

1. Stereo Vision and Pose Estimation

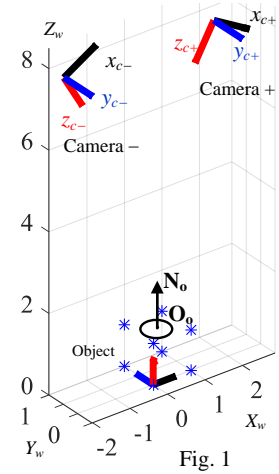
- (a) Camera Model. Write a program (CameraModel.m) to transform 8 features image -points indicated as “*” in Fig. 1 from the 3D world coordinate ($X_w Y_w Z_w$) to the 2D undistorted image coordinate ($u_1 v_1, u_2 v_2$) for Camera+ and Camera- (both with the same length $f=1.2$), the coordinates of which are defined by

Camera \pm : Start with $[\mathbf{R}_x(180^\circ)]$ and $\mathbf{T}=[\pm 2.0.5 \ 8]^T$, then $[\mathbf{R}_y(\mp 30^\circ)]$.

Given the location/orientation of the two cameras and the 8 physical points (Table 1), determine and show the same points but in the corresponding image planes ($u_1 v_1, u_2 v_2$).

Table 1 Features

	Eight feature points								O_o	N_o
X_w	0	1	1	0	0	1	1	0	0.5	0
Y_w	0	0	1	1	0	0	1	1	0.5	0
Z_w	0	0	0	0	1	1	1	1	1	1

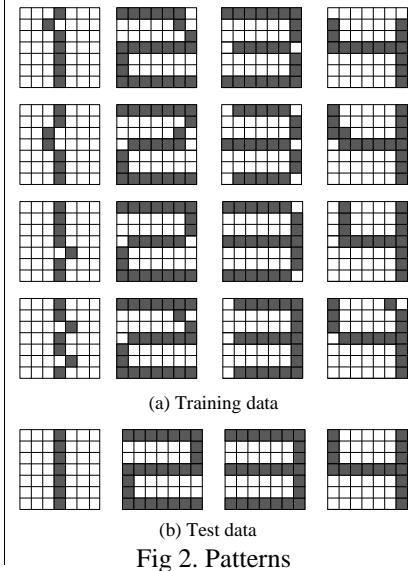


- (b) Stereo Vision. Using the computed image points in Part (a), write a program (StereoVision.m) to reconstruct the 8 physical feature points in the world coordinates.
- (c) Pose Estimation [1]. Write a program (PoseEstimation.m) to calculate the rotation matrix \mathbf{R} and translation vector \mathbf{T} from the world coordinates to Camera +. You may use the following steps:
- Step 1: Determine these 8 feature points and the centroid and normal vectors of the circle (Table 1) in Camera + coordinates.
- Step 2: Form the matrix $[\mathbf{W}]=[\mathbf{M} \ \mathbf{Q} \ \mathbf{Q}^T]$ to solve for \mathbf{V} (that contains the elements of \mathbf{R} and \mathbf{T}) from $[\mathbf{W}]\mathbf{V}=\mathbf{b}$, where \mathbf{M} and \mathbf{Q} characterize the point and the ellipse-circle correspondences; and \mathbf{Q}^T takes into account the orthonormal constraints.

2. Neural Network [2]

Design an artificial neural network (ANN) based on back-propagation learning for number recognition. To illustrate your understanding, the ANN is designed to have 49 inputs and 4 outputs to recognize the numerical characters (1, 2, 3 and 4) in a binary 7x7 image.

- (a) Derive the weight-update rule for the ANN assuming that each of the processing elements has a uni-polar sigmoid function.
- (b) Write a Matlab program NN_training.m to train the ANN using the training patterns in training_data.m (Fig. 2a). Show the convergence curve (mean squared error vs. number of epoch). Save the weights of nodes in the file NN_weights.mat.
- (c) Write a Matlab program NN_test.m (that reads the weights of nodes in NN_weights.mat) to test the ANN using data in test_data.m (Fig. 2b). Show the values of four output nodes for each testing pattern.



3. Color

Artificial Color Contrast (ACC) [3] increases color contrast using the DoG functions to facilitate feature classification as discussed in class:

$$h_i(x, y) = G_{\sigma_c} * f_j(x, y) - G_{\sigma_s} * f_k(x, y)$$

Derive $h_i(x, y)$ with $f_j(x, y) = R(x, y)$ and $f_k(x, y) = -[R(x, y) - G(x, y)]$ where $R(x, y)$ and $G(x, y)$ correspond to the red and green component images respectively. Use this single channel to filter out blood stains and find the fanbone in the image 'fanbone.bmp' (Fig. 3).

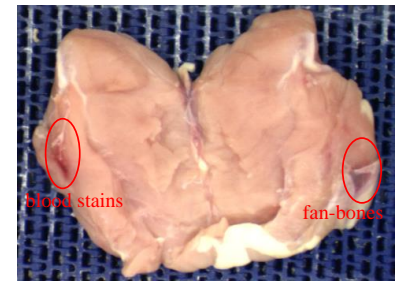


Fig. 3 fanbone.bmp

Principle component analysis (PCA): Use the RGB image 'fanbone.bmp' for the following.

- Determine the covariance matrix of data.
- Derive the components (eigenvectors) with eigenvalues arranged in a descending order.
- Obtain the maximum and minimum values of three component matrices. Show these three matrices (images) with linear mapping from the minimum and maximum values to the range of (0-255).

Color-based Image Segmentation: Color is an important information. Same objects commonly have their domain color. L - a - b color system is the color-opponent space with the L lightness dimension and a - b color-opponent dimensions. The color-based image segmentation can be performed by applying the clustering method on the points in a - b domain. Transfer pixels from RGB to Lab color system and followed by applying k -means clustering on data in a - b domain with cluster number ($k=3$) for segmentation of the RGB image 'fanbone.bmp'.

4. Morphology

- As in the case of dilation and erosion, opening and closing are duals of each other with respect to set complementation and reflection. Prove validity of the duality expression: $(A \bullet B)^c = (A^c \circ B)$
- Write a MATLAB program to perform a morphological filtering on the head CT image in Fig. 4(a) using the structuring element in Fig. 4(b). Denoting the head CT image as f and the structuring element as b , perform morphological gradient operation on f by following the steps:
 - $f \oplus b$
 - $f \ominus b$
 - $(f \oplus b) - (f \ominus b)$

Show the corresponding image obtained in each step



Fig. 4(a)

1	1	1
1	1	1
1	1	1

Fig. 4(b)

Suggested MATLAB function: *imdilate.m*, *imerode.m*

Reference:

- [1] Qiang Ji, Mauro Costa, Robert Haralick, and Linda Shapiro, "An Integrated Linear Technique for Pose Estimation from Different Features," *International Journal of Pattern Recognition and Artificial Intelligence*, Vol. 13, No. 5, 1999
- [2] Simpson, P. (1992). Foundations of neural networks, Chapter 1 in *Artificial Neural Networks*, Lau, C. & Sanchez-Sinencio, E. (Eds.), pp. 2-13, IEEE Press, New York, NY
- [3] Lee, K.-M.; Li, Q.; Daley, W., "Effects of Classification Methods on Color-Based Feature Detection With Food Processing Applications," *IEEE Trans. on Automation Science and Engineering*. Vol. 4, No. 1, Jan. 2007 pages: 40-51