

Medical Image Analysis (EE61008)

Mid-Semester Examination

Spring, 2015-16

(1 mark)

(1 mark)

(1 mark)

(1 mark)

Credits: 4 Full Marks: 70 Date: Friday, 19 February 2016, 9-11 AM

Duration of Examination: 2 hours

Instructions:

- 1. All guestions are compulsory. Marks are indicated in parentheses. Clearly state all assumptions in your answer.
- 2. Please write your name, roll number, subject name and code, date and time of examination on the answer script before attempting any solution.
- 3. Use of only electronic calculators is permitted.
- 4. No extra resources viz. graph papers, log-tables, trigonometric tables would be required.

Question 1:

Identify the following in Fig. 1

- (a) Organ? (b) Plane of imaging? (Transverse, Sagittal, Coronal)
- (c) Modality?
- (d) Structure / anatomical location of interest?



Question 2:

The following matrix in *I* is a snippet of an RGB image from a digital histology sample and the seeds for the two classes are marked in the matrix M as 1 and 2 and the unmarked pixels are in 0. Compute the following for estimating the response of the random walker analogue of an electrical network, assuming 8-connected relationship between the pixels, and the edge conductance between the *i*th and *j*th pixel in *I* defined as $w_{i,i} = e^{-\beta \|\mathbf{c}_i - \mathbf{c}_j\|}$ where \mathbf{c}_i is the (r,g,b) colour triplet vector representing the i^{th} pixel and $\|.\|$ is the Euclidean distance measure. Also consider that $\beta = 100$.

$I = \begin{bmatrix} (100,0,0) \\ (0,0,100) \end{bmatrix}$	(0,1,0) (1,1,1)	(0,100,0) (100,0,0)	$M = \begin{bmatrix} 0\\ 0 \end{bmatrix}$	1 0	0 0		
[(100,0,0)	(1,0,0)	(0,100,0)]	LO	2	01		
a. Draw the graph representation of the random walks solver?							(2 marks)

- b. Mark the edge conductivity (weights) on the graph?
- (4 marks) What is the probability for each class associated with the underlined pixel in I as c. estimated by solving the electrical network equivalent of the random walks? (10 marks)

Question 3:

The following matrix consists of the readings of a CT scanner in Hounsfield unit (HU) as recorded in the matrix H and the ground truth for 2 classes as marked in the matrix G. Using the following standard range of HU readings, can you predict the class labels for the two tissue types, i.e. 1 - fat where HU is in range (-200 to -50) and 2 - brain tissue HU in range (30 to 40) and answer the following questions:

 $G = \begin{bmatrix} 1 & 1 & 2 & 2 & 2 \\ 1 & 1 & 2 & 2 & 2 \\ 1 & 1 & 2 & 2 & 2 \\ 1 & 1 & 2 & 2 & 2 \\ 1 & 1 & 2 & 2 & 2 \end{bmatrix}$ -12032 33 34 40 -115 -123 31 37 37 H =-100 -100 36 39 -132 32 39 38 -198 - 143

Compute the (a) TP, (b) TN, (c) FP, (d) FN, (e) accuracy, (f) sensitivity, (g) specificity, (h) precision, (i) F1 score and (j) Kappa score. (20 marks)

Question 4:

- a. Construct a binary decision tree using the information provided in *H* and *G*. Provide all necessary details of the steps involved in the process. (10 marks)
- b. Comment on the node purity of the leaf nodes in terms of the posterior probability of each class in the leaf nodes of the decision tree. (5 marks)

Question 5:

A standard 7 T MRI machine has an effective imaging volume dimension of 60 cm x 60 cm along the transverse section and a length of 200 cm. The slice encode gradient is of 40 mT/m and the frequency encode gradient is of 40 μ T/m. Compute the following:

- a. Range of magnetic field strength to be experienced within the effective imaging volume? (2 marks)
- b. Range of Larmor frequencies to be experienced within the effective imaging volume by a hydrogen atom with a gyromagnetic ratio of $\gamma = 42.58 \text{ MHz/T}$? (2 marks)
- c. Consider that the voxel resolution of the MRI imaging device is 10 mm x 10 mm x 10mm. What is the minimum frequency hop between two neighboring voxels in the volume?

(5 marks)

Question 6:

Consider a vector **x** to be the input to a perceptron model such that it produces a vector $\hat{\mathbf{y}}$ as the output. They are related as $\hat{\mathbf{y}} = f_{\text{NL}}(\mathbf{w}, [\mathbf{x}; 1])$, and the non-linear activation function is $f_{\text{NL}}(z) = \frac{1}{1+e^{-z}}$, and **w** is termed as the connection weights relating the input to the output. Say that during training of the network using the gradient descent rule, for a set of given inputs $\{\mathbf{x}_i, \mathbf{y}_i\}$ the error between the observed as the predicted is written as the cost function of the learning rule as

$$J(\mathbf{w}) = \frac{1}{N} \sum_{i=1}^{N} \|\hat{\mathbf{y}}_i - \mathbf{y}_i\|$$

then derive the analytical form of the following?

$$\frac{\partial J(\mathbf{w})}{\partial \mathbf{w}}$$

(6 marks)