

2D Shape Matching

Computer Vision
Exercise session 6

TA: Yawei Li

Assignment Tasks:

1. Shape Matching
2. Shape based Image Classification

Hand-out: 23-11-2017

Hand-in: 08-12-2017 13:00

Task 1: Shape Matching

Objectives

1. Compute shape context descriptors
2. Match a template shape to a target set of points using shape contexts



Overview of Algorithm

Given a set of template and target points:

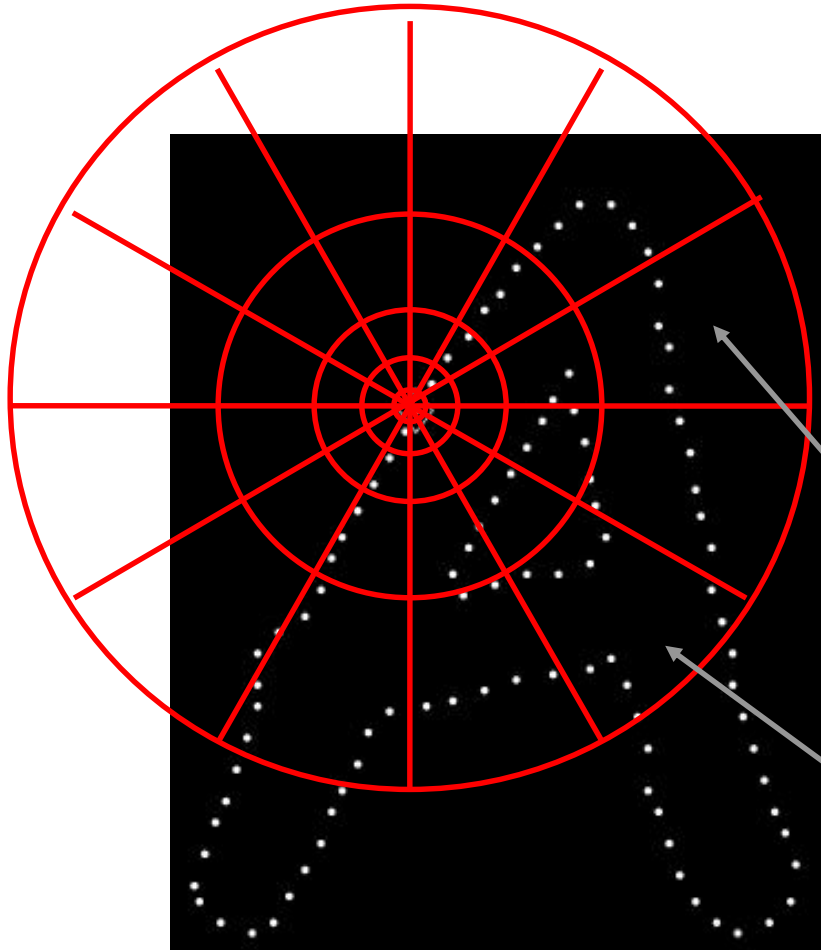
- a. Compute shape context descriptors for both sets of points
- b. Estimate cost matrix between two sets of descriptors
- c. Use cost matrix to solve the correspondence problem between two sets of descriptors (e.g. with Hungarian algorithm)
- d. From the correspondence, estimate a transformation from template to target points (e.g. with Thin Plate Splines) and perform this transformation on the template points
- e. Iterate steps a-d.

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Shape Context Descriptor

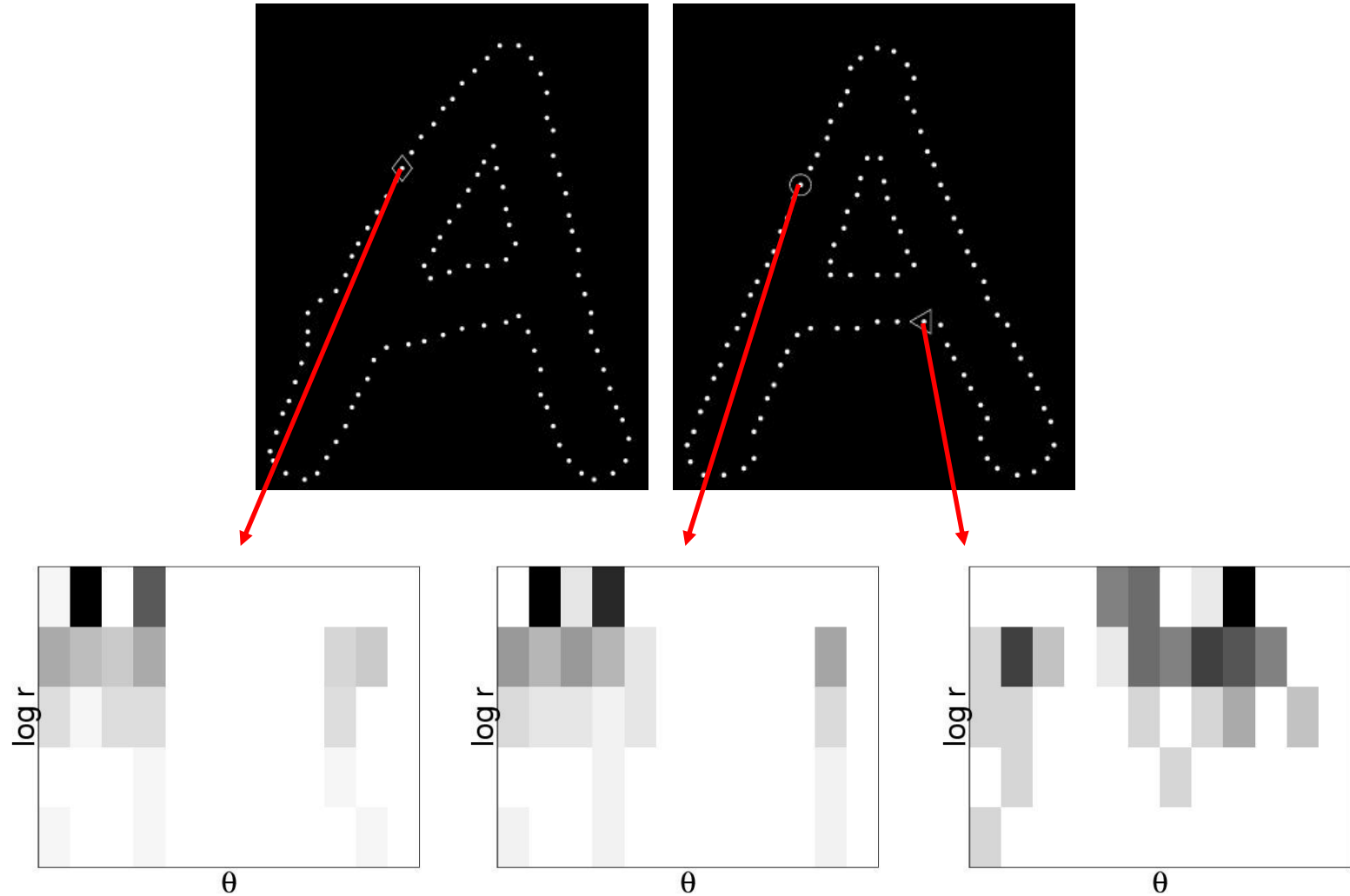


- Count number of points inside each bin
- Compact representation of distribution of points relative to each point

Count = 4

Count = 10

Shape Context Descriptor (2)



Task 1

Belongie, et al, PAMI 2002, [Shape matching and object recognition using shape contexts](#)

Overview of Algorithm

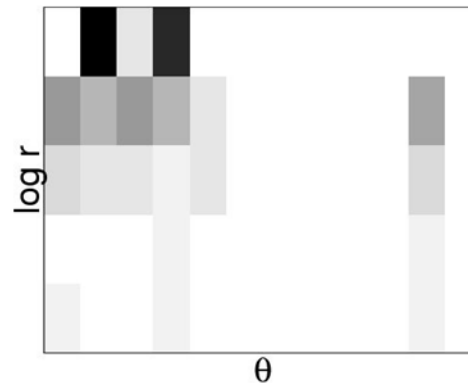
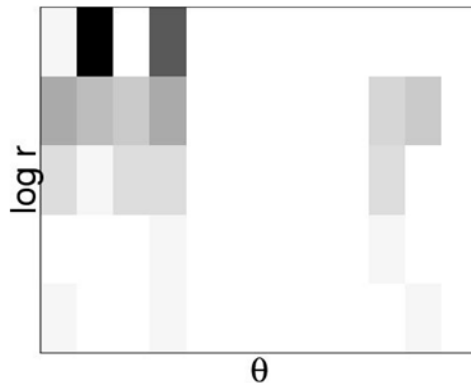
Given a set of template and target points:

- a. Compute shape context descriptors for both sets of points
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Matching Costs

Chi-squared
distance between
descriptors i and j .

$$C_{ij} \equiv C(p_i, p_j) = \frac{1}{2} \sum_{k=1}^K \frac{[p_i(k) - p_j(k)]^2}{p_i(k) + p_j(k)}$$

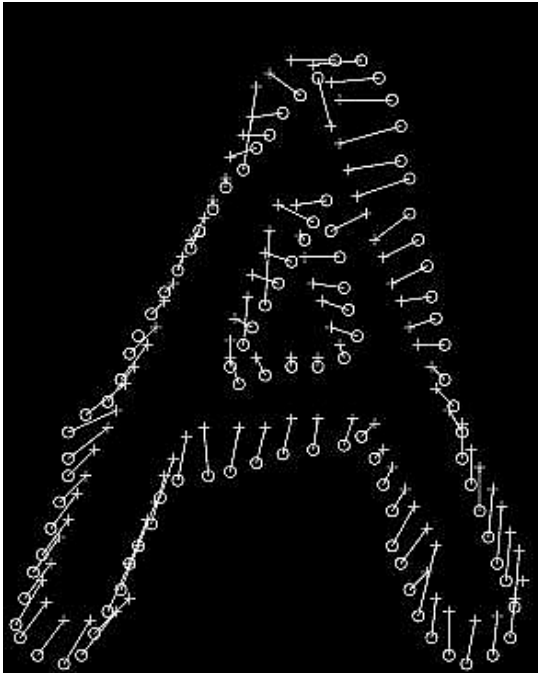
 p_i p_j 

Overview of Algorithm

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Correspondence Problem



- Minimize total cost of matching such that matching is one-to-one
- E.g. with Hungarian algorithm

Overview of Algorithm

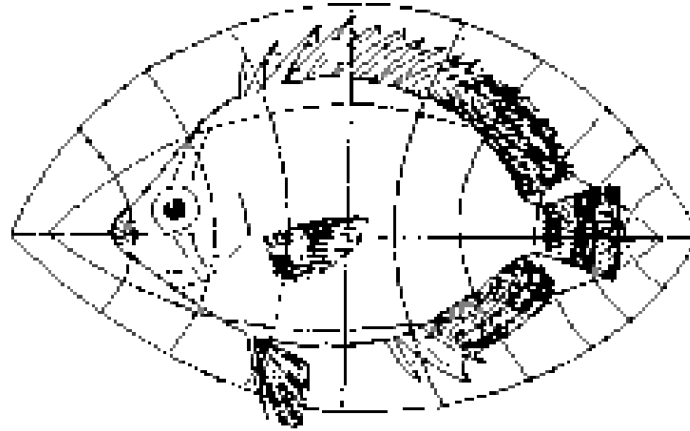
Given a set of template and target points:

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Transformation

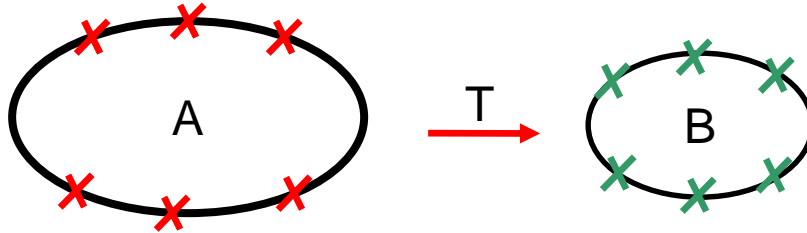


Model



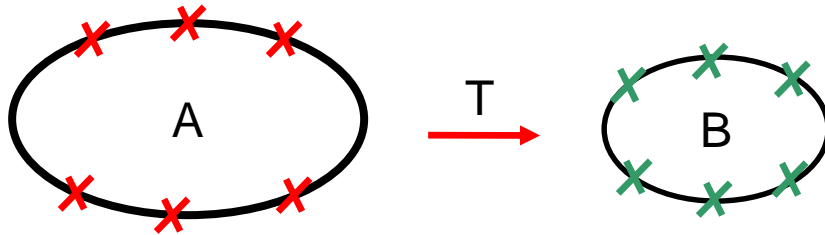
Target

Thin Plate Splines(1)

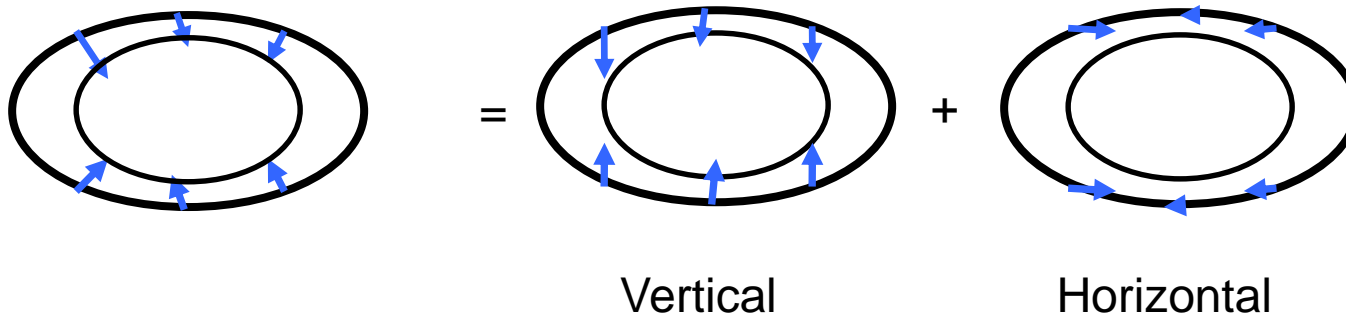


- We are given a set of correspondences
- We want to estimate the function $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ that transforms A into B

Thin Plate Splines(2)

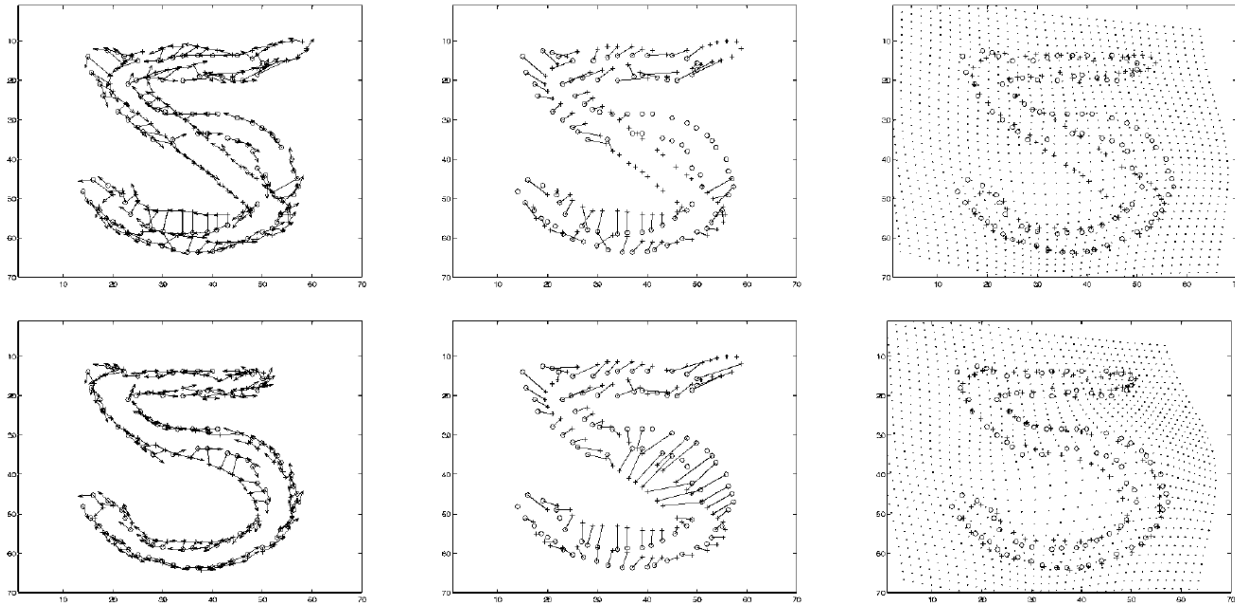


- From the correspondences, we get a displacement:



- Each component (vertical and horizontal) is a single function that we want to interpolate with a TPS.

Thin Plate Splines(3)

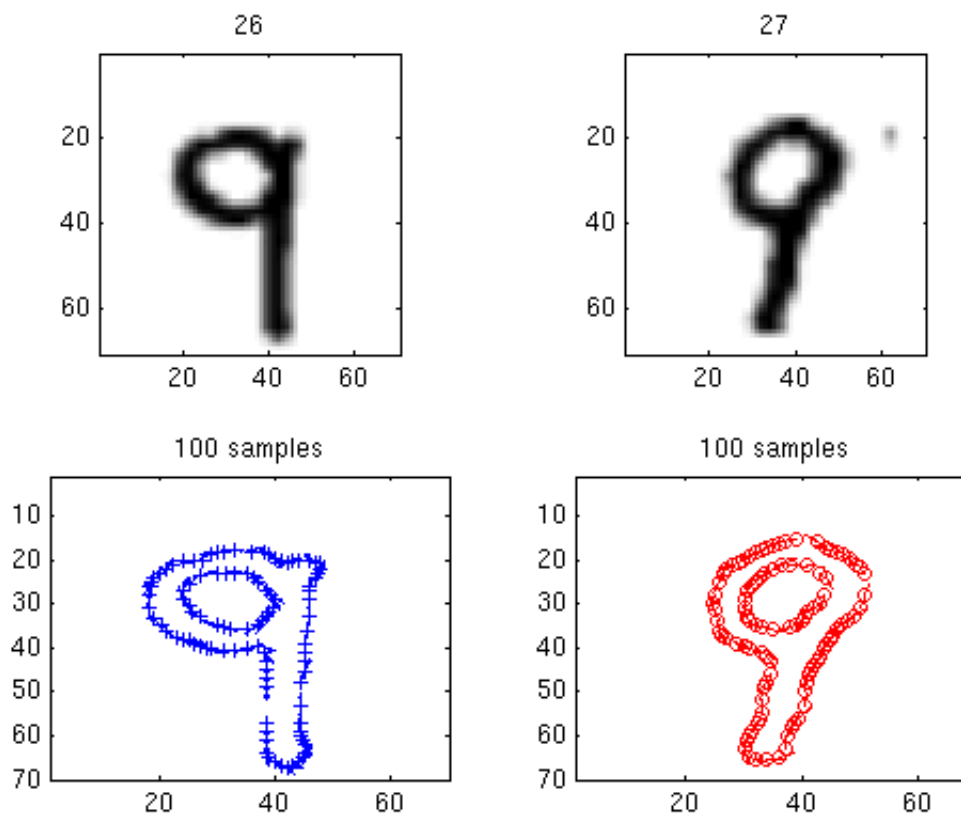


Overview of Algorithm

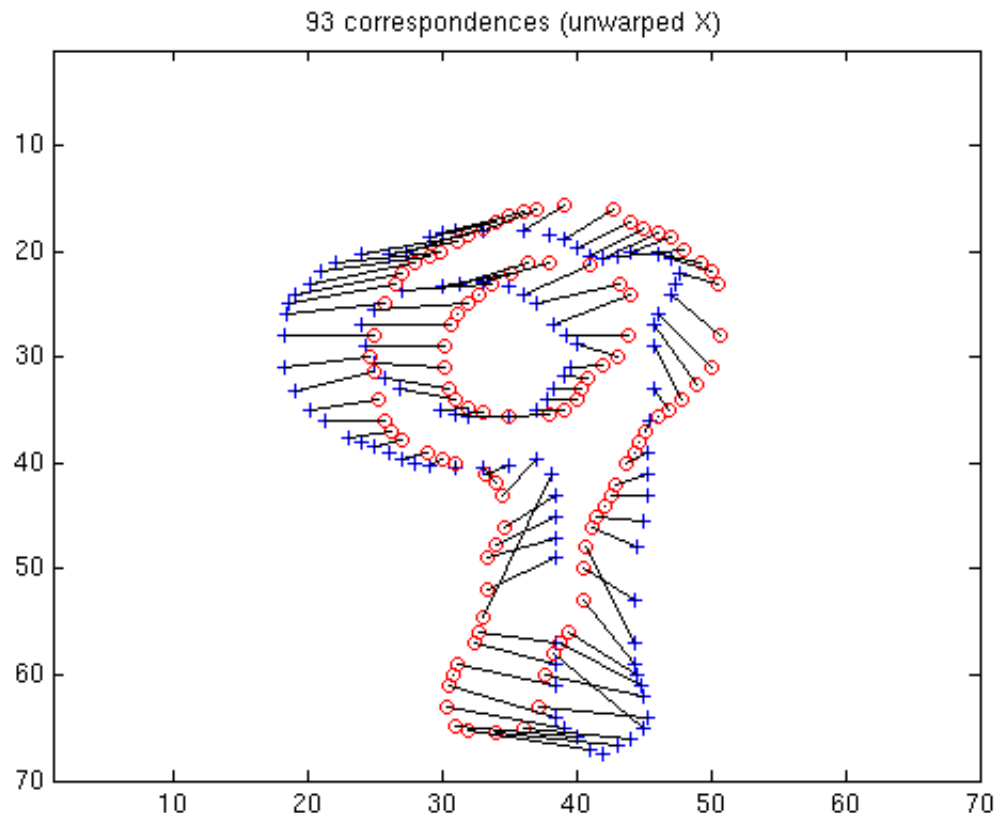
Given a set of template and target points:

- a. Compute shape context descriptors for both sets of points
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- c. Use cost matrix to solve assignment problem between two sets of descriptors (e.g. with Hungarian algorithm)
- d. From the assignment, estimate a transformation from template to target points (e.g. with Thin Plate Splines)
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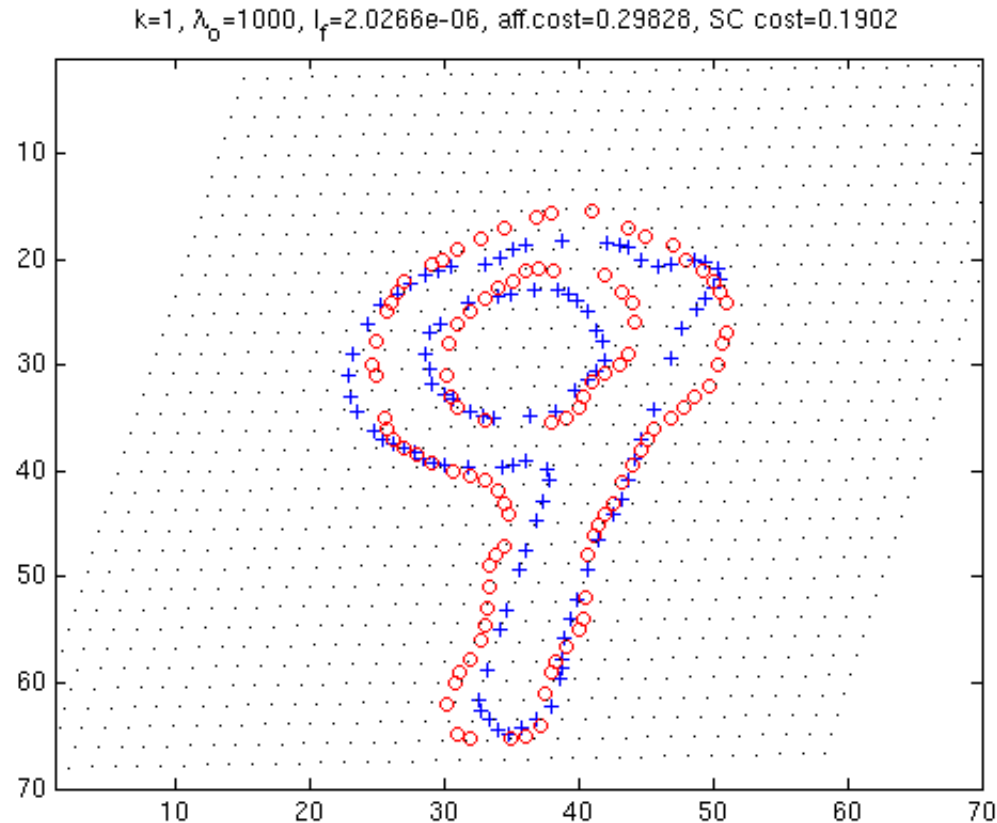
Example 1 - Numbers



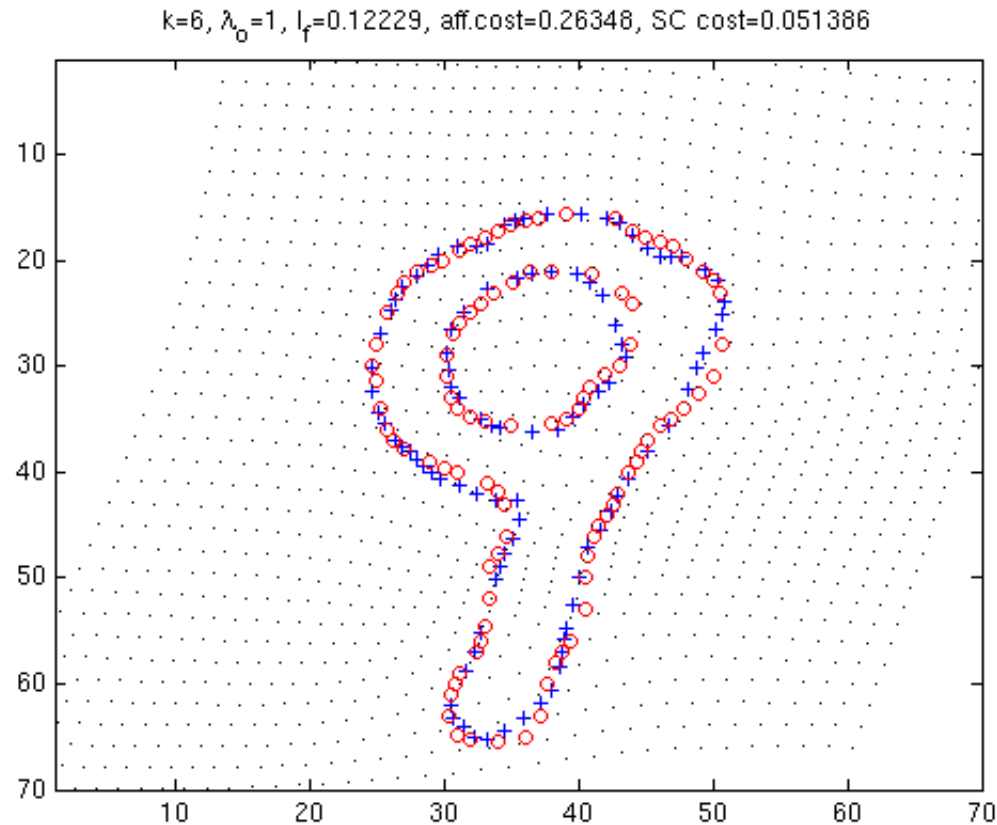
Example 1 - Numbers



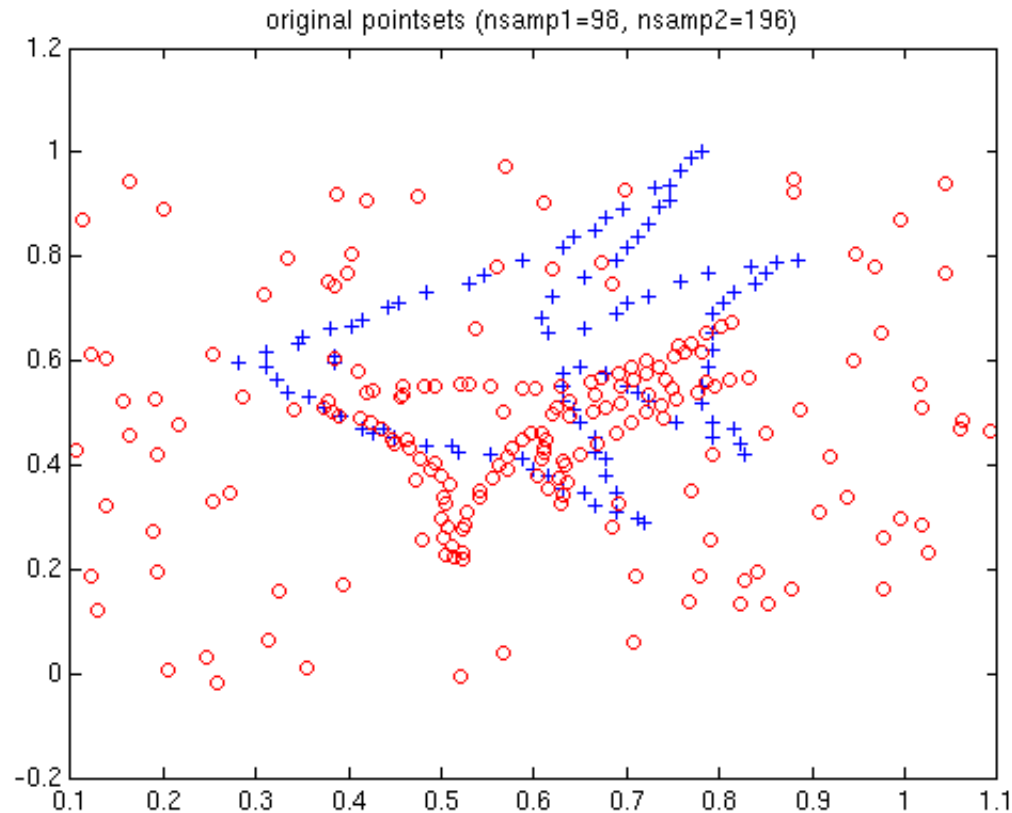
Example 1 - Numbers



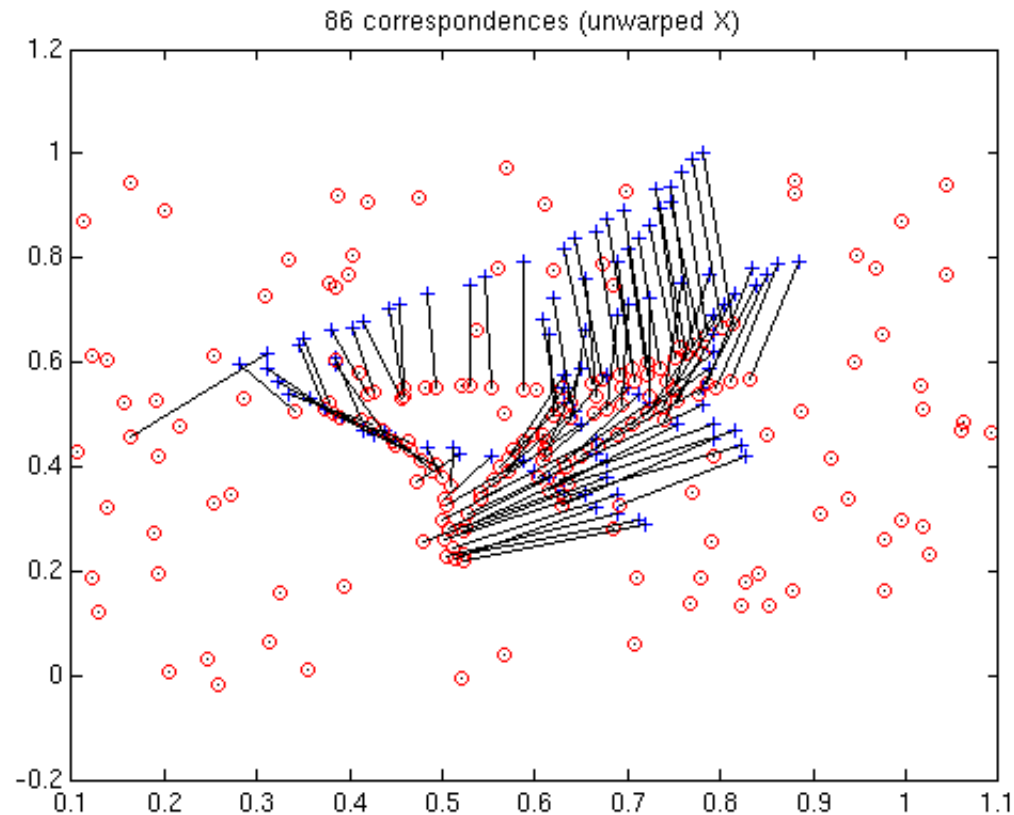
Example 1 - Numbers



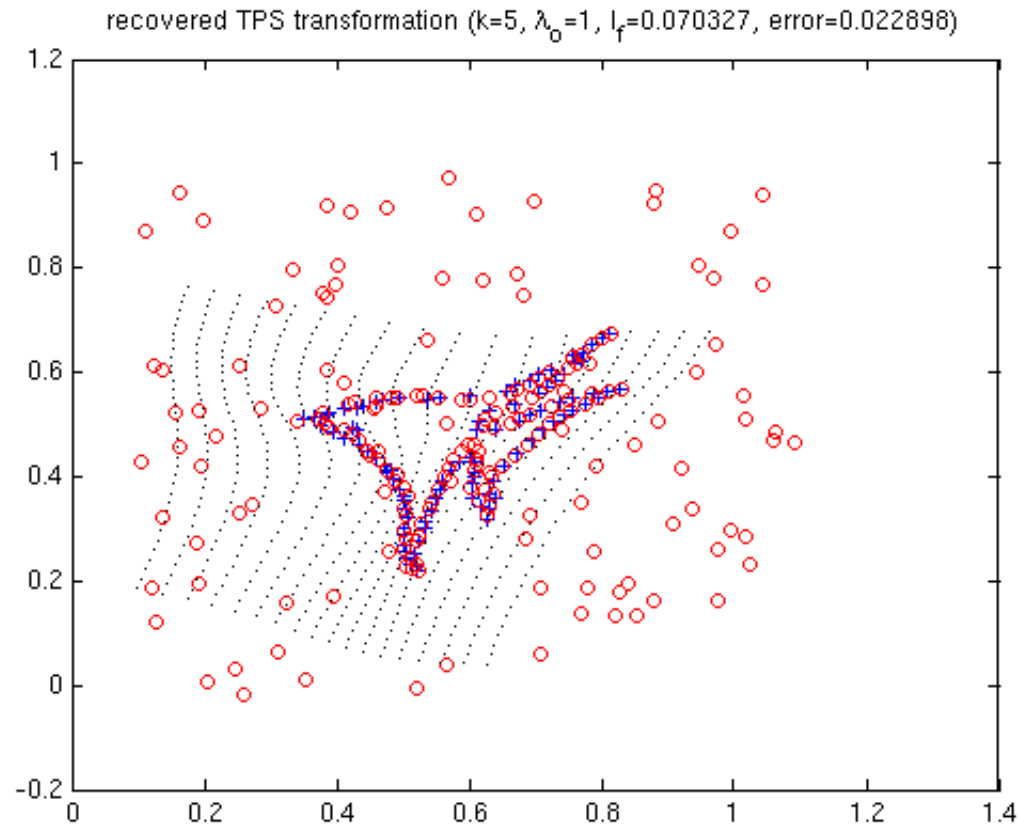
Example 2 - Fish



Example 2 - Fish



Example 2 - Fish



Task 2: Shape based Image Classification Objectives

1. Design a k-nearest-neighbour classifier
2. Perform shape based image classification
 - using the k-nearest-neighbour classifier
 - leave-one-out cross validation scheme

Dataset

- 15 images containing instances of 3 classes:

- heart,



- fork



- watch



Design a k-nearest-neighbour classifier

- 2D shape matching for a considered image shape with all the other 14 image shapes
- compute the bending energies E_1, \dots, E_{14}
- define the k-nearest-neighbours of the image shape based on E_1, \dots, E_{14}
- classify the considered image using the labels of the k-nearest-neighbours

Shape based Image Classification

- for every image in the dataset (15 images)
 - k-nearest-neighbour classifier using the other 14 images (labels are known)
- classification performance:
inferred label = ground truth label?

- Send your reports
 - explain main steps of your implementation
 - comment the results
 - answer the questions
- yawli@vision.ee.ethz.ch
- Them: Exercise 9: Shape Context + your name