Automatic System for Classification of Precipitation Cells

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Abstract: This paper presents an automatic system for classification of the precipitations cells, conceived around a graphic interface. This interface is based on the fractal geometry and particularly on fractal dimension and the fractal lacunarity. We have initially analyzed these two parameters and we showed that they can be useful as discriminating parameters. Then, we developed a graphical interface which makes possible to identify in real time the type of cells. This tool was tested on different areas from the earth and showed its efficiency whatever the studied site. This system can be used in weather radar for the improvement of the precipitations estimations and in telecommunication for the correction of the signal for the microwave links.

Keywords: Fractal Lacunarity, Fractal Dimension, Convective cells, Stratiform cells, Graphical Interface.

1. Introduction

Precipitations are generally considered to be two clearly distinguishable types, stratiform and convective (A. Tokey and D. A. Short; 1996). This classification of precipitation is useful in a variety of meteorological applications (N. E. Anagnostou; 2004), (M. I. Biggerstaff and Listemaa S. A.; 2000), (M. Steiner and al; 1995). It is important in understanding cloud physics, as the two precipitation systems are characterized by different precipitation growth mechanisms (J. Gao and D. J. Stensrud; 2012), (C. Gao and al; 2006), (R. A. Houze, Jr.; 1993), and it is also important in quantitative precipitation estimation from both ground and space-based instruments (E. N. Anagnostou, and W. F. Krajewski; 1998), (Y. Yang and al; 2013).

Stratiform precipitations fall generally from nimbostratus clouds, and are of low quantities but can persists for hours, leading to a significant rain accumulation. In the other hand, convective precipitations are caused by cumulus and cumulonimbus clouds (R. A. Houze, Jr.; 1993), (A. Tokey and D. A. Short; 1996), they are characterized by strong narrow updraft, i.e. convection, where raindrops can grow rapidly. The resulting convective rainfall is intense, relatively short in duration, and highly fluctuating (F. Tridon and al; 2010).

This paper aims to make a difference between convective and stratiform cells in radar images, using a graphical application programmed under MATLAB called interface "GUI". This approach is based on the fractal geometry and particularly on fractal dimension and fractal lacunarity.

2. Data Base

We considered in this study four different weather radar sites namely, Dakar (Senegal), Bordeaux (France), Melbourne (Florida, USA) and State College PA (Pennsylvania, USA).

2.1. Region of Dakar

Dakar is the capital of the Republic of Senegal. It is located at the western extremity of Africa, on the narrow peninsula of Cape Verde.

Being in a tropical semi-desert, Dakar has a microclimate of coastal type, influenced by the monsoon trade winds and sea. The hot wet season extends from June to October with temperatures of about 27°C and a peak of precipitations in August (250mm).

The radar of Dakar is installed on a tower of 30m above the sea level at Yoff Airport, in the east of Dakar. Its geographical coordinates are 14°44’ North latitude and 17°29’ West longitude. It is operated by both the National Meteorological Office of Senegal and ASECNA (Association for the Safety of Air Navigation). This radar works only during storm periods. It also has a chain SANAGA. The average intensity of rainfalls varies from 300mm to 1500mm, collected from Saint Louis to Cape Skirring (A. Nzeukou and H. Sauvageot; 2002). The technical characteristics of the weather radar of Dakar are given in Table 1.

Table 1. Technical characteristics of the weather radar of Dakar region

| Technical characteristic | Value |
|--------------------------|-------|
| Diameter of antenna      | 2.5 m |
| Transmission frequency   | 5.7 GHz |
| Transmission power       | 250 Kilowatts peak |
| Spatial resolution       | 1 Km x 1 Km |
| Pulse duration           | 3 μs |
| Repetition sensitivity   | -110 dBm |
| Beam width to 3 dB       | 1.6° |
| Antenna gain             | 40 dB |

2.2. Region of Bordeaux

The city of Bordeaux is located in the south western of France. The Bordeaux radar is located in the Bordeaux-Mérignac airport, whose geographical coordinates are 44°49’54” North latitude and 0°41’30” West longitude. It is one of 24 metropolitan France radars. The digital analysis of the reflectivity of 8 bits
The radar is part of the French network (ARAMIS) managed by Météo-France. The area is almost flat. Table 2 gives the technical characteristics of the radar of Bordeaux.

Table 2. Technical characteristics of the weather radar of Bordeaux region

| Technical characteristic | Value               |
|--------------------------|---------------------|
| Type of antenna          | Parabolic           |
| Transmission frequency   | 3 GHz               |
| Transmission power       | 500 Kilowatts peak  |
| Spatial resolution       | 1 Km x 1 Km         |
| Pulse duration           | 5 μs                |
| Repetition frequency     | 300 Hz              |
| Beam width to 3 dB       | 1.8°                |

2.3. Region of Melbourne

Melbourne is a city of the United States of America located on the east coast of Florida. It is situated to the east of Orlando, and south of Cape Canaveral. It has a humid subtropical climate, typical of the Gulf and South Atlantic states.

The radar used is Nexrad WSR-88D type, which works on the coherent Doppler principle. The geographical position of the radar of Melbourne is 28.109 ° N, 80.650 ° W. It allows to observe the precipitation in the oceanic, littoral and terrestrial part (D. B. Wolff and al; 2005). The Melbourne climate is subtropical, with hot and humid summers and cool winters.

The technical characteristics of the weather radar of Melbourne are given in Table 3.

Table 3. Technical characteristics of the weather radar of Melbourne region

| Technical characteristic | Value               |
|--------------------------|---------------------|
| Diameter of antenna      | 18.5m               |
| Maximum rotation rate    | Azimuth 30°s-1 Elevation 30°s-1 |
| Maximum acceleration     | Azimuth 15°s-1 Elevation 15°s-1 |
| Mechanical limits        | -1° to +60°         |
| Polarization             | Horizontal linear  |
| Beam width to 3 dB       | 0.93° (to 2850 MHz) |
| Antenna gain             | 45 dBz              |

2.4. Region of State College

State College is located in state of Pennsylvania in the east side of the United States of America. State College is situated at an elevation of approximately 1,200 feet (370 m) above sea level. It is surrounded by large tracts of farmland, and an expanse of mountains and forests. State College has a humid continental climate. Temperatures average is −3.7 °C in January and 21.8 °C in July. Annual precipitation is just less than 40 inches (1,000 mm), and 46.3 inches (118 cm) of snow a year, falls in the city.

The radar used is Nexrad WSR-88D type called KCCX, which works on the coherent Doppler principle. Its technical characteristics are given in Table 4.

Table 4. Technical characteristics of the weather radar of State College region

| Technical characteristic | Value               |
|--------------------------|---------------------|
| Type of antenna          | Parabolic           |
| Longitude position       | 40° 55' 23" N (deg min sec) |
| Latitude position        | 78° 0’ 14" W (deg min sec) |
| Elevation                | 758 meters          |
| Beam width to 3 dB       | 0.93° (to 2850 MHz) |

3. Theoretical Concepts

3.1. Fractal Dimension

The concept of fractals was first introduced by Mandelbrot (B.B. Mandelbrot; 1983), (K. I. Kilic and R. H. Abiyev; 2011), as an indicator of surface roughness. Fractal theory has been widely applied to different scientific areas, particularly to image processing. The fractal dimension contains information about geometric structures of fractals and can also be used for non-fractals structures. It is a measure used to quantize the complexity of the geometrical objects. Many approaches were proposed for estimating fractal dimension of images. In our case, we used the Box-Counting approach (S. Lovejoy and D. Schertzer; 1990). This method is based on the recovery of the space in which the object is included in order to calculate the fractal dimension Df by a grid consisting of square (or "boxes") of side ε. The number N(ε) represents the number of boxes which are used to pave the object. The fractal dimension Df is then defined by:

\[ D_f = \lim_{\varepsilon \to 0} \left[ \frac{ \ln N(\varepsilon) }{ \ln(1/\varepsilon) } \right] \]

(1)

3.2. Fractal Lacunarity

Lacunarity is another concept introduced by Mandelbrot (B.B. Mandelbrot; 1983), in an effort to quantize the nature of gaps in texture images. It is inspired from the Latin (lacuna) origin of the English word “lake”, defines a fractal as being “lacunar”, when its gaps tend to be large. If a fractal has large gaps or holes, it has high lacunarity (K. I. Kilic and R. H. Abiyev; 2011). Visually different images sometimes may have similar values for their fractal dimensions. But lacunarity estimation can help us to distinguish such images. It is a measure for spatial heterogeneity. So higher values of lacunarity implies more heterogeneity, it means that an object is more homogeneous when the lacunarity is low (K. Falconer; 1990).

There are in the literature several methods for calculating the fractal lacunarity. In our case, we use the method of Allain and Cloitre, based on the sliding box algorithm, where the lacunarity Λ(L) is defined as a function of the scale L, by (C. Allain and M. Cloitre; 1991):

\[ \Lambda(L) = \left( \bar{s}^2(L) / s^2 (L) \right) + 1 = \left( \sigma^2 / \mu^2 \right) + 1 \]

(2)

Where: \( \bar{s}(L) \) = the average of masses per box, noted μ ; \( s^2 \) = variance of masses per box, noted \( \sigma^2 \).
Primary, we select the images to be treated. If the precipitation is composed of only one type of cells, the interface makes possible to identify the precipitation type without ambiguity. On the other hand, if the two types of cells are present, we must proceed by thresholding, by clicking on one of the choices given in the menu-bar “thresholding”. The choice is made depending on the type of cell we want to highlight and on the studied site.

After applying thresholding on the image presented in (Figure.1), we get the thresholded image as shown in (Figure.2). Proposed Methodology

Data base

In the present work, tasks like image acquisition, preprocessing, feature extraction, classification are carried out. The classification tree is given in (Figure.3). The detailed block diagram of adopted methodology is shown in (Figure.4).

4.1. Fractal Dimension

(Figure.3 & 4) show that when we click on the button “Fractal Dimension”, we obtain a straight curve which represents the fractal dimension depending on the size. The type of cells is shown in the part “result and discussion-Fractal dimension” after having clicked on the button “cells type”. In a previous work, we have demonstrated that the fractal dimension of the convective cells in the region of Dakar is included in the interval \([1.14 ; 1.16]\), while that of the stratiform cells is situated between 1.41 and 1.54 (N. Azzaz and B. Haddad; 2013). The classification of precipitations in real time by our GUI using the concept of fractal dimension is based on the comparison between the fractal dimension values obtained in our process and the parameters already computed in the last work and inserted in our GUI as comparison thresholds. The value of the fractal dimension for convective cells is still lower than that of the stratiform ones. This shows us that the stratiform cells are more deformed than the convective cells. As the fractal dimension varies with the nature of the cells and with the prevailing climate in the region considered, we conclude that the precipitations have a multifractal structure; this result is in good agreement with the literature (N. Azzaz and B. Haddad;
2013), (D. Schertzer and S. Lovejoy; 1992).

4.2 Fractal Lacunarity

After clicking on the button “Fractal Lacunarity” in the part “type of processing”, we obtain a power curve which represents the fractal lacunarity depending on the size.

When we click on the button “cells type” in the part “result and discussion-Fractal Lacunarity”, we see the type of the cells studied as it is shown in (Figure 5 & 6).

To classify the precipitations in convective and stratiform cells in real time, the GUI uses the concept of fractal lacunarity on comparing values of homogeneity factors “a” obtained in our process and those calculated in previous work and inserted in our GUI. The algorithms used in this work are implemented using MATLAB 7.0.

We now that the parameter “a” and the lacunarity value of convective cells are greater than that of stratiform ones, whatever the study region (N. Azzaz and B. Haddad; 2013). This is due to the fact that convective cells are less homogeneous than stratiform cells.

5. Conclusion

In conclusion, our study shows the importance of our graphical interface GUI to facilitate anyone the classification of the precipitation in two types, convective and stratiform, in real time. The GUI uses the fractal dimension and the fractal lacunarity as powerful discriminators between stratiform cells and convective cells.

This automatic system, tested on areas where prevail different climates, contributes to the improvement of the precipitation estimations through the relation Z-R for the weather radars, or to avoid the natural disasters caused especially by the violent storms. It can also be used in telecommunication for the correction of the signal attenuation for the microwave links.

This interface can be inserted without any difficulty on any weather radar.

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