

ENGINEERING TRIPOS PART IIA

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Thursday 7 May 2009 9 to 10.30

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Module 3F3

SIGNAL AND PATTERN PROCESSING

*Answer not more than **three** questions.*

*All questions carry the same number of marks.*

*The **approximate** percentage of marks allocated to each part of a question is indicated in the right margin.*

*There are no attachments.*

STATIONERY REQUIREMENTS

Single-sided script paper

SPECIAL REQUIREMENTS

Engineering Data Book

CUED approved calculator allowed

**You may not start to read the questions  
printed on the subsequent pages of this  
question paper until instructed that you  
may do so by the Invigilator**

1 The Discrete Time Fourier Transform (DTFT) of a sequence  $\{x_n\}$ ,  $n = 0, \pm 1, \pm 2, \dots, \pm \infty$  is given by

$$X(e^{j\omega T}) = \sum_{n=-\infty}^{\infty} x_n e^{-jn\omega T}$$

where  $T$  is the sample period.

(a) Explain why the DTFT cannot in practice be evaluated on a digital computer and how this problem can be overcome. Show how the Discrete Fourier Transform (DFT)

$$X_p = \sum_{n=0}^{N-1} x_n e^{-j\frac{2\pi np}{N}}$$

where  $N$  is the transform size, can be derived directly from the DTFT expression by application of a rectangular window function sequence  $\{w_n\}$ ,  $n = 0, \pm 1, \pm 2, \dots, \pm \infty$

$$w_n = \begin{cases} 1, & \text{for } n = 0, 1, \dots, N-1 \\ 0, & \text{otherwise} \end{cases}$$

and discretization of the frequency variable.

[30%]

(b) Show that the relationship between the DFT spectrum values  $X_p$  and the true DTFT spectrum  $X(e^{j\omega T})$  may be expressed as a frequency domain convolution of the true spectrum and the spectrum  $W(e^{j\theta})$  of the window function

$$\frac{1}{2\pi} \int_0^{2\pi} W(e^{j\theta}) X(e^{j(\omega T - \theta)}) d\theta$$

Determine an expression for the window spectrum  $W(e^{j\theta})$ .

[30%]

(c) Explain how the radix-2 Fast Fourier Transform (FFT) algorithm allows very efficient implementation of the DFT above when  $N$  is a power of 2. Your description should include: the number of stages in the radix-2 FFT algorithm and the number of “butterfly” computations required in relation to the transform size  $N$ ; the “butterfly” structure and the computations required to compute one “butterfly” operation; bit reversal operations and in-place computation. Show that the total number of real operations (that is, real multiplications, additions or subtractions) required is approximately  $5N \log_2 N$ .

[40%]

2 (a) List the advantages and disadvantages of IIR filters in comparison with FIR filters. [20%]

(b) Describe how the bilinear transform

$$s = \frac{1 - z^{-1}}{1 + z^{-1}}$$

can be used to convert analogue filters to digital filters. State the type of filters that can usefully be designed using this method, and any distortions that are introduced. [20%]

(c) A bandpass filter is required in a channelised radio communications system with sampling rate 8kHz and 3dB cutoff frequencies of 2kHz and 3kHz. We are interested in designing such a filter from an analogue lowpass prototype given by

$$H(s) = \frac{1}{1 + s}$$

which has a 3dB cutoff frequency of 1 rad/sec.

Using the lowpass to bandpass transformation

$$s' = \frac{s^2 + \omega_l \omega_u}{s(\omega_u - \omega_l)}$$

with lower cutoff at  $\omega_l$  and upper cutoff at  $\omega_u$  together with the bilinear transform, design the required digital filter. [40%]

(d) The filter is to be implemented in fixed precision digital hardware. Explain briefly some potential undesirable consequences of this implementation, and suggest strategies for overcoming them. [20%]

(TURN OVER

3 The autoregressive moving-average (ARMA) model is a wide sense stationary process  $\{x_n\}$  satisfying the equation:

$$\sum_{p=0}^P a_p x_{n-p} = \sum_{q=0}^Q b_q w_{n-q}$$

where  $\{w_n\}$  is a white process with unity variance, and without loss of generality we assume  $a_0 = b_0 = 1$ .

(a) By interpreting  $\{x_n\}$  as a result of filtering and assuming that the poles of the ARMA process all lie within the unit circle, derive the power spectrum of the above ARMA process. [30%]

(b) Consider the following stationary random process as a particular case of the ARMA model presented above:

$$x_n = \sum_{q=0}^Q b_q w_{n-q}$$

where  $\{w_n\}$  is a white process with unity variance, and without loss of generality we assume  $b_0 = 1$ . What type of process is this? Show that the autocorrelation function of this process is given by

$$r_{XX}[k] = \sum_{q=0}^Q b_q r_{XW}[k-q]$$

where  $r_{XW}[k]$  is the crosscorrelation between  $x_n$  and  $w_n$  at lag  $k$ . [20%]

(c) Further show that the autocorrelation above can be expressed as: [20%]

$$r_{XX}[k] = \begin{cases} \sum_{l=0}^{Q-k} b_{l+k} b_l, & \text{for } k = 0, 1, \dots, Q \\ 0, & \text{for } k > Q \end{cases}$$

(d) Given the following estimates of the autocorrelation function of a particular signal:

$$\begin{aligned} r_{XX}[1] &= -\frac{1}{2} \\ r_{XX}[k] &= 0 \text{ for } |k| > 1 \end{aligned}$$

determine the coefficient values for this particular model and roughly sketch the signal power spectrum. [30%]

4 Consider the k-means clustering algorithm which seeks to minimise the cost function

$$C = \sum_{n=1}^N \sum_{k=1}^K s_{nk} \|x_n - m_k\|^2$$

where  $m_k$  is the mean (centre) of cluster  $k$ ,  $x_n$  is data point  $n$ ,  $s_{nk} = 1$  signifies that data point  $n$  is assigned to cluster  $k$ , and there are  $N$  data points and  $K$  clusters.

(a) Given all the assignments  $\{s_{nk}\}$ , derive the value of  $m_k$  which minimises the cost  $C$  and give an interpretation in terms of the k-means algorithm. [30%]

(b) Give a probabilistic interpretation of k-means and describe how it can be generalised to unequal cluster sizes (number of data points per cluster) and non-spherical (elongated) clusters as shown in Fig. 1 below. [30%]

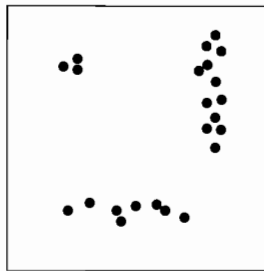


Fig. 1

(c) In many real-world applications, data points arrive sequentially and one wants to cluster them as they come in. Devise a sequential variant of the k-means algorithm which takes in one data point at a time and updates the means  $\{m_1, \dots, m_K\}$  sequentially without revisiting previous data points. Describe your sequential algorithm. [40%]

**END OF PAPER**