



# INTRODUCTION TO COMPUTER VISION

## ECSE-415

**(There are 10 questions in total, each is worth 10 marks)**

This take-home exam is to be done online, on mycourses.

The exam will be available on mycourses for a 48-hour period. You can submit your answers at any time during this period. You can do multiple submissions, but only the last one will be graded.

The exam is open-book, but you must do the exam by yourself, with no outside assistance.

## 1. Cameras

Suppose you are at the camera store and are aiming to buy a new lens for your digital camera. The choices available at the store are:

- a. 135mm focal length, f-stop of  $f/2.0$
  - b. 24mm focal length, f-stop of  $f/2.8$
- 
- a) Which lens would you choose if you wanted the widest field of view?
  - b) Which lens would you choose if you wanted the shallowest depth of field?
  - c) Which lens would be the least expensive?
  - d) Which lens provides the most light at the camera's image sensor (when viewing a scene with uniform brightness)?
  - e) Which lens would give more perspective distortion?

## 2. Filtering

Given the five convolution kernels shown in a) through e) apply each convolution mask to the central pixel of the image patch depicted in diagram f) below (i.e. get a single number for each).

a)

1	1	1
1	1	1
1	1	1

b)

-1	-2	3
-1	2	-1
3	2	1

c)

-1	-2	-1
0	0	0
1	2	1

d)

-1	0	1
-2	0	2
-1	0	1

e)

0	1	0
1	-4	1
0	1	0

f)

7	3	1
7	5	0
8	4	1

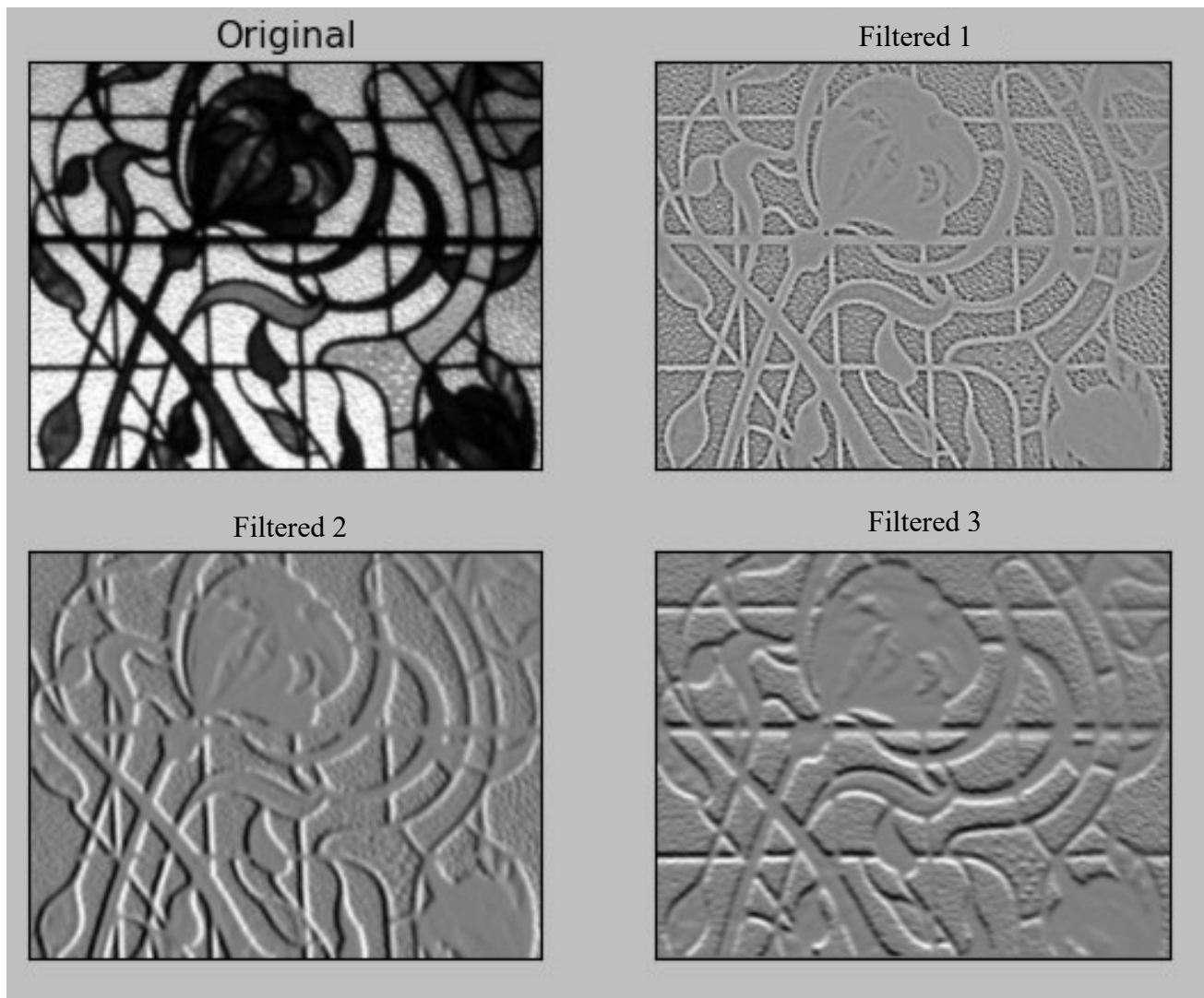
Value for a): .....

Value for b): .....

Value for c): .....

Value for d): .....

Value for e): .....



Match each filtered image with the convolution kernel that was used to generate it (from the previous page)

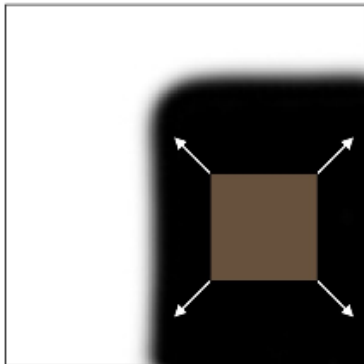
Filtered 1: convolution kernel ..... (out of choices a, b, c, d, e )

Filtered 2: convolution kernel ..... (out of choices a, b, c, d, e )

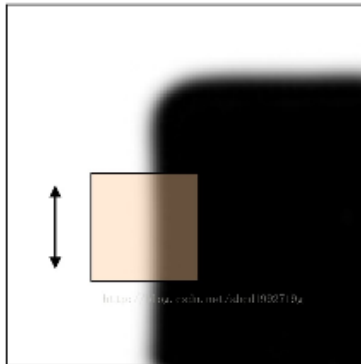
Filtered 3: convolution kernel ..... (out of choices a, b, c, d, e )

### 3. Harris Corner Features

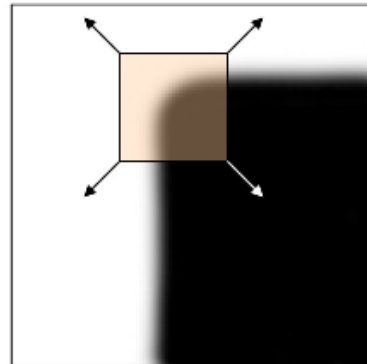
a) Which of the following image patch locations would produce the highest response for a Harris operator? Answer = .....(a,b or c)



(a)



(b)



(c)

b) The Harris operator is given by:

$$f = \frac{\lambda_- \lambda_+}{\lambda_- + \lambda_+}$$

Suppose we have the following 4x4 pixel image patch:

3	4	3	1
4	6	5	0
6	8	4	1
6	9	7	2

Compute the value of  $f$  for the 2x2 pixel window in the middle of this patch.

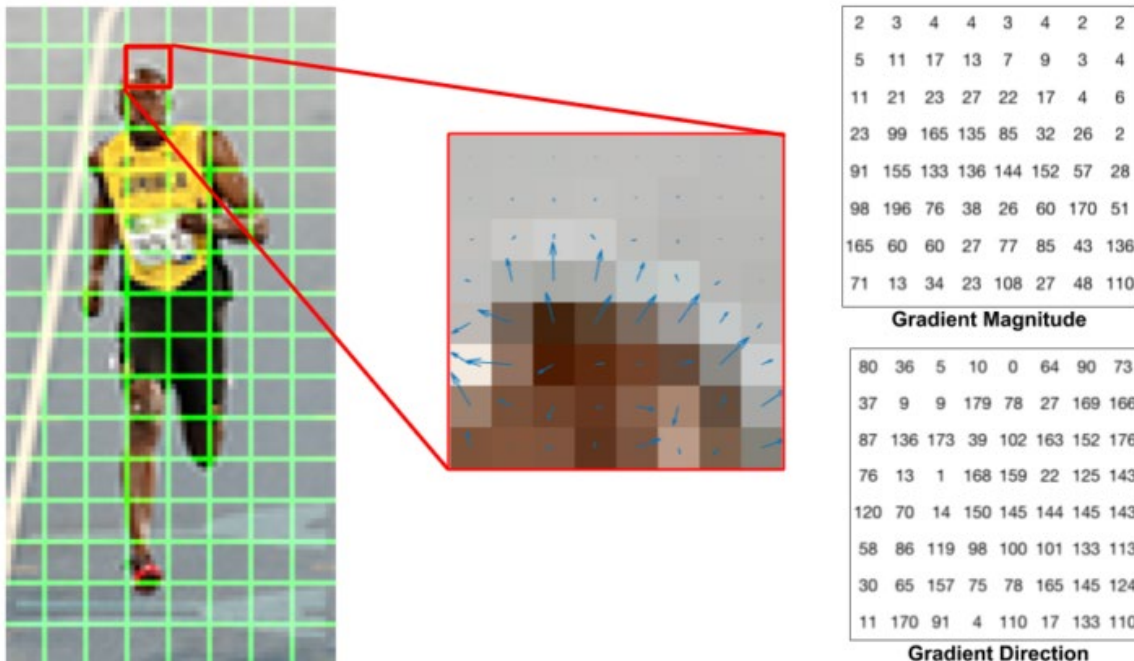
Answer = .....

## 4. SIFT/HoG Descriptors

a) What image transformations is SIFT invariant to? (select all that apply)

- a. Rotation in the image plane
- b. Translation
- c. Non-rigid object deformation
- d. Affine intensity change
- e. Logarithmic image intensity compression
- f. Image intensity inversion (e.g.  $255-I(x,y)$ )
- g. Mirror reflections
- h. Scaling (changes in size of objects)
- i. Similarity transforms
- j. Perspective projections

b) In the image block shown highlighted below, the gradient magnitude and gradient directions have been computed. Assuming the orientation histogram has 9 bins covering the range of angles from 0 to 180 degrees: (0-19, 20-39, 40-59, 60-79, 80-99, 100-119, 120-139, 140-159, 160-179) Compute the 9 entries of the HOG histogram for this block, weighted by the gradient magnitudes (i.e. increment the corresponding bin by an amount equal to the gradient magnitude).



Answer = [.....]

## 5. Homography and Image matching

- a) Match the following homography matrices to the type of transformation that they produce.

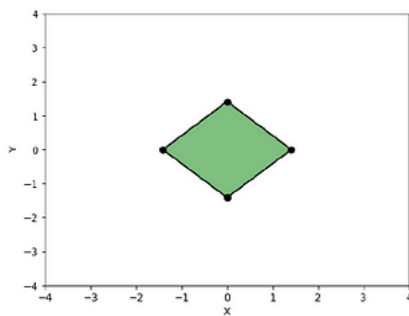
1	0	1
0	1	2
0	0	1

0.866	-0.5	0
0.5	0.866	0
0	0	1

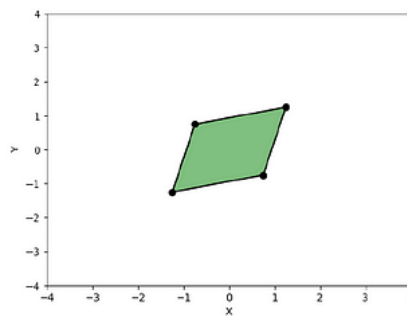
1	0.25	0
0.25	1	0
0	0	1

2	0	0
0	2	0
0	0	1

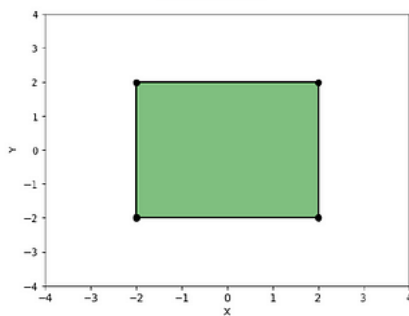
0	-1	0
1	0	0
0	0	1



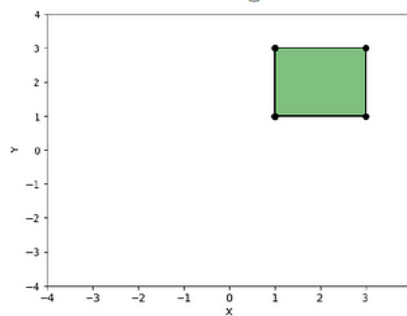
Rotation



Skewing



Scaling



Translation

b) RANSAC: *for each of the 10 pairs of points* in the dataset of 5 different 3D features shown below, *find the number of inliers* to the line between those points. Define an inlier as a point that has a distance of 1.0 units or less to the line (hint: look back at your math class notes to remember how to compute the distance between a point and a line).

Point 1: (1.0, 1.0, -2.2)

Point 2: (0.0, 1.0, -1.4)

Point 3: (1.0, -1.0, 2.0)

Point 4: (2.0, 0.2, -2.0)

Point 5: (-1.0, 0.8, 0.6)

Pair [Point1, Point2] # inliers = .....

Pair [Point1, Point3] # inliers = .....

Pair [Point1, Point4] # inliers = .....

Pair [Point1, Point5] # inliers = .....

Pair [Point2, Point3] # inliers = .....

Pair [Point2, Point4] # inliers = .....

Pair [Point2, Point5] # inliers = .....

Pair [Point3, Point4] # inliers = .....

Pair [Point3, Point5] # inliers = .....

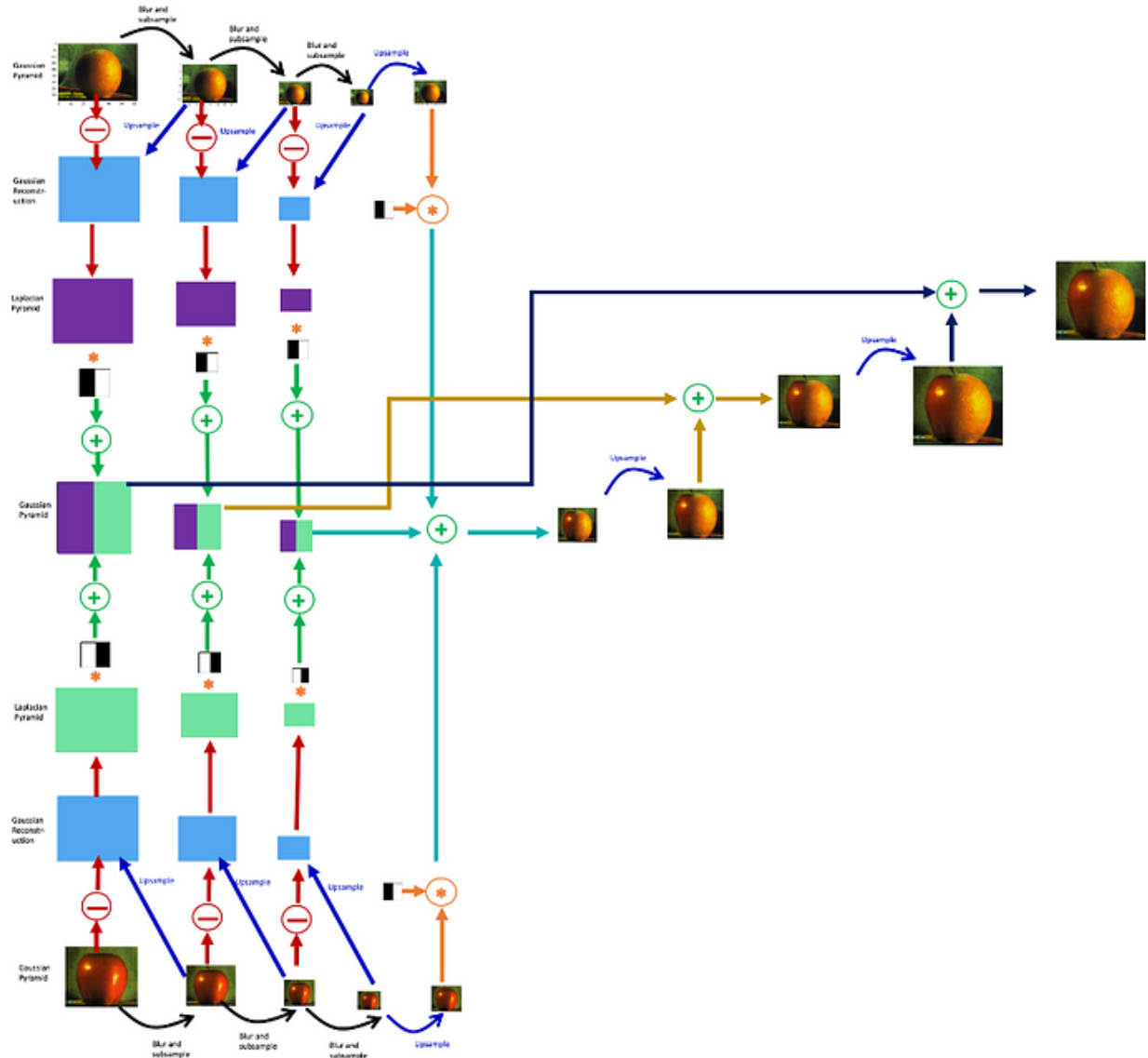
Pair [Point4, Point5] # inliers = .....



## **6. Scale Space**

- a) Explain what a Laplacian or DoG pyramid is and how it can be constructed.

b) With the aid of the diagram below explain how multi-scale image blending works, such as used in making panoramic images.





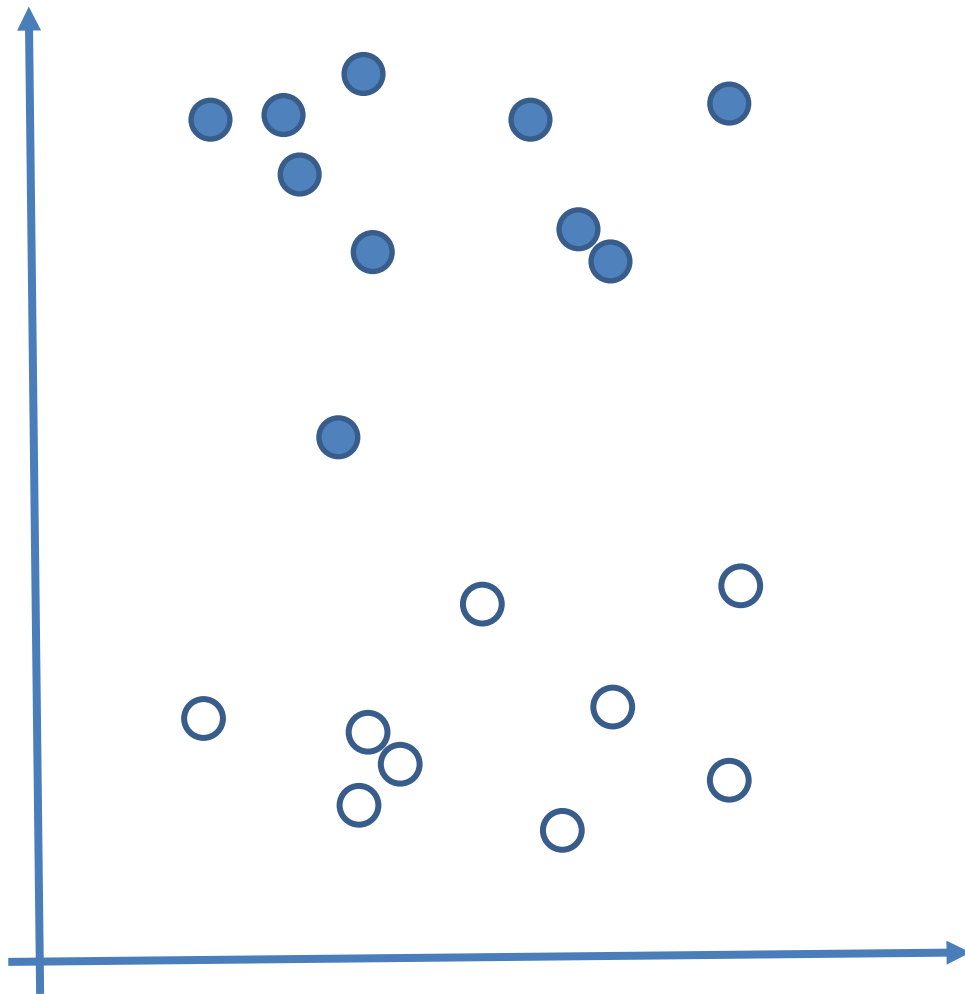
## 7. Faces

- a) What are two characteristics of a good image representation for face detection?
  
  
  
  
  
  
  
  
  
  
- b) Explain the difference between face detection and face recognition.
  
  
  
  
  
  
  
  
  
  
- c) Suppose you want to represent a face image patch with three principal components. How do you select the principal components that would give the best reconstruction of the original face image patch?

## 8. Support Vector Machines

Consider a 2-dimensional 2-class classification problem, with samples (training data) shown in the diagram below.

- Draw the *maximum margin classifier's* decision boundary and show the *margin* (or at least where you think they should be, approximately)
- Circle at least one *support vector* on the diagram



- c) Suppose that for a given 2-D feature vector  $(x_1, x_2)$  the training data can be nonlinearly separated by the curve defined by  $x^3 + y^3 + xy = 0$ .

Define a new set of features  $(z_1, z_2)$  for which the data can be linearly separated by the hyperplane  $z_1 + z_2 = 0$ .

$z_1 =$

$z_2 =$

- d) Which one of these images is the most likely to be a support vector if they were used as training examples for an SVM classifier? Explain your answer.

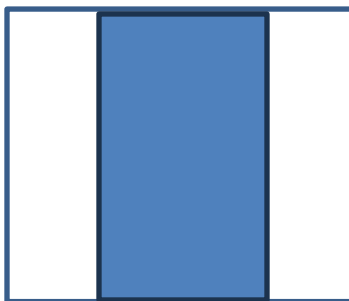


## 9. Haar features

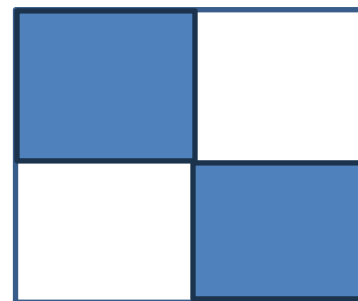
Shown below are 3 different Haar features:  
(on a 4x4 grid of pixels, blue = -1, white = +1)



Haar Feature A



Haar Feature B



Haar Feature C

- a. In the box on the right, fill in the values of the INTEGRAL IMAGE corresponding to the small image whose values are shown on the left.

1	3	0	7
1	1	2	3
4	1	2	2
4	3	0	1


- b. Compute the value for each of the Haar features for the image above.

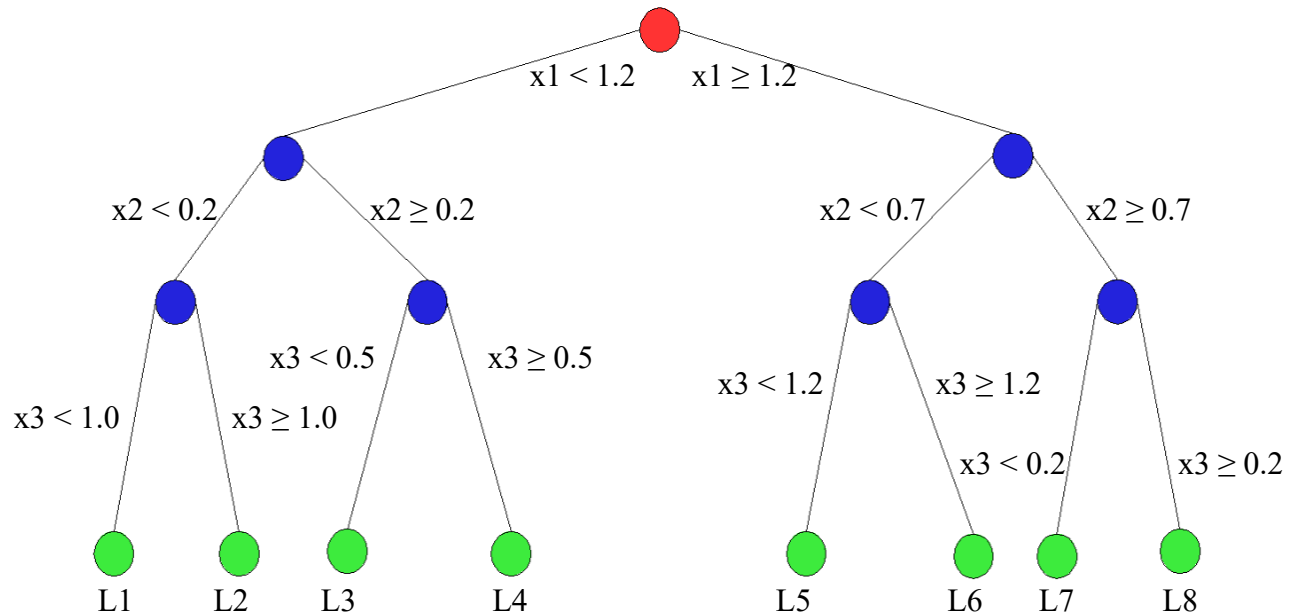
Answer: Haar Feature A Value = .....

Answer: Haar Feature B Value = .....

Answer: Haar Feature C Value = .....

## 10. Decision Tree Classification:

Consider the decision tree shown below which uses three features  $x_1$ ,  $x_2$ ,  $x_3$  to classify into one of three classes  $C_1$ ,  $C_2$ ,  $C_3$ .



Suppose we are given the following training data:

$x_1$	$x_2$	$x_3$	Class	Leaf Node
0.0	0.0	0.0	$C_1$	
1.0	1.0	1.5	$C_3$	
1.5	0.1	0.9	$C_1$	
1.8	0.1	1.1	$C_2$	
1.3	0.8	2.0	$C_1$	
0.6	0.6	1.7	$C_3$	
0.4	1.2	0.7	$C_1$	
2.2	0.0	1.0	$C_2$	
0.0	0.8	1.3	$C_2$	
1.2	0.5	1.1	$C_3$	

- Fill in the last column of the table (indicate which leaf node is selected).
- Given new feature values  $x_1=2.0$ ,  $x_2=0.5$ ,  $x_3=0.9$ , what is the probability of each class?







