between -3 and 3. For all shifts outside this range, the inner product is zero. Finally, we plot the non-zero values of the linear autocorrelation.

$\vec{s}_1[n]$	0		0		0		2		-2		2		-2		0		0		0	
$\vec{s}_1[n+3]$	2		-2		2		-2		0		0		0		0		0		0	
$\langle \vec{s}_1[n], \vec{s}_1[n+3] \rangle$	0	+	0	+	0	+	-4	+	0	+	0	+	0	+	0	+	0	+	0	= -4

$\vec{s}_1[n]$	0		0		0		2		-2		2		-2		0		0		0	
$\vec{s}_1[n+2]$	0		2		-2		2		-2		0		0		0		0		0	
$\langle \vec{s}_1[n], \vec{s}_1[n+2] \rangle$	0	+	0	+	0	+	4	+	4	+	0	+	0	+	0	+	0	+	0	= 8
	'																			
$\vec{s}_1[n]$	0		0		0		2		-2		2		-2		0		0		0	
$\vec{s}_1[n+1]$	0		0		2		-2		2		-2		0		0		0		0	
$\langle \vec{s}_1[n], \vec{s}_1[n+1] \rangle$	0	+	0	+	0	+	-4	+	-4	+	-4	+	0	+	. 0	+	0	+	0	= -12
( - 2 3 / - 2 3 /	1																			
$\vec{s}_1[n]$	0		0		0		2		-2		2		-2		0		0		0	
$\frac{\vec{s}_1[n]}{\vec{s}_1[n+0]}$	0		0		0		$\frac{2}{2}$		-2		2		-2		0		0		0	
$\frac{\vec{s}_1[n+0]}{\langle \vec{s}_1[n], \vec{s}_1[n+0]\rangle}$	0	+	0	+	0	+	4	+	4	+	4	+	4	+	0	+		+	0	= 16
$\langle s_1[n], s_1[n+0] \rangle$	0	Т	U	т	U	т	7	Т	7	Т	7	Т	7	Т	U	Т	U	Т	U	- 10
$\vec{s}_1[n]$	0		0		0		2		-2		2		-2		0		0		0	
$\vec{s}_1[n-1]$	0		0		0		0		2		-2		2		-2		0		0	
$\overline{\langle \vec{s}_1[n], \vec{s}_1[n-1] \rangle}$	0	+	0	+	0	+	0	+	-4	+	-4	+	-4	+	0	+	0	+	0	= -12
$\vec{s}_1[n]$	0		0		0		2		-2		2		-2		0		0		0	
$\vec{s}_1[n-2]$	0		0		0		0		0		2		-2		2		-2		0	
$\langle \vec{s}_1[n], \vec{s}_1[n-2] \rangle$	0	+	0	+	0	+	0	+	0	+	4	+	4	+	0	+	0	+	0	= 8
(11)/11/1/	I																			
₹.[n]	0		0		Λ		2		2		2		2		Λ		Λ		Λ	
$\vec{s}_1[n]$					0		2		-2		2		-2		0		0		0	
$\vec{s}_1[n-3]$	0		0		0		0		0		0		2		-2		2		-2	
$\overline{\langle \vec{s}_1[n], \vec{s}_1[n-3] \rangle}$	0	+	0	+	0	+	0	+	0	+	0	+	-4	+	0	+	0	+	0	= -4



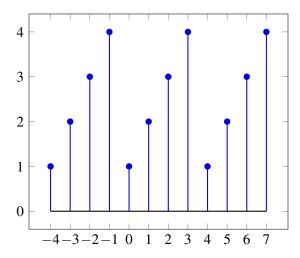
(a) Using the procedure demonstrated above, compute  $\operatorname{corr}_{\vec{s}_1}(\vec{s}_2)[k]$ , the linear cross-correlation of  $s_2$  with  $s_1$ . Like the example, use tables like the one given below for k=-3 and plot the resulting correlation.

$\vec{s}_1[n]$	0	0	0	2	-2	2	-2	0	0	0	
$\vec{s}_2[n+3]$	1	2	3	4	0	0	0	0	0	0	
$\overline{\langle \vec{s}_1[n], \vec{s}_2[n+3] \rangle}$											

(b) Will the linear cross-correlation of  $s_2$  with  $s_1$  ( $\operatorname{corr}_{\vec{s}_1}(\vec{s}_2)[k]$ ) be the same as the cross-correlation of  $s_1$  with  $s_2$  ( $\operatorname{corr}_{\vec{s}_2}(\vec{s}_1)[k]$ )? You can use the iPython notebook to figure this out. How are they related to each other?

Now, we will review the procedure to perform linear cross-correlation between one signal that is periodic with a period of 4 and another that is finite length and extended with zeros as in the previous parts. As an example, we will compute the linear correlation  $\operatorname{corr}_{\vec{p}_2}(\vec{s}_1)[k]$  between the periodic signal  $\vec{p}_2$  (with period 4), formed by repeating  $\vec{s}_2$ , and the finite length signal  $\vec{s}_1$  extended with zeros. The result will be a periodic signal with period 4.

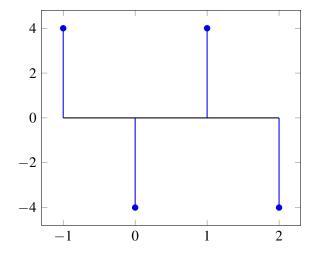
The periodic signal,  $\vec{p}_2$ , formed by repeating  $\vec{s}_2$  is plotted below for indices -4 to 7. It is defined and non-zero for all indices from  $-\infty$  to  $+\infty$ .



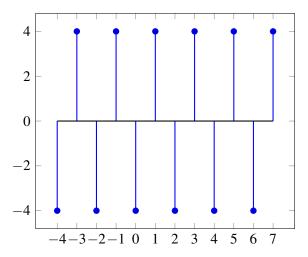
We compute one period of the result of the cross-correlation by starting at a shift of k = -1 and ending at a shift of k = 2.

$\vec{p}_2[n]$	2		3		4		1		2		3		4		1		2		3	
$\vec{s}_1[n+1]$	0		0		2		-2		2		-2		0		0		0		0	
$\langle \vec{p}_2[n], \vec{s}_1[n+1] \rangle$	0	+	0	+	8	+	-2	+	4	+	-6	+	0	+	0	+	0	+	0	= 4
$ec{p}_2[n]$	2		3		4		1		2		3		4		1		2		3	
$\vec{s}_1[n+0]$	0		0		0		2		-2		2		-2		0		0		0	
$\langle \vec{p}_2[n], \vec{s}_1[n+0] \rangle$	0	+	0	+	0	+	2	+	-4	+	6	+	-8	+	0	+	0	+	0	= -4
$ec{p}_2[n]$	2		3		4		1		2		3		4		1		2		3	
							1		_		5		•		-		_		9	
$\vec{s}_1[n-1]$	0		0		0		0		2		-2		2		-2		$\frac{2}{0}$		0	
	0	+	0	+		+		+		+		+		+	-2 -2	+		+		= 4
$\vec{s}_1[n-1]$	_	+		+	0	+	0	+	2	+	-2	+	2	+		+	0	+	0	= 4
$\frac{\vec{s}_1[n-1]}{\langle \vec{p}_2[n], \vec{s}_1[n-1]\rangle}$	0	+	0	+	0	+	0	+	2 4	+	-2 -6	+	8	+		+	0	+	0	= 4

The computed single period of the resulting linear cross correlation is plotted below.



The resulting linear cross correlation for shifts from k = -4 to k = 7 is plotted below.



(c) Repeat the procedure described above to compute the correlation  $\operatorname{corr}_{\vec{p}_1}(\vec{s}_1)[k]$  between a periodic signal  $\vec{p}_1$  (with period 4), formed by repeating  $s_1$ , and the finite-length signal  $s_1$  extended with zeros. Like the example, evaluate tables like the one below for k=-3 for different shifts and plot a single period of the result.

$ec{p}_1[n]$	-2	2	-2	2	-2	2	-2	2	-2	2	
$\vec{s}_1[n+3]$	2	-2	2	-2	0	0	0	0	0	0	
$\overline{\langle \vec{p}_1[n], \vec{s}_1[n+3]\rangle}$											

## 2. Audio File Matching

Lots of different quantities we interact with every day can be expressed as vectors. For example, an audio clip can be thought of as a vector. Each element of the vector might be a sound pressure value or voltage recorded by a microphone, while the index of the element indicates the time at which this value was recorded. This simple series of numbers can completely capture the sound played by a speaker that we hear with our ears.

Many audio processing application rely on representing audio files as vectors. In this problem we explore how we could possibly use the idea of inner products to build an application like *Shazam*.

Audio files can naturally be represented as vectors. Every component of the vector determines the sound we hear at a given time. We'll also refer to these vectors as the audio *signal*. We will use inner products to determine if a particular audio clip was part of a longer song. The ideas here are similar to the themes of the Acoustic Positioning System in the lab.

For this problem we make the *simplification* that the magnitude of each vector determines the volume and the angle of each vector captures the tune.

Let us consider a very simplified model for an audio signal. At each time, the audio signal can be either -1 or +1. A vector of length N makes up the audio file.

- (a) Say we want to compare two audio files of the same length N to decide how similar they are. First, consider two vectors that are exactly identical, namely  $\vec{x}_1 = \begin{bmatrix} 1 & 1 & \cdots & 1 \end{bmatrix}^T$  and  $\vec{x}_2 = \begin{bmatrix} 1 & 1 & \cdots & 1 \end{bmatrix}^T$ . What is the inner product of these two vectors? What if  $\vec{x}_1 = \begin{bmatrix} 1 & 1 & \cdots & 1 \end{bmatrix}^T$  but  $\vec{x}_2$  oscillates between 1 and -1? Assume that N, the length of the two vectors, is an even number.
  - Use this to suggest a method for comparing the similarity between a generic pair of length-N vectors.
- (b) Next, suppose we want to find a short audio clip in a longer one. We might want to do this for an application like *Shazam*, which is able to identify a song from a short clip. Consider the vector of length 8,  $\vec{x} = \begin{bmatrix} -1 & 1 & 1 & -1 & 1 & -1 & 1 \end{bmatrix}^T$ . Let us label the elements of  $\vec{x}$  so that

$$\vec{x} := \begin{bmatrix} x[0] & x[1] & x[2] & x[3] & x[4] & x[5] & x[6] & x[7] \end{bmatrix}^T$$

We want to find the short segment  $\vec{y} := \begin{bmatrix} y[0] & y[1] & y[2] \end{bmatrix}^T = \begin{bmatrix} 1 & 1 & -1 \end{bmatrix}^T$  in the longer vector. To do this, perform the linear cross correlation between these two finite length sequences and identify at what shift(s) the linear cross correlation is maximized. Apply the same technique to identify what shift(s) gives the best match for  $\vec{y} = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}^T$ .

(If you wish, you may use iPython to do this part of the question, but you do not have to.)

- (c) Now suppose our audio vector is represented using integers beyond simply just 1 and -1. Find the short audio clip  $\vec{y} = \begin{bmatrix} 1 & 2 & 3 \end{bmatrix}^T$  in the song given by  $\vec{x} = \begin{bmatrix} 1 & 2 & 3 & 1 & 2 & 2 & 3 & 10 \end{bmatrix}^T$ . Where do you expect to see the peak in the correlation of the two signals? Is the peak where you want it to be, i.e. does it pull out the clip of the song that you intended? Why?
  - (If you wish, you may use iPython to do this part of the question, but you do not have to.)
- (d) (**Optional:**) Let us think about how to get around the issue in the previous part. Cross-correlation compares segments of  $\vec{x}$  of length 3 (which is the length of  $\vec{y}$ ) with  $\vec{y}$ . Instead of directly taking the cross correlation, we want to normalize each inner product computed at each shift by the magnitudes of both segments, i.e. we want to consider  $\frac{\langle \vec{x}_k, \vec{y} \rangle}{\|\vec{x}_k\| \|\vec{y}\|}$ , where  $\vec{x}_k$  is the length 3 segment starting from the k-th index of  $\vec{x}$ . This is referred to as normalized cross correlation. Using this procedure, now which segment matches the short audio clip best?
- (e) (**Optional:**) We can use this on a more 'realistic' audio signal see part 2e in the IPython notebook, where we use normalized correlation on a real song. Run the cells to listen to the song we are searching through, and add a simple comparison function vector\_compare to find where in the song the clip comes from. Running this may take a couple minutes on your machine, but note that this computation can be highly optimized and run super fast in the real world! Also note that this is not exactly how Shazam works, but it draws heavily on some of these basic ideas.

## 3. Homework Process and Study Group

Who else did you work with on this homework? List names and student ID's. (In case of homework party, you can also just describe the group.) How did you work on this homework?