

國立高雄應用科技大學



100-101年度獎勵科技大學及技術學院

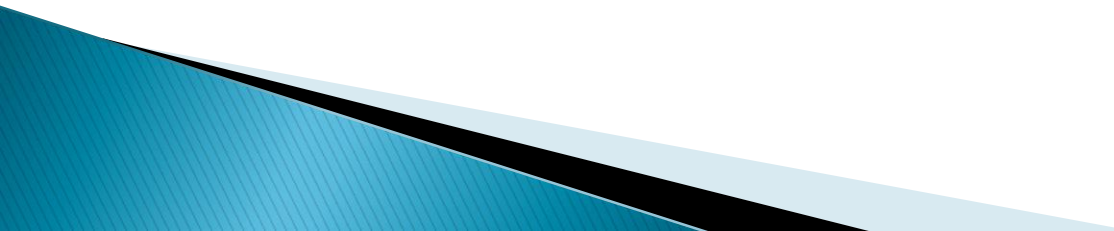
教學卓越計畫

產學及研究成果轉課程製作教材

電機工程系 黃敬群教授 學生黃得凱

A study of Computer-Aided Detection on
digital chest radiographs

Outline

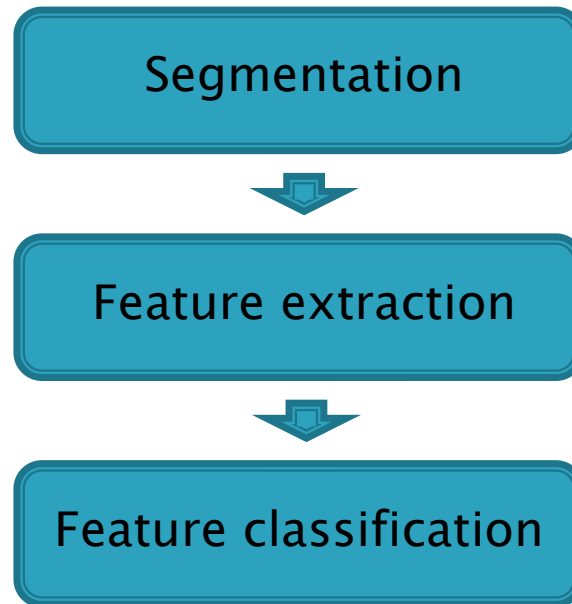
- ▶ Goal
 - ▶ Introduction
 - ▶ Methods
 - Segmentation
 - The ASM algorithm
 - Feature extraction
 - Gabor
 - Feature classification
 - SVM
 - ▶ Experiment
 - ▶ Test result
- 

Goal

- ▶ This paper presents an automatic computer-aided detection scheme on digital chest radiographs to detect disease.

Introduction

- ▶ Computer-Aided Detection (CAD)
 - Generally, a complete CAD scheme for chest consists of **three main steps**:



Segmentation



Segmentation is a necessary prerequisite process for **quantitative analysis applications**.



Feature extraction



Describing the image feature using geometrical feature, texture feature



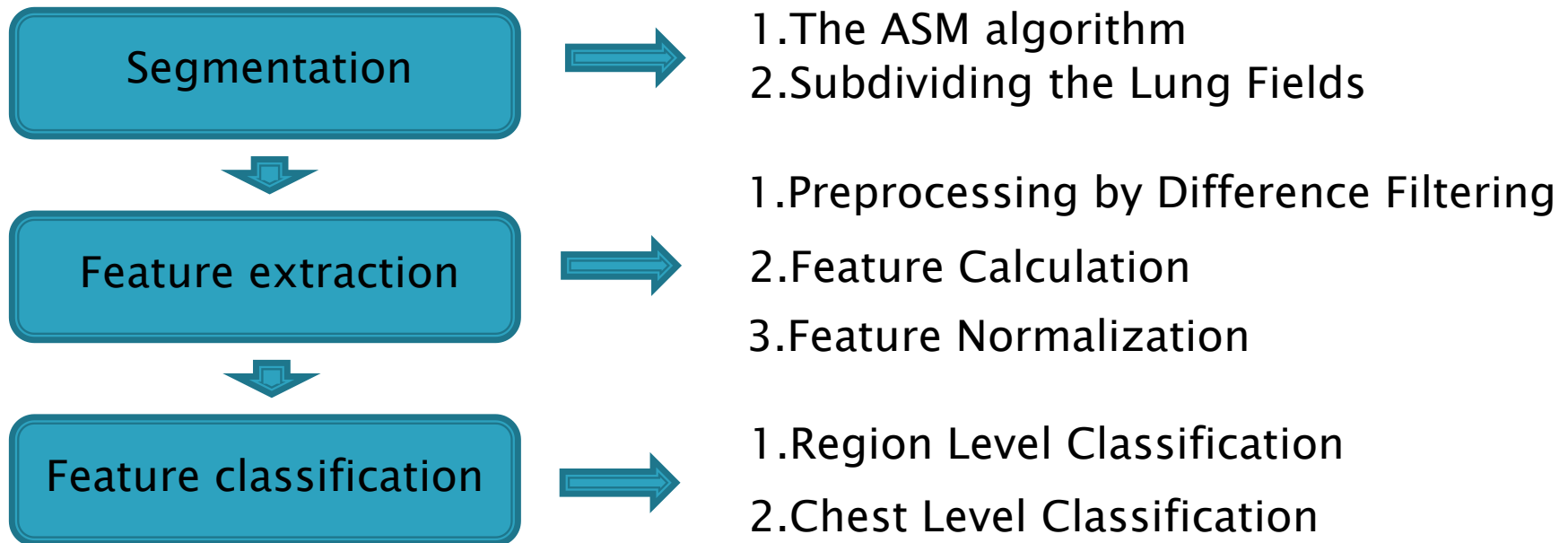
Feature classification



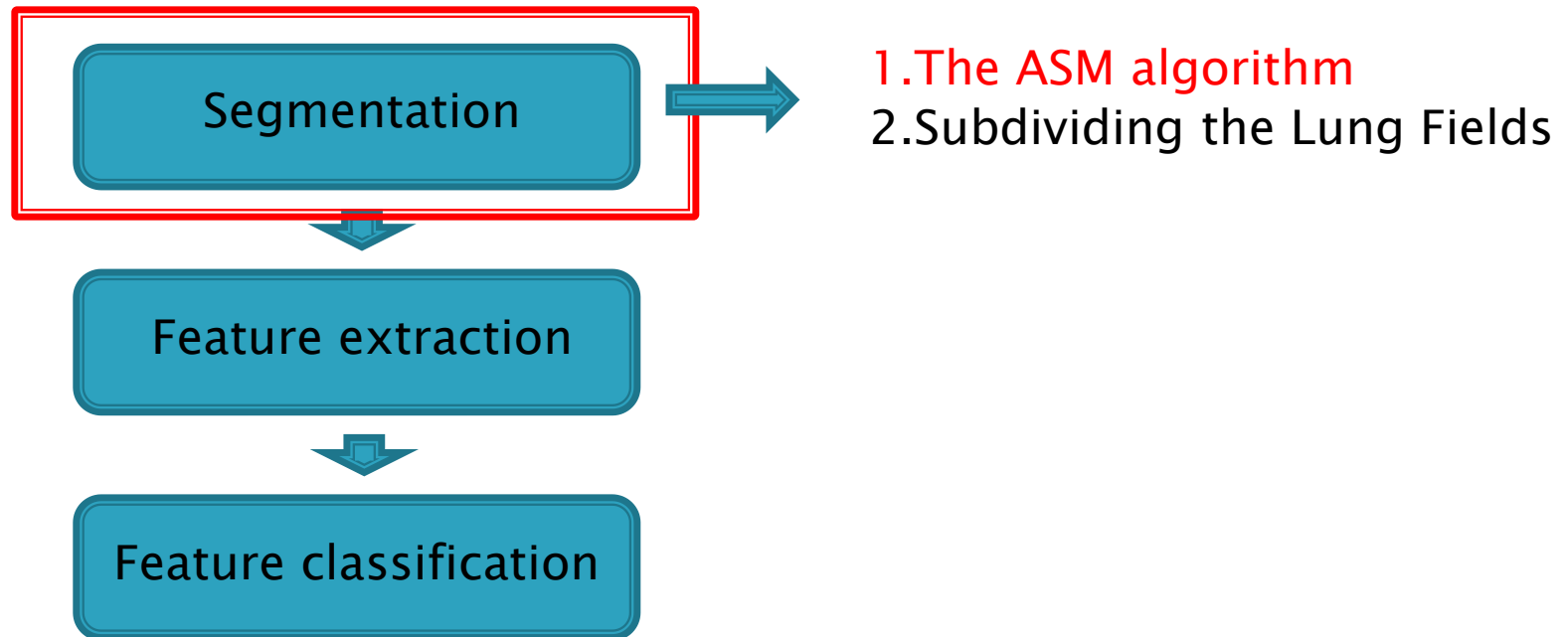
Could be employed to discriminate normal samples from abnormal.

Methods

In this paper, each main step have independent algorithm in block.

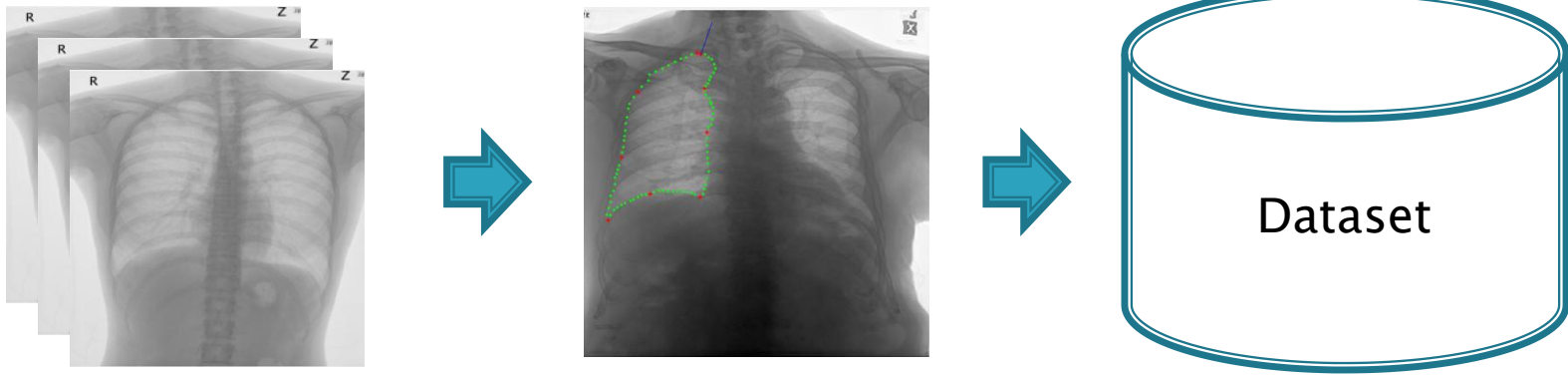


- ▶ Computer-Aided Detection (CAD)
 - Generally, a complete CAD scheme for chest consists of **three main steps**:



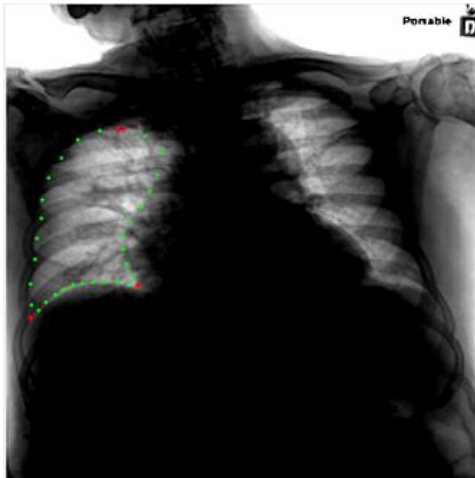
- ▶ The ASM algorithm could be mainly divided into two steps, training and segmentation.

Training phase



Label the lung contour landmarks on the training image

Example1 training

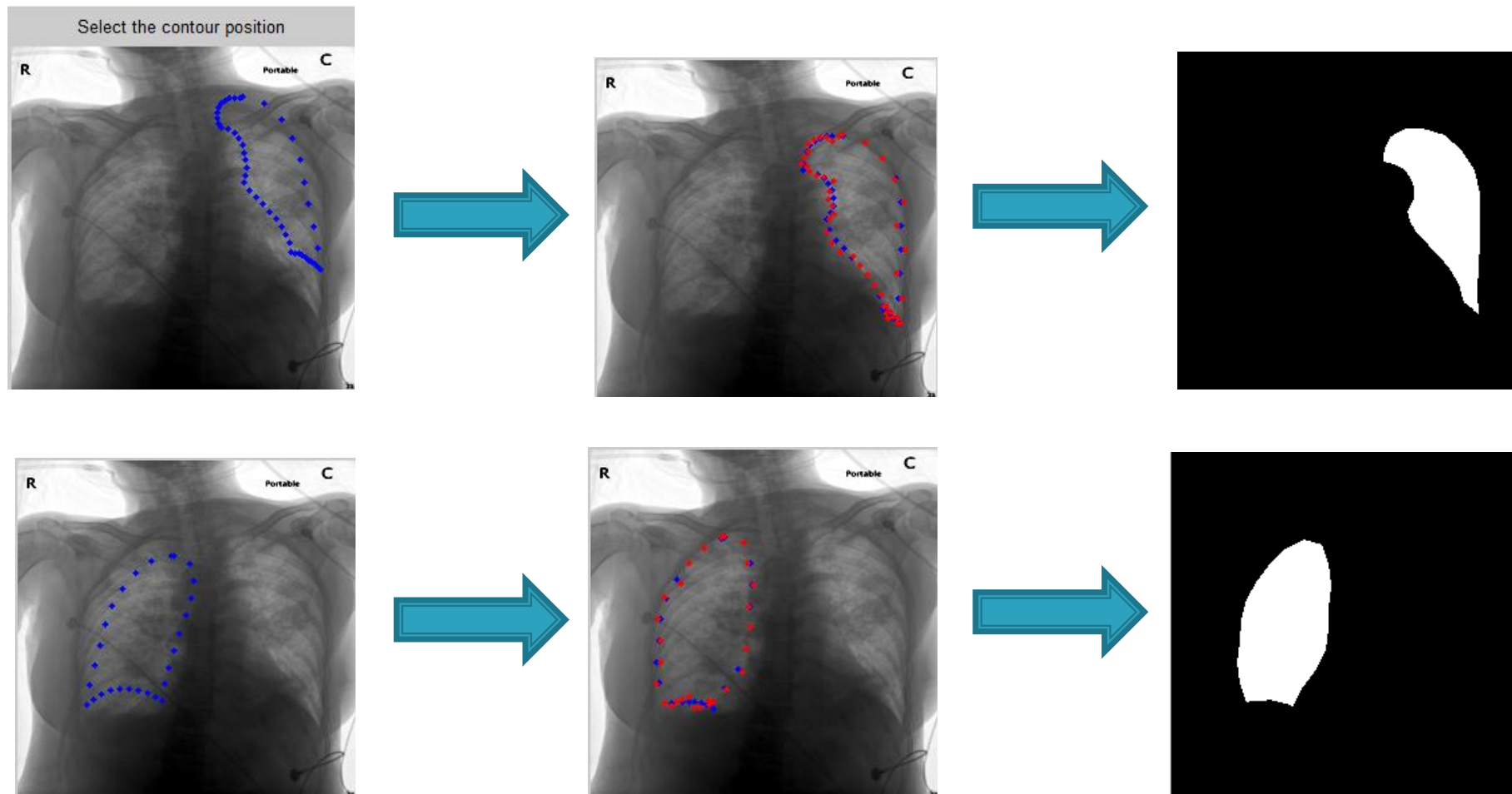


Right Lung



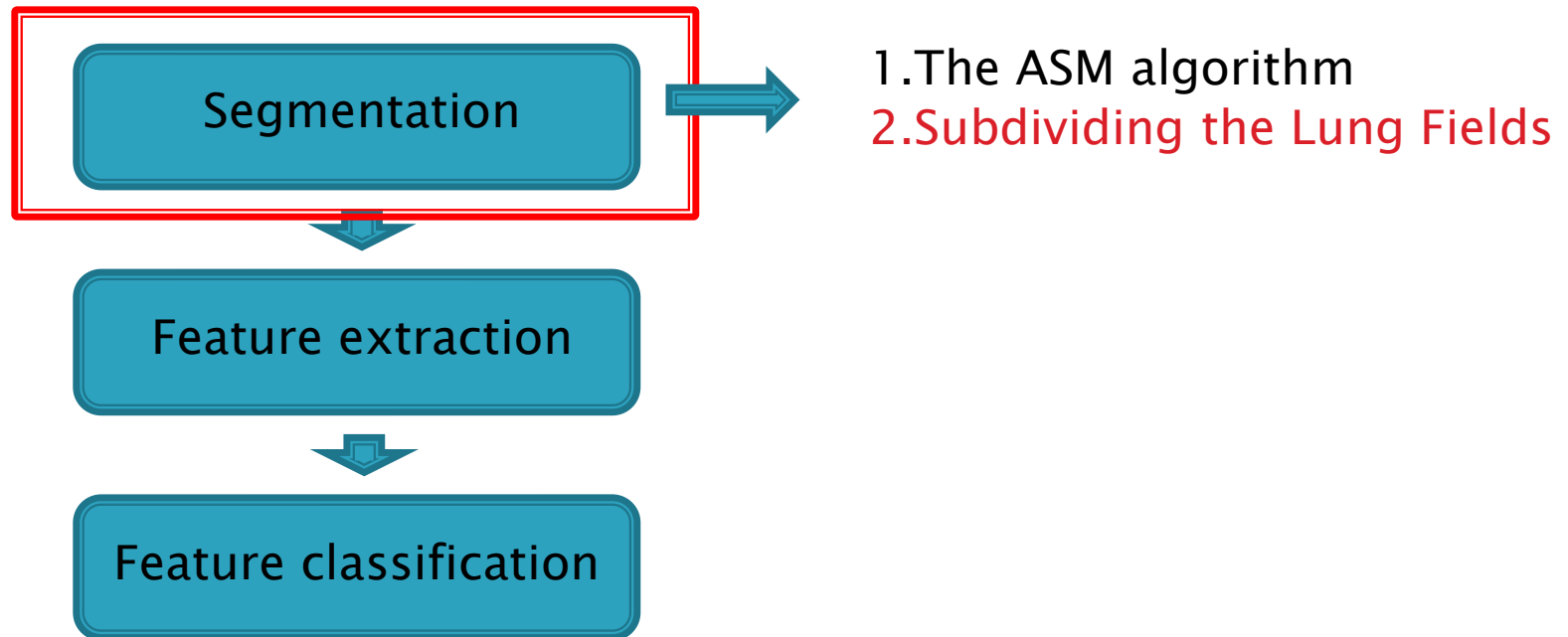
left Lung

Example2 testing



Finding the best position to recognition lung region.

- ▶ Computer-Aided Detection (CAD)
 - Generally, a complete CAD scheme for chest consists of **three main steps**:

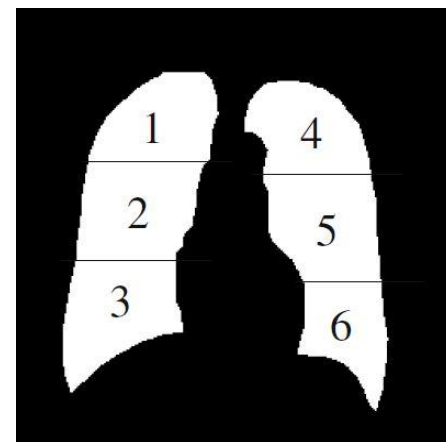


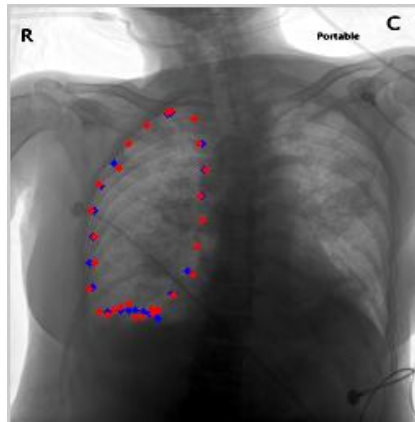
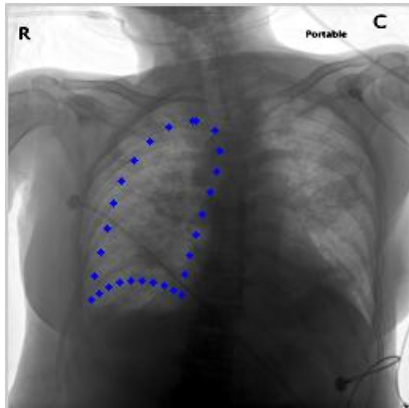
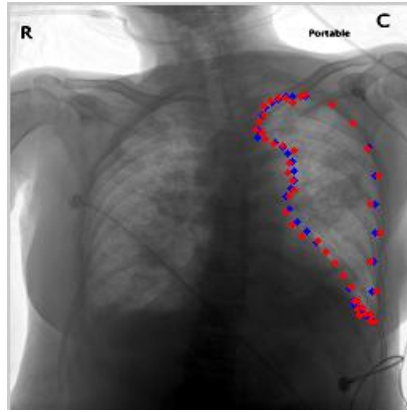
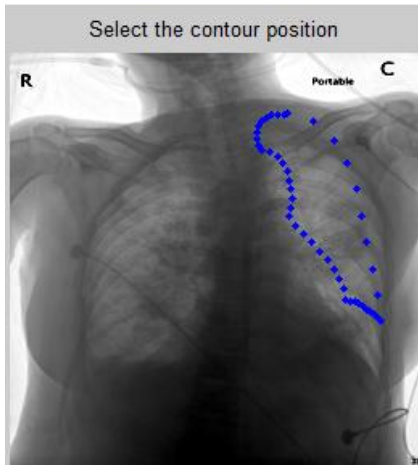
► Subdividing the Lung Fields into Six Zones

- Algorithm:

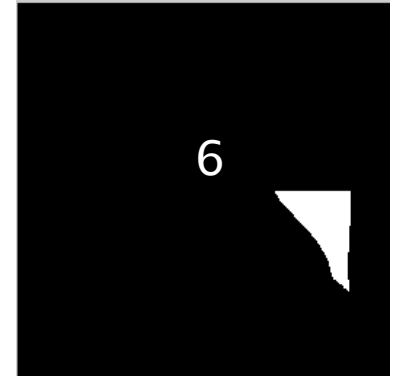
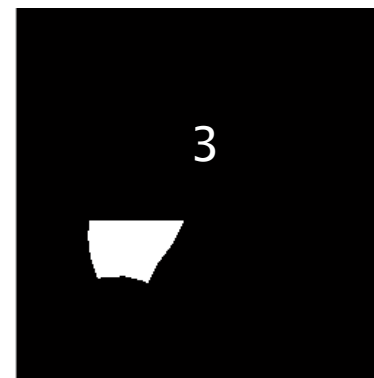
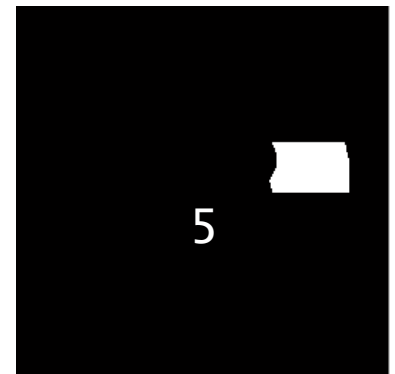
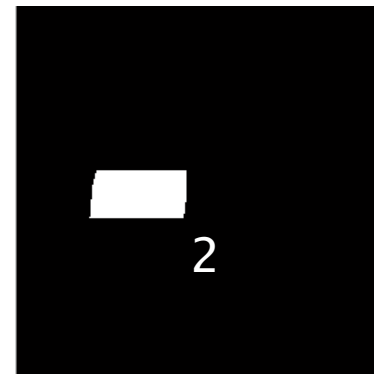
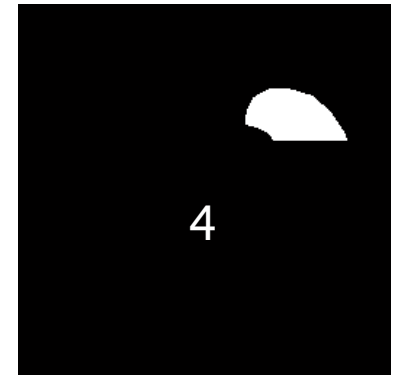
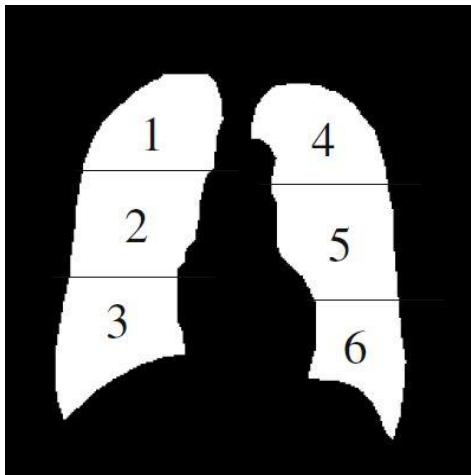
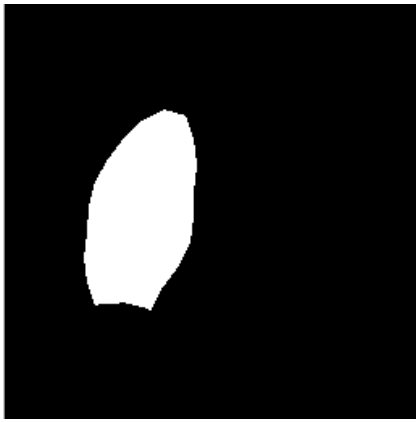
- 1. Computing lung area.
- 2. Subdividing the Lung Fields

| left lung | right lung |
|--------------------------|-----------------------------|
| apex and bottom position | apex and diaphragm position |
| horizontal lines | horizontal lines |



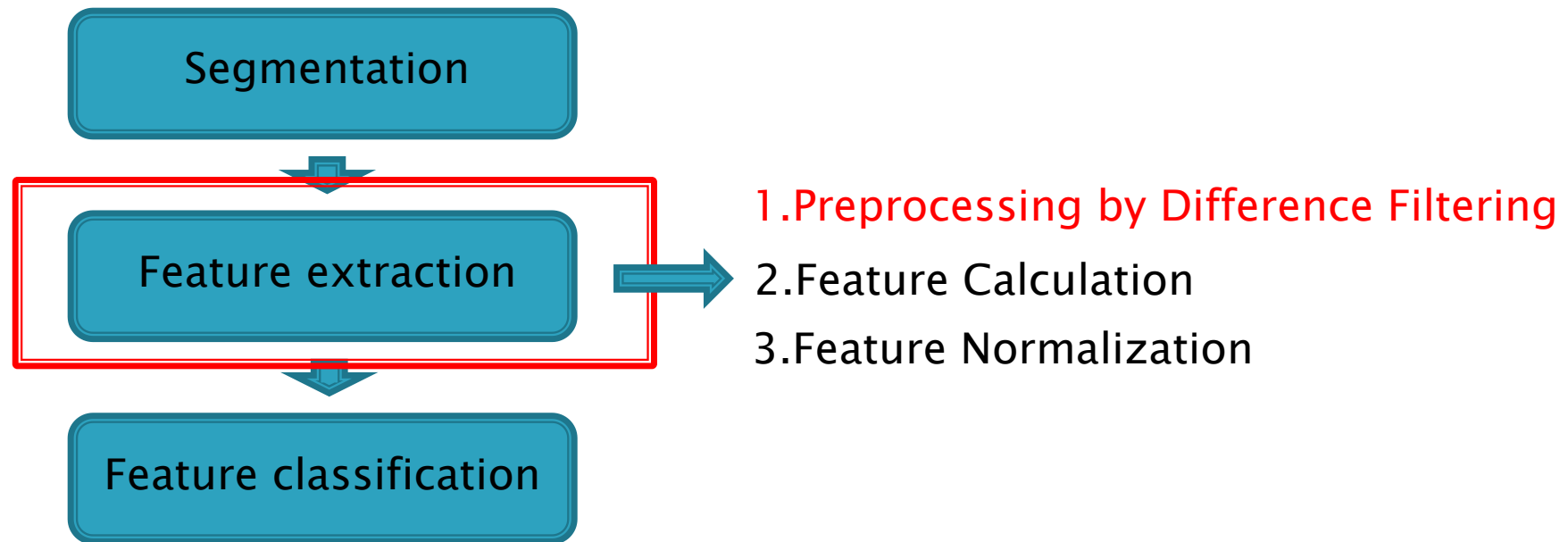


Finding the best position to recognition lung region.



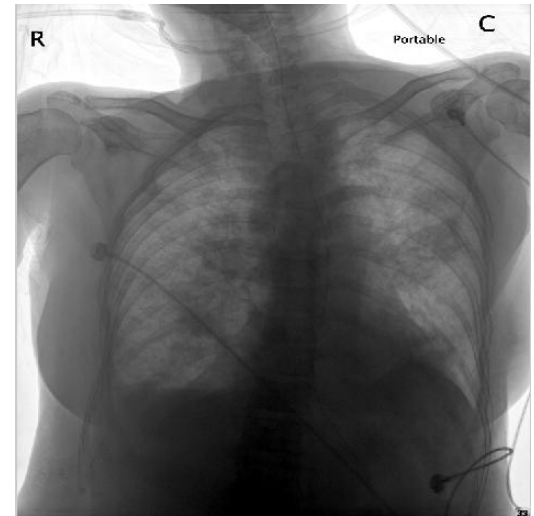
► Computer-Aided Detection (CAD)

- Generally, a complete CAD scheme for chest consists of three main steps:



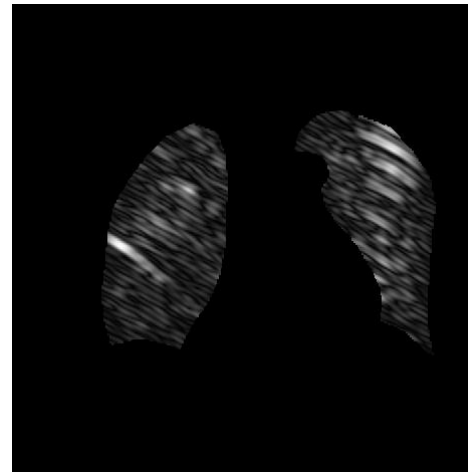
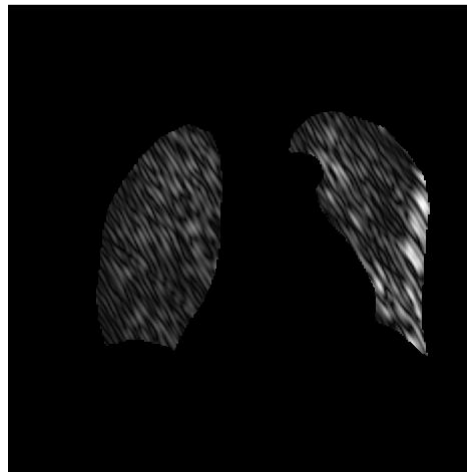
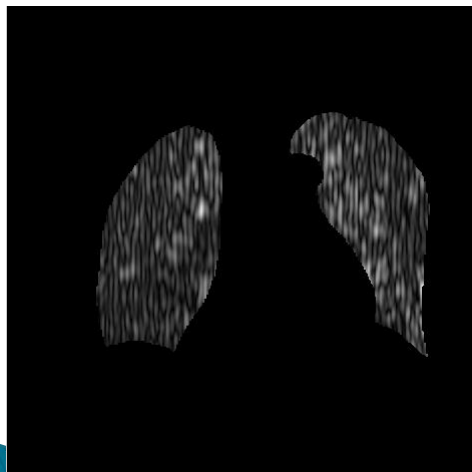
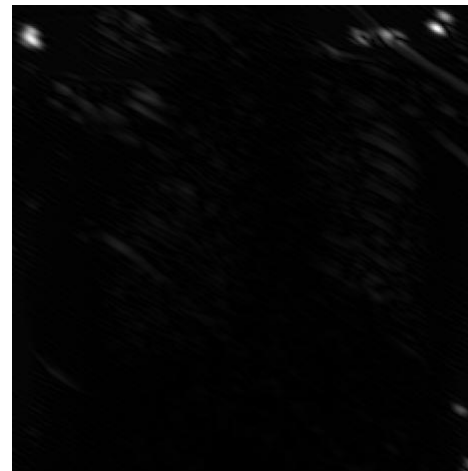
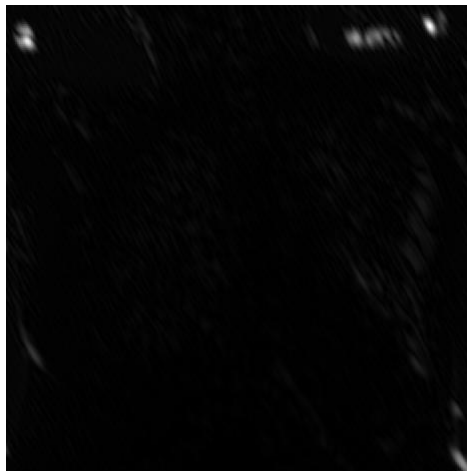
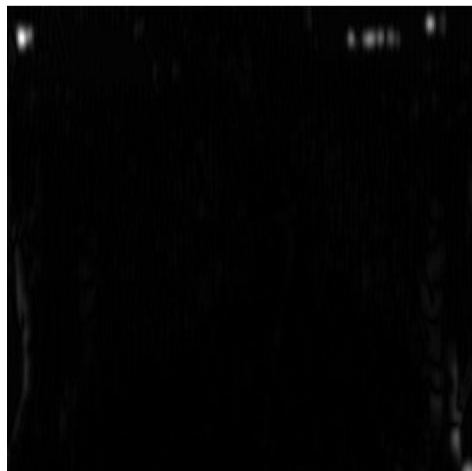
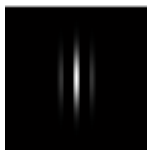
Preprocessing

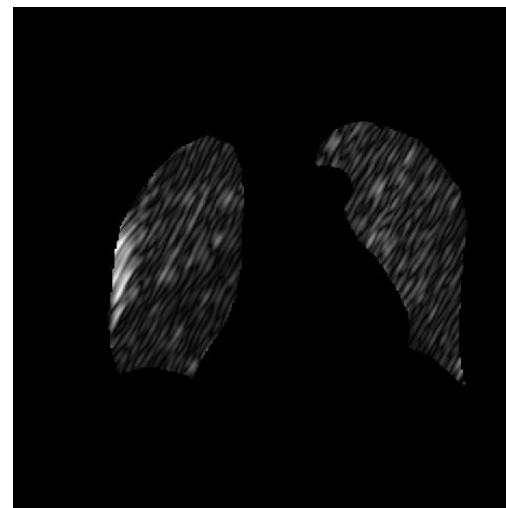
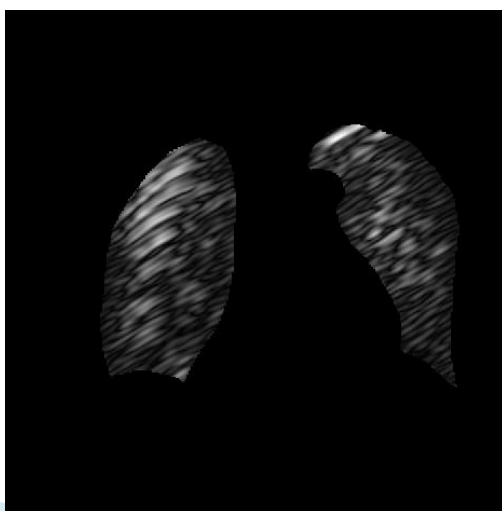
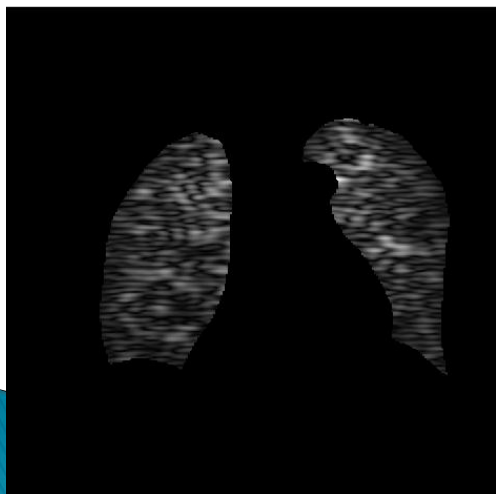
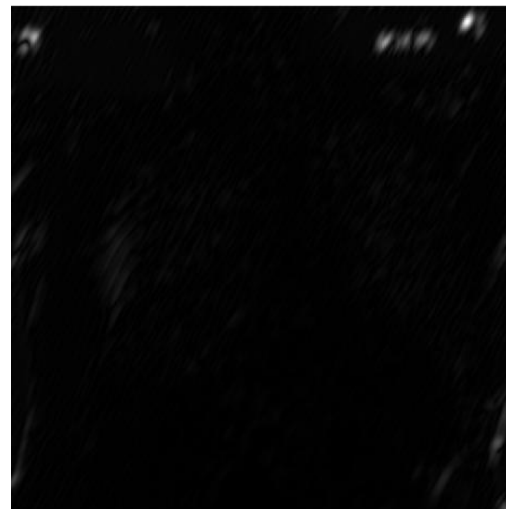
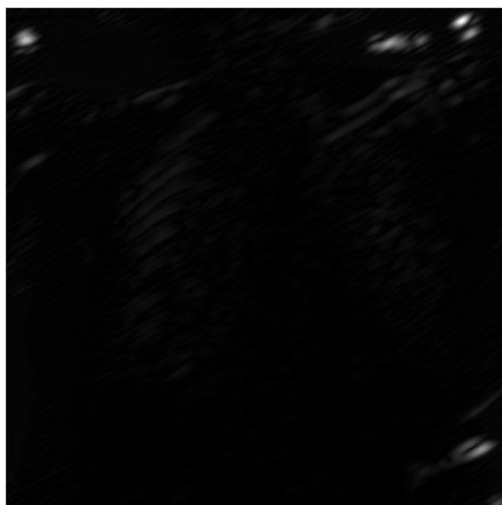
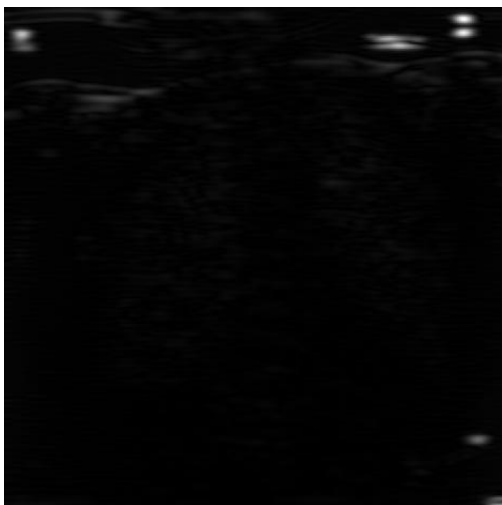
- ▶ Size normalization
 - ▶ 700*700
- ▶ Gabor filter bank
 - Six kernel:
 - $\text{Theta} = 0, \pi/6, \pi/3, 3\pi/2, 2\pi/3, 5\pi/6$
 - $\text{Lambda} = 8$



Kernel

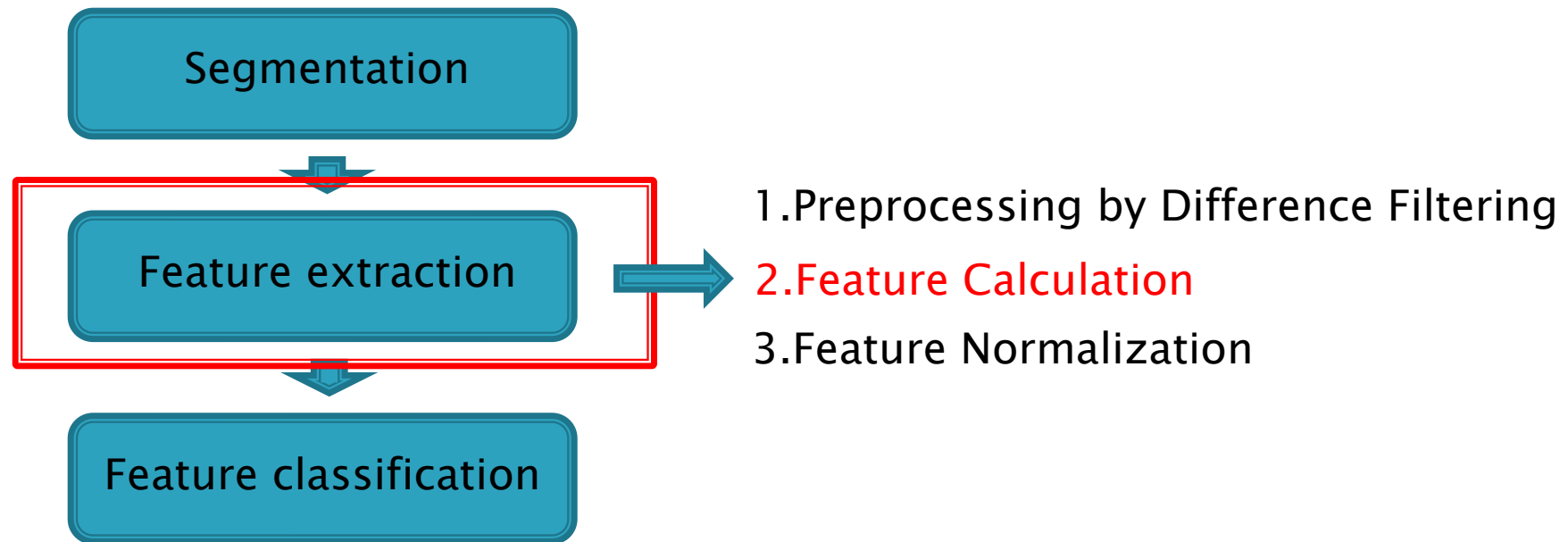






► Computer-Aided Detection (CAD)

- Generally, a complete CAD scheme for chest consists of three main steps:



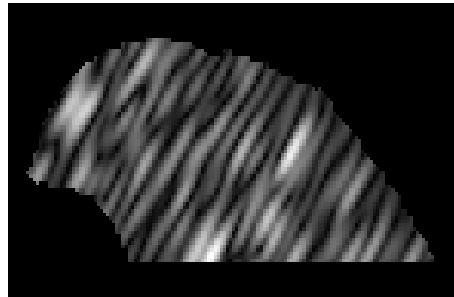
Feature Calculation

- ▶ Moment:
 - MEAN, Standard deviation, Variance, Skewness

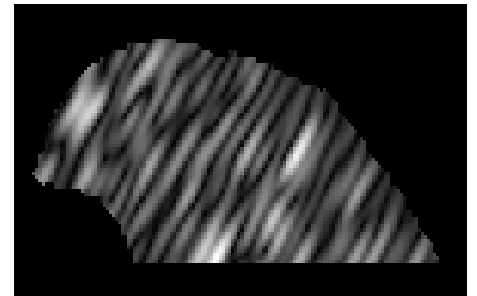
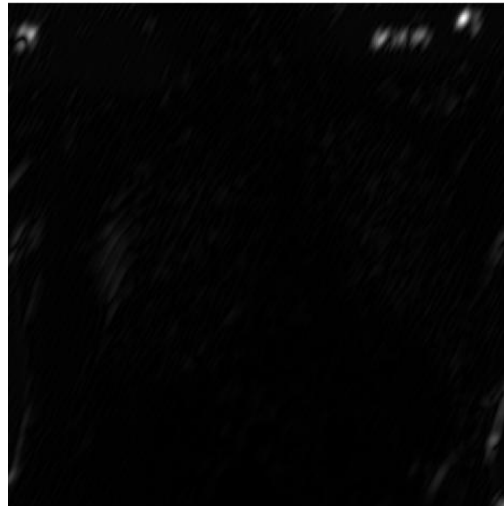
| Feature | Equation |
|--------------------|---|
| Mean | $M_1 = \sum i \cdot h(i)$ |
| Standard deviation | $M_2 = \sqrt{\sum (i - M_1)^2 \cdot h(i)}$ |
| Skewness | $M_3 = \frac{1}{M_2^3} \sum (i - M_1)^3 \cdot h(i)$ |
| Kurtosis | $M_4 = \frac{1}{M_2^4} \sum (i - M_1)^4 \cdot h(i) - 3$ |

Mean for each regions

▶ Feature 1:

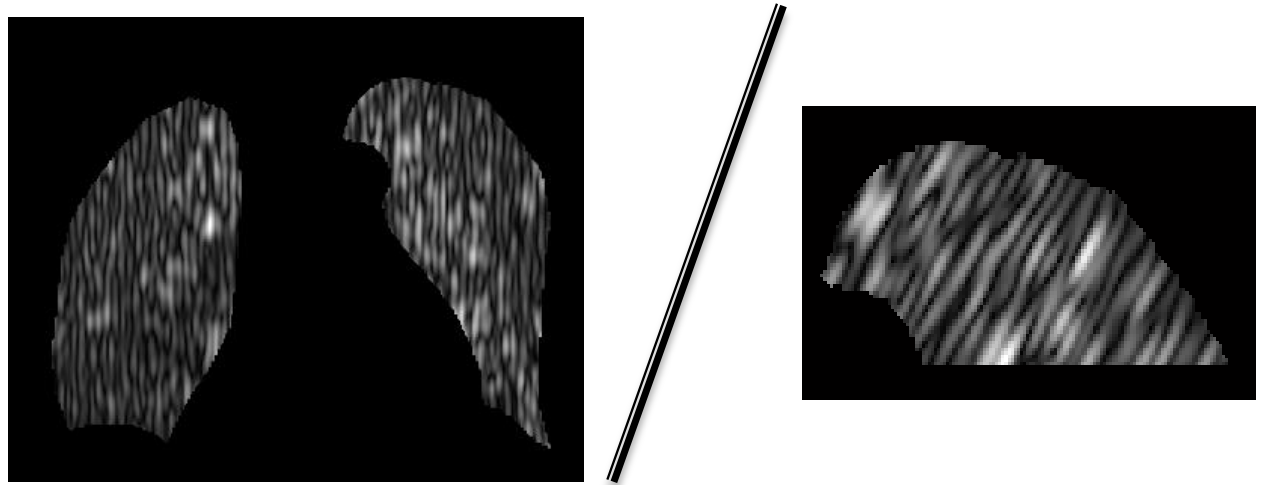


▶ Feature 2:

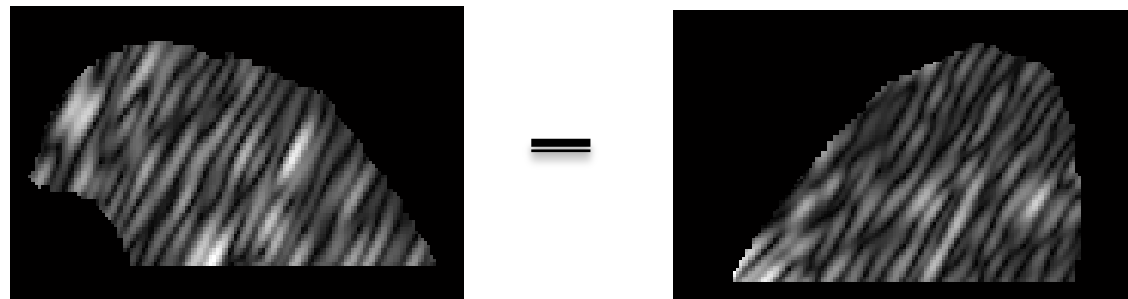


Mean for each regions

► Feature 3:



► Feature 4:



Data saving

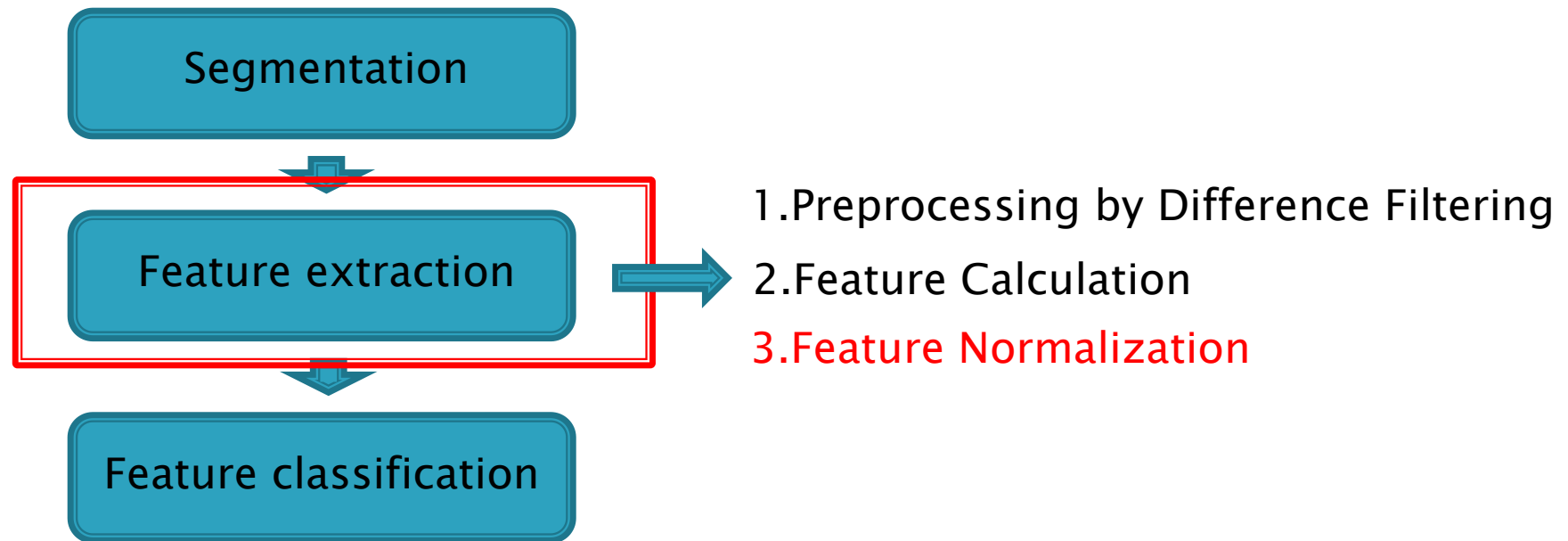
Each region field of lung



| | moment1 | moment2 | moment3 | moment4 |
|---------|-----------|-----------|-----------|-----------|
| kernel1 | Feature 1 | Feature 2 | Feature 3 | Feature 4 |
| kernel2 | Feature 1 | Feature 2 | Feature 3 | Feature 4 |
| kernel3 | Feature 1 | Feature 2 | Feature 3 | Feature 4 |
| kernel4 | Feature 1 | Feature 2 | Feature 3 | Feature 4 |
| kernel5 | Feature 1 | Feature 2 | Feature 3 | Feature 4 |
| kernel6 | Feature 1 | Feature 2 | Feature 3 | Feature 4 |

► Computer-Aided Detection (CAD)

- Generally, a complete CAD scheme for chest consists of three main steps:



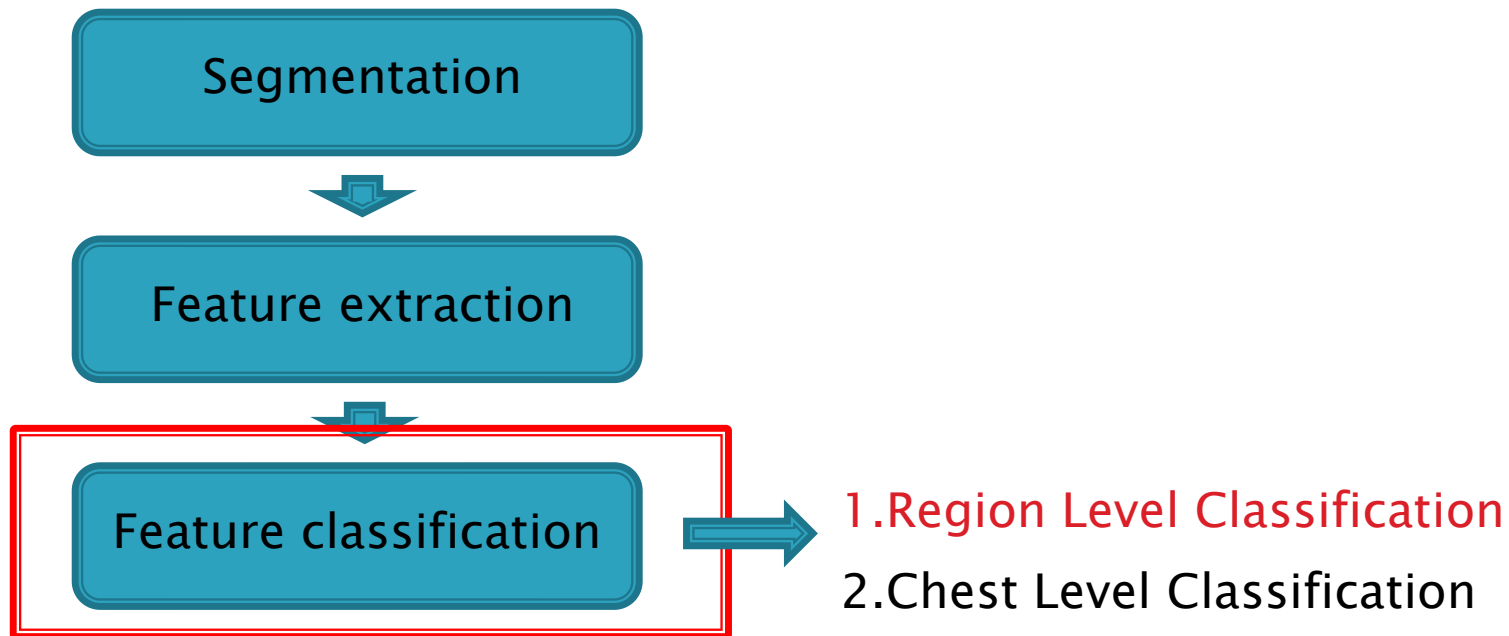
Feature Normalization

- ▶ Feature Normalization

- Equation:

$$X_{normal} = \frac{2X - (X_{max} + X_{min})}{X_{max} - X_{min}}$$

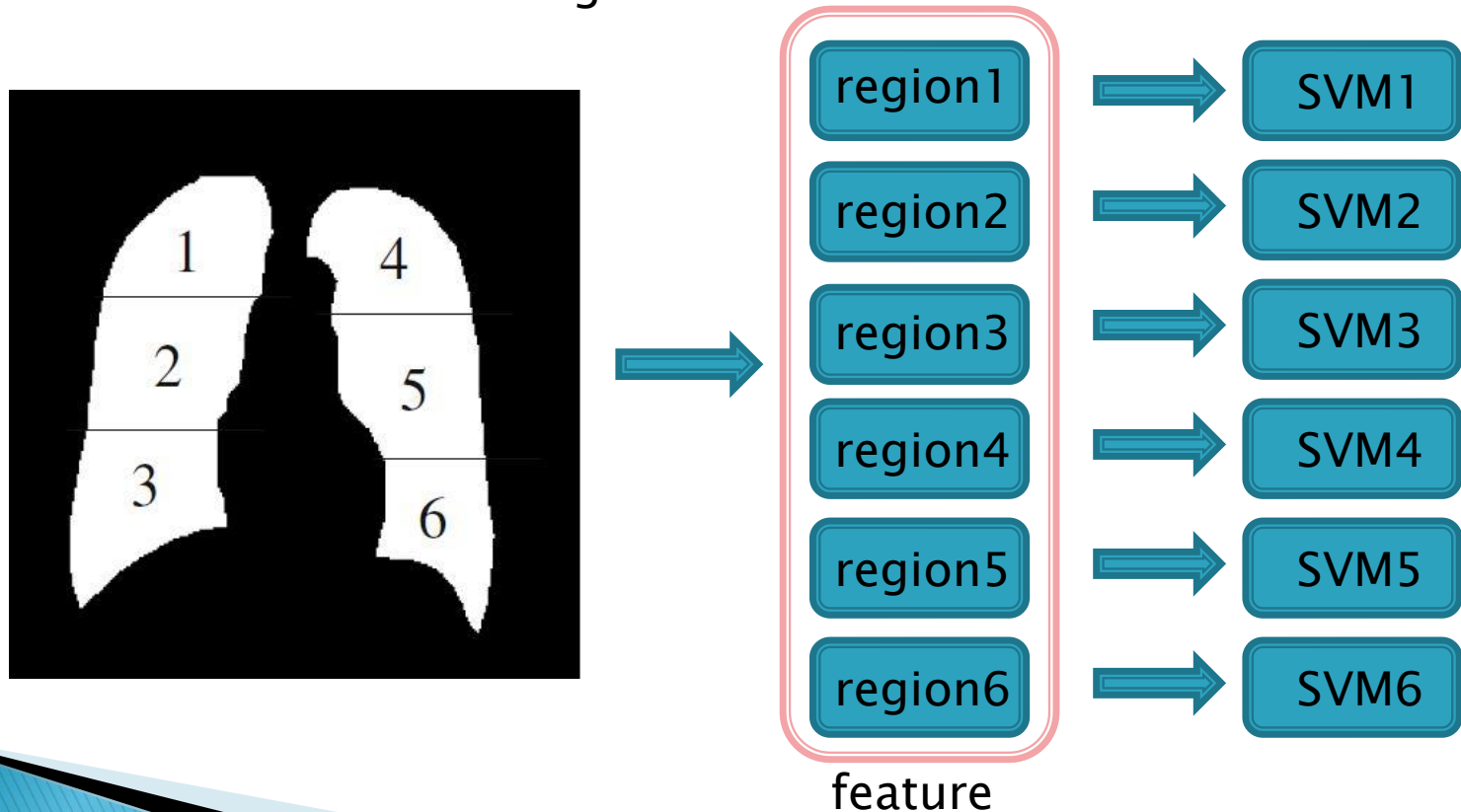
- ▶ Computer-Aided Detection (CAD)
 - Generally, a complete CAD scheme for chest consists of three main steps:



Feature classification

▶ Region Level Classification

These feature vectors are used by the classifier to estimate the abnormalities of each region.



Feature classification

► Chest Level Classification

- A chest level classifier is employed to integrate the probabilities of the six regions to obtain the classification result for the whole image.

| Methods | Weighted | Probability |
|--------------------------|--|---|
| The weighted voting | $W_i = \frac{A_{v_i}}{\sum_{i=1}^N A_{v_i}}$ | $\text{Pr} = \sum_{i=1}^N W_i P_i$ |
| The weighted multiplying | $f_i = \text{MAX} \left(\frac{A_{v_i} - T_{Av}}{1 - T_{Av}}, 0 \right)$ | $\text{Pr} = 1 - \prod_{i=1}^N (1 - f_i P_i)$ |

SYMBOL

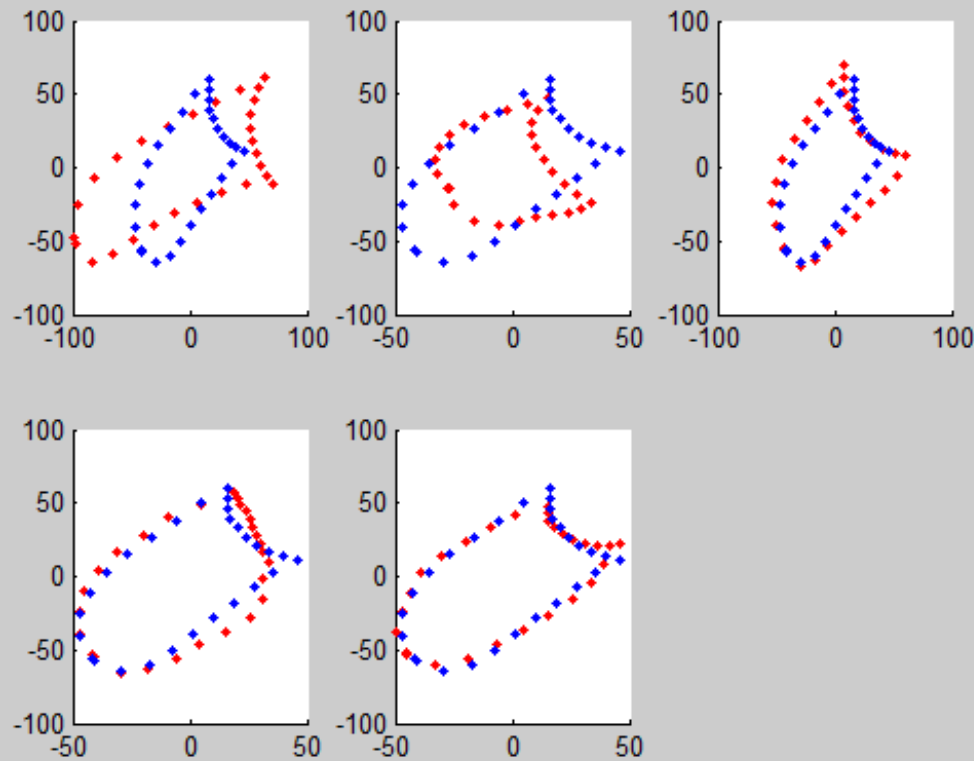
The area under the ROC curve (AUC) value is denoted as **Av**.(range from 0 to 1)

T_{Av} denote the threshold to determines the minimum value A_{v_i} .

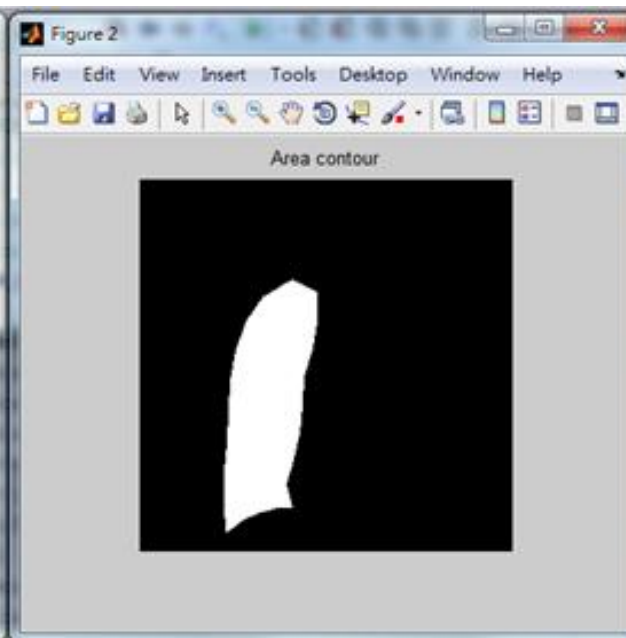
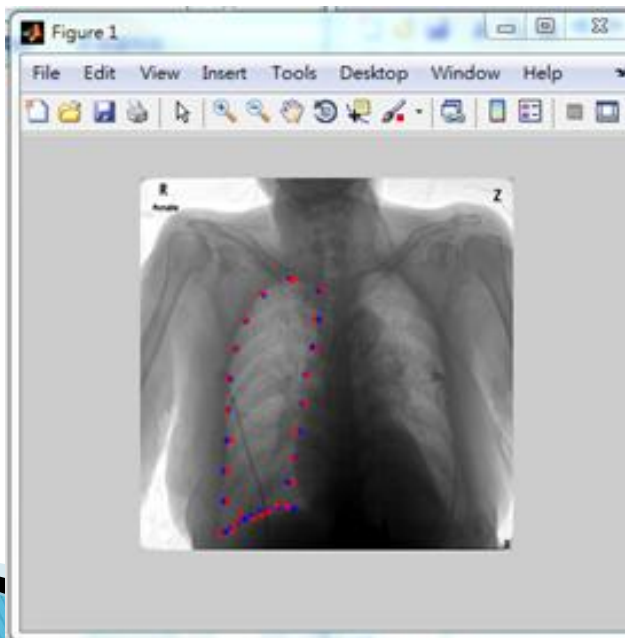
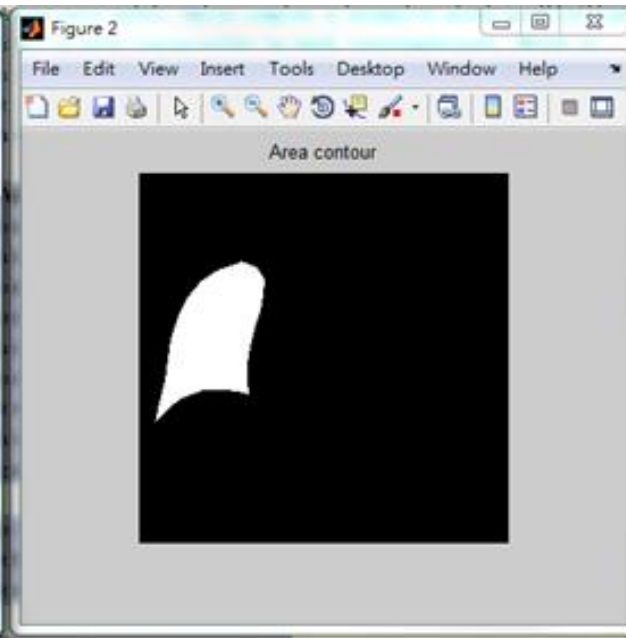
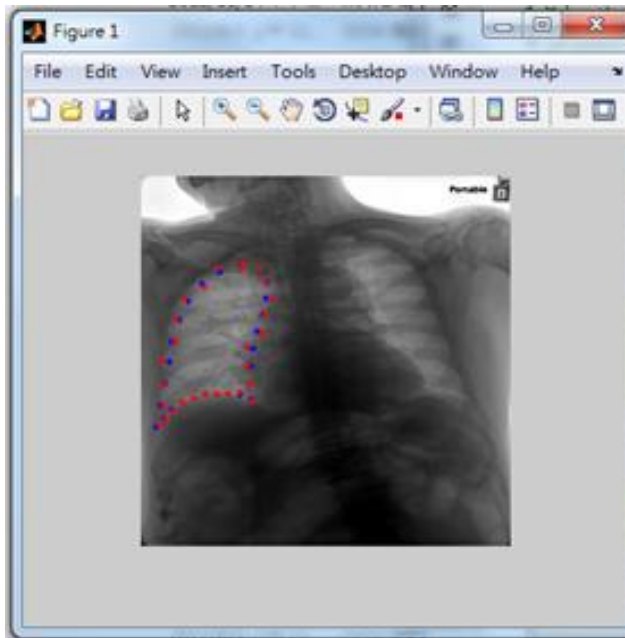
| Methods | Weighted & Weighted factor | Probability |
|--------------------------|--|---|
| The weighted voting | $W_i = \frac{A_{v_i}}{\sum_{i=1}^N A_{v_i}}$ | $\text{Pr} = \sum_{i=1}^N W_i P_i$ |
| The weighted multiplying | $f_i = \text{MAX} \left(\frac{A_{v_i} - T_{Av}}{1 - T_{Av}}, 0 \right)$ | $\text{Pr} = 1 - \prod_{i=1}^N (1 - f_i P_i)$ |

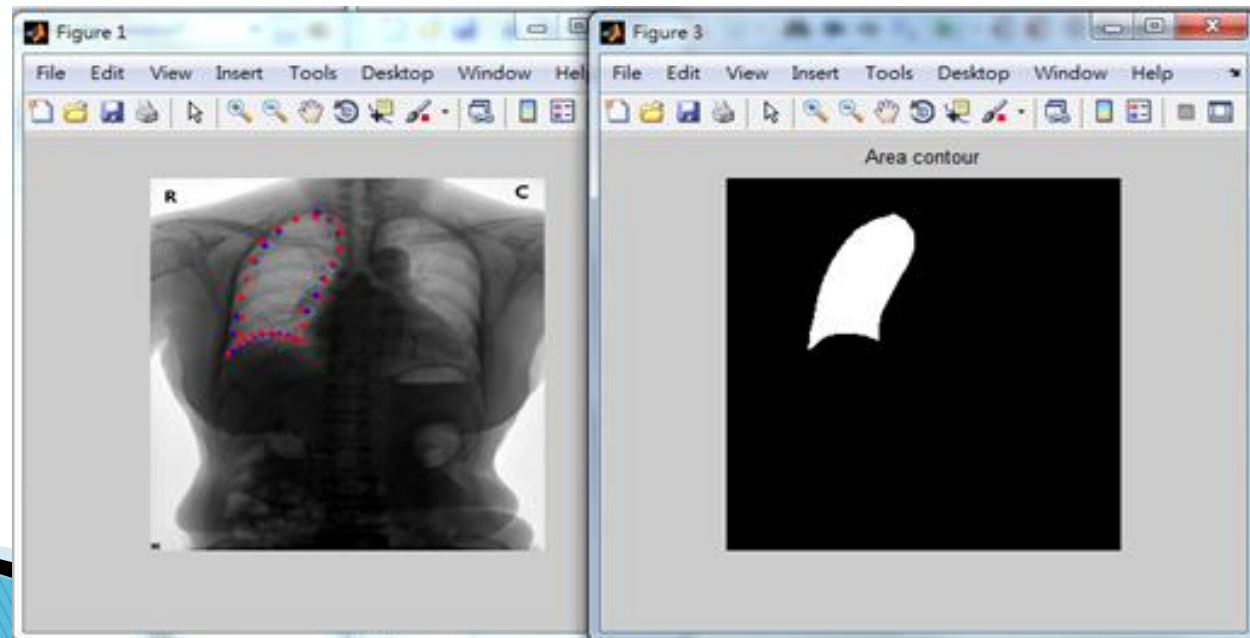
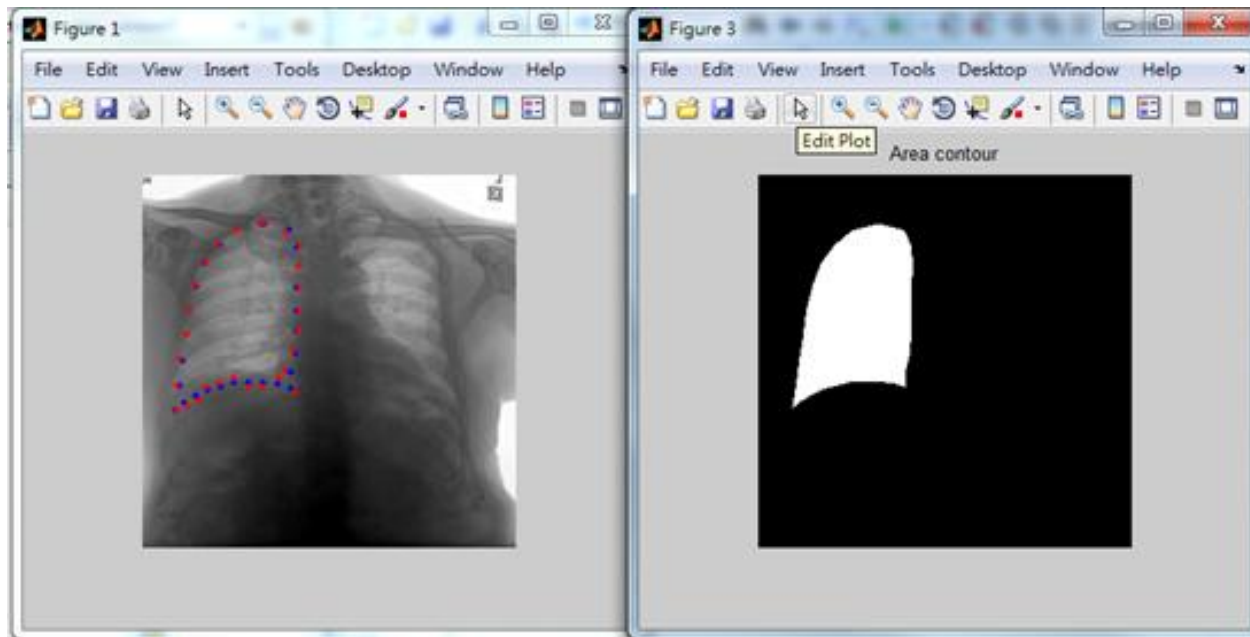
Experiment

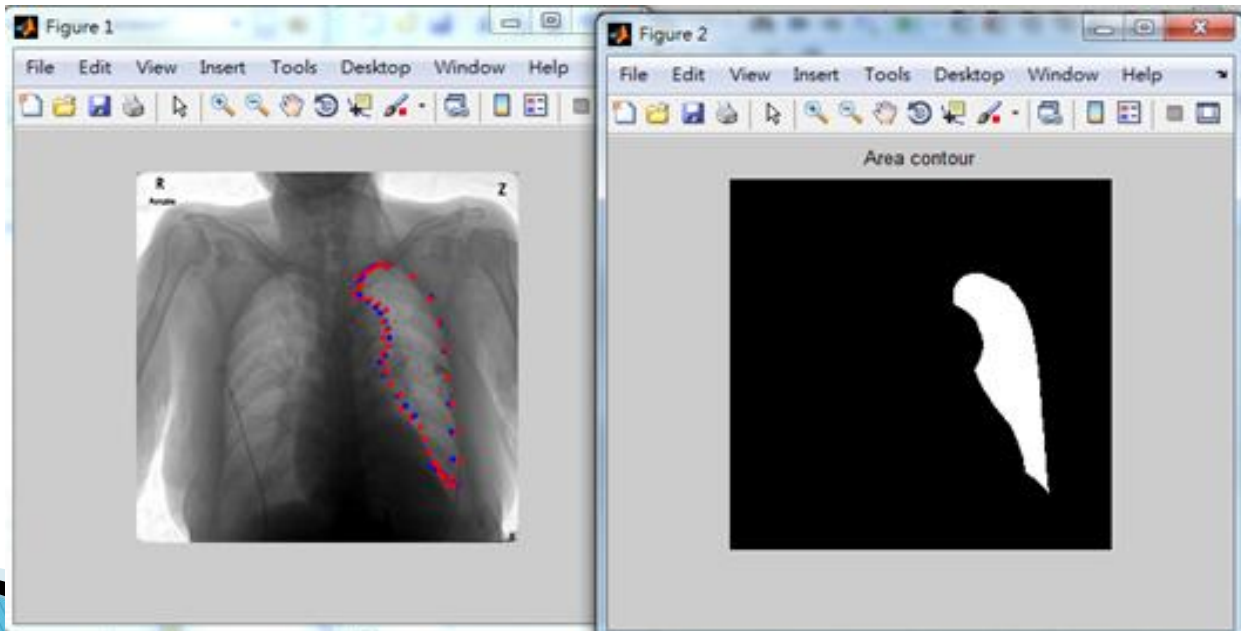
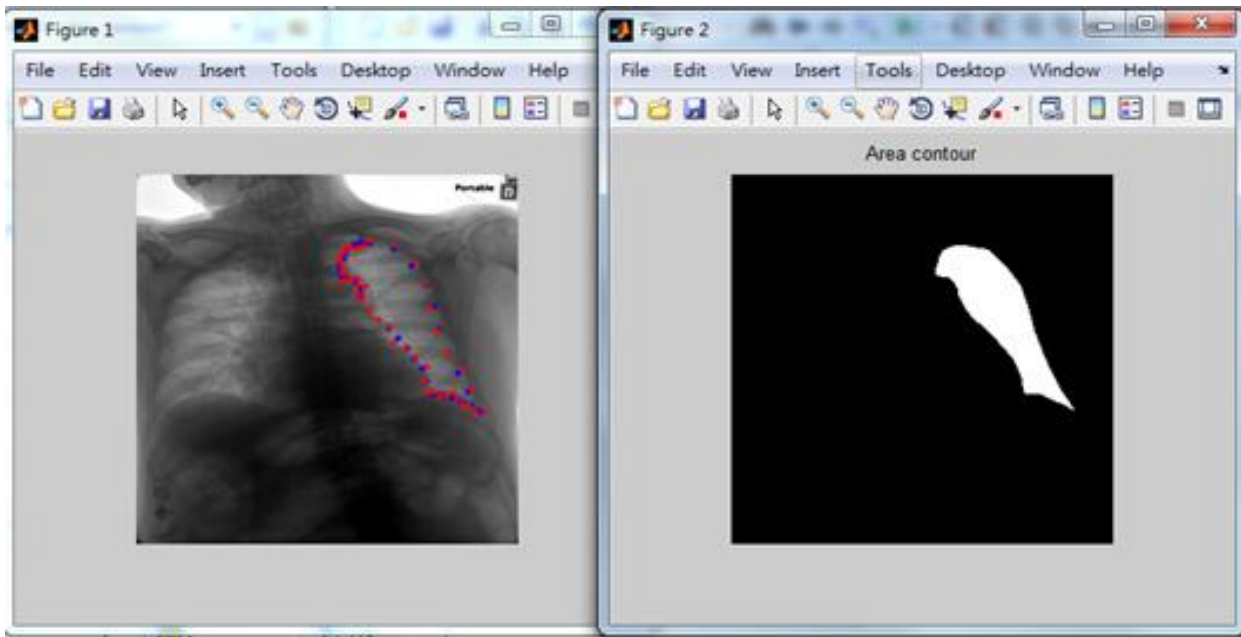
► ASM Right Lung

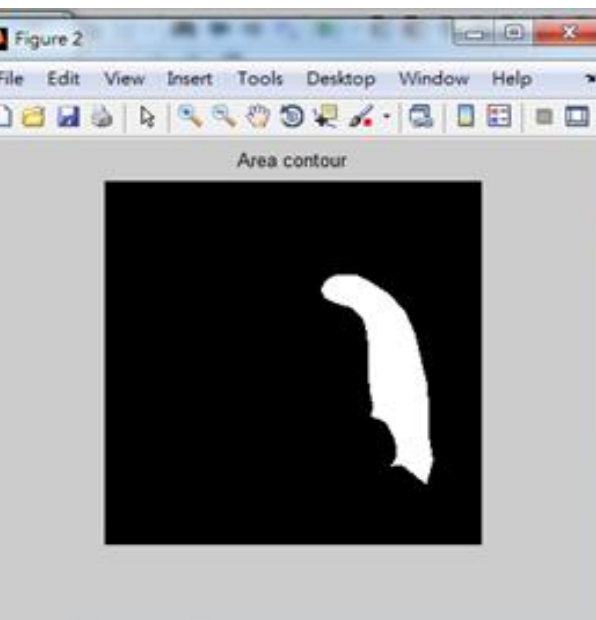
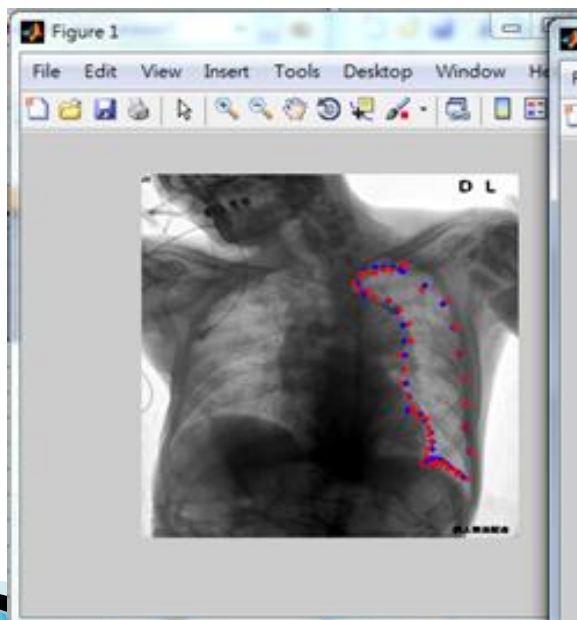
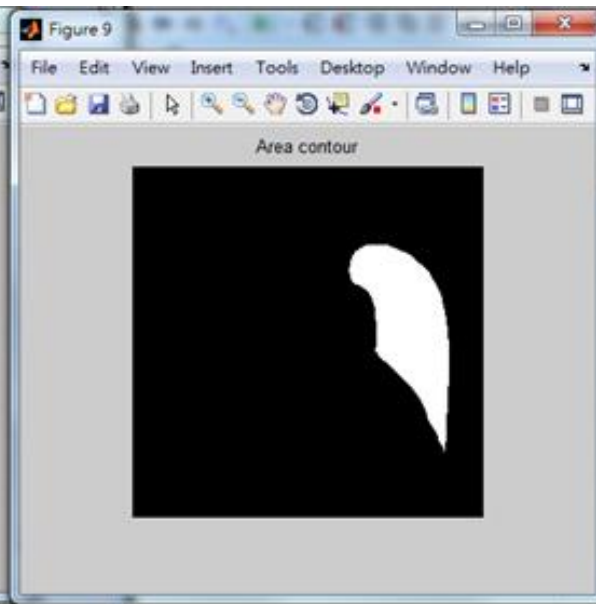
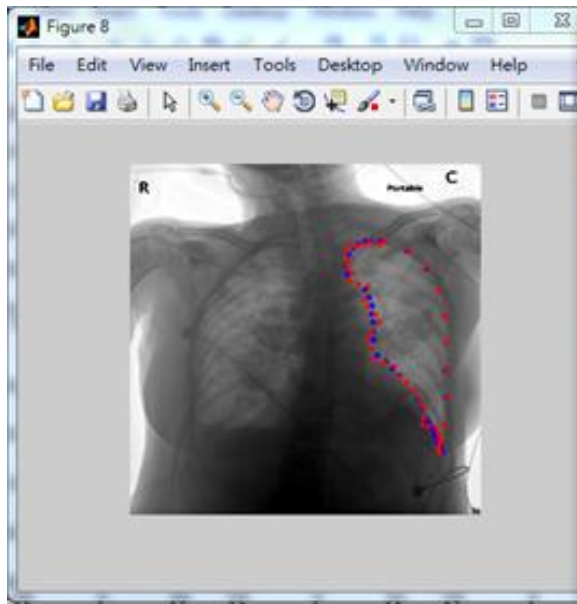


Red : shape model
Blue : mean model







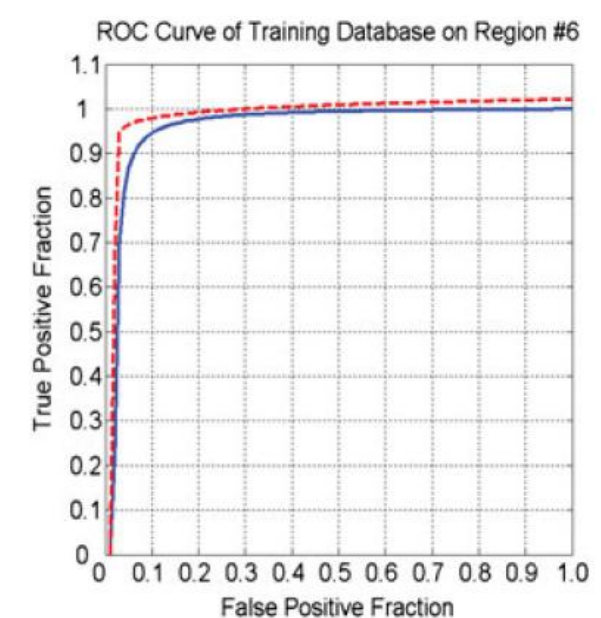
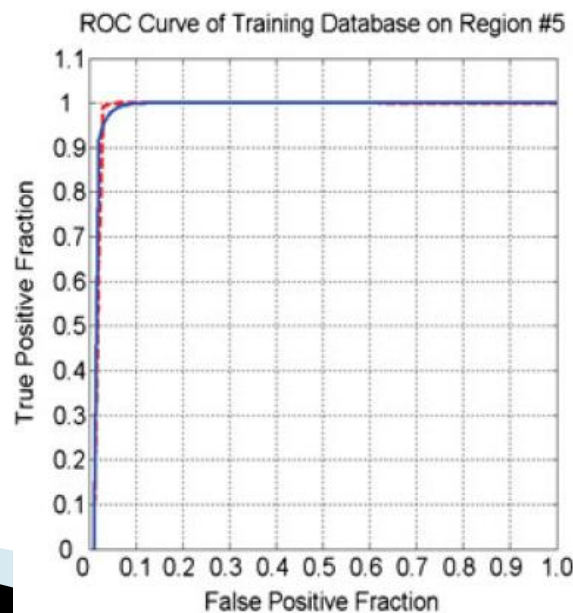
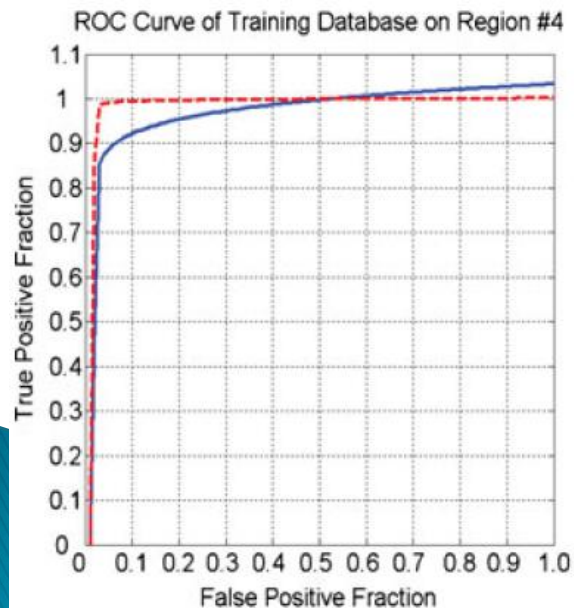
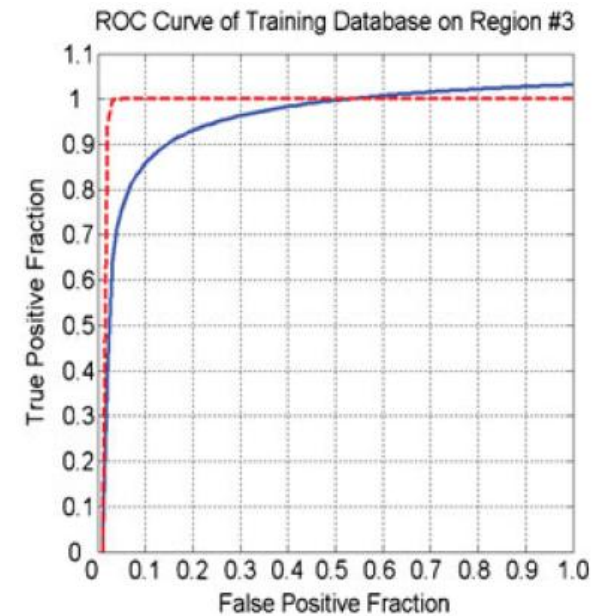
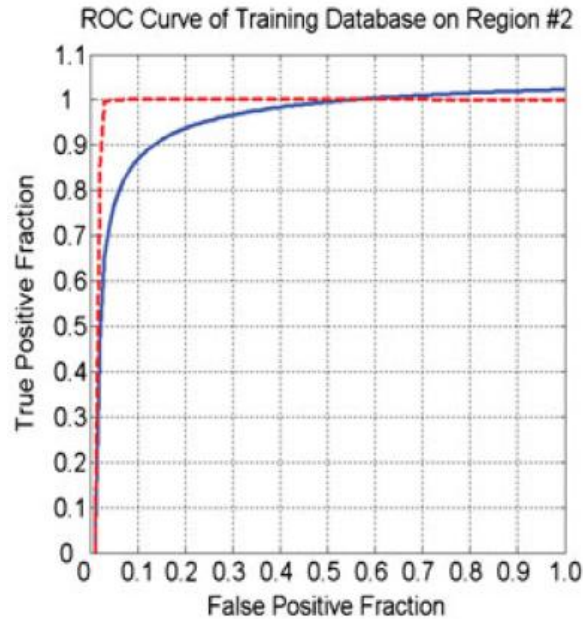
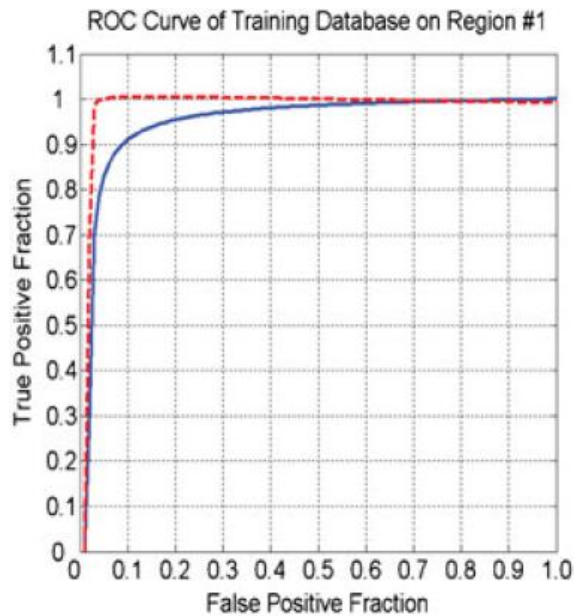


Test result

| Feature Vector | Region # | lgC | lg Γ | SSAve | Acc | Ses | Spe | AUC value |
|----------------|----------|-----|-------------|-------|-------|-------|-------|-----------|
| Full | R1 | 9 | -11 | 0.912 | 0.908 | 0.920 | 0.903 | 0.998 |
| Selected | R1 | 11 | 5 | 0.878 | 0.882 | 0.862 | 0.894 | 0.968 |
| Full | R2 | 7 | -9 | 0.923 | 0.925 | 0.920 | 0.926 | 0.999 |
| Selected | R2 | 11 | -7 | 0.879 | 0.889 | 0.851 | 0.907 | 0.970 |
| Full | R3 | 9 | -9 | 0.928 | 0.922 | 0.944 | 0.913 | 0.999 |
| Selected | R3 | 7 | -5 | 0.884 | 0.892 | 0.862 | 0.905 | 0.971 |
| Full | R4 | 7 | -9 | 0.895 | 0.892 | 0.904 | 0.887 | 0.998 |
| Selected | R4 | 11 | -5 | 0.873 | 0.875 | 0.866 | 0.879 | 0.988 |
| Full | R5 | 11 | -11 | 0.924 | 0.924 | 0.921 | 0.927 | 0.998 |
| Selected | R5 | 11 | -3 | 0.889 | 0.880 | 0.911 | 0.867 | 0.998 |
| Full | R6 | 9 | -11 | 0.896 | 0.892 | 0.905 | 0.885 | 0.993 |
| Selected | R6 | 11 | -5 | 0.878 | 0.875 | 0.886 | 0.870 | 0.979 |

Classification Results of Training Set on Feature Vectors for Each Region Using Leave-One-Out Cross-Validation Method.

ROC curves for the training set for the six regions on feature vectors.



REFERENCES

- ▶ [1] Ginneken BV, Romeny BM, Viergever MA: Computeraided diagnosis in chest radiography: a survey. IEEE Trans MedImag 20(12):1228–1241, 2001.
- ▶ [2] Peichun Yu, Hao Xu, Ying Zhu, Chao Yang, Xiwen Sun, and Jun Zhao1 An Automatic Computer Aided Detection Scheme for Pneumoconiosis on Digital Chest Radiographs.
- ▶ [3] Jen Hong Tan, U. Rajendra Acharya, Collin Tan, K. Thomas Abraham, Choo Min LimComputer-Assisted Diagnosis of Tuberculosis: A First Order Statistical Approach to Chest Radiograph.
- ▶ [4] Kobatake H, Ohishi K, Miyamichi J: Automatic diagnosis of pneumoconiosis by texture analysis of chest X-ray images. IEEE ICASSP 12:610–613, 1987.
- ▶ [5] Soliz P, Pattichis MS, Ramachandran J. Computerassisted diagnosis of chest radiographs for pneumoconioses. Proceedings of SPIE 667–675, 2001