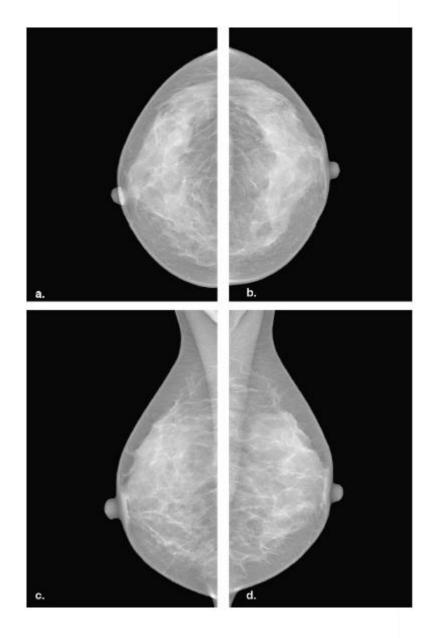
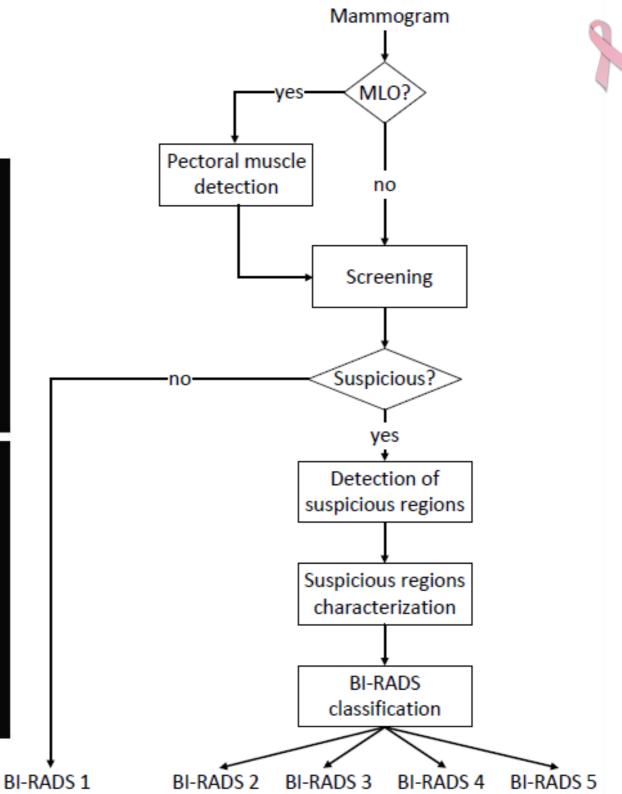
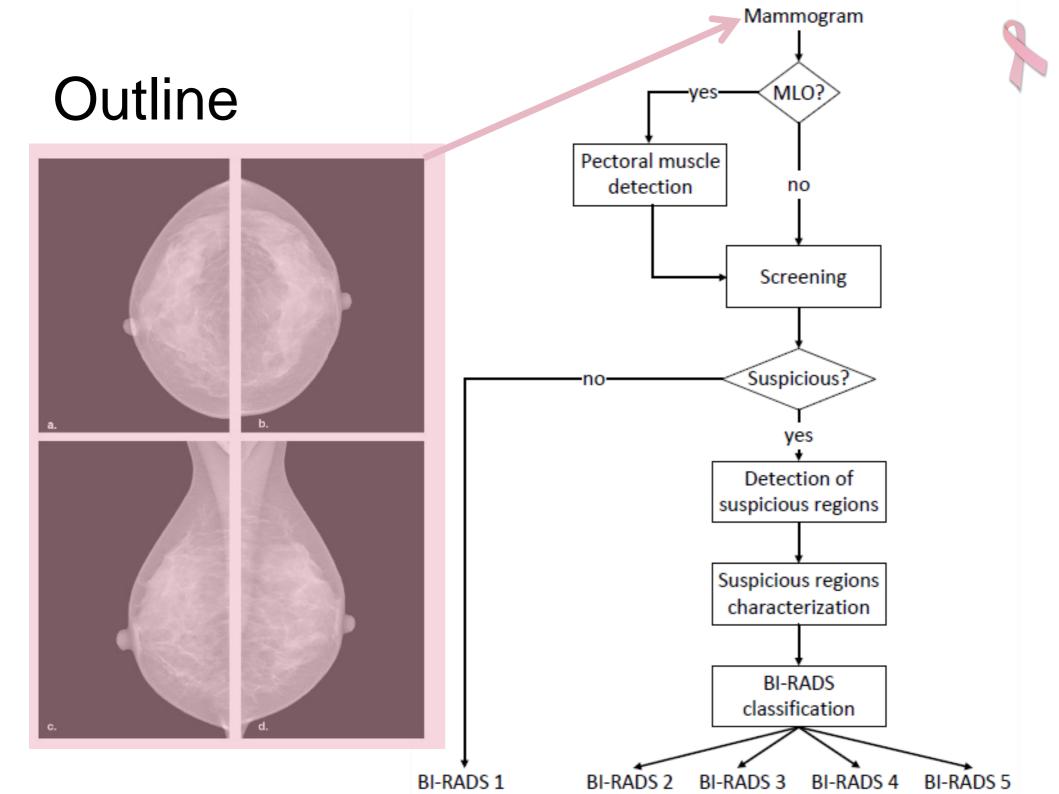
An automatic mammogram system: from screening to diagnosis

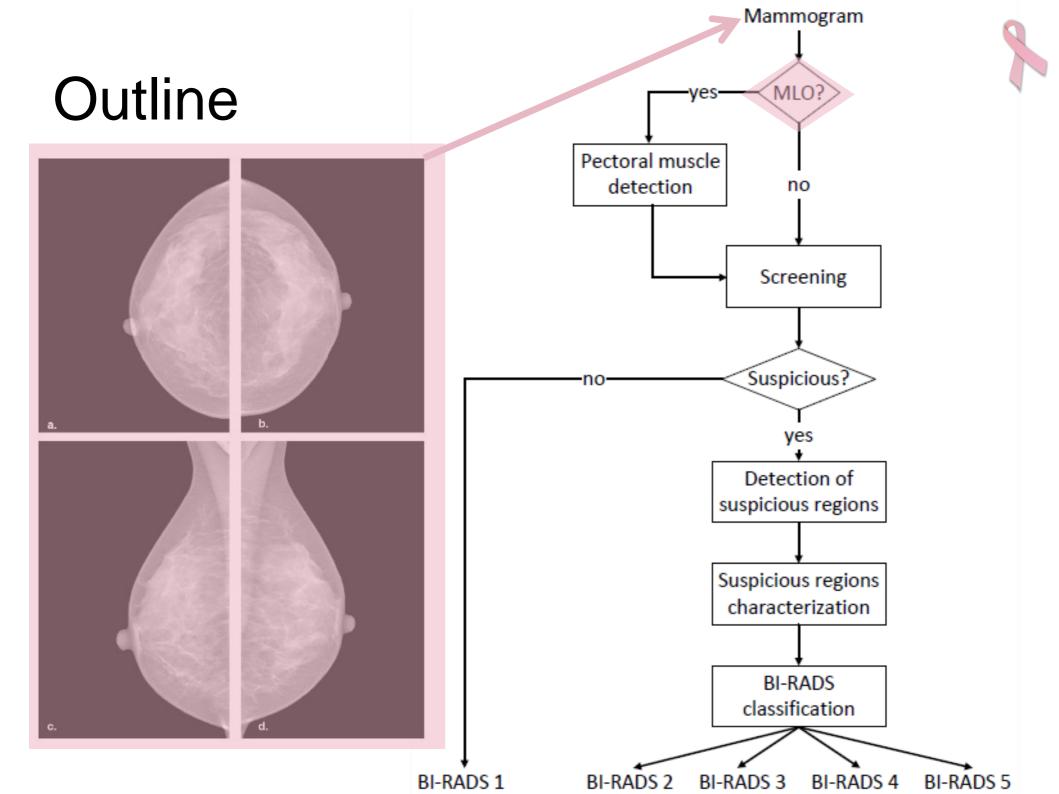
Inês Domingues

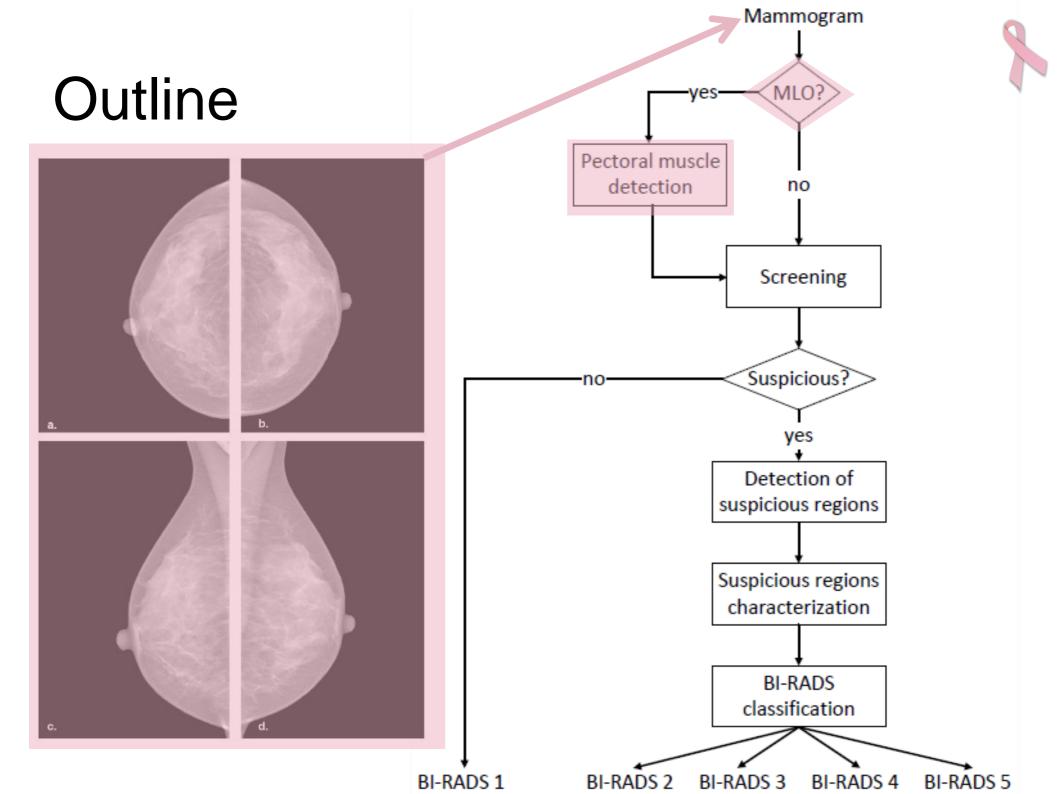
Outline

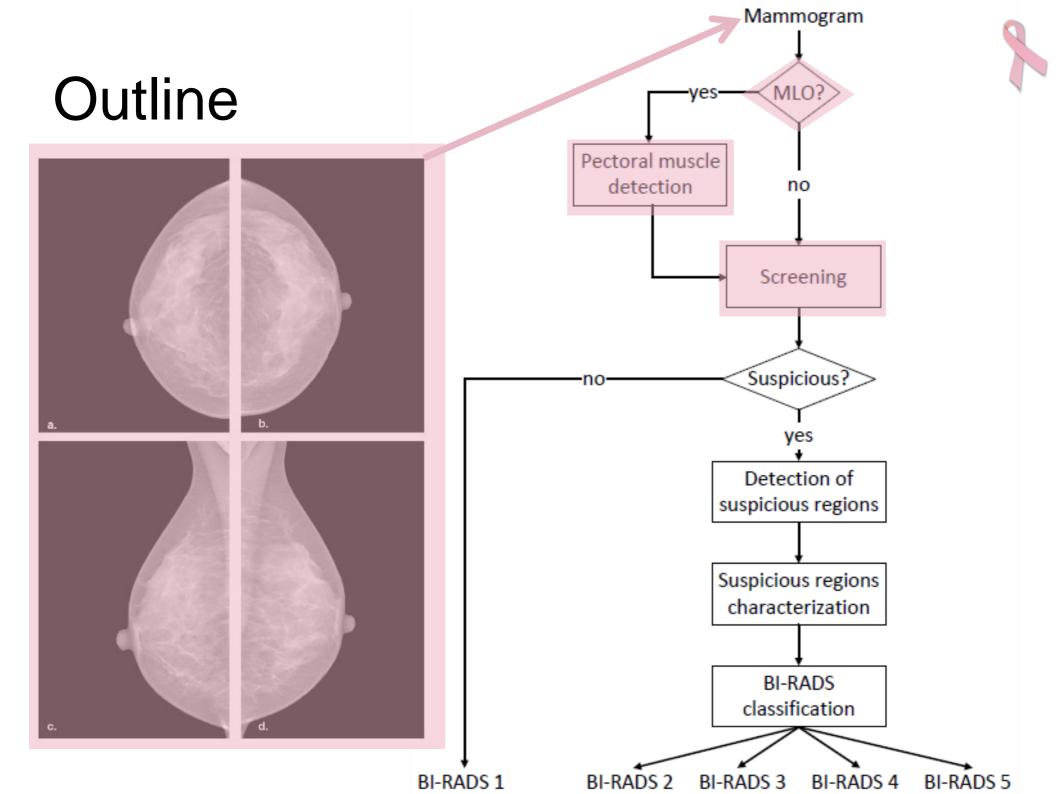


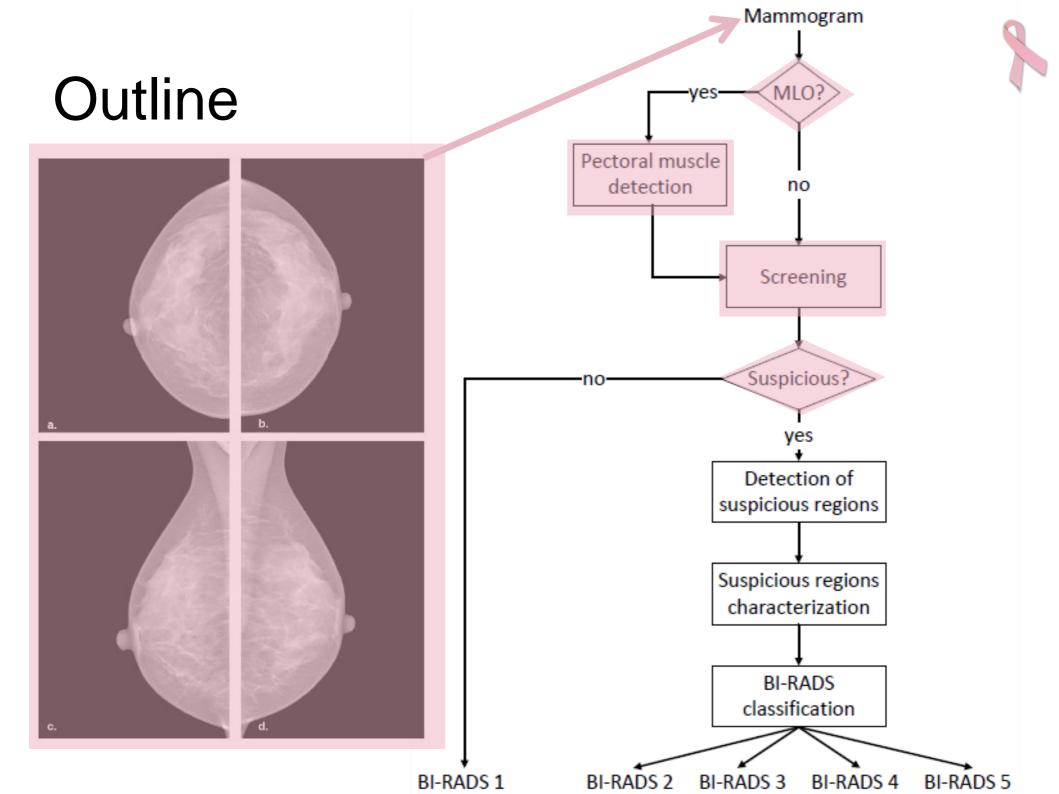


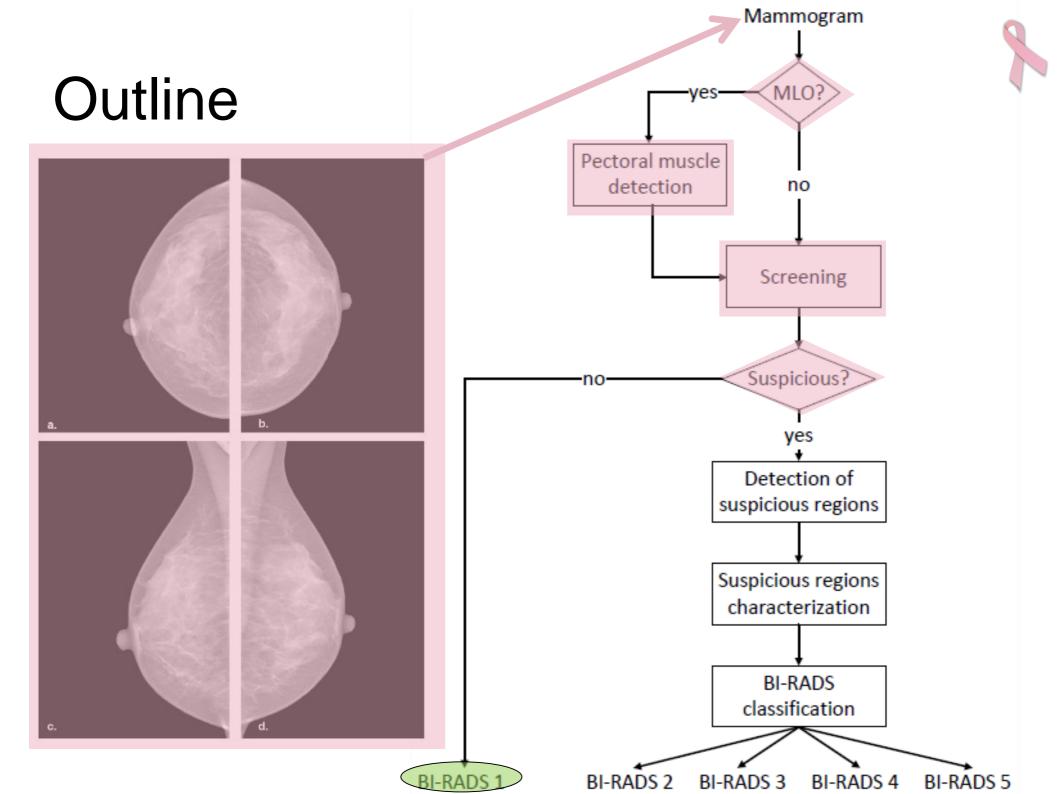


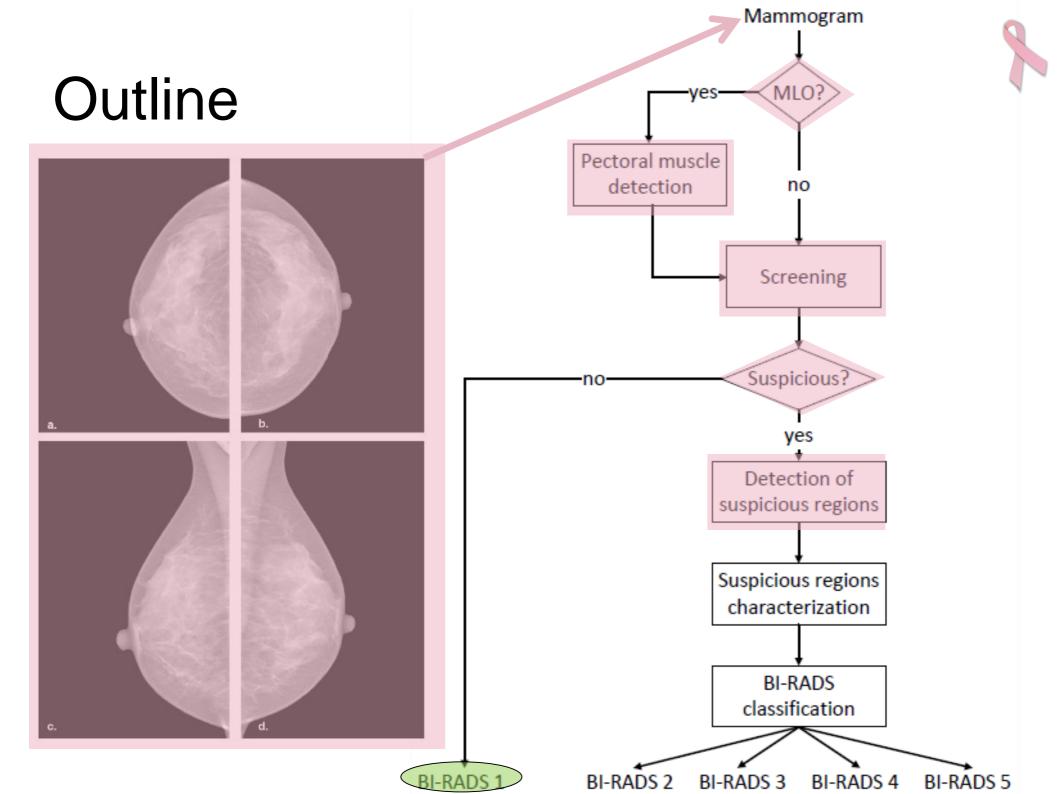


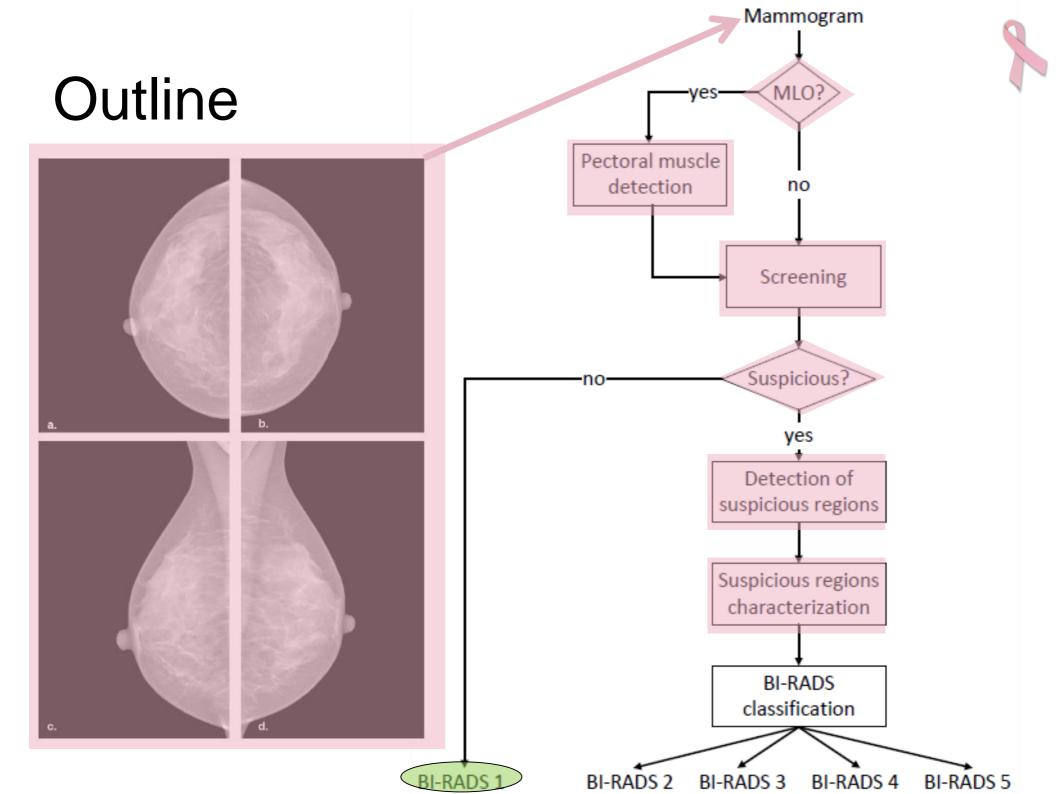


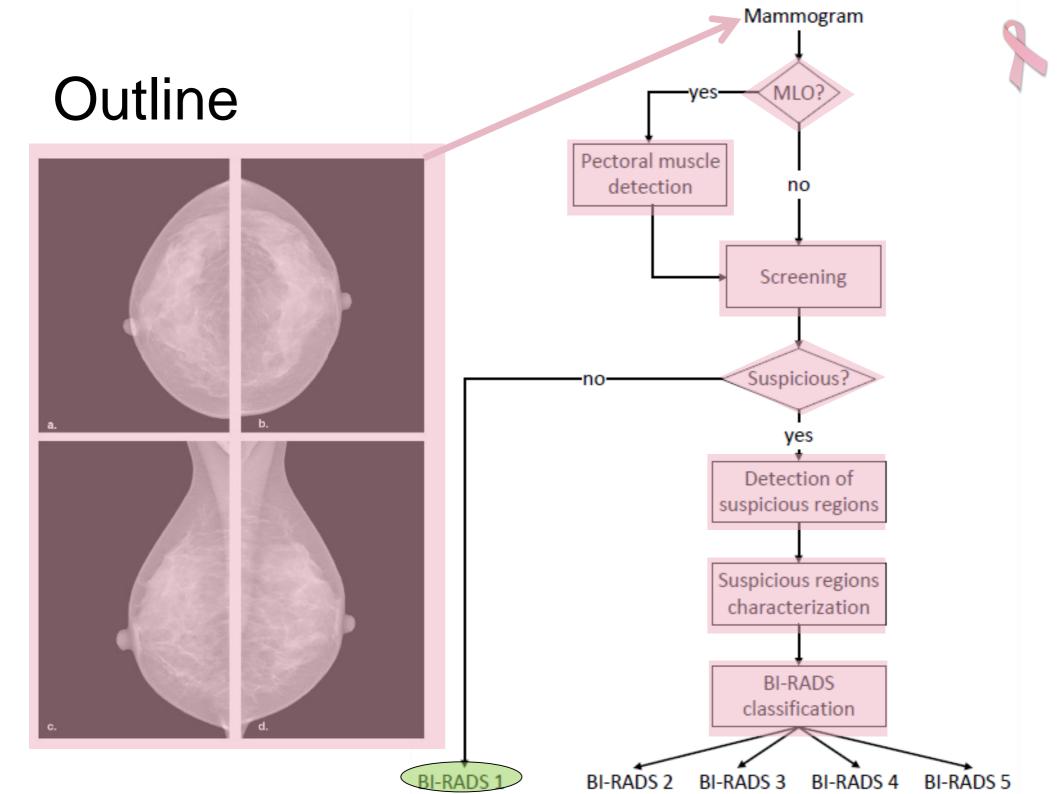






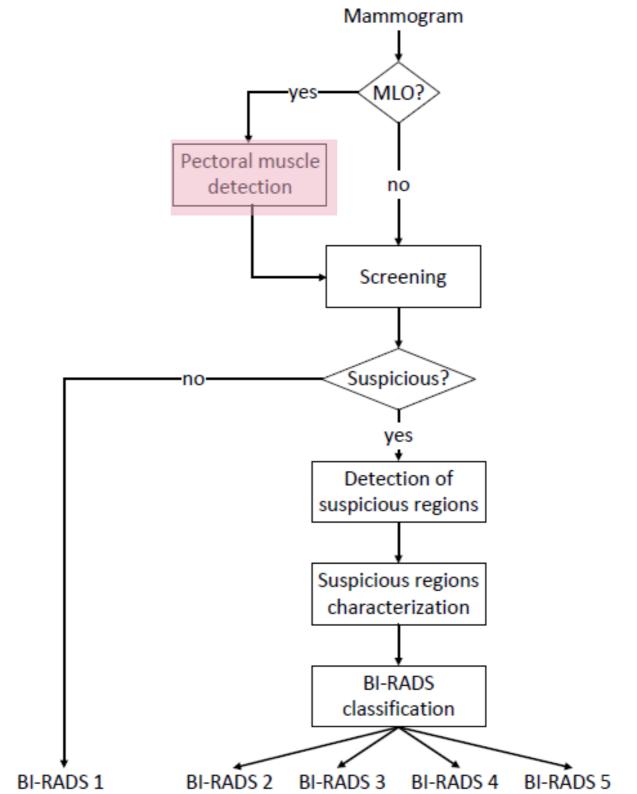






Mammogram Outline MLO? Pectoral muscle detection no Screening Suspicious? -no yes Detection of suspicious regions Suspicious regions characterization **BI-RADS** classification BI-RADS 1 BI-RADS 2 BI-RADS 5 BI-RADS 3 BI-RADS 4

Outline

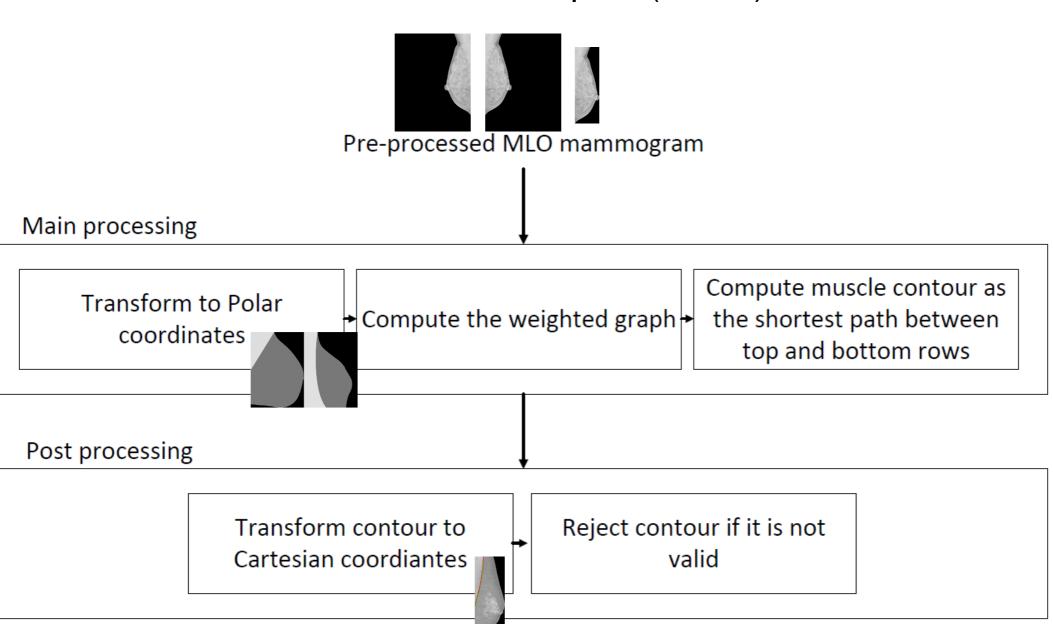




Pectoral muscle detection

7

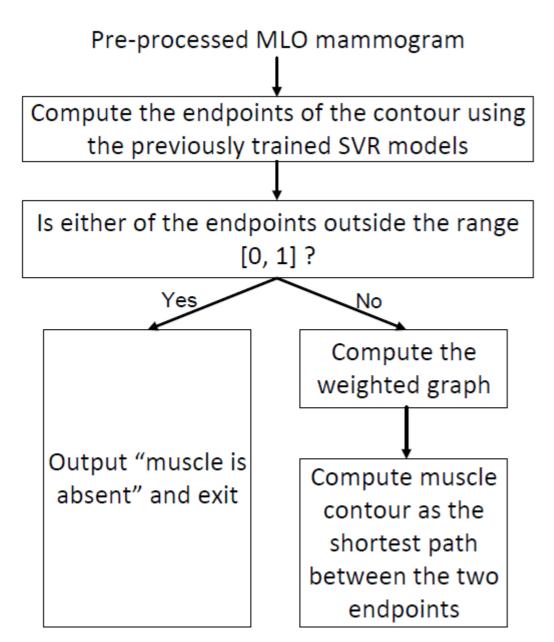
Polar coordinates and the shortest path (SPPC)

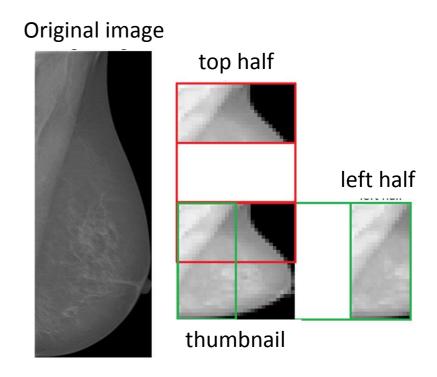


Pectoral muscle detection

7

Shortest path with endpoints learnt by SVMs (SPLE)

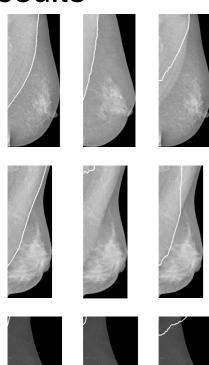




Pectoral muscle detection



Results



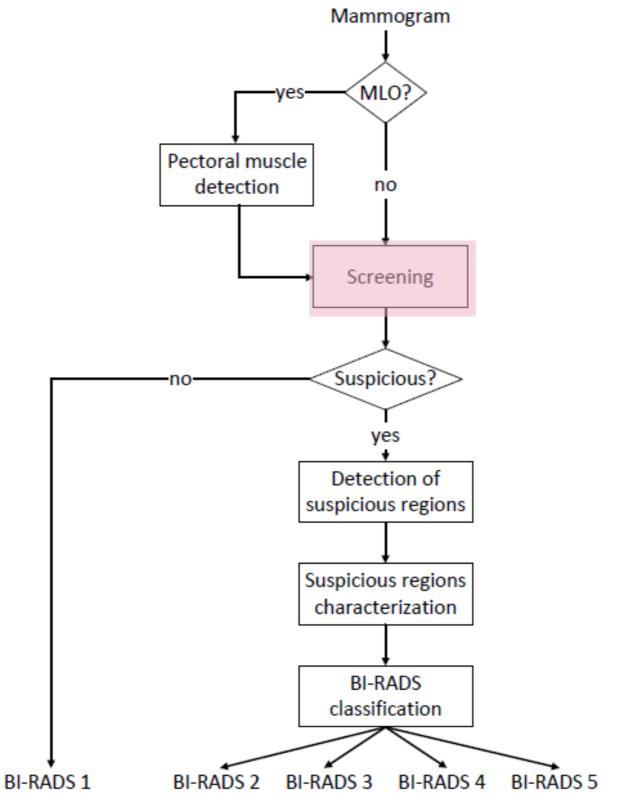
 Differences between SPPC and SPLE are <u>not</u> significant

SPLE

- if a robust estimation of the endpoints can be achieved
- the pectoral muscle boundary can be effectively predicted
- the prediction of the endpoints seems to be the main source of errors

Outline

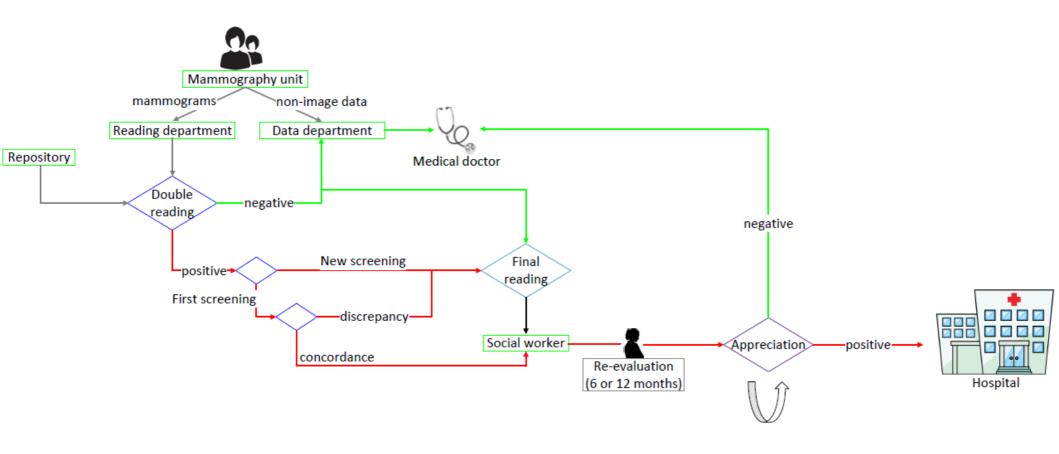




Screening

7

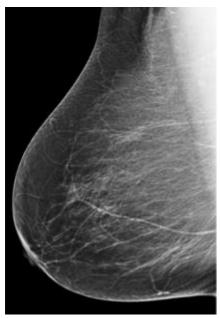
Portuguese Breast Cancer Screening Program



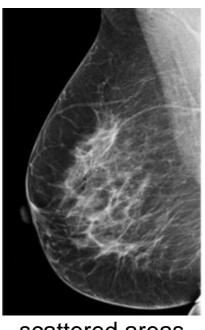
Screening

7

Breast density



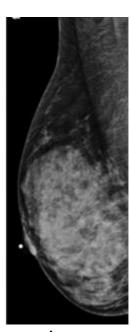
almost entirely fatty



scattered areas of fibroglandular density



heterogeneously dense

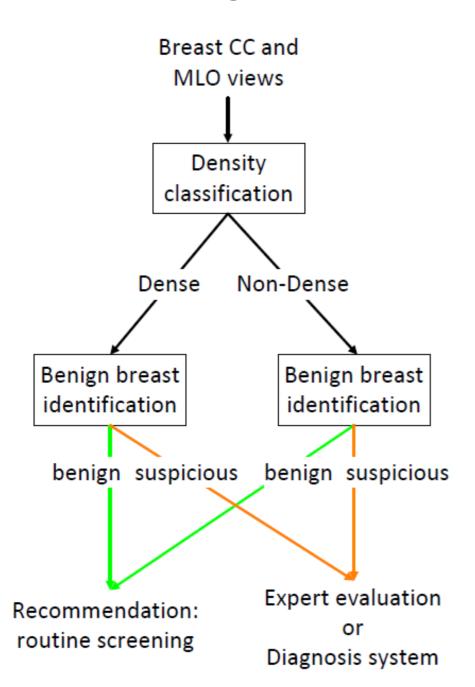


dense

- density has been associated with a higher risk of cancer
- masses and calcifications are harder to detect in dense breasts
- density decreases the sensitivity of automatic systems

Screening

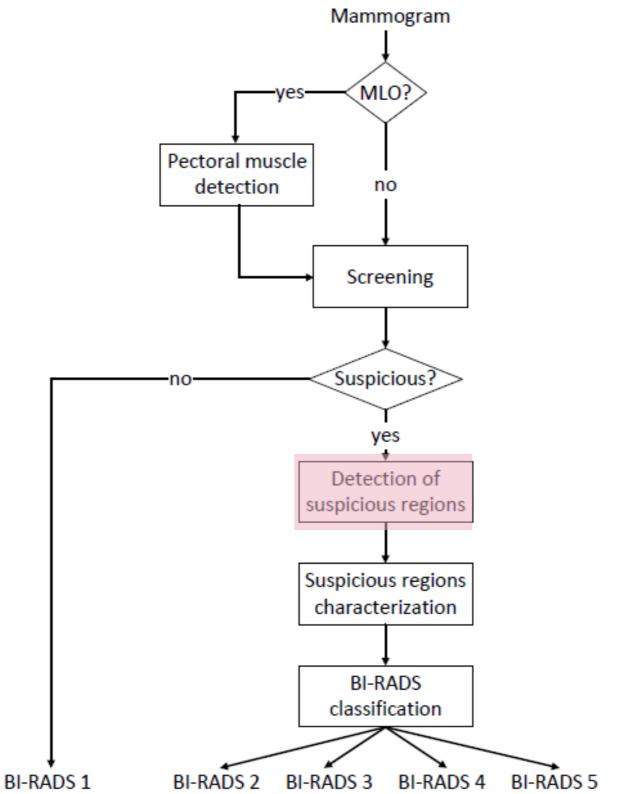




- sensitivity and FNr better than reported for human specialists
 - real clinical setting example
 - replace one of the radiologists during the double-reading
 - if a disagreement exists, the exam is sent for further investigation

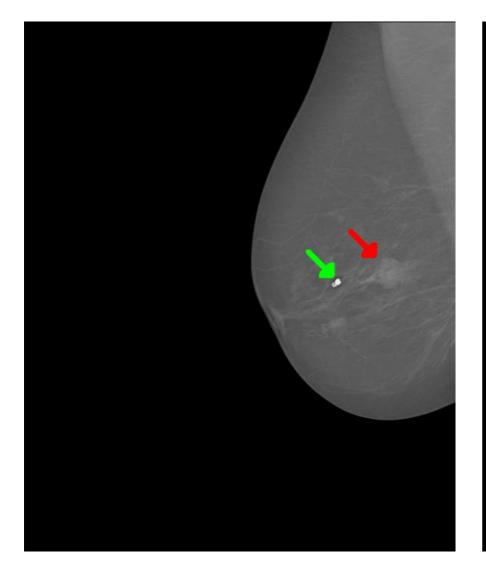
Outline

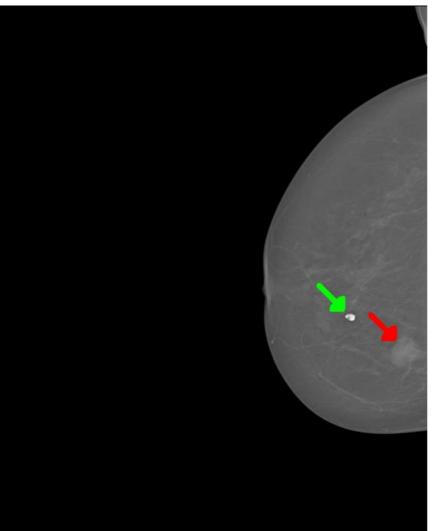




Detection of suspicious regions

Some types of suspicious regions



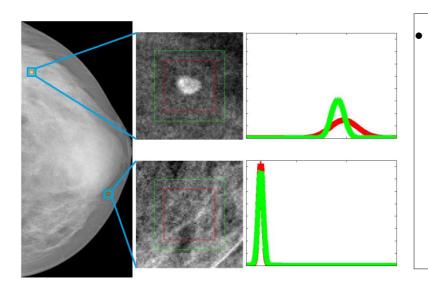




Detection of suspicious regions



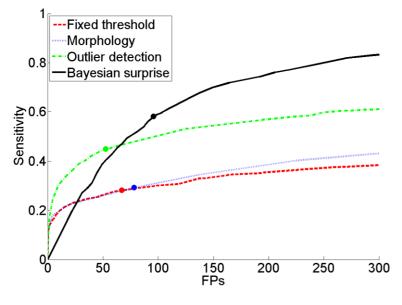
Calcifications

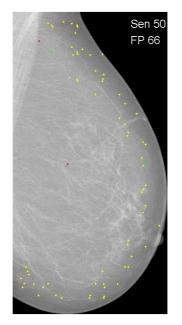


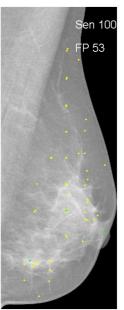
- for each patch of the image
- compute surprise

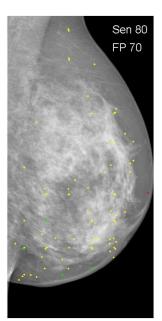
$$S(D, \mathcal{M}) = K(P(M)||P(M|D)) = \sum_{M \in \mathcal{M}} P(M) \log \frac{P(M)}{P(M|D)}$$

- if surprise > threshold
 - calcification





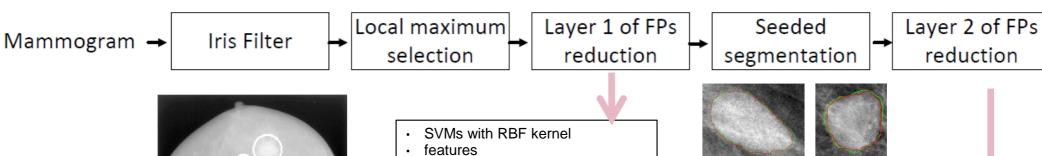


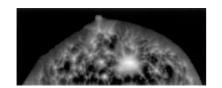


Detection of suspicious regions



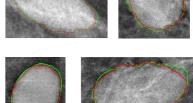
Masses

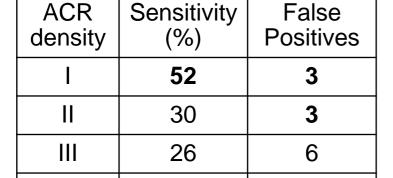




9

- original images
- intensity value
- Patch standard deviation
- o Patch 25th percentile
- Patch median value
- Patch mean value
- Patch 75th percentile
- Patch maximum intensity
- Iris filtered images
- Patch 25th percentile
- Patch median value
- Patch maximum value





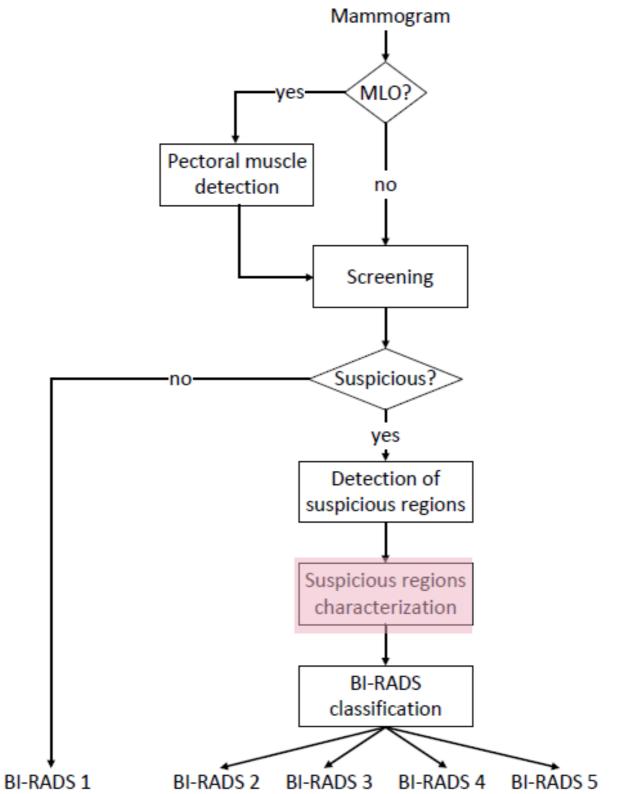
- SVMs with RBF kernel
- 9 shape features
 - area of the segmented region
 - o area of the bounding box of the region
 - area of the region's convex hull
 - eccentricity
 - length of the major axis of the ellipse that has the same normalized 2nd-moments as the region
 - length of the minor axis of the ellipse that has the same normalized 2nd-moments as the region
 - o diameter of a circle with the same area as the region, orientation
 - Perimeter
- 1 feature that uses both shape and intensity information
 - o distance between the centroid and the weighted centroid

overall performance: Sensitivity = 38% with 5 false positives

IV

Outline





Characterization of suspicious regions

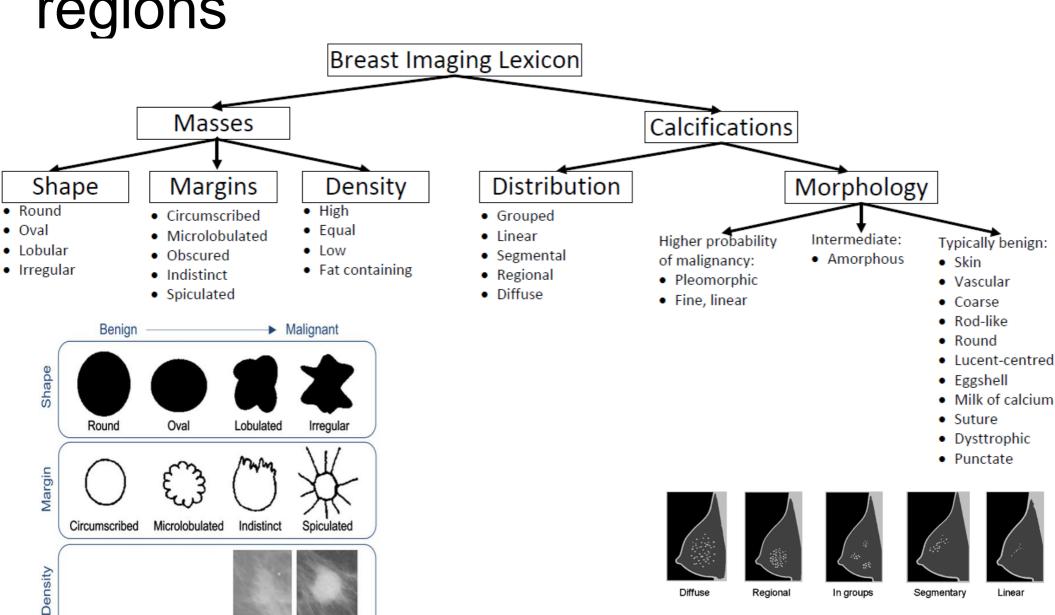
High-density

Isodense



Linear

Segmentary



Diffuse

Regional

In groups





Pearson correlation, distance correlation and Maximal Information Coefficient

7 calcification features:

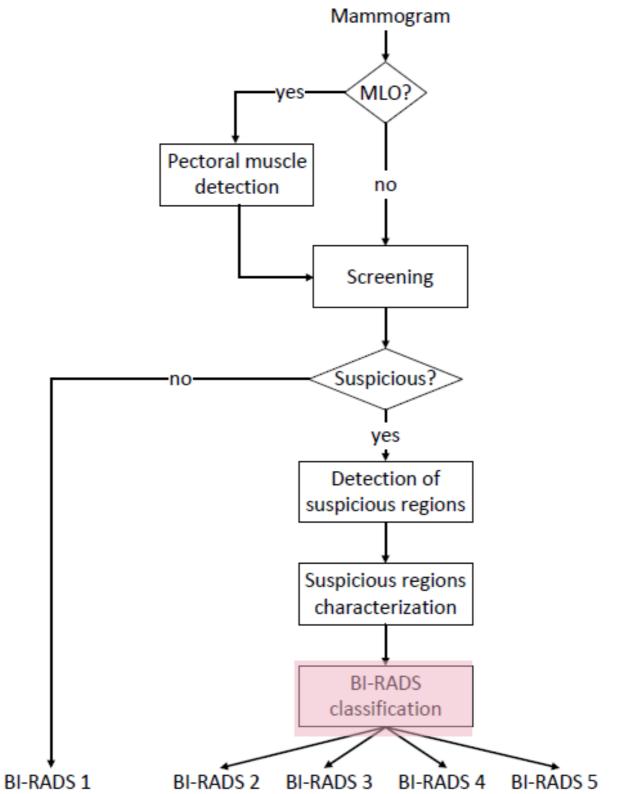
- 1. Zernike moment of order 3 and repetition +3
- 2. Zernike moment of order 4 and repetition 0
- 3. Zernike moment of order 4 and repetition -4
- 4. Eccentricity extracted from the Spatial Density Function
- 5. Minimum of the mean intensities of the calcifications
- 6. Intensity std
- 7. Std of the mean intensities of the calcifications

9 mass features

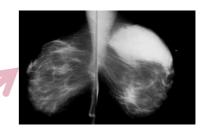
- 1. Solidity
- 2. Compactness
- 3. Thinness ratio
- 4. Skeleton end points
- 5. Shape Index
- 6. Convexification
- 7. Extent
- 8. Contained lines
- 9. $CC_2 = \sqrt{\text{(Rmin / Rmax)}}$

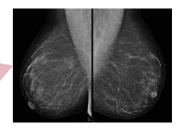
Outline

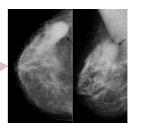


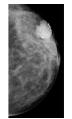


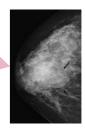
BI-RADS	Description		
0	the exam is not conclusive		
1	no findings		
2	benign findings		
3	probably benign findings		
4	suspicious findings		
5	high probability of malignancy		
6	proved cancer		



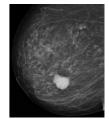


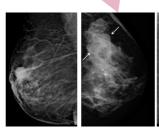


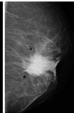




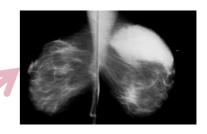


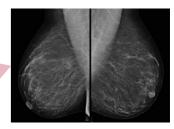


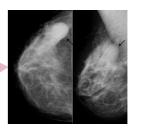


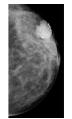


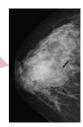
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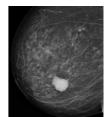


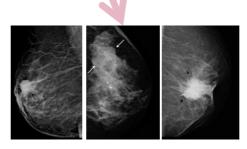




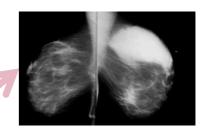


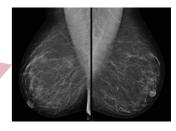


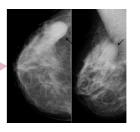


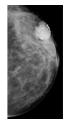


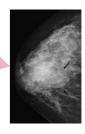
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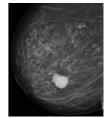


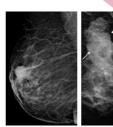


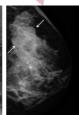


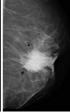




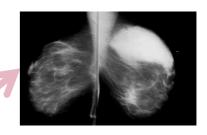


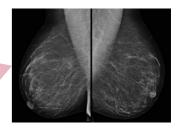


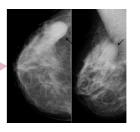


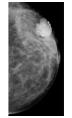


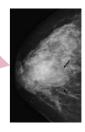
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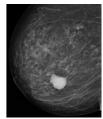


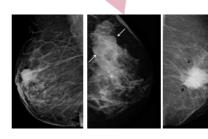




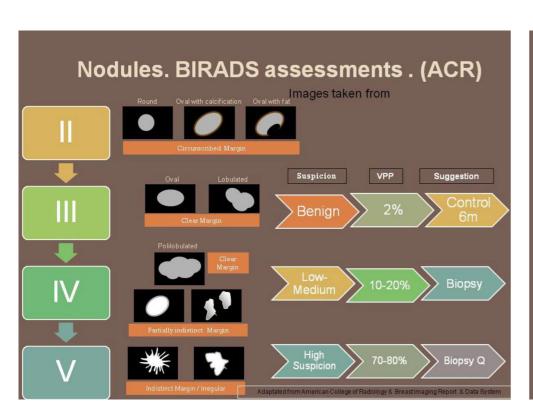


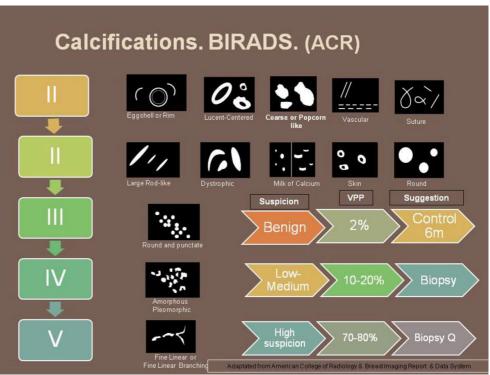






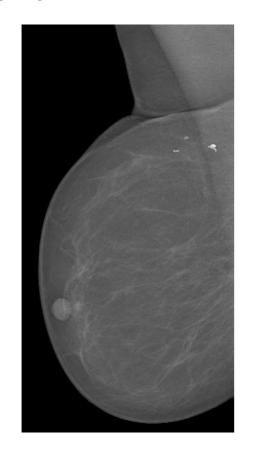




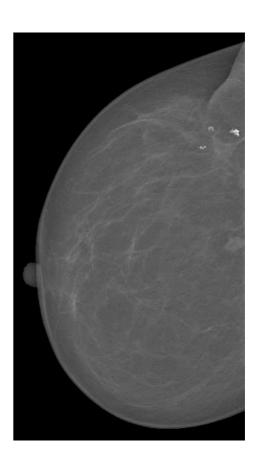


7

Motivation





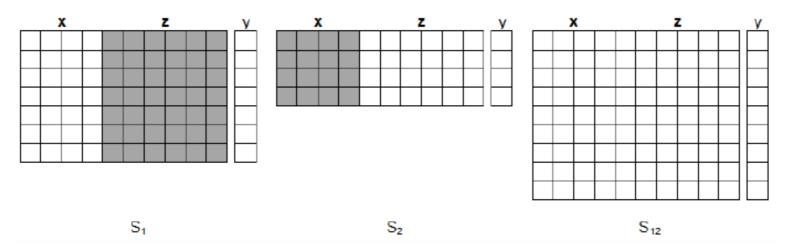


When more than one finding is present in the mammogram, the overall BIRADS in the medical report corresponds to the finding with the highest BI-RADS

7

Methods

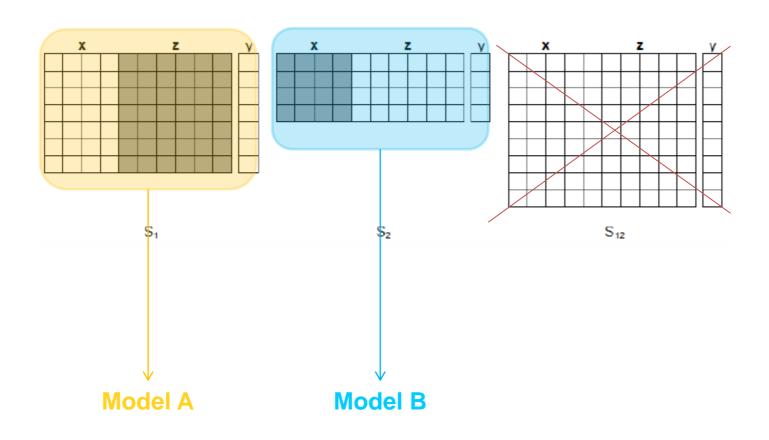
- Max Ordinal Learning (MOL)
 - MOL.LA
 - MOL.CD



Training set illustration
White represents observed and gray not present features

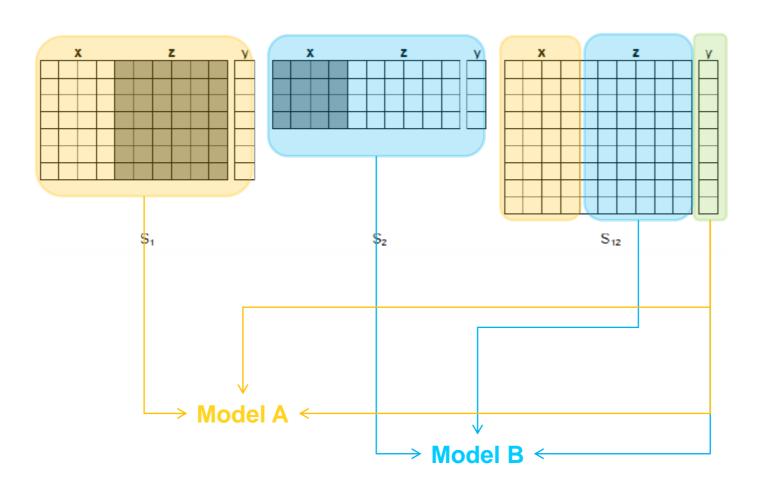
7

non-MOL



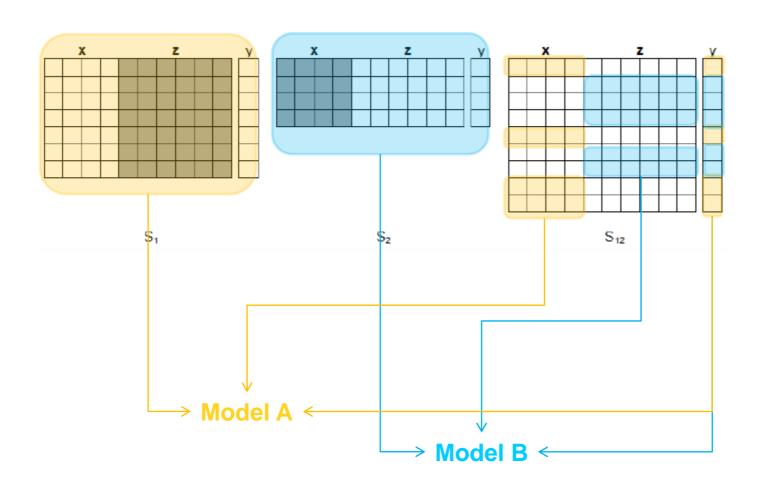
7

MOL.LA initialization



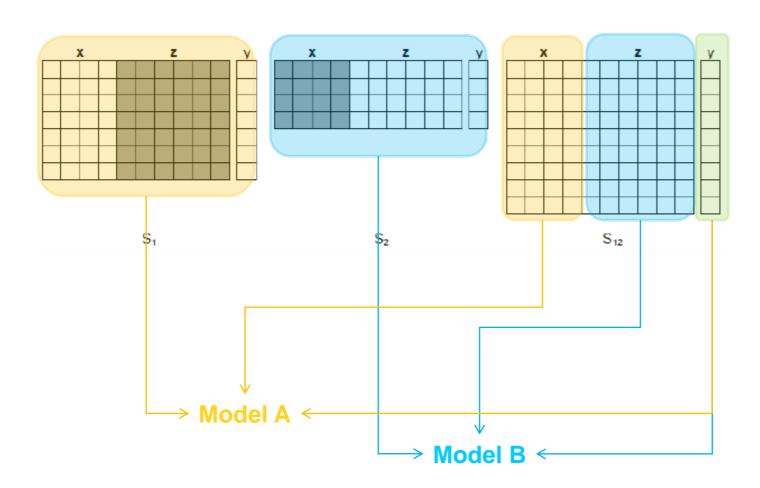
7

MOL.LA subsequent iterations



7

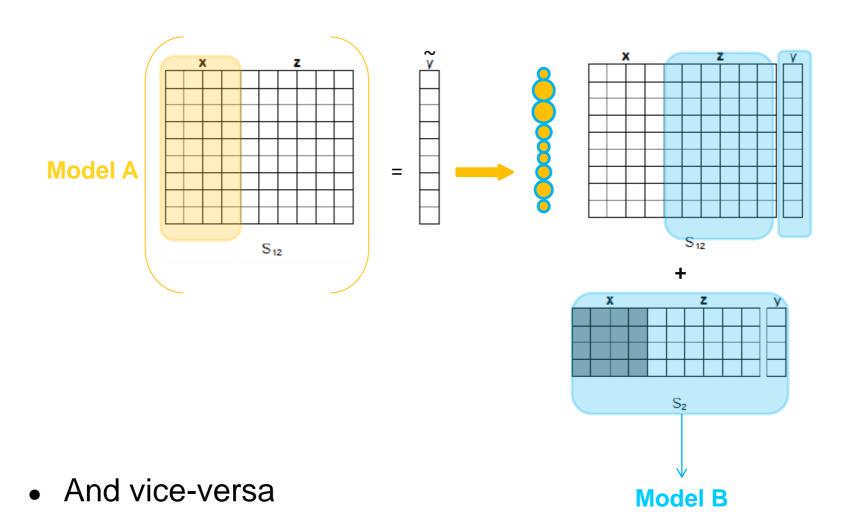
MOL.CD initialization



7

MOL.CD subsequent iterations

Consider Model A fixed and update Model B



7

Experiments

- Two kernels
 - Linear & Radial Basis Function
- Model parameterization selection
 - two-fold cross-validation
- Non-ordinal extension from binary to multi-class
 - one-against-one
 - instantiated with SVMs
- Ordinal methods
 - Frank and Hall
 - instantiated with SVMs
 - Data replication
 - instantiated with SVMs
 - KDLOR
 - instantiated LDA



7

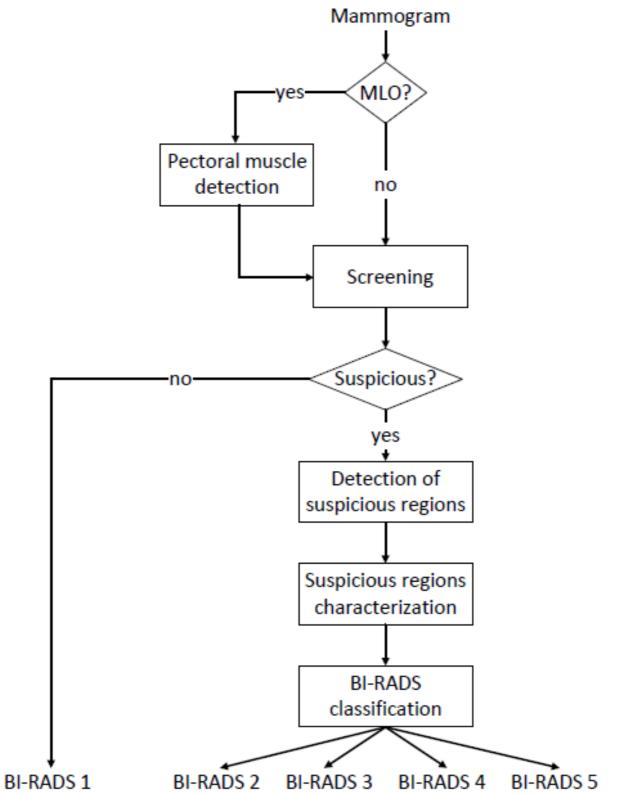
Results

		Mass contorus	
		Ground truth	CaPTOR
Baseline	Standard Model	15	13
techniques	Tri- Training	17	16
	Non- ordinal	10	7
MOL.LA	Frank&Hall	9	7
	Data Replication	7	8
	Frank&Hall	9	7
MOL.CD	Data Replication	9	7

- Automatic
 segmentation does
 not seem to
 negatively affect
 classification results
 Both the MOL.LA and
 MOL.CD techniques
 perform better than
 the standard methods
- It is sufficient to test and compare MOL.LA and MOL.CD

Outline





Putting all together



Component	Ground truth	Automatic	
	AOM = 0.65		
	CM = 0.77		
Pectoral muscle detection	AD = 0.06		
	AMED = 0.07		
	HD = 0.17		
	TPr = 0.92	TPr = 0.82	
Screening	TNr = 0.18	TNr = 0.33	
Screening	FNr = 0.08	FNr = 0.17	
	FPr = 0.82	FPr = 0.67	
Calcification detection	Sensitivity = 56.4 %	Sensitivity = 63.8 %	
Calcincation detection	FP = 47	FP = 49	
Mass detection	Sensitivity = 47.6 %	Sensitivity = 48.8 %	
IVIASS GELECTION	FP = 4	FP = 4	
BI-RADS classification	MAE = 10 %	MAE = 88 %	

Thank you!

Questions?

