

# Radiomic features and segmentation: comparison between two readers

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# 01

# INTRODUCTION

Background and aim of the study

# RADIOMICS: BACKGROUND

## OBJECT OF THE ANALYSIS

Data extracted from mammograms DCIS cases interpreted blindly by two radiologists

## AIM (1)

Assess the agreement between the two readers in terms of segmentation

## AIM (2)

Assess the reproducibility and robustness of radiomic features extracted from mammograms

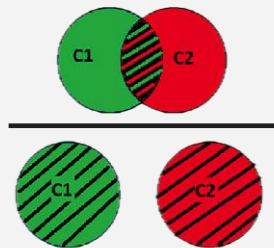
02

# STATISTICAL METHODS

# METHODOLOGY (I)

## SEGMENTATION AGREEMENT BETWEEN RADIOLOGISTS

- ★ Delineation and identification of Regions Of Interest (ROI)
- ★ Dice Similarity Coefficient (DSC):
  - Index of spatial overlap between two sets of segmentation results
  - Ranges from 0 (= absence of spatial overlap) to 1 (= complete overlap)
  - A DSC higher than 0.7 represents an adequate agreement
  - Calculated within patient



# METHODOLOGY (II)

## ROBUSTNESS OF RADIOMIC FEATURES

- ★ Quantitative features extracted from each mammogram
- ★ Intraclass Correlation Coefficient (ICC)
  - Proportion of the total variability due to between-subjects variability
  - Single measurement per subject, fixed number of readers ( $k = 2$ ), absolute agreement → two-way mixed effects model
  - Ranges from 0 (= absence of robustness) to 1 (= perfect reproducibility)
  - Calculated within features

# 03

# RESULTS



# RESULTS (1): SAMPLE DESCRIPTION

- ★ The sample consists of 394 women with diagnosis of DCIS
- ★ The mammograms were performed with five different mammographs

Mammograph	Patients	
ESS_IEO1	28	
ESS_IEO2	61	
ESS_SLU	33	
SENO_2000	4	272
SENO_DS	268	
Overall	394	

Table 1: Number of patients per mammograph

- ★ 59 radiomic features were extracted, but 56 were included in the analysis since we excluded the zero-variance
- ★ Features grouped in seven categories

Label	Features
CONVENTIONAL	10
DISCRETIZED	14
GLCM	7
GLRLM	11
GLZLM	10
NGLDM	3
SHAPE	1
Overall	56

Table 2: Number of radiomic features per category

# RESULTS: SEGMENTATION AGREEMENT

Table 3: Mean and Median DSC per mammograph

Mammograph	DSC: Mean (SD)	DSC: Median (IQR)
ESS_IEO1	0.79 (0.08)	0.80 (0.12)
ESS_IEO2	0.79 (0.12)	0.82 (0.11)
ESS_SLU	0.78 (0.10)	0.79 (0.13)
SENO	0.80 (0.12)	0.83 (0.13)
Overall	0.79 (0.11)	0.82 (0.12)

Figure 1: Dice Similarity Coefficient distribution in the sample

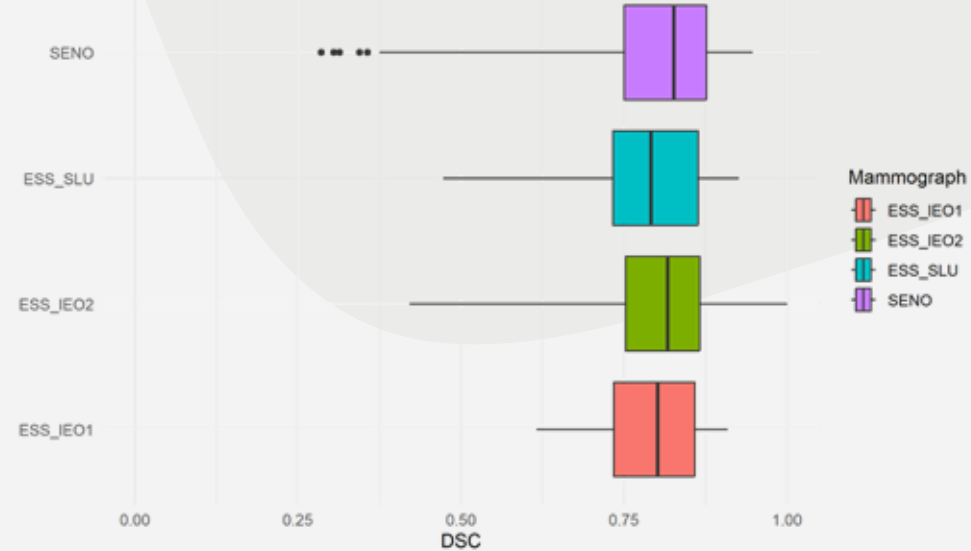
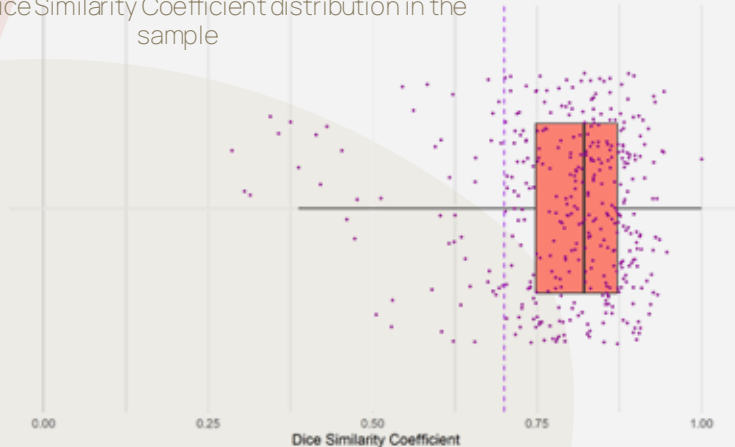


Figure 2: Boxplot of Dice Similarity Coefficient across mammographs

# RESULTS: OVERALL FEATURES ROBUSTNESS

Category	ICC	
	Mean (SD)	Median (IQR)
Conventional	0.98 (0.023)	0.99 (0.96-0.99)
Discretized	0.93 (0.027)	0.93 (0.91-0.95)
GLCM	0.95 (0.017)	0.96 (0.94-0.96)
GLRLM	0.93 (0.037)	0.93 (0.91-0.96)
GLZLM	0.93 (0.041)	0.95 (0.93-0.95)
NGLDM	0.91 (0.040)	0.91 (0.89-0.93)
SHAPE	0.95 (0.00)	0.95 (0.95-0.95)
Overall	0.94 (0.035)	0.95 (0.92-1.00)

Table 4: Mean and Median ICC per radiomic category.

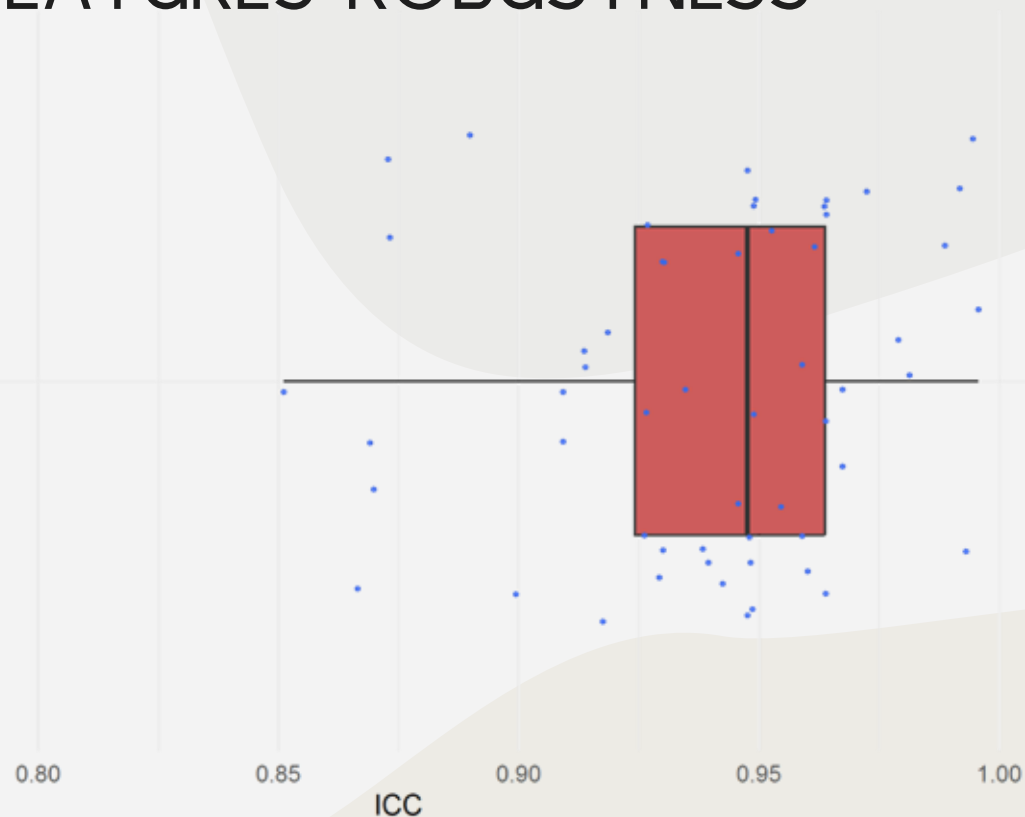


Figure 3: Boxplot of the Intraclass Correlation Coefficients overall

# RESULTS: FEATURES ROBUSTNESS

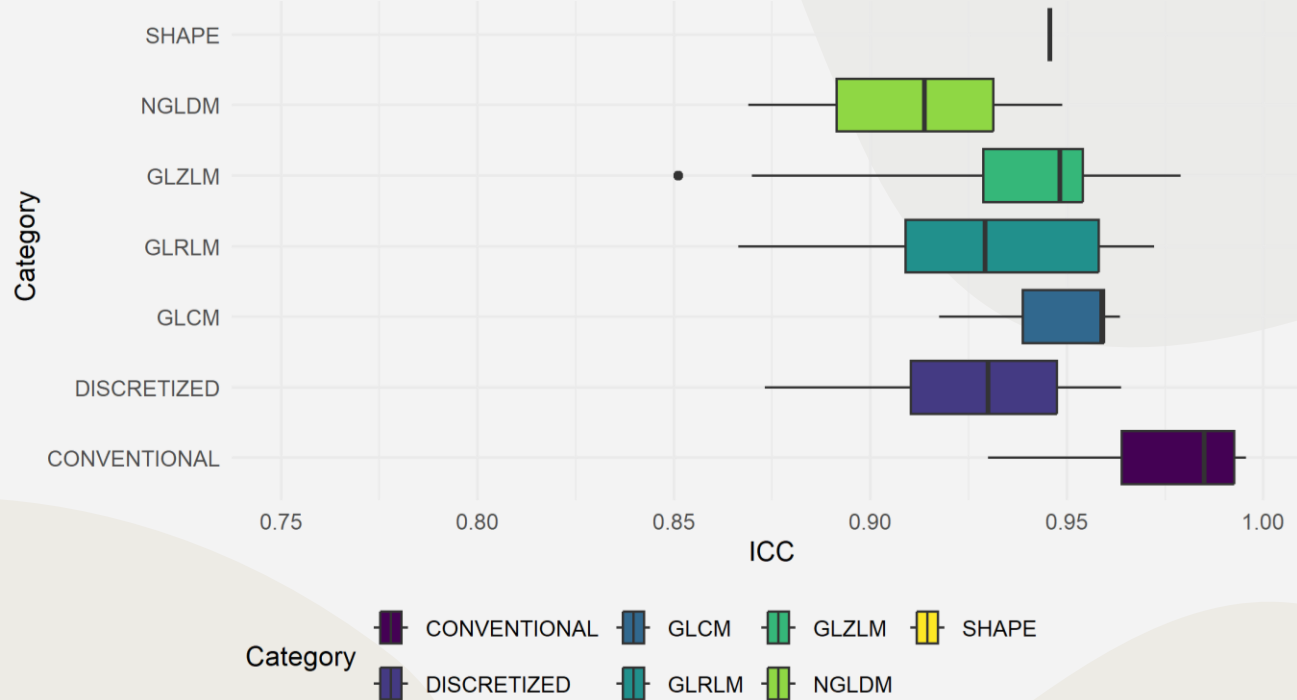


Figure 4: Boxplots of Intraclass Correlation Coefficients across features categories

# RESULTS: FEATURES ROBUSTNESS BY MAMMOGRAPH AND CATEGORY

Table 5: Mean and Median ICC per radiomic category and mammograph

		MAMMOGRAPH				
		<i>ESS_IEO1</i>	<i>ESS_IEO2</i>	<i>ESS_SLU</i>	<i>SENO</i>	Overall
CATEGORY	<i>Conventional</i>	0.97 (0.98)	0.98 (0.99)	0.95 (0.98)	0.97 (0.98)	0.98 (0.99)
	<i>Discretized</i>	0.91 (0.91)	0.94 (0.95)	<b>0.88 (0.89)</b>	0.93 (0.93)	0.93 (0.93)
	<i>GLCM</i>	0.96 (0.95)	0.95 (0.95)	0.95 (0.95)	0.95 (0.96)	0.95 (0.96)
	<i>GLRLM</i>	0.90 (0.93)	0.95 (0.95)	<b>0.88 (0.85)</b>	0.93 (0.93)	0.93 (0.93)
	<i>GLZLM</i>	0.91 (0.93)	0.94 (0.95)	<b>0.88 (0.92)</b>	0.94 (0.95)	0.93 (0.93)
	<i>NGLDM</i>	<b>0.89 (0.88)</b>	0.92 (0.90)	0.91 (0.90)	0.91 (0.92)	0.91 (0.91)
	<i>SHAPE</i>	<b>0.89 (0.89)</b>	0.98 (0.98)	0.92 (0.92)	0.95 (0.95)	0.95 (0.95)
	Overall	0.92 (0.94)	0.95 (0.95)	0.90 (0.93)	0.94 (0.95)	0.94 (0.95)

# RESULTS: FEATURES ROBUSTNESS BY MAMMOGRAPH AND CATEGORY

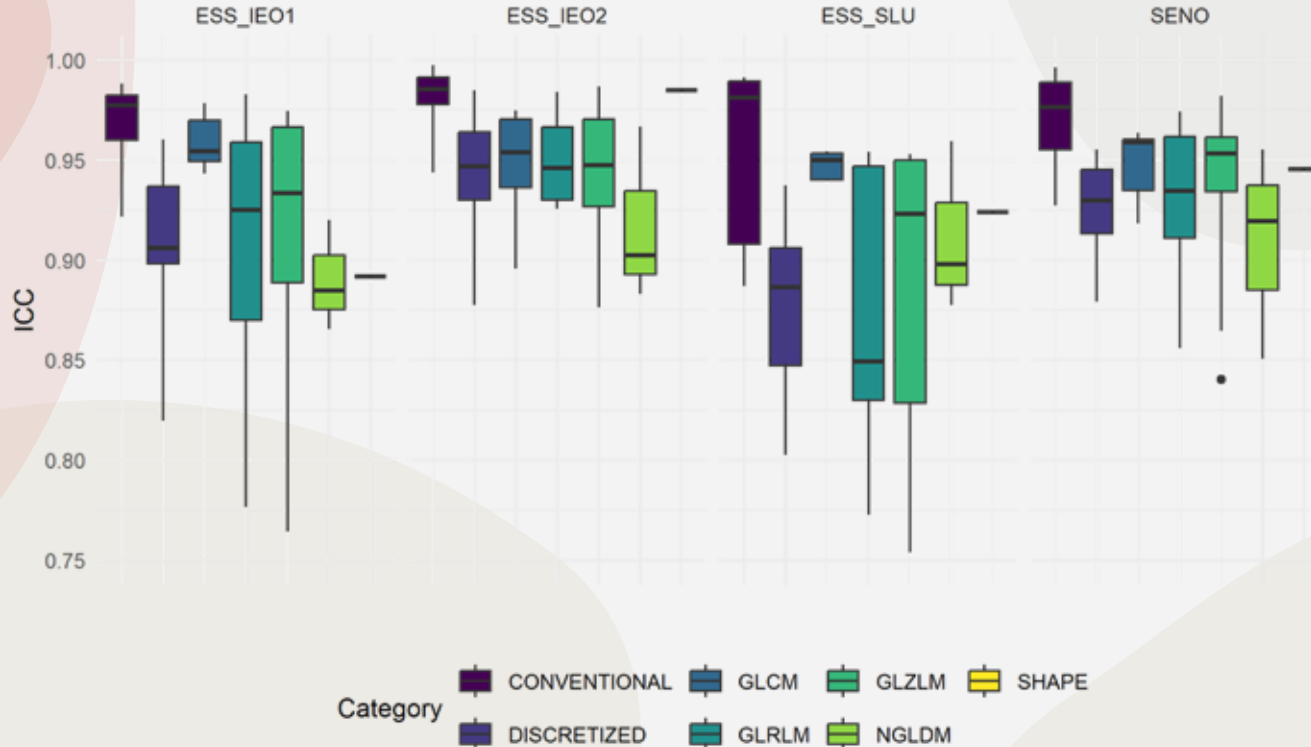


Figure 5: Boxplots of Intraclass Correlation Coefficients across features categories and mammographs

# 04

## DISCUSSION

Limits and further developments

# CONCLUSIONS

- ★ A good agreement has been observed, related both to the segmentation and the features extracted
- ★ LIMITATIONS
  - ★ Limited number of radiomic features to analyse
  - ★ Features are imbalanced between category, especially for the shape
- ★ NEXT STEP
  - ★ Employ radiomic features as predictors in models to predict the upstaging of DCIS to invasive breast cancer.
  - ★ Prognostic models to identify high-risk patients integrating both clinical and radiomic features



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  - ★ Fisica Medica (IEO): Radiomic Features extraction and computation of Dice Similarity Coefficient
  - ★ Molecular and Pharmacoepidemiology Unit, Department of Experimental Oncology (IEO): Analysis of segmentation agreement and features robustness via Intraclass Correlation Coefficient

THANK YOU FOR YOUR  
ATTENTION!

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