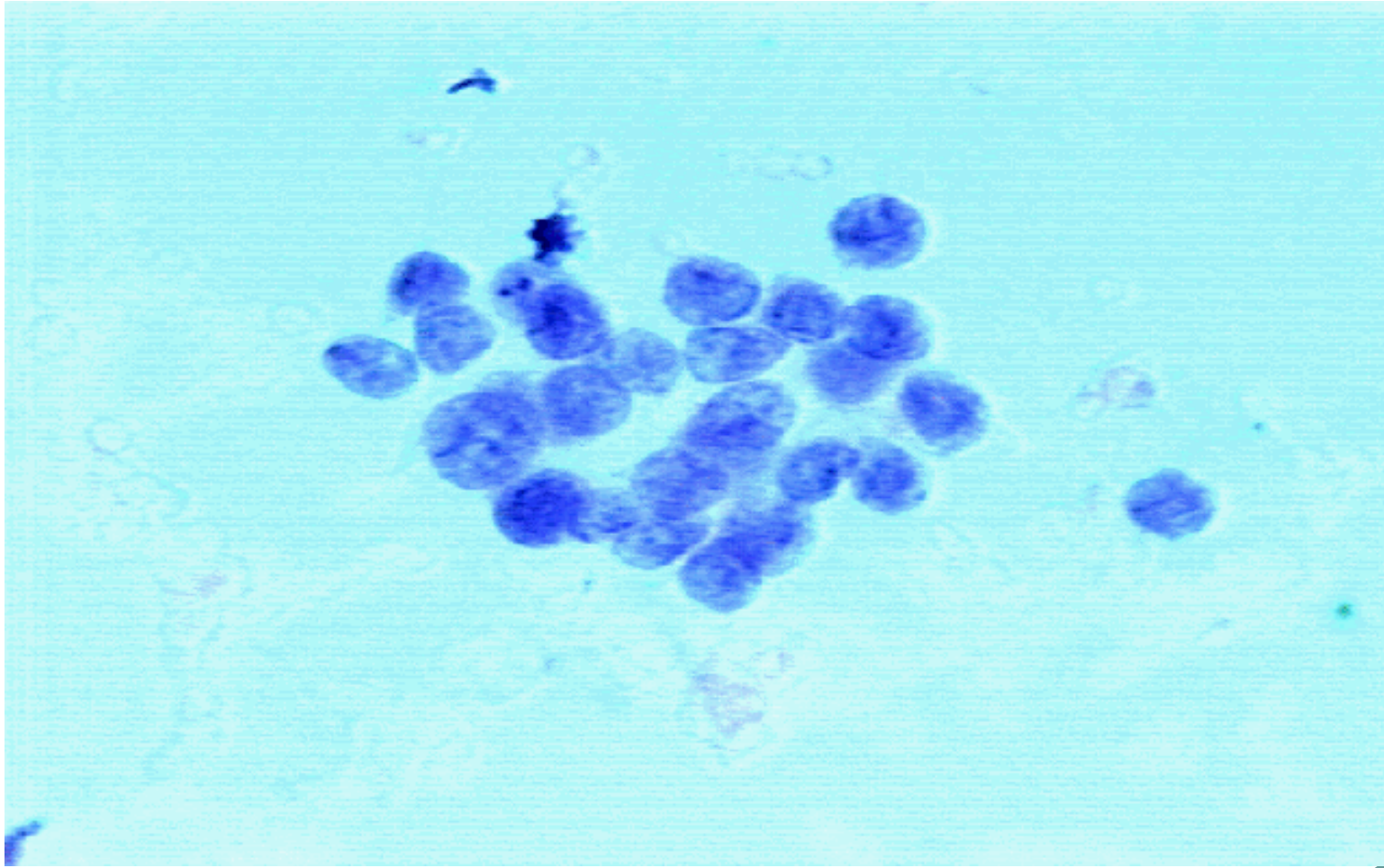
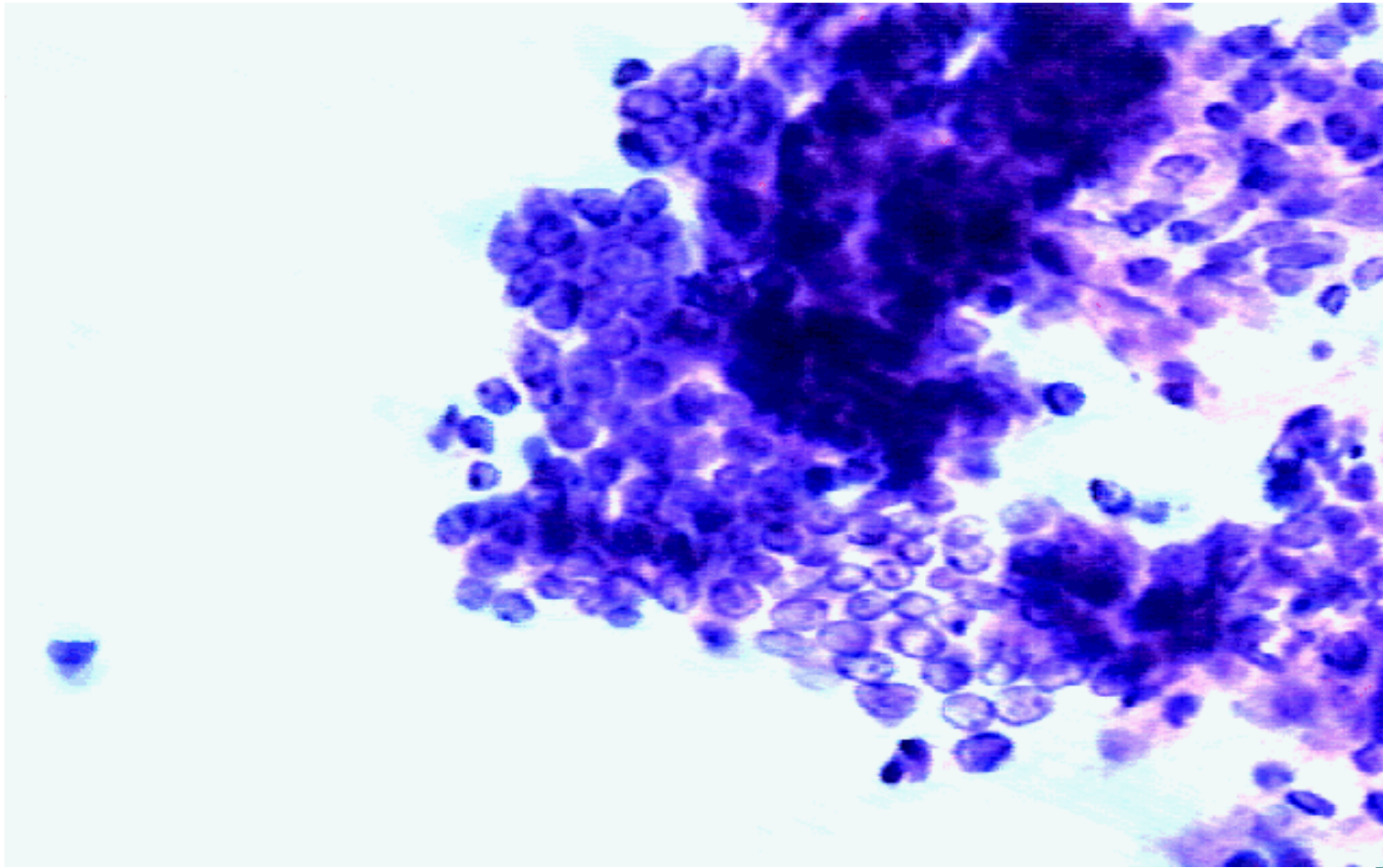


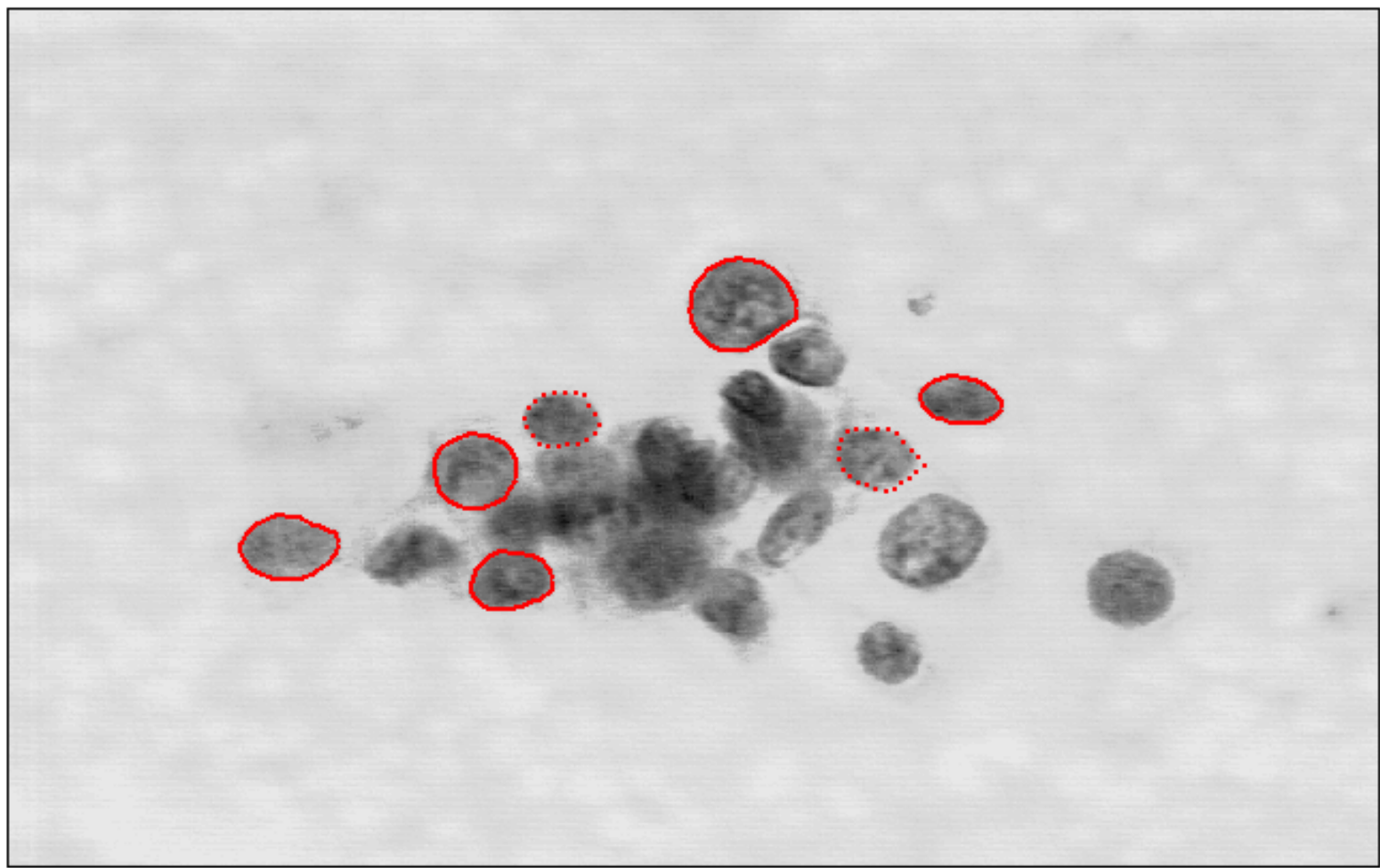
# Breast Cancer Diagnosis

# Wisconsin Breast Cancer Database

- Walberg and Mangasarian 1990
- 699 instances, 2 classes
  - 458 instances in the benign category
  - 241 instances in the malignant category







Features

Diagnosis

Prognosis

Quit

# Predictor

- 9 features to predict one of benign and malignant categories.

- clump thickness
- uniformity of cell size
- uniformity of cell shape
- marginal adhesion
- single epithelial cell size
- bare nuclei
- bland chromatin
- normal nucleoli and mitoses

# Features

- (a) Radius was computed by averaging the length of radial line segments from the center of the nuclear mass to each of the points of the nuclear border.
  
- (b) Perimeter was measured as the distance around the nuclear border.

# Features

(c) Area was measured by counting the number of pixels in the interior of the nuclear border and adding one-half of the pixels on the perimeter.

(d) Perimeter and area were combined to give a measure of the compactness of the cell nuclei using the following formula:  
 $\text{perimeter}^2/\text{area}$ .



# Features

(e) Smoothness was quantified by measuring the difference between the length of each radius and the mean length of adjacent radii.

(f) Concavity was determined by measuring the size of any indentations in the nuclear border.

# Features

(g) Concave points counted the number of points on the nuclear border that lie on an indentation.

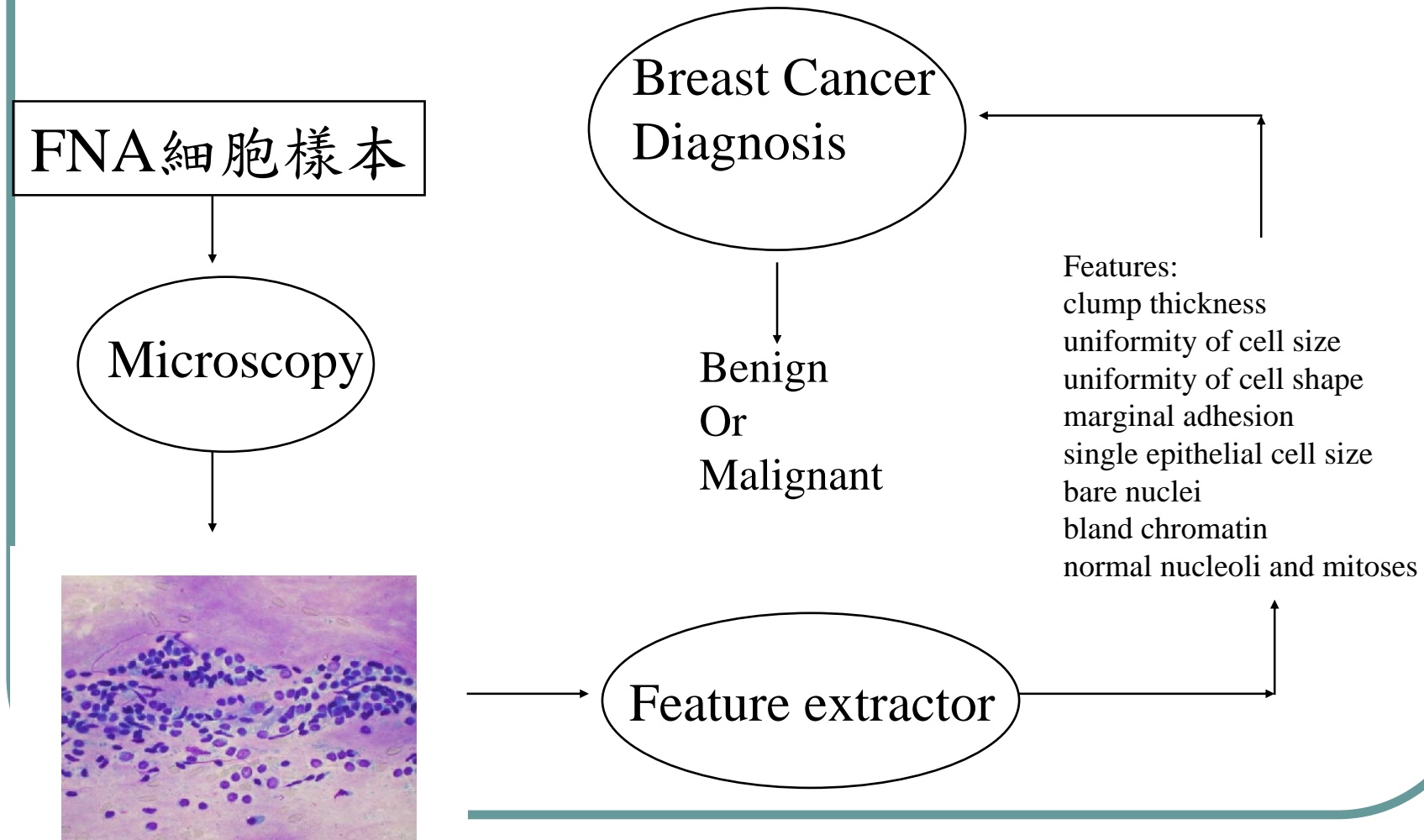
(h) Symmetry was measured by finding the relative difference in length between line segments perpendicular to and on either side of the major axis.

# Features

(i) Fractal dimension was approximated using the "coastline approximation" described by Mandelbrot [\(12\)](#) that measured nuclear border irregularity.

(j) Texture was measured by finding the variance of the gray scale intensities in the component pixels.

# Breast Cancer Classification



# Input Data

```
TrainFile = 'data\\breast_train.dat';  
TestFile = 'data\\breast_test.dat';  
TrainSet = load(TrainFile, '-ascii');  
TestSet = load(TestFile, '-ascii');
```

# Data flow

Training data  
 $x: dxN, d=9, N=483$   
 $Y: 1xN$

```
[E,NetDef,W1,W2]=noisecan(x',y',M);  
[Y_sim,E,PI] = nneval(NetDef,W1,W2, x, y,1);
```

Testing data  
 $x\_test: 9x216$   
 $y\_test: 1x216$

```
[Y_sim,E,PI] = nneval(NetDef,W1,W2, x_test, y_test,1);
```

Testing errors

# Exercise

## Breast cancer classification by learning MLP

- Input data
- Form training data:  $x$  and  $y$
- Form testing data:  $x_{\text{test}}$ ,  $y_{\text{test}}$
- Train an MLP network
- Test
- Output testing errors

## Simulation Results

- ❖ Walberg and Mangasarian 1990

error rate for testing  $> 6\%$

- ❖ 683 instances of the database by Malini Lamego(2001)

	PottsDA(42)	Neural Net with algebraic loops
Train(483)	1.4%	2.3%
Test(200)	1%	4.5%

- ❖ For the 219-case test set, the RBF method with 80 kernels and the SVM method result in error rates 4.17% and 4.63% for testing.