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# Tsallis Entropy-based mass and non-mass classification

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## 1 Introduction

Breast cancer is the most frequent disease among women. In 2012 about 1.7 millions of new cases were registered around the world [1]. Screening mammography provides a significant reduction on breast cancer mortality rate and computational methods are improving this exam efficiency. This paper presents a new approach for mass and non-mass classification based on Tsallis entropy feature extraction from singular value decomposition (SVD).

## 2 Methodology

Figure 1 shows the pipeline of the process of this work.

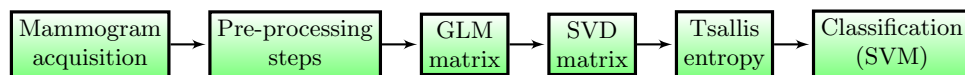


Figura 1: Flowchart

Based on [2], gray-level co-occurrence matrix (GLCM) was calculated for each quantized image on database fixing a distance  $d$  and varying the direction  $\theta$ . From GLCM results, SVD was applied using Equation (1).

$$A = U_{(G \times G)} \Sigma_{(G \times G)} V_{(G \times G)}^T \quad (1)$$

where  $G$  corresponds to gray-level value ( $G \in \{2^3, 2^4, 2^5, 2^6, 2^7, 2^8\}$ ),  $U$  and  $V$  matrices are composed by eigenvectors of  $AA^T$  and  $A^T A$ , respectively.  $\Sigma$  matrix represents the

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diagonal matrix ( $\Sigma = \text{diag}(\sigma_1, \sigma_2, \sigma_3, \dots, \sigma_G)$ ) composed by all singular values. These values were used then to calculate Tsallis entropy [3] through Equation (2)

$$S = \frac{1 - \sum_{i=0}^G (\sigma_i)^q}{q - 1} \quad (2)$$

where  $q$  is an real number that can be changed according to the application.

### 3 Results

Exhaustive tests were developed to refine q-index range and the best results were provided by  $q \in [1.71, 1.79]$ . The best results are described on Table 1.

Five metrics statistics were calculated over classification results: accuracy, sensibility, specificity, positive predict value and negative predict value.

Tabela 1: Best results for  $q \in [1.71, 1.79]$ .

$q$	$d$	Statistics (%)				
		Acc.	Sens.	Spec.	PPV	NPV
1.73	1	82.2	79.4	84.6	83.8	80.4
1.76	1	84.2	79.8	88.7	87.6	81.4
1.79	1	82.6	74.5	90.7	88.9	78.0

### 4 Conclusions

The main proposal was to provide a novel approach for mass and non-mass classification based on Tsallis entropy extraction from SVD of GLCM matrices. To improve accuracy rate, q-index value were adjusted and the best value was provided for pair  $(q, d) = (1.76, 1)$  where accuracy was 84.2%.

### Referências

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