

# Visión Artificial

## Práctica 4: El clasificador de Bayes

**David Martín Gómez**

Laboratorio de Sistemas Inteligentes

Universidad Carlos III de Madrid

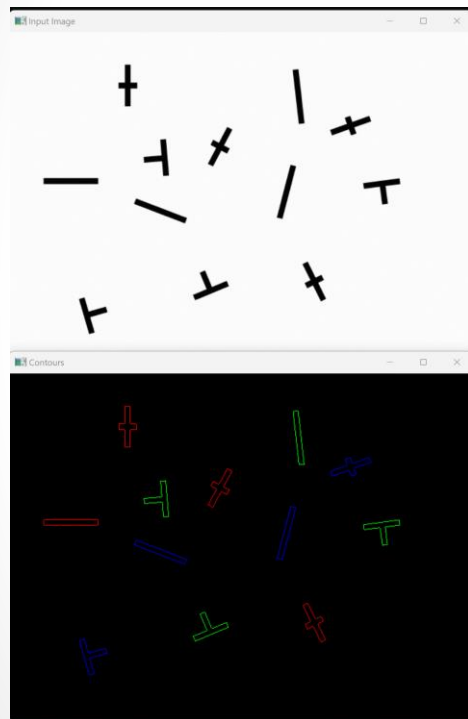
# Índice

- Objetivos
- Etiquetado de objetos
- Obtención de descriptores
- El clasificador de Bayes

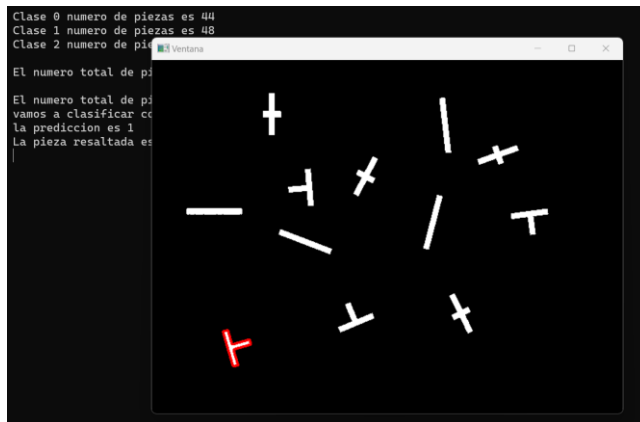
# Índice

- Objetivos
- Etiquetado de objetos
- Obtención de descriptores
- El clasificador de Bayes

# Objetivos



	A	B	C	D	E	F	G	H	I
1	perimetro	area	Hu1	Hu2	Hu3	Hu4	Hu5	Hu6	Hu7
2	164	540	0.862	0.714454	4.73E-05	3.97E-05	1.72E-09	3.35E-05	-9.62E-12
3	165	535	0.861	0.713894	6.41E-06	2.40E-06	8.56E-12	1.21E-06	-3.89E-12
4	168	533	0.873	0.733276	1.02E-05	2.26E-06	6.34E-12	3.11E-07	-8.85E-12
5	168	532	0.873	0.734509	6.56E-06	5.65E-06	3.44E-11	4.76E-06	5.20E-13
6	167	521	0.884	0.753504	1.70E-05	1.14E-05	1.59E-10	9.63E-06	9.88E-12
7	166	516	0.899	0.780266	3.78E-05	2.83E-05	9.22E-10	2.36E-05	-5.08E-11
8	168	520	0.888	0.760736	8.37E-06	1.26E-06	-1.47E-13	-4.05E-07	-4.09E-12
9	168	519	0.886	0.75758	2.45E-05	3.60E-06	-3.31E-12	-1.37E-06	3.36E-11
10	172	519	0.888	0.759855	2.48E-06	1.44E-06	2.70E-12	9.69E-07	4.05E-13
11	173	520	0.889	0.761395	1.37E-06	8.94E-07	9.84E-13	6.69E-07	-9.35E-14
12	174	518	0.889	0.761703	3.74E-05	2.88E-05	9.42E-10	2.37E-05	-4.57E-11



# Etiquetado de objetos

- Imágenes
  - Pieza1.png,
  - Pieza2.png,
  - Pieza3.png,
  - Piezas.png

```
using namespace cv;
using namespace std;

int main() {

    Mat image = imread("../imagenes/Piezas.png");

    if (!image.data) {
        cout << "Error!!\n";
        return 1;
    }
    imshow("Input Image", image);

    //Umbralizamos
    cvtColor(image, image, COLOR_BGR2GRAY);
    threshold(image, image, 128, 255, THRESH_BINARY_INV);

    //Encontramos los contornos
    vector<vector<Point>> > contours;
    Mat contourOutput = image.clone();
    findContours(contourOutput, contours, RETR_LIST, CHAIN_APPROX_NONE);

    //Los dibujamos
    Mat contourImage(image.size(), CV_8UC3, Scalar(0, 0, 0));
    Scalar colors[3];
    colors[0] = Scalar(255, 0, 0);
    colors[1] = Scalar(0, 255, 0);
    colors[2] = Scalar(0, 0, 255);
    for (size_t idx = 0; idx < contours.size(); idx++)
        drawContours(contourImage, contours, idx, colors[idx % 3]);
    imshow("Contours", contourImage);

    waitKey(0);
    return 0;
}
```

# Etiquetado de objetos

## ◆ findContours() [1/2]

```
void cv::findContours ( InputArray      image,
                       OutputArrayOfArrays contours,
                       OutputArray     hierarchy,
                       int              mode,
                       int              method,
                       Point            offset = Point() )
```

### Python:

```
cv.findContours( image, mode, method[, contours[, hierarchy[, offset]]] ) -> contours, hierarchy
```

```
#include <opencv2/imgproc.hpp>
```

Finds contours in a binary image.

The function retrieves contours from the binary image using the algorithm [239]. The contours are a useful tool for shape analysis and object detection and recognition. See squares.cpp in the OpenCV sample directory.

### Note

Since opencv 3.2 source image is not modified by this function.

### Parameters

- image** Source, an 8-bit single-channel image. Non-zero pixels are treated as 1's. Zero pixels remain 0's, so the image is treated as binary. You can use `compare`, `inRange`, `threshold`, `adaptiveThreshold`, `Canny`, and others to create a binary image out of a grayscale or color one. If mode equals to `RETR_CCOMP` or `RETR_FLOODFILL`, the input can also be a 32-bit integer image of labels (CV\_32SC1).
- contours** Detected contours. Each contour is stored as a vector of points (e.g. `std::vector<std::vector<cv::Point> >`).
- hierarchy** Optional output vector (e.g. `std::vector<cv::Vec4i>`), containing information about the image topology. It has as many elements as the number of contours. For each *i*-th contour `contours[i]`, the elements `hierarchy[i][0]`, `hierarchy[i][1]`, `hierarchy[i][2]`, and `hierarchy[i][3]` are set to 0-based indices in contours of the next and previous contours at the same hierarchical level, the first child contour and the parent contour, respectively. If for the contour *i* there are no next, previous, parent, or nested contours, the corresponding elements of `hierarchy[i]` will be negative.

## ◆ RetrievalModes

```
enum cv::RetrievalModes
```

```
#include <opencv2/imgproc.hpp>
```

mode of the contour retrieval algorithm

### Enumerator

RETR_EXTERNAL Python: cv.RETR_EXTERNAL	retrieves only the extreme outer contours. It sets <code>hierarchy[i][2]=hierarchy[i][3]=-1</code> for all the contours.
RETR_LIST Python: cv.RETR_LIST	retrieves all of the contours without establishing any hierarchical relationships.
RETR_CCOMP Python: cv.RETR_CCOMP	retrieves all of the contours and organizes them into a two-level hierarchy. At the top level, there are external boundaries of the components. At the second level, there are boundaries of the holes. If there is another contour inside a hole of a connected component, it is still put at the top level.
RETR_TREE Python: cv.RETR_TREE	retrieves all of the contours and reconstructs a full hierarchy of nested contours.
RETR_FLOODFILL Python: cv.RETR_FLOODFILL	

## ◆ ContourApproximationModes

```
enum cv::ContourApproximationModes
```

```
#include <opencv2/imgproc.hpp>
```

the contour approximation algorithm

### Enumerator

CHAIN_APPROX_NONE Python: cv.CHAIN_APPROX_NONE	stores absolutely all the points. The resulting contour will be either horizontal, vertical, or diagonal.
CHAIN_APPROX_SIMPLE Python: cv.CHAIN_APPROX_SIMPLE	compresses horizontal, vertical, and diagonal segments of the contour to their endpoints.
CHAIN_APPROX_TC89_L1 Python: cv.CHAIN_APPROX_TC89_L1	applies one of the flavors of the Teukolsky-Chang-Kanani (TC89) algorithm.
CHAIN_APPROX_TC89_KCOS Python: cv.CHAIN_APPROX_TC89_KCOS	applies one of the flavors of the Teukolsky-Chang-Kanani (TC89) algorithm.

# Índice

- Objetivos
- Etiquetado de objetos
- Obtención de descriptores
- El clasificador de Bayes

# Obtención de descriptores

- [cv::arcLength](#)
- [cv::contourArea](#)
- [cv::convexHull](#)
- [cv::HuMoments](#)
- [cv::minAreaRect](#)
- [cv::minEnclosingCircle](#)
- [cv::minEnclosingTriangle](#)
- [cv::moments](#)



# Perímetro y área

## ◆ `arcLength()`

```
double cv::arcLength ( InputArray curve,
                      bool          closed
                      )
```

Python:

```
cv.arcLength( curve, closed ) -> retval
```

## ◆ `contourArea()`

```
double cv::contourArea ( InputArray contour,
                         bool          oriented = false
                         )
```

Python:

```
cv.contourArea( contour[, oriented] ) -> retval
```

# Momentos

## ◆ moments()

```
Moments cv::moments ( InputArray array,
                        bool binaryImage = false
                      )
```

Python:

```
cv.moments( array[, binaryImage] ) -> retval
```

## ◆ HuMoments() [1/2]

```
void cv::HuMoments ( const Moments & moments,
                     double hu[7]
                   )
```

Python:

```
cv.HuMoments( m[, hu] ) -> hu
```

# ..y más

## ◆ convexHull()

```
void cv::convexHull ( InputArray points,
                    OutputArray hull,
                    bool clockwise = false,
                    bool returnPoints = true )
```

## ◆ minAreaRect()

```
RotatedRect cv::minAreaRect ( InputArray points )
```

Python:

```
cv.minAreaRect( points ) -> retval
```

## ◆ minEnclosingCircle()

```
void cv::minEnclosingCircle ( InputArray points,
                             Point2f & center,
                             float & radius )
```

## ◆ minEnclosingTriangle()

```
double cv::minEnclosingTriangle ( InputArray points,
                                   OutputArray triangle )
```

Python:

```
cv.minEnclosingTriangle( points[, triangle] ) -> retval
```

## ◆ boxPoints()

```
void cv::boxPoints ( RotatedRect box,
                    OutputArray points )
```

Python:

```
cv.boxPoints( box[, points] ) -> points
```

```
#include <opencv2/imgproc.hpp>
```

Finds the four vertices of a rotated rect. Useful to draw the rotated rectangle.

The function finds the four vertices of a rotated rectangle. This function is useful to draw the rectangle. In C++, instead of using this function, you can directly use `RotatedRect::points` method. Please visit the [tutorial on Creating Bounding rotated boxes and ellipses for contours](#) for more information.

### Parameters

**box** The input rotated rectangle. It may be the output of `cv::minAreaRect`.

**points** The output array of four vertices of rectangles.

# Índice

- Objetivos
- Etiquetado de objetos
- Obtención de descriptores
- El clasificador de Bayes

# El clasificador de Bayes

- Hay que crear las estructuras del clasificador.
  - Primero los datos

```
// creación de las matrices para almacenar la información
Mat train_data(nentre, NUMDESCRIPTORES, CV_32FC1);
Mat response_data(nentre, NUMDESCRIPTORES, CV_32FC1);

//Convertir los datos de una matriz al tipo TrainData
static Ptr<ml::TrainData> prepare_train_data(const Mat& data, const Mat& responses, int ntrain_samples)
{
    Mat sample_idx = Mat::zeros(1, data.rows, CV_8U);
    Mat train_samples = sample_idx.colRange(0, ntrain_samples);
    train_samples.setTo(Scalar::all(1));

    int nvars = data.cols;
    Mat var_type(nvars + 1, 1, CV_8U);
    var_type.setTo(Scalar::all(ml::VAR_ORDERED));
    var_type.at<uchar>(nvars) = ml::VAR_CATEGORICAL;

    return ml::TrainData::create(data, ml::ROW_SAMPLE, responses,
                                  noArray(), sample_idx, noArray(), var_type);
}
```

# El clasificador de Bayes

- Hay que crear las estructuras del clasificador.
  - Luego entrenamos

```
//ponemos la informacion en el formato que le gusta a Opencv
Ptr<ml::TrainData> tdata = prepare_train_data(train_data, response_data, nentre);

//Creacion del clasificador de Bayes
Ptr<ml::NormalBayesClassifier> bayes = ml::NormalBayesClassifier::create();
// Entrenamos
bayes->train(tdata);
```

## Member Function Documentation

### ◆ create()

static Ptr<NormalBayesClassifier> cv::ml::NormalBayesClassifier::create ( )

Python:

```
cv.ml.NormalBayesClassifier.create( ) -> retval
cv.ml.NormalBayesClassifier_create( ) -> retval
```

Creates empty model Use StatModel::train to train the model after creation.

### ◆ load()

static Ptr<NormalBayesClassifier> cv::ml::NormalBayesClassifier::load ( const String & filepath,  
const String & nodeName = String()  
)

Python:

```
cv.ml.NormalBayesClassifier.load( filepath[, nodeName] ) -> retval
cv.ml.NormalBayesClassifier_load( filepath[, nodeName] ) -> retval
```

Loads and creates a serialized **NormalBayesClassifier** from a file.

Use **NormalBayesClassifier::save** to serialize and store an **NormalBayesClassifier** to disk. Load the **NormalBayesClassifier** from this file again, by calling this function with the path to the file. Optionally specify the node for the file containing the classifier

#### Parameters

**filepath** path to serialized **NormalBayesClassifier**  
**nodeName** name of node containing the classifier

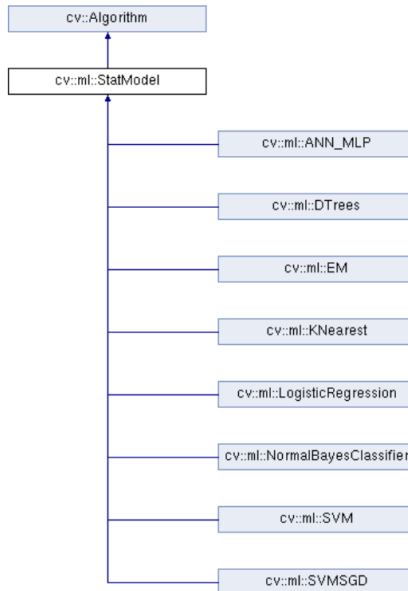
# El clasificador de Bayes

## cv::ml::StatModel Class Reference

Base class for statistical models in OpenCV ML. [More...](#)

```
#include <opencv2/ml.hpp>
```

Inheritance diagram for cv::ml::StatModel:



### Public Member Functions

virtual float **calcError** (const **Ptr< TrainData >** &data, bool test, **OutputArray** resp) const  
Computes error on the training or test dataset. [More...](#)

virtual bool **empty** () const **CV\_OVERRIDE**  
Returns true if the **Algorithm** is empty (e.g. in the very beginning or after unsuccessful read. [More...](#)

virtual int **getVarCount** () const =0  
Returns the number of variables in training samples. [More...](#)

virtual bool **isClassifier** () const =0  
Returns true if the model is classifier. [More...](#)

virtual bool **isTrained** () const =0  
Returns true if the model is trained. [More...](#)

virtual float **predict** (**InputArray** samples, **OutputArray** results=noArray(), int flags=0) const =0  
Predicts response(s) for the provided sample(s) [More...](#)

virtual bool **train** (const **Ptr< TrainData >** &trainData, int flags=0)  
Trains the statistical model. [More...](#)

virtual bool **train** (**InputArray** samples, int layout, **InputArray** responses)  
Trains the statistical model. [More...](#)