SOLUTIONS

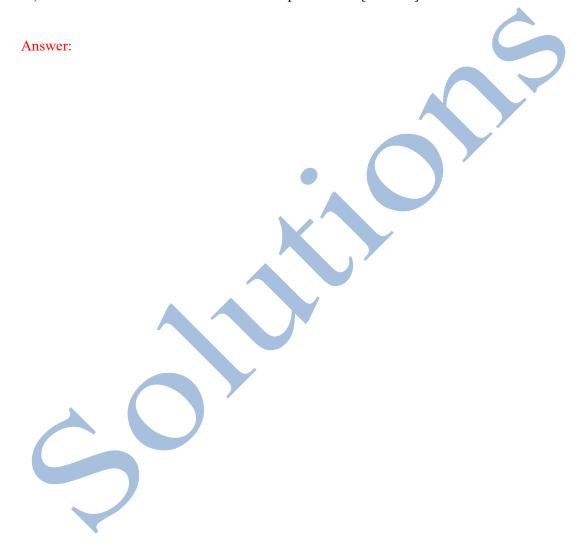
Module:	Advanced Transform Methods					
Module Code	EBU6018	Paper	Α			
Time allowed	2hrs	Filename	Solutions_	201920	EBU6018	_A
Rubric	ANSWER ALL FOUR QUESTIONS					
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EBU6018 Solutions A

Question 1.

- (a) The matrix A is: $A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$
- i) Find an orthogonal basis of the NULL space of A. [10 marks]
- ii) Find the RANK of A. [2 marks]
- iii) Find an orthonormal basis of the ROW space of A. [5 marks]



(i) NULL SPACE CONSISTS OF THE SOLUTIONS OF Ax=0 [1 MARK]

LET X = X T [1 MARK]

THEN O O D T X = O T [MARK]

GIVING $X_1 = -X_3$ $X_2 = 0$ $X_3 = 0$ $X_4 = 0$

THEN X = [-X3] = X3[0] [IMARK]

SO DC = 0 IS A BASIS FOR NULL SPACE OF 1

[1 MARK] FOR DRING NORMANY, FIND NORM OR DE [1 MARK)

11x11= Ja [1 MARN]

SO ORTHONORMA BASIS REQUIRED IS 1 [MARK)

A HAS 2 NON-ZERO ROWS [I MARK] (ii) SO LAND IS 2 [I MAIN]

FOR ROW SPACE OF A, NON-2500 ROWS ARE A BASIS [1 MARK]

SO BASIS IS {[6], [6]} [1 MARY]

DOT PRODUCT IS O, ! ORTHOGONA [1 MARK]

NORMS ARE JZ AND I RESPECTIVELY [1 MARK]

SO ORTHONORMAL BASIS FOR ROW SPACE OR A 15:

S 5 9 , [677 [1 marn]

- (b) An FFT is a fast algorithm for implementing a DFT.
- i) Estimate the approximate number of computations that are required to perform the FFT of an 8-point sequence. [2 marks]

ii) One FFT structure is radix-2 decimation-in-time. Illustrate this FFT structure using the following 8-point sequence:

$$S[n] = [2, 6, 3, 9, 7, 4, 1, 11]$$

[6 marks]

Answer: i) number of computations is $Nlog_2N$ [1 mark] So no of computations = 8x3 = 24. [1 mark]

S[n] = [7, 3, -5, 2, 6, 4, -1, 8]

STEP 1: [7, -5, 6, -1][3, 2, 4, 8]

STEP 2: [7, 6][-5, -1][3, 4][2, 8]

STEP 3: [7][6][-5][-1][3][4][2][8]

[6 marks: 2 for each step]

Question 2.

(a) With the aid of a suitable diagram, briefly explain the application of Linear Transform Coding (LTC) to the processing of images. [10 marks]

Answer:

(a)

Divide image (or signal) into P blocks of N pixels (samples). [1 mark] The kth block is now an N-dimensional vector: [1 mark]

$$\mathbf{x}_{k} = (x_{1,k}, x_{2,k}, ..., x_{N,k})^{T}$$

The image (signal) is now a sequence of vectors $\{\mathbf{x}_k\}$.[1mark] We now transform each \mathbf{x} by multiplying by a linear matrix

$$y = Ax[1mark]$$

transmit the first M coefficients $\hat{\mathbf{y}} = (y_1, ..., y_M)^T$, [1 mark] discarding the remaining N - M coeffs $y_{M+1} \cdots y_N[1 \text{mark}]$ We then reconstruct the image block using another matrix

$$\hat{\mathbf{x}} = \mathbf{B}\hat{\mathbf{y}}[1mark]$$



[3 marks for diagram, 1 for each block]

- (b) i) Briefly explain the use of the Karhunen-Loeve Transform (KLT) in the compression of images. [8 marks]
- iii) List and state the advantage and THREE (3) disadvantages of the KLT for the compression of images, and THREE (3) advantages of using the Discrete Cosine Transform (DCT). [7 marks]

Answer:

i) Uses Principal Component Analysis (PCA) Multivariate statistics. [1 mark]

Finds a projection of the observations onto orthogonal axes contained in the space defined by the original variables. [1 mark]

Correlated variables transformed into uncorrelated variables. [1 mark]

Ordered by reducing variability. [1 mark]

Computes compact, optimal description of data set. [1 mark]

Rotates data so that maximum variabilities projected onto the axes. [1 mark]

Rotates existing axes to new positions in the space defined by the original variables. [1 mark]

Uses PCA is to reduce dimensionality of a data set while retaining as much information as is possible. [1mark]

ii)

Advantage: KLT maximises the coding gain, i.e. maximises the SNR after a given level of compression. [1 mark]

Disadvantages: [3 marks, 1 mark for any three of the following list]

- Estimate of correlation can be unwieldy
- Solution of eigenvector decomposition is computationally intensive (i.e. slow)
- Calculation of forward and inverse transforms is O(MN) for each image block
- Transmission of data-dependent basis A is required
- The technique is linear, therefore any non-linear correlation between variables will not be captured.

Comparison: [3 marks: 1 mark for any three of the following list]

DCT has fixed basis functions. a good approximation to KLT for typical images. needs no eigenvalue decomposition. transform is $O(N \log N)$.

Question 3.

(a) Compare Short-time Fourier Transform (STFT) ($\gamma(t)$) and Continuous Wavelet Transform (CWT) ($\psi(t)$) in terms of

- i) similarities and
- ii) differences. [5 marks]

Answer:

- i) Similarities [3 marks, 1 mark each]
- Signal is multiplied by a function, and the transform is computed separately for different segments of signals. [1mark]
- Both can be written in inner product form

$$STFT(b,\omega) = \left\langle s(t), \gamma(t-b)e^{j\omega t} \right\rangle$$
 $CWT(b,a) = \left\langle s(t), \frac{1}{\sqrt{a}}\psi\left(\frac{t-b}{a}\right) \right\rangle$ [Imark

- Time-frequency window area remains constant [1mark]
- ii) Differences [2 marks, 1 mark each]
- Fixed time duration and freq bandwidths of $\gamma(t)$ [1 mark]
- Variable time duration and bandwidth of $\psi(t)$ [1 mark]
- b) Describe the problems with Continuous Wavelet Transform (CWT) [5 marks]

Answer:

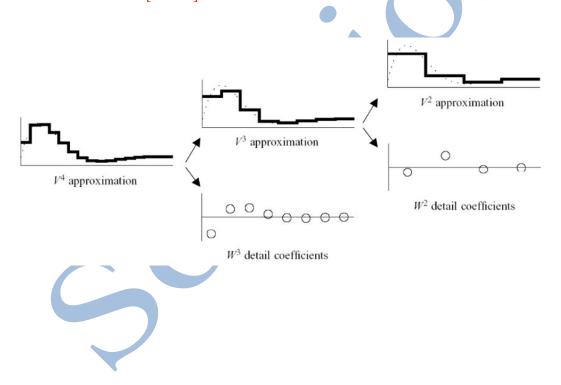
Described 1 problem: [1 mark] Described 2 problems: [3 marks] Described 3 problems: [5 marks]

- 1. Redundancy
- : Basis functions for CWT are shifted and scaled versions off each other. Cannot form a very orthonormal base.
- 2. Infinite solution space
- : The result holds an infinite number of wavelets: hard to solve and hard to find the desired results out of the transformed data.
- 3. Efficiency
- : Most transforms cannot be solved analytically. Solutions must be calculated numerically: time-consuming.

- (c) Describe multi resolution analysis in terms of its
 - i) objective
 - ii) concept with a diagram describing piecewise approximation of a signal. [5 marks]

Answer:

- i) objective [2 marks]
- : To analyze a complicated function by dividing it into several simpler ones and studying them separately.
- ii) concept with a diagram describing piecewise approximation of a signal [3marks]
- [1 mark] for the description [2 marks] for the diagram
- Decompose a fine-resolution signal into a coarse-resolution version of the signal and the differences left over. [1 mark]



d) Apply the transform defined by

$$x_{n-1,i} = (x_{n,2i} + x_{n,2i+1})/2$$

 $d_{n-1,i} = (x_{n,2i} - x_{n,2i+1})/2$

to the sequence

$$[x_{n,i}] = [3,4,2,3,4,2,2,3],$$

where i = 0.....7, is the index position in the sequence, and n is the level. The next level is n - 1. At each level, calculate the sequences for $x_{n-1,i}$ and $d_{n-1,i}$. Continue till no further levels are possible.

- i) State the significance of the first element in the final level
- ii) Has any information lost in the process?
- iii) Describe how this process could be used to compress the data. [10 marks]

Answer:

Applying the transform:

n=3 [3.0, 4.0, 2.0, 3.0, 4.0, 2.0, 2.0, 3.0],

n=2 [3.5, 2.5, 3.0, 2.5, -0.5, -0.5, 1.0, -0.5]

n=1 [3.0, 2.75, 0.5, 0.25, -0.5, -0.5, 1.0, -0.5]

n=0 [2.875, 0.125, 0.5, 0.25, -0.5, -0.5, 1.0, -0.5]

[6 marks: 2 for each row]

- i) The first element is the average of all the elements in the original sequence [1 mark].
- ii) No information has been lost [1 mark]
- iii) Because most of the values in the final level are small, potentially fewer bits would be required to store it [1 mark]. Where there are zeroes, they do not need to be stored, although the positions of the other values would need to be stored [1 mark]. Small values could be replaced by zeroes without significant loss of detail [1 mark], this can be done by applying a threshold value, the bigger the threshold the greater the loss of detail [1 mark].

Question 4.

(a) A Haar wavelet transform is implemented using an analysis filterbank using normalised low-pass and high-pass filters:

$$h_0 = \left[\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right], \quad h_1 = \left[\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right]$$

Calculate the Haar Transform for 2 levels of decomposition for the input sequence: s[n] = [2, 0, 2, 4]. [10 marks]

Answer:

[5 marks] 1st level

$$\left[\frac{2+0}{\sqrt{2}}, \frac{2+4}{\sqrt{2}}; \frac{2-0}{\sqrt{2}}, \frac{2-4}{\sqrt{2}}\right]$$

$$-\left[\frac{2}{\sqrt{2}}, \frac{6}{\sqrt{2}}; \frac{2}{\sqrt{2}}, \frac{-2}{\sqrt{2}}\right]$$
[5 marks] 2nd level

$$\left[\frac{2+6}{\sqrt{2}\cdot\sqrt{2}}, \frac{2-6}{\sqrt{2}\cdot\sqrt{2}}; \frac{2-2}{\sqrt{2}\cdot\sqrt{2}}, \frac{2+2}{\sqrt{2}\cdot\sqrt{2}}\right]$$

$$= \left[\frac{8}{2}, \frac{-4}{2}; 0, \frac{4}{2}\right] = [4, -2; 0, 2]$$

(b) Filter banks are used to implement wavelet transforms.

The analysis low pass filter is referred to as H_0 , and the high pass filter as H_1 .

The synthesis low pass filter is referred to as γ_0 and the high pass filter as γ_1 . For orthogonal analysis filters:

$$H_1(z) = (-z)^{-N} H_0(-z^{-1})$$

And for the synthesis filters

$$\gamma_0(z) = H_1(-z)$$

$$\gamma_1(z) = -H_0(-z)$$

Daubechies wavelets are orthogonal. For a Daubechies 2nd order wavelet, determine

- i) $H_1[n]$,
- ii) $\gamma_0[n]$,
- iii) $\gamma_1[n]$,

in terms of the lowpass filter $H_0[n] = [0.4830, 0.8365, 0.2241, 0.1294]$. [9 marks]

Answer:

$$\begin{split} H_0[n] &= \big[H_0[0], H_0[1], H_0[2], H_0[3]\big] \\ H_1[n] &= \big[H_0[3], -H_0[2], H_0[1], -H_0[0]\big] = \big[0.1294, -0.2241, 0.8365, -0.4830\big] \quad [3 \text{ marks}] \\ \gamma_0[n] &= \big[H_0[3], H_0[2], H_0[1], H_0[0]\big] = \big[-0.1294, 0.2241, 0.8365, 0.4830\big] \quad [3 \text{ marks}] \\ \gamma_1[n] &= \big[-H_0[0], H_0[1], -H_0[2], H_0[3]\big] = \big[-0.4830, 0.8365, -0.2241, -0.1294\big] \quad [3 \text{ marks}] \end{split}$$

c) In the wavelet transform, the scaling function coefficients $c_{m,n}$ and wavelet series coefficients $d_{m,n}$ can be calculated recursively according to the following equations:

$$c_{m-1,n} = \sqrt{2} \sum_{i} h_0[i-2n] c_{m,i}$$

$$d_{m-1,n} = \sqrt{2} \sum_{i} h_1[i-2n] c_{m,i}$$

Explain how this can be interpreted in terms of filtering and downsampling, and hence leads to the concept of an *analysis filterbank*. Sketch a diagram to illustrate this filterbank. [6 marks]

Answer:

If $c_{m,i}$ are the scaling function coefficients at level m, we can calculate the scaling function coefficients and wavelet series coefficients $d_{m-1,n}$ at level m-1 recursively from these. [2 marks]

The signals h_0 and h_1 are (time-reversed) low-pass and high-pass filters. [1 mark] Steps of 1 in n correspond to steps of 2 in i, corresponding to downsampling by a factor of 2 from n to i. [1 mark]

Therefore these equations represent filtering followed by downsampling, as shown in the diagram:

