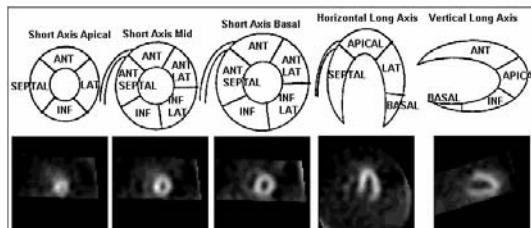
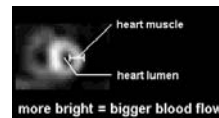
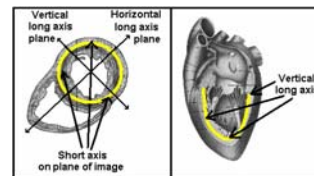




# Ensemble of Classifiers to Improve Accuracy of SPECT Heart Image Analysis System

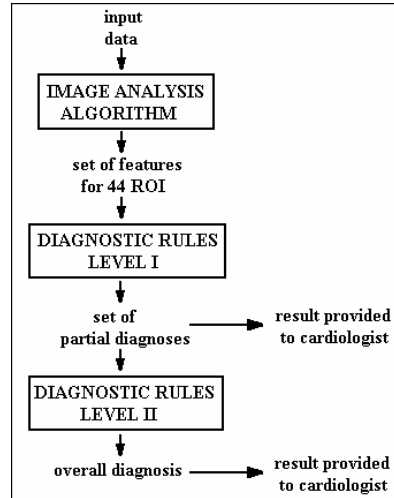
## SPECT Heart Image Analysis System

- SPECT (Single Photon Emission Computed Tomography)
- Two 3D sets of images, backprojected into six 2D sets of images
- Provides information about the left ventricle perfusion only
- Low resolution (64x64), b&w
- 2 series (rest & stress study)
- 22 defined regions of the heart LV muscle (partial diagnoses)
- Five images for each study are chosen from 2D sets



# SPECT Heart Image Analysis System

- **GOAL:** semi-automation of cardiac SPECT diagnostic process
- **Image analysis in combination with ML used to mimic a diagnostic process performed by a cardiologist**
- **Image analysis identifies key features from cardiac SPECT images**
  - 10 images (stress and rest studies)
  - patient sex
- **Next, two types of diagnostic rules are generated**
  - **for partial diagnoses**
    - input: features extracted from SPECT images
    - rules generated heuristically
  - **for the overall diagnosis using**
    - input: partial diagnoses
    - **rules generated by the CLIP3 ML algorithm**



# SPECT Heart Image Analysis System

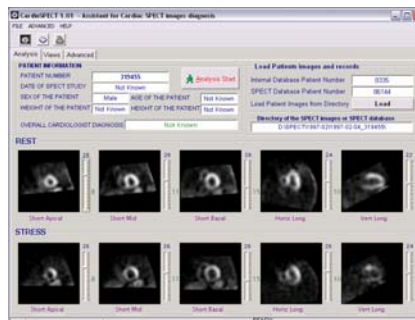
## Results for partial diagnoses

	[%]	Correct / Total
Overall diagnosis: NL (49 pat.)	<b>89.52</b>	965 / 1078
Overall diagnosis: IS (27 pat.)	<b>82.32</b>	489 / 594
Overall diagnosis: INF (47 pat.)	<b>81.53</b>	843 / 1034
Overall diagnosis: ART (31 pat.)	<b>81.52</b>	556 / 682
Overall diagnosis: IS-IN (48 pat.)	<b>77.65</b>	820 / 1056
ENTIRE NEW DB (267 pat.)	<b>81.34</b>	4778 / 5874

## Results for overall diagnoses

Sensitivity		Specificity		Predictive Accuracy		# rules
[%]	Correct / Total	[%]	Correct / Total	[%]	Correct / Total	
<b>80.00</b>	12 / 15	<b>84.30</b>	145 / 172	<b>83.96</b>	157 / 187	<b>3</b>

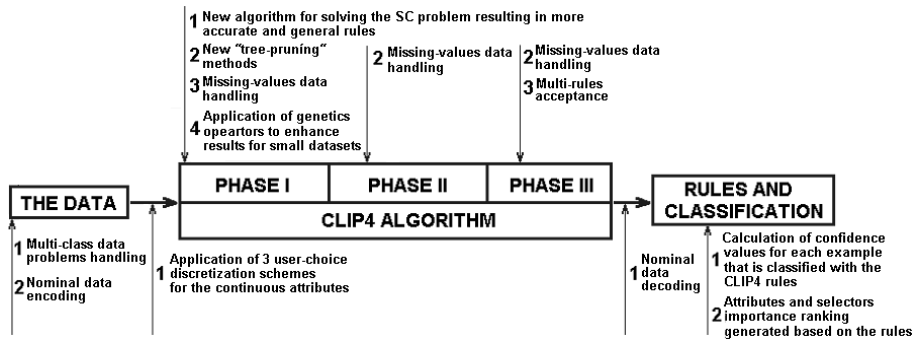
- **Simple, easy to understand, and highly accurate diagnostic rules**
- **System can be used as an assistant tool by cardiologists to help them to make more consistent diagnosis of cardiac SPECT studies**



# CLIP4 Algorithm

## • The algorithm

- partitions data into subsets using a tree structure and then generates production rules only from subsets stored at the leaf nodes
- generates model of the data that consists of
  - well-generalized classification rules
  - ranked attributes and selectors



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# Ensemble of CLIP4 Classifiers

- An ensemble of classifiers is generated by injecting randomness into the CLIP4 algorithm
  - Multiple, different classifiers are generated
  - The classifiers are combined using weighted voting scheme

## • The ensemble of CLIP4 classifiers algorithm

**Given:** Training data, where  $x_i \in X$  is an example, and  $y_i \in Y$  is its corresponding class label

**Initialize:** Set the semi-random mode of solving the SC problem,  $T$  - the number of the classifiers to be generated

For  $t = 1, \dots, T$

1. Generate a set of rules (classifier) using the CLIP4 algorithm
2. Assign a weight  $\alpha_t$  to the generated classifier  $c_t : X \rightarrow Y$

where:  $\alpha_t$  = accuracy of  $c_t$  on the entire training data set

**Classify the examples using:**

$$y_i = \arg \max_j \left( \sum_{j=1}^T \alpha_j c_j \text{ if } c_j = y_i \right)$$

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# Ensemble of CLIP4 Classifiers

- Ensemble classifier generation is based on the semi-random method for obtaining solutions for the set covering problem

Given: BINARY matrix

Initialize: Remove all empty rows from the BINARY matrix; if the matrix has no ones then return error

- Select active rows that have the minimum number of 1's in rows – min-rows
- Select the columns that have the maximum number of 1's within the min-rows – max-columns
- Select randomly one of the max-columns
- Add the selected column to the solution
- Mark the inactive rows, if all the rows are inactive then terminate, otherwise go to 1

- Each classifier is generated separately using the same training data set

FIRST ITERATION

1	0	1	1	0
0	1	1	0	1
0	0	1	1	1
0	1	1	0	1
0	1	1	0	1
1	0	0	1	0
1	0	0	1	0

1) min-row (2)  
1) min-row (2)  
2) 2)  
max-columns (2)  
3)  
max-max-columns (3)  
[ 0 0 0 1 0 ] SOLUTION

SECOND ITERATION

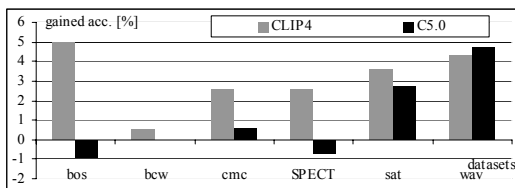
1	0	1	1	0
0	1	1	0	1
0	0	1	1	1
0	1	1	0	1
0	1	1	0	1
1	0	0	1	0
1	0	0	1	0

inactive row  
1) min-row (3)  
inactive row  
1) min-row (3)  
1) min-row (3)  
inactive row  
inactive row  
2)2) 2)  
max-columns (3)  
3)3) 3)  
max-max-columns (3)  
4)  
min inactive ones (0)  
THE FINAL SOLUTION  
[ 0 1 0 1 0 ]

# Ensemble of CLIP4 Classifiers

Dataset	CLIP4 with ensemble accuracy. [%]	CLIP4 with ensemble max. accuracy [%]	CLIP4 single classifier accuracy [%]	Gained accuracy [%]
SPECT	88.7	90.4	86.1	2.6
sat	83.1	84.0	79.5	3.6
wav	78.8	78.8	74.5	4.3
bos	75.5	N/A	70.5	5.0
bcw	95.7	N/A	95.2	0.5
cmc	49.5	N/A	46.9	2.6

- Significant improvement, when comparing to the accuracy of a single CLIP4 classifier
  - average achieved gain of accuracy was about 3.1%, which corresponds to about 15% decrease of the error rate
- Testing results show that a significant gain of accuracy is achieved for the ensemble of 3 to 4 classifiers
  - high usefulness of the proposed method



- Increase of accuracy is achieved for all datasets
  - in contrary to boosting technique

# Ensemble of CLIP4 Classifiers and SPECT Heart Image Analysis System

- ◆ Using the ensemble of classifiers method, the accuracy of the rules used to generate overall diagnosis for the diagnostic system was improved
  - ◆ a single CLIP4 classifier achieved accuracy of **86.1%**
  - ◆ the ensemble of 4 CLIP4 classifier achieved accuracy of **90.4%**
  - ◆ the gained 4% accuracy **reduces the previous error rate by 31%**
    - significant and essential improvement for the entire medical diagnostic system

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