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# Inteligência Artificial

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Reconhecimento de  
Movimentos de Mãos

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# Objetivo

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Através de exemplos de vídeos (explodidos em frames) de 9 tipos diferentes de movimentos de mãos, extrair características usando a biblioteca OpenCV, e usar a base de dados gerada para treinar algoritmos de machine learning com auxílio do Weka.

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# Seleção de Frames

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O programa extrator recebe como input o número de frames a serem extraídos de cada exemplo

Por exemplo,  $N=5$ :

0%, 25%, 50%, 75%, 100%

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# Segmentação das Imagens

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Usando HSV com a faixa de cor da pele



$[0, 30, 60]$  to  $[50, 150, 255]$

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# Extração de Características

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1. Achar o contour da imagem segmentada com maior área
2. Derivar primitivas: retângulo de área, mínima, elipse englobante, convex hull
3. Derivar características:

area, perimeter, convex\_hull\_area, solidity, rect\_center\_x, rect\_center\_y, rect\_width, rect\_height, rect\_angle, rect\_aspect\_ratio, ellipse\_center\_x, ellipse\_center\_y, ellipse\_major\_axis, ellipse\_minor\_axis, ellipse\_angle, farthest\_convex\_defect, hu\_moment\_1, hu\_moment\_2, hu\_moment\_3, hu\_moment\_4, hu\_moment\_5, hu\_moment\_6, hu\_moment\_7

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# Arquivo .arff

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Script que faz a extração das características e escreve num arquivo.

Neste exemplo, usa 5 frames por exemplo e usa todos os sets (1 a 5), salvando num arquivo:

```
$ python extractor/arff_generator.py 5 1 2 3 4 5 > All.arff
```

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# Testando as Funções

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Os seguintes testes usam somente 2 frames por exemplo.

Os parâmetros foram variados agressivamente.

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# Árvore de Decisão

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`**binarySplits**` had no influence on the results.

`**confidenceFactor**` did not impact much on classification performance, but higher levels slowed down the classification quite a bit. We maintained it at 0.25.

`**minNumObj**` was optimal at 10.

`**reducedErrorPruning**` deemed worse results, so we turned it off.

Time taken to build model: 0.33 seconds

Correctly Classified Instances	634	70.4444 %
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Kappa statistic	0.6675
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# K-Nearest Neighbors

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`**KNN**` was optimal at 1.

`**distanceWeighting**` had no influence on the results.

`**meanSquares**` had no influence on the results.

`**nearestNeighbourSearchAlgorithm**` we tried different ones, but all yielded the same result.

Time taken to build model: 0 seconds

Correctly Classified Instances	712	79.1111 %
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Kappa statistic	0.765
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# SVM

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**\*\*SVMType\*\*** nu-CSV was better than C-CSV.

**\*\*coef0\*\*** 1.0 (small influence).

**\*\*cost\*\*** 1.0 (no influence).

**\*\*degrees\*\*** was optimal at 3

**\*\*nu\*\*** was optimal at 0.1

Time taken to build model: 1.75 seconds

Correctly Classified Instances	771	85.6667 %
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Kappa statistic	0.8388
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# Neural Network

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`**hiddenLayers** t (attrs+classes)`

`**learningRate** 0.3`

`**momentum** 0.5`

`**trainingTime** 500 (diminishing returns).`

Time taken to build model: 22.3 seconds

Correctly Classified Instances	780	86.6667 %
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Kappa statistic	0.85
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# Testando as Características

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Usando os parâmetros ótimos encontrados anteriormente, variamos as características para descobrir se há alguma que influencia negativamente os resultados.

Para isso, foi usando o Set1 com 2 frames.

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# Testando as Características

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A retirada de características se mostrou prejudicial ao desempenho dos algoritmos de classificação.

Exceto por “`farthest_convex_defect`”, que indica o maior defeito de convexidade.

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# Farthest Convex Defect

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Rede Neural

Com: 96.1111 %

Sem: 96.6667 %

Outros classificadores apresentaram resultados similares

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# Maximizando

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Somente com o Set 1, com 10 frames/exemplo, usando os parâmetros relatados anteriormente.

Mostramos também as matrizes de confusão.

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# Árvore de Decisão

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Time taken to build model: 0.15 seconds

Correctly Classified Instances 141 78.3333 %

Kappa statistic 0.7563

	a	b	c	d	e	f	g	h	i	<-- classified as
15	1	1	0	0	0	0	3	0	0	a = 0000
0	16	0	0	1	0	0	0	2	1	b = 0001
1	0	16	0	0	0	0	2	0	1	c = 0002
1	0	0	18	0	0	0	1	0	0	d = 0003
0	0	1	0	18	0	0	0	0	1	e = 0004
0	1	0	0	3	13	1	0	2	0	f = 0005
1	0	0	3	0	1	15	0	0	0	g = 0006
0	0	1	0	2	0	0	15	2	0	h = 0007
0	0	0	0	1	1	2	1	15	0	i = 0008



# K-Nearest Neighbors

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Time taken to build model: 0 seconds

Correctly Classified Instances 164 91.1111 %

Kappa statistic 0.9

	a	b	c	d	e	f	g	h	i	<-- classified as
17	0	0	2	0	0	1	0	0	0	a = 0000
0	19	0	0	1	0	0	0	0	0	b = 0001
0	0	19	0	0	0	0	0	0	1	c = 0002
0	0	0	17	0	0	3	0	0	0	d = 0003
0	1	0	0	18	0	0	1	0	0	e = 0004
0	1	0	0	0	18	0	0	1	0	f = 0005
0	0	0	0	0	0	20	0	0	0	g = 0006
0	0	0	0	0	0	0	18	2	0	h = 0007
0	0	0	0	0	0	1	1	18	0	i = 0008

# SVM

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Time taken to build model: 1.13 seconds

Correctly Classified Instances 172 95.5556 %

Kappa statistic 0.950

a	b	c	d	e	f	g	h	i	<-- classified as
20	0	0	0	0	0	0	0	0	a = 0000
0	19	0	0	0	1	0	0	0	b = 0001
0	0	19	0	0	0	0	0	1	c = 0002
0	0	0	19	0	0	1	0	0	d = 0003
0	0	0	0	19	0	0	1	0	e = 0004
0	0	0	0	0	20	0	0	0	f = 0005
0	0	0	0	0	0	20	0	0	g = 0006
0	0	0	0	0	0	0	20	0	h = 0007
0	0	0	0	0	0	0	0	20	i = 0008

# Neural Network

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Time taken to build model: 89.95 seconds

Correctly Classified Instances 174 96.6667 %

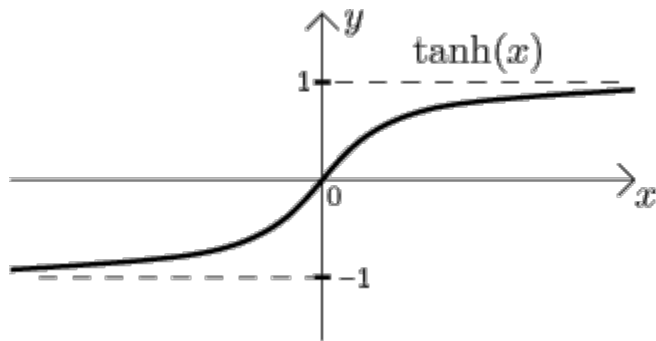
Kappa statistic 0.9625

a	b	c	d	e	f	g	h	i	<-- classified as
20	0	0	0	0	0	0	0	0	a = 0000
0	19	0	0	0	1	0	0	0	b = 0001
0	0	20	0	0	0	0	0	0	c = 0002
0	0	0	19	0	0	1	0	0	d = 0003
0	0	0	0	19	0	0	1	0	e = 0004
0	0	0	0	0	20	0	0	0	f = 0005
0	0	0	0	0	0	20	0	0	g = 0006
0	0	0	0	0	0	0	20	0	h = 0007
0	0	1	0	0	0	2	0	17	i = 0008

# Implementação

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Implementação em Ruby de um algoritmo de Rede Neural com backpropagation



# Implementação - XOR

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2 input, 1 output, 2 hidden cells in 1 hidden layer

Uso:

```
xor = [  
    [[0,0], [0]],  
    [[0,1], [1]],  
    [[1,0], [1]],  
    [[1,1], [0]]]
```

```
NeuralNetwork  
.new(2, 2, 1)  
.train(xor)  
.test(xor)  
.internals
```

Resultado:

Got 4/4 -> 100.0%

Input Activation:

1.0  
1.0  
1.0

Hidden Layer Activation:

-0.9999648742143428  
-0.9678124516804766

Output Activation:

-0.19664136237293744

Input Weights:

[[-1.9706870887241765, -3.794446904016083],  
 [-1.9823678135520832, -3.811420648109506],  
 [1.9021664183757332, 2.1829678415844036]]

Output Weights:

[[2.6446273179756994],  
 [-2.810229779367159]]

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