

Medical Image Analysis

Guest Lecture,
2018-10-04

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Advertisement

- **FMAN30**
- **Medical Image Analysis**
- **7.5hp**
- **Study period 2**
- **4 hand-ins**
- **<http://www.maths.lth.se/course/medim/2018/>**



Learning objectives

- **Describe** different image acquisition techniques used in medical imaging, e.g. X-Ray, CT, MR, Ultrasound, PET, Scint and SPECT.
- **Explain** and **use** medical image analysis algorithms to perform registration, segmentation and classification
- **Decide** on appropriate algorithms for solving medical image analysis problems
- **Implement** automated medical analysis systems
- **Validate** the results of automated medical analysis systems



Research on Medical Image Analysis

- Mathematical imaging group
 - <http://www2.maths.lth.se/matematiklth/vision/>
- Cardiac MR Group
 - http://www.med.lu.se/klinvetlund/klinisk_fysiologi/forskning/cardiac_mr_group

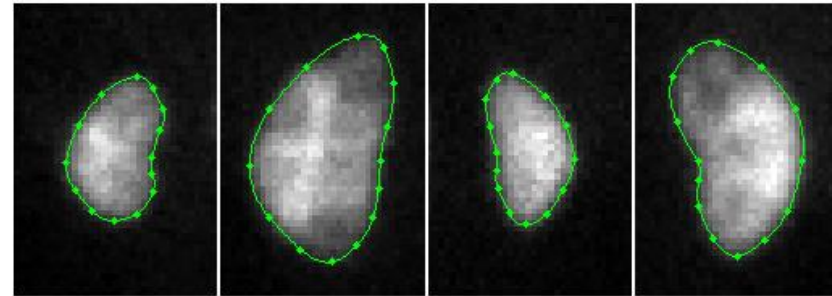
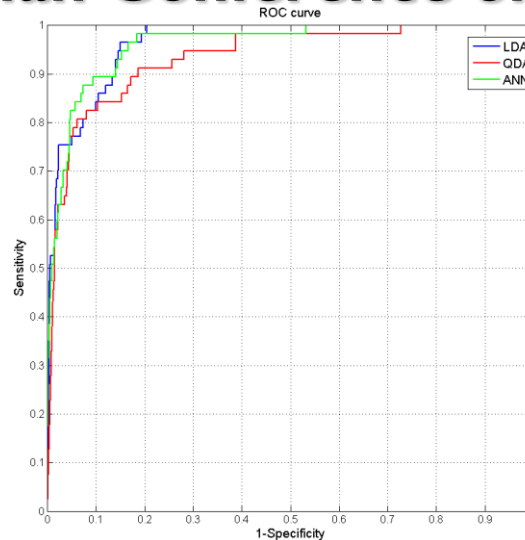
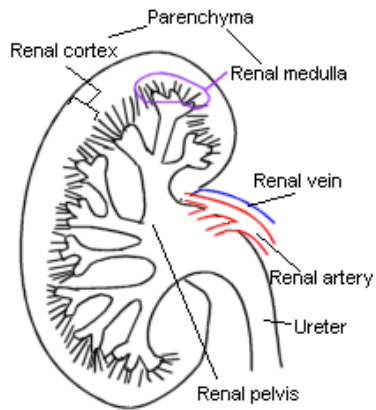


Examples



Detection and Diagnosis of Kidney Lesions

Scandinavian Conference on Image Analysis, 2011

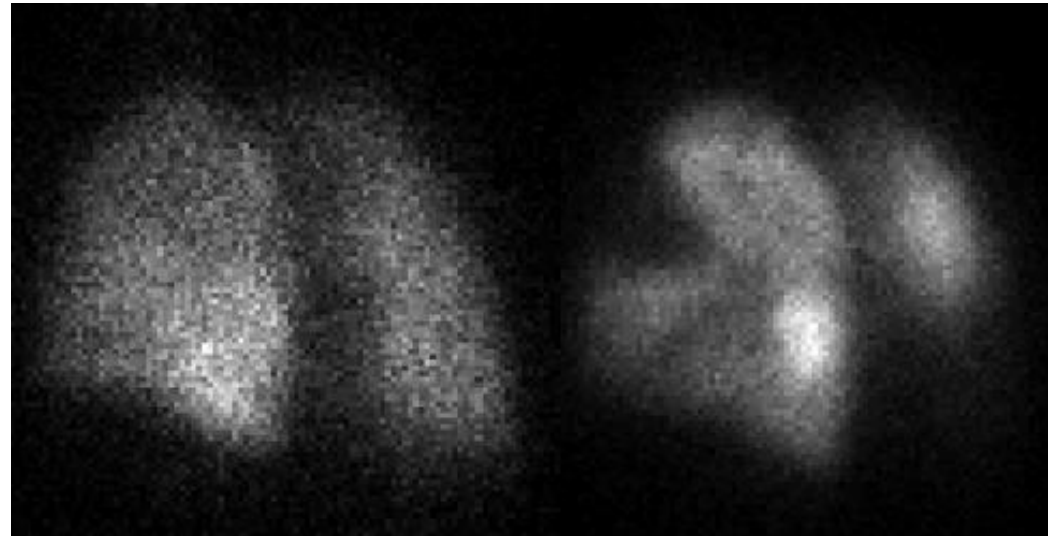
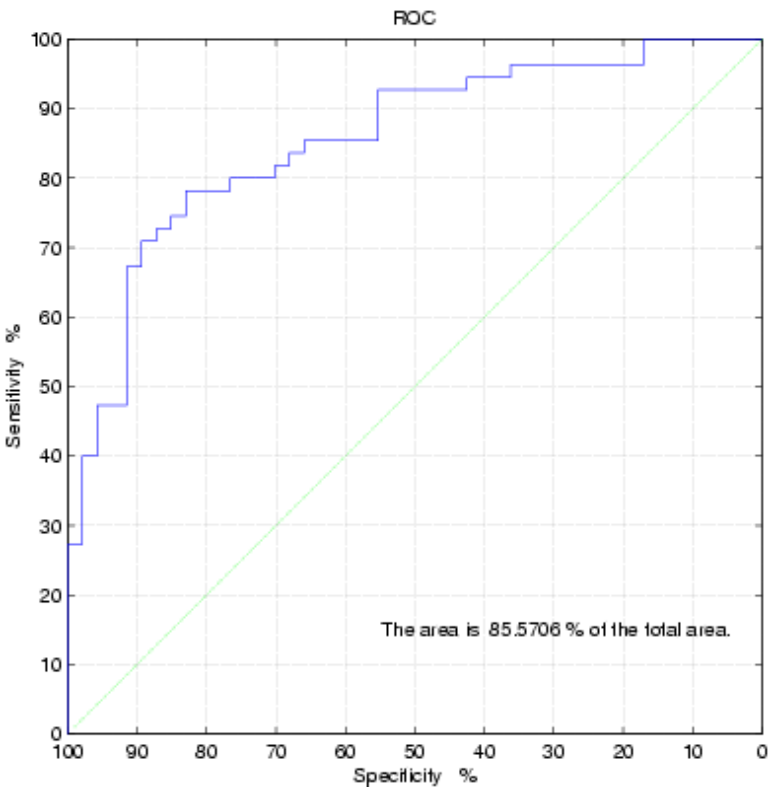


	LDA	QDA	ANN
Area under ROC curve (AUC)	0.964	0.935	0.960
Sensitivity (%)	96.5	96.5	96.5
Specificity (%)	84.8	61.2	83.4
Positive Predictive value (%)	35.0	17.4	32.9
Negative Predictive value (%)	99.7	99.5	99.6
Mis-classification rate (%)	14.2	36.0	15.6



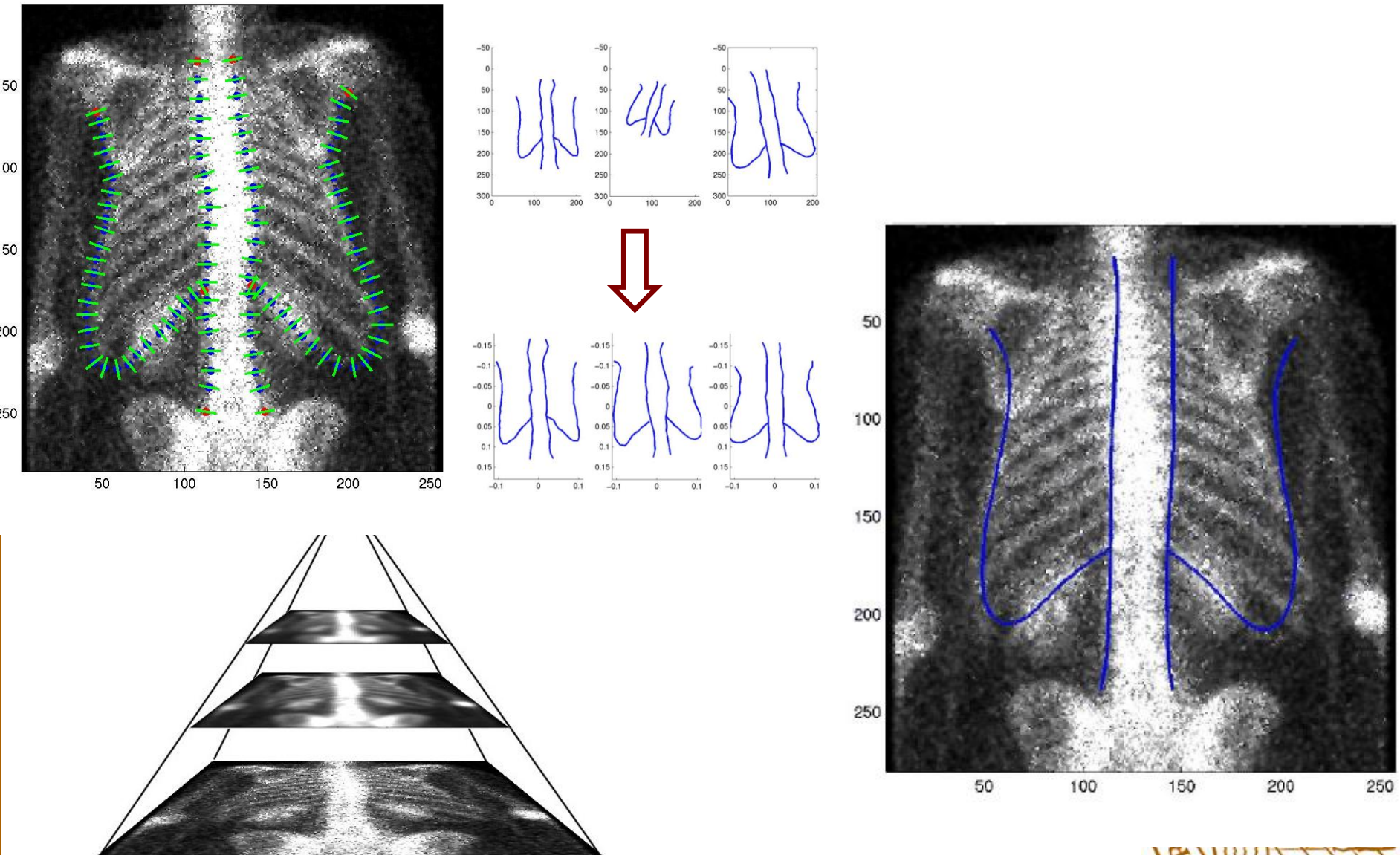
Diagnosis of Pulmonary Embolism

European Journal of Nuclear Medicine, 2000

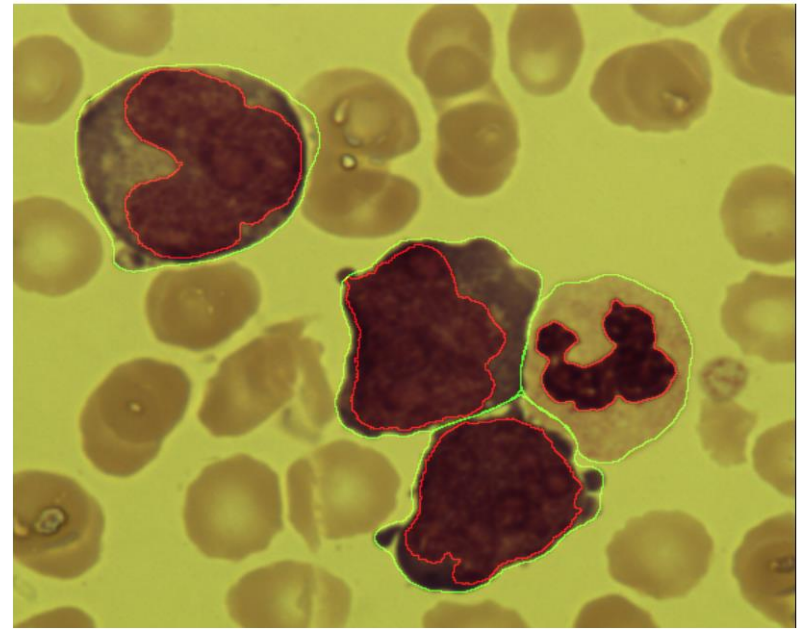
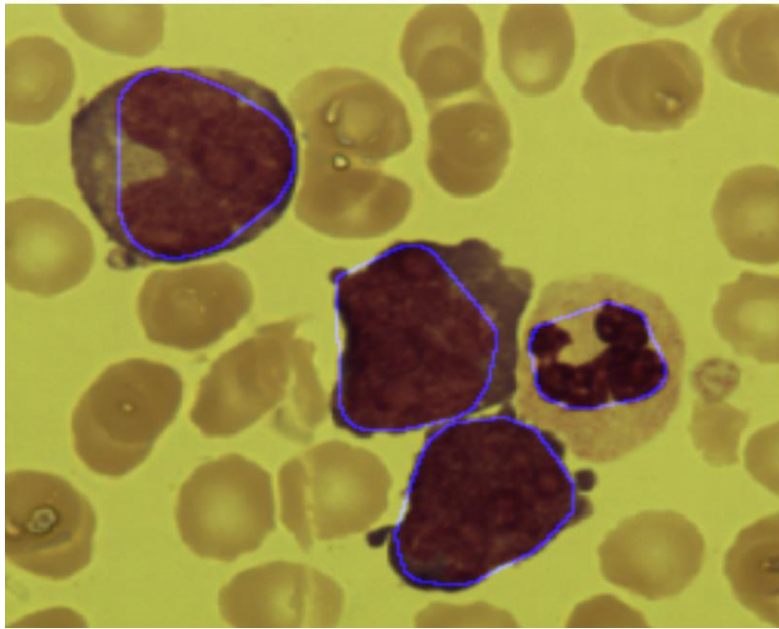


Segmentation – shape variation methods

Understanding both appearance and shape



Segmenting white blood cells



Septum segmentation in ultrasound images

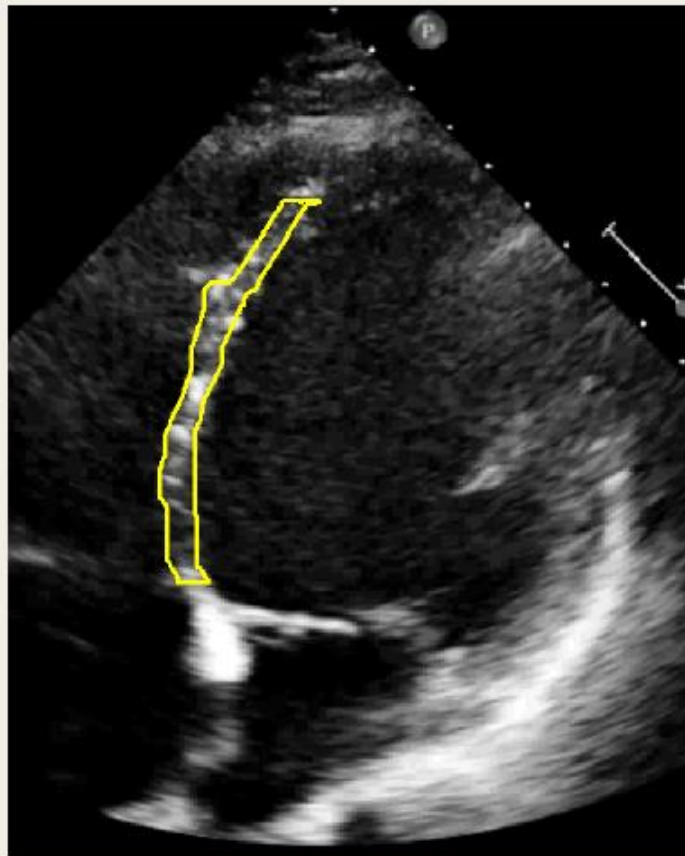


Fig. 2. The inter-ventricular septum marked with a yellow line obtained with our shortest path segmentation

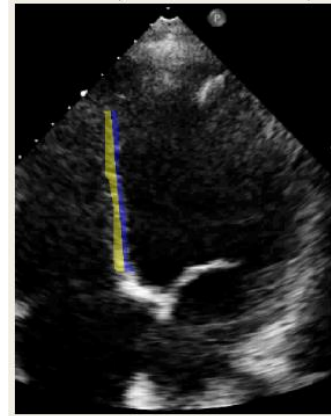


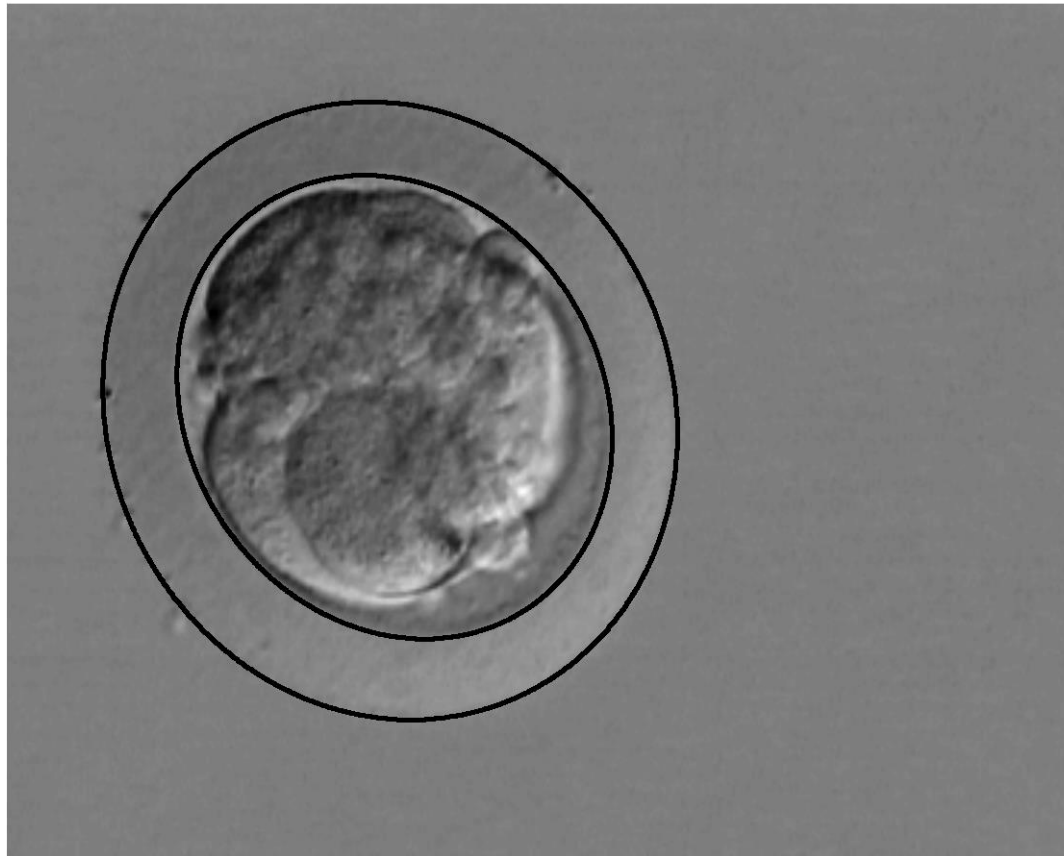
Fig. 8. Assessment: straight. $s = 0.279$.



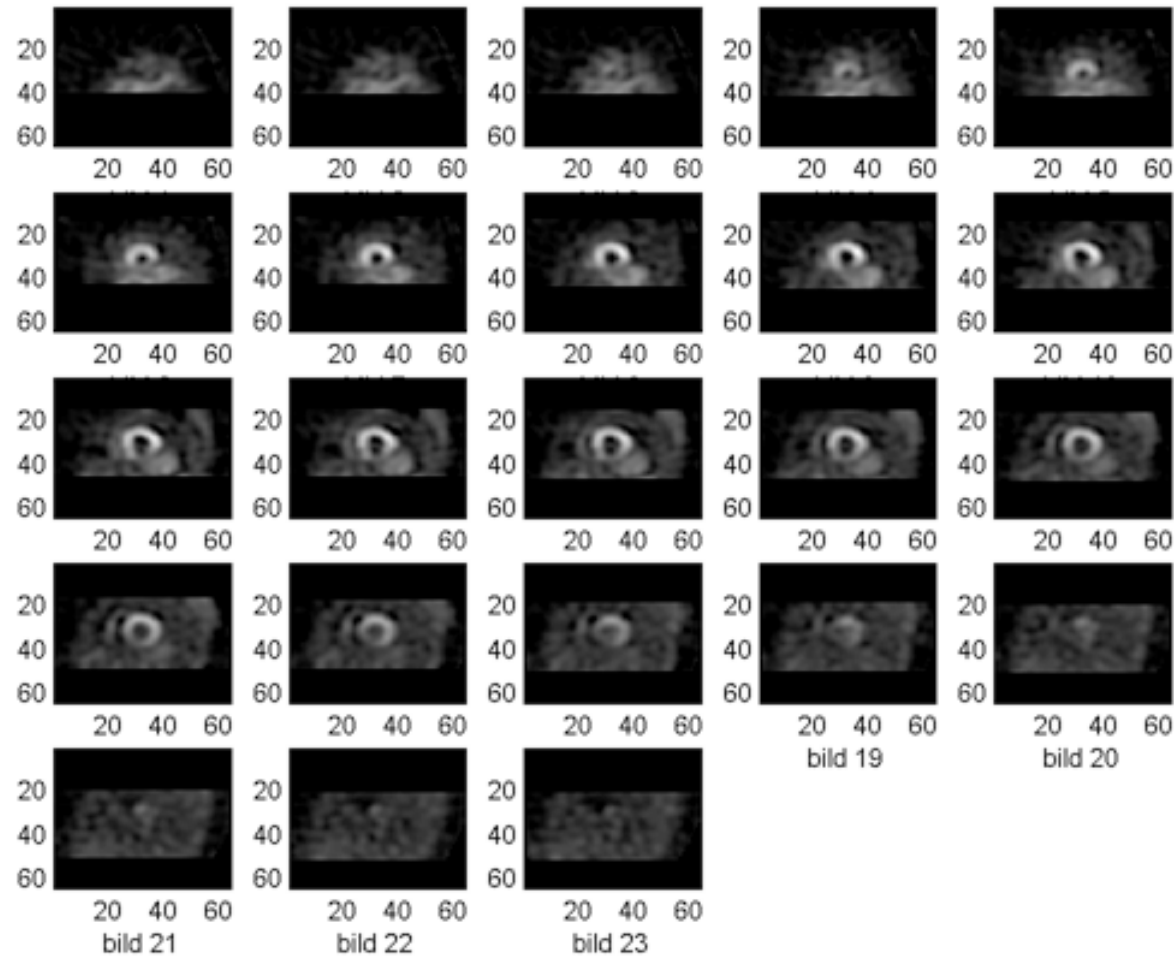
Fig. 10. Assessment: bulges towards the right ventricle. $s = 0.896$.



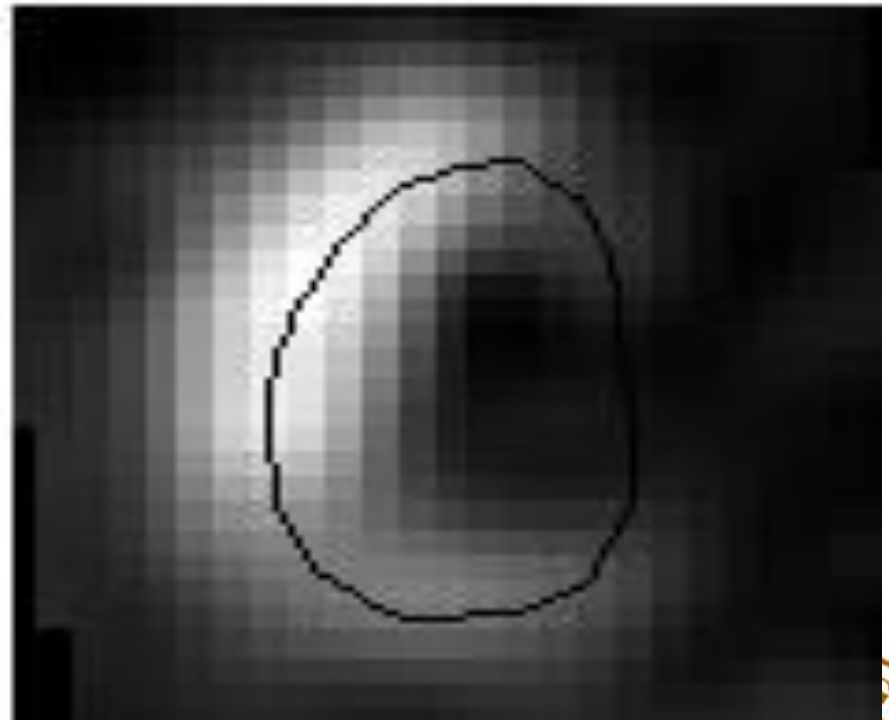
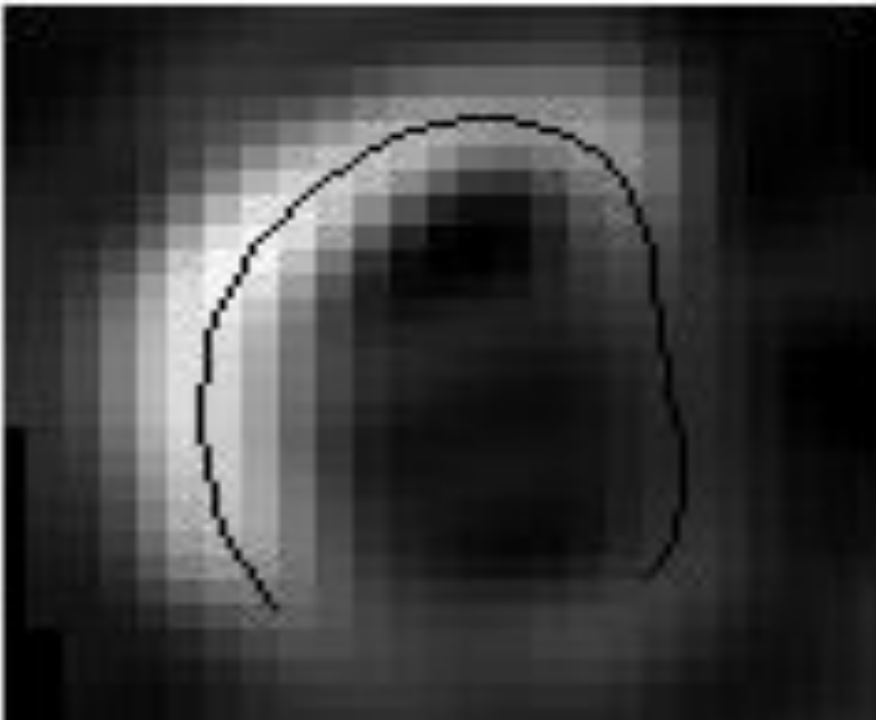
Segmentation of the Zona Pellucida



SCINT - Heart



Segmentation results



Exini Diagnostics

CARE File Help

Bullseye v1-6

DICOM HD

Patient's ID	Status
19200130	Done
19510402	Done
19350312	Done
19551209	Done
19620706	Done

Clear

Patient's ID	Patient's Name	Series Date	Series Description	Image ID	Study Description
19510402	Frau Wiess		oblique	oblique	NC_stress
19510402	Frau Wiess		oblique	oblique	NC_rest
19200130	Mrs Cole		oblique	oblique	NC_rest
19200130	Mrs Cole		oblique	oblique	NC_stress
19270803	Frau Gruber		oblique	oblique	NC_stress
19270803	Frau Gruber		oblique	oblique	NC_rest

Evaluate

Patient's ID 19510402 **Comment** **Feedback** **Snapshot**

Date **WeAidU Ref.** 6977

DISCLAIMER: This is a computer-based interpretation and WeAidU takes no responsibility for its use.

Infarct

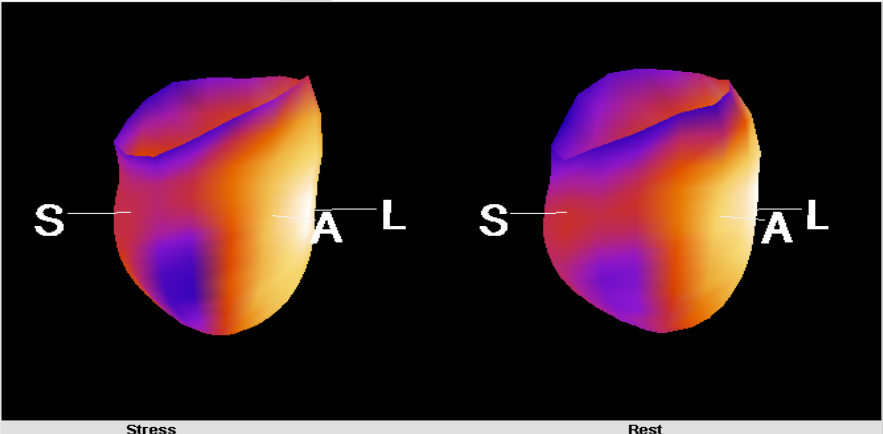
Area	Physician	WeAidU
Anterior	Absent	
Inferior	Absent	
Apical	Absent	
Septal	Absent	
Lateral	Absent	

Ischemia

Area	Physician	WeAidU
Anterior	Absent	
Inferior	Absent	
Apical	Absent	
Septal	Absent	
Lateral	Absent	

Compare diagnosis

BULLSEYES SA VLA HLA 3D

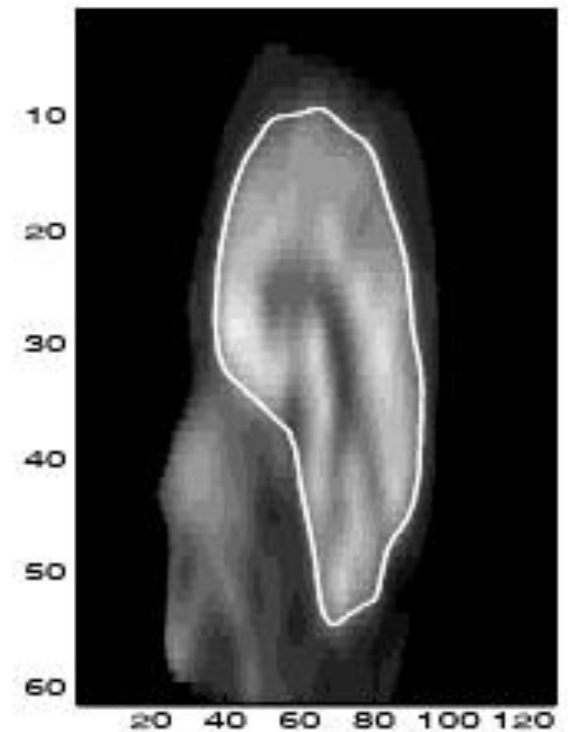
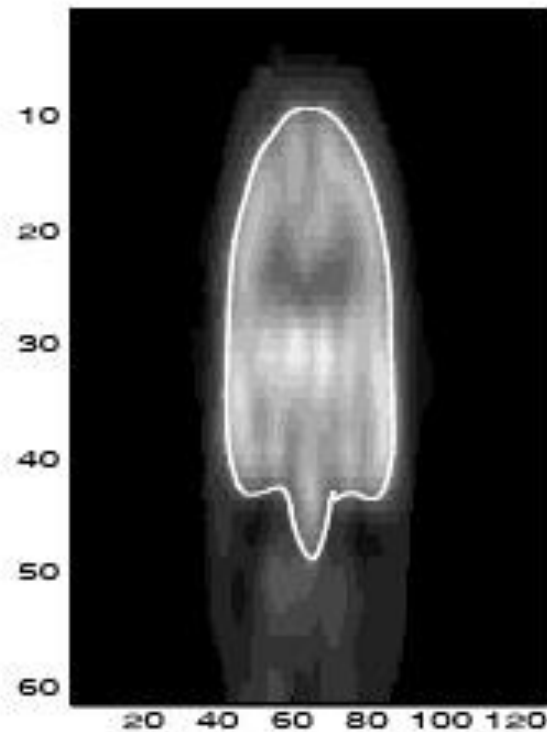
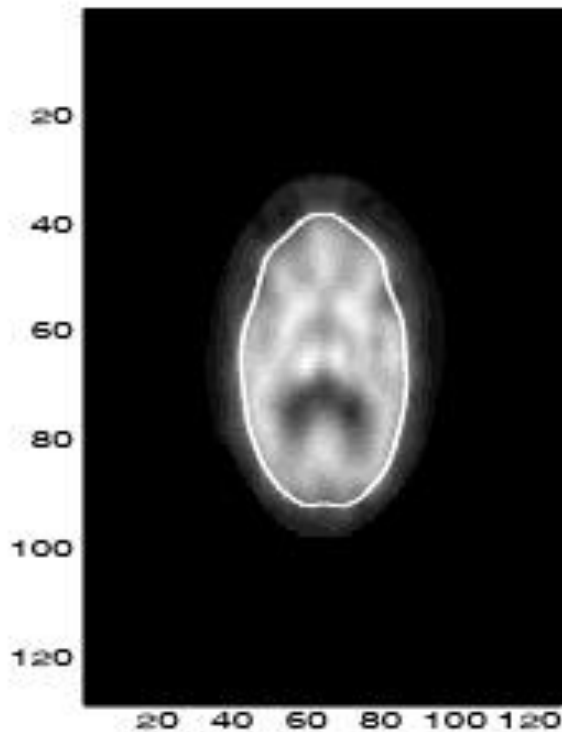


Stress **Rest**

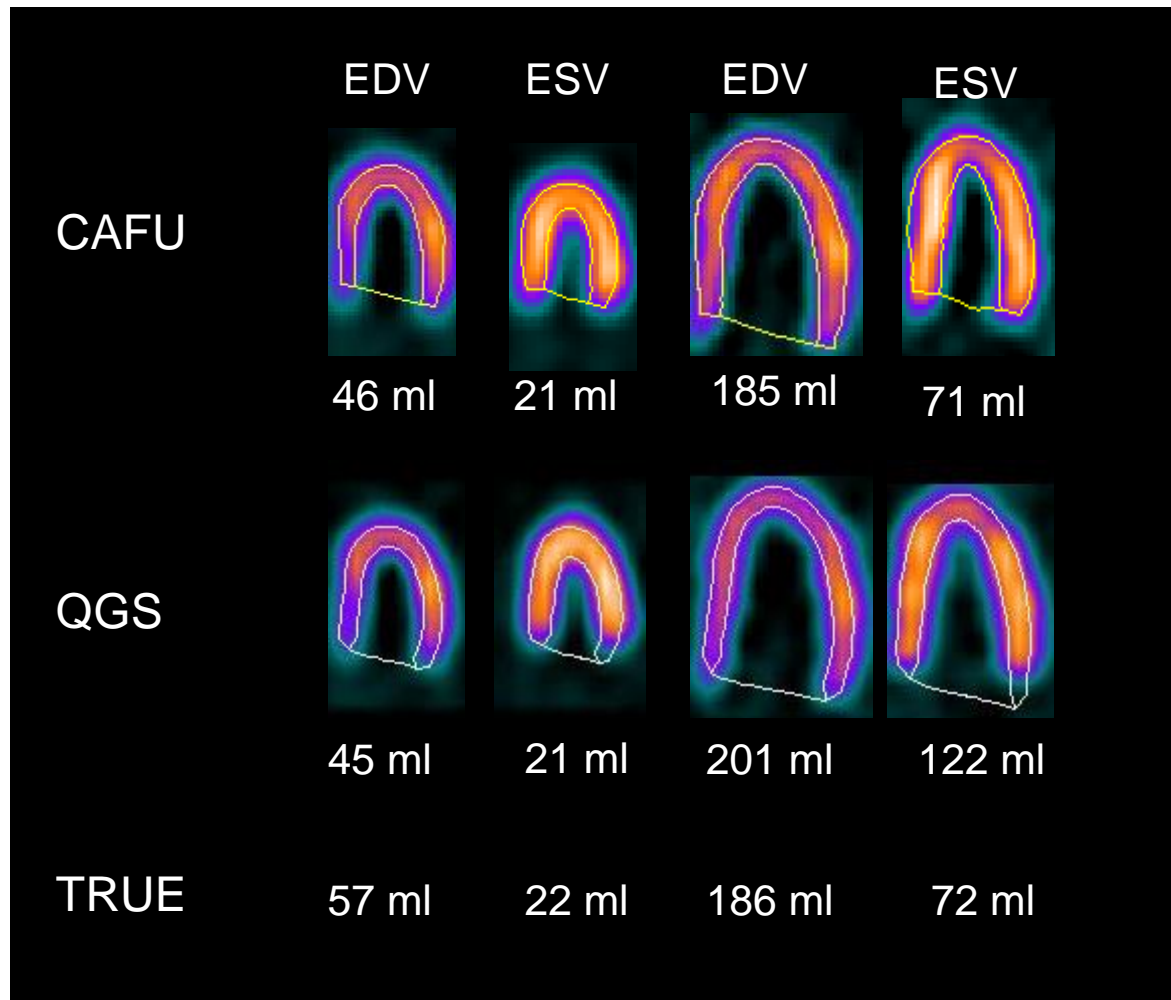
MR – knee injuries



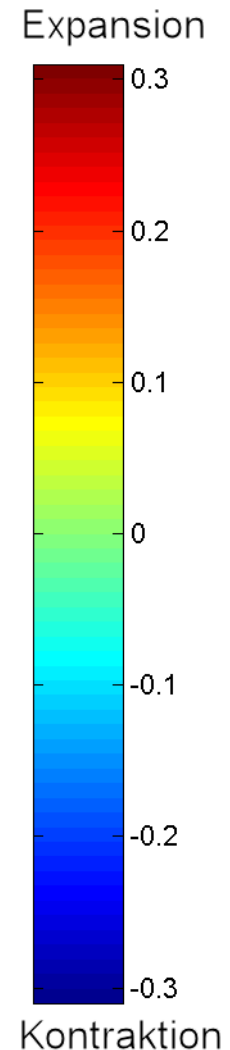
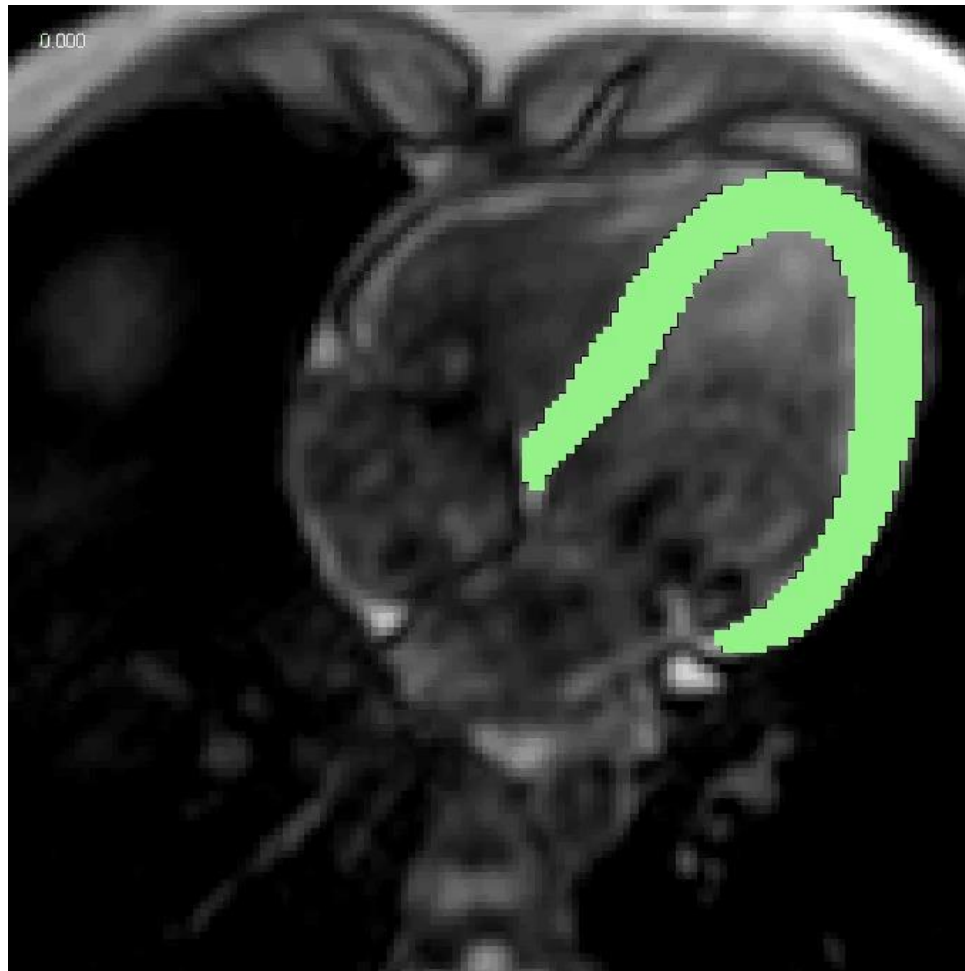
SPECT – brain (dementia)



Gated SCINT - heart



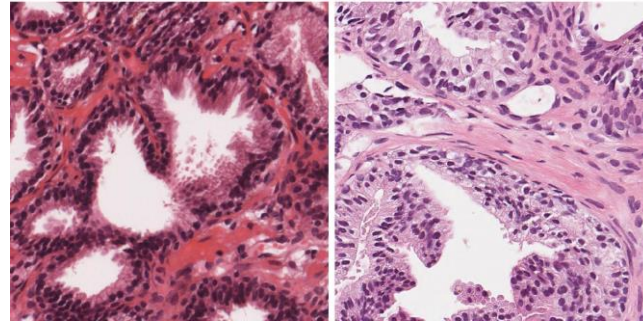
Longitudinal expansion of left ventricle



Digital Pathology (prostate biopsies)

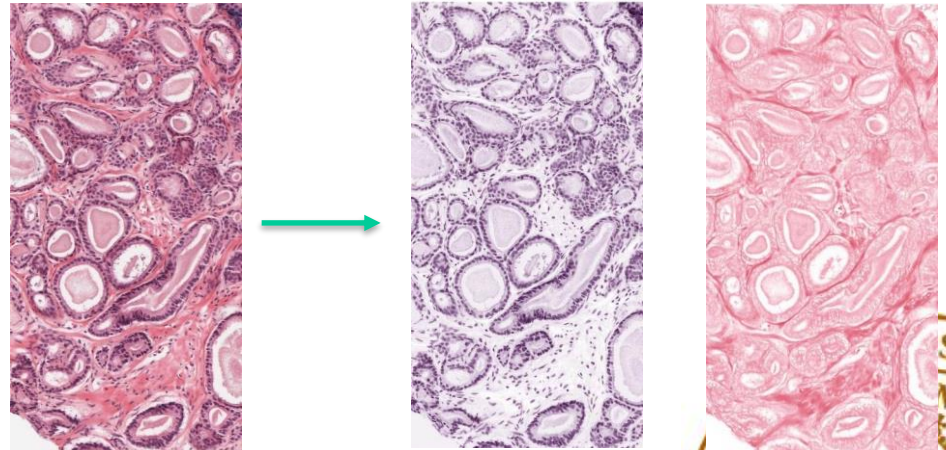
Digital stain separation to reduce variations between different labs

Typical stain variation:



Digital stain separation:

This makes it possible to normalize the stains individually, thus making different stain concentrations look similar.



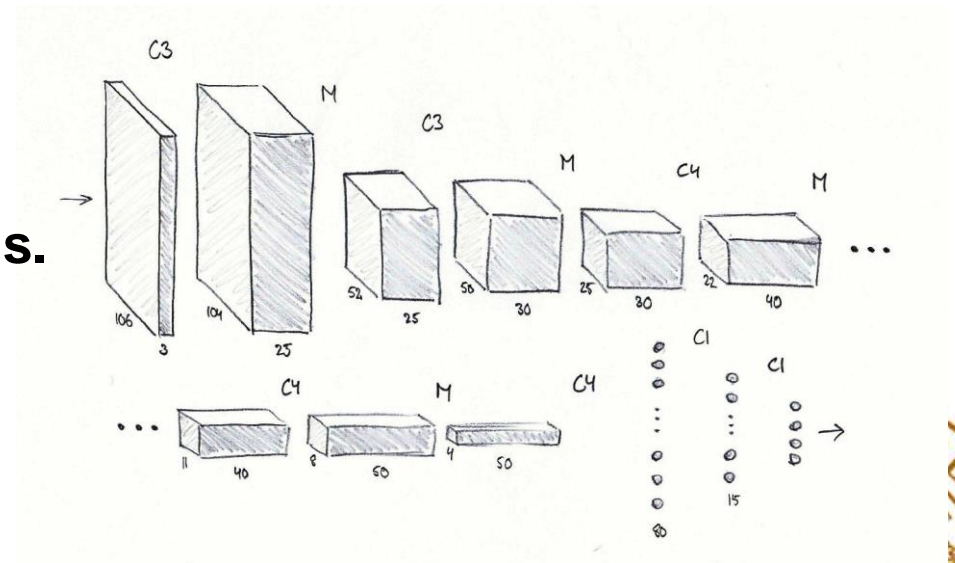
Deep Learning Approach (CNN)

Idea for Deep Learning:

Let the computer itself optimize an algorithm by training on a large dataset.

Typically optimizes 100 000 to 100 000 000 parameters.

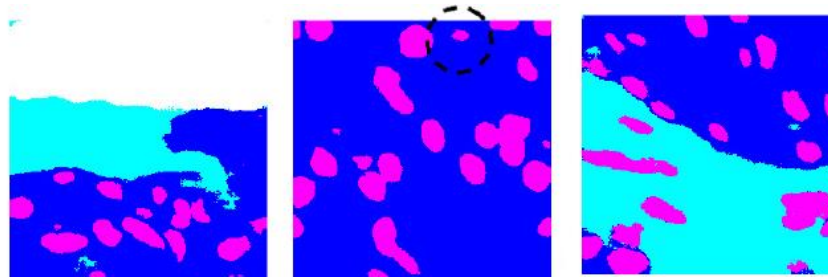
**Convolutional Neural Networks
specially designed for image analysis.**



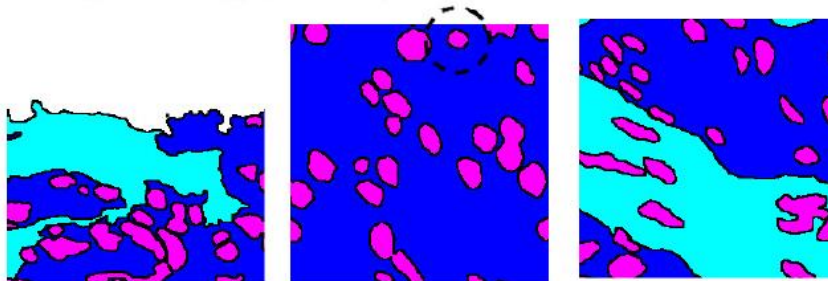
Segmentation Results

Segmentation into: Background, Stroma, Epithelial Cytoplasm and Nuclei

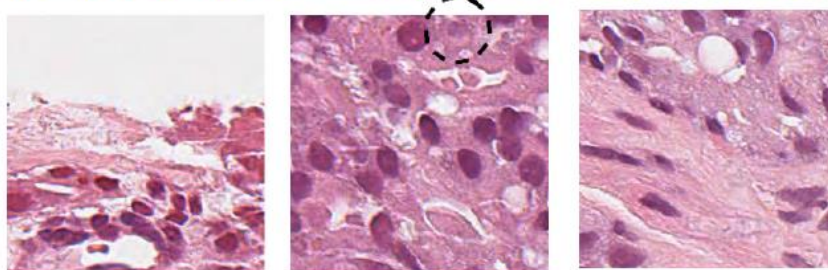
Original
Images:



Ground Truth
Segmentation:



Results:



Classification Results

**Classification into four classes:
Benign, Gleason 3, Gleason 4 and Gleason 5**

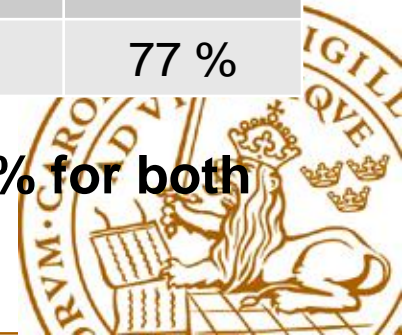
**Accuracies when RGB
images are used as
input:**

Validation data	Training data	
	Dataset A	Dataset B
Dataset A	93 %	32 %
Dataset B	60 %	75 %

**Accuracies when stain
separated images are
used as input:**

Validation data	Training data	
	Dataset A	Dataset B
Dataset A	91 %	46 %
Dataset B	60 %	77 %

**Accuracies for when the datasets are combined are 84 % for both
types of inputs.**



Details on Registration



RANSAC-based registration

- **Given a tentative list of point correspondences and a geometric transformation**
- **Select at random a minimum number of point correspondences needed to determine the transformation**
- **Calculate the transformation**
- **Determine the number of inliers and outliers**
- **Repeat a pre-determined number of iterations**
- **Estimate the best transformation from all inliers in the set containing most inliers**



Euclidean (rigid)

- Two point correspondences needed
- The rotation angle can be determined by the angle between the lines between the points
- The translation can be determined from a linear system of equations



Similarity (rigid+scale)

- Two point correspondences needed
- The scale can be determined from the quotient of the distances between the points
- The rotation angle can be determined by the angle between the lines between the points
- The translation can be determined from a linear system of equations



Affine

- Two affine transformation contains 6 parameters
- Thus three point correspondences are needed
- The transformation can be obtained directly from a linear set of equations obtained from $y = Ax + t$ for the three different point correspondences.

