# MONITORING HYDROLOGICAL VARIATIONS USING MULTISPECTRAL SPOT-5 DATA: REGIONAL CASE OF JALISCO IN MEXICO

Ivan E. Villalon-Turrubiates, Member, IEEE

Universidad de Guadalajara, Centro Universitario de los Valles Carretera Guadalajara-Ameca Km. 45.5, 46600 Ameca Jalisco, México, villalon@ieee.org

# ABSTRACT

The extraction of hydrological characteristics from a particular geographical region through remote sensing (RS) data processing allows the generation of electronic signature maps, which are the basis to create a high-resolution collection atlas processed in time for a particular geographical zone. This can be achieved using a developed tool for supervised segmentation and classification of hydrological remote sensing signatures (HRSS) via the combination of both statistical strategies defined as the Weighted Order Statistics (WOS) and the Minimum Distance to Means (MDM) techniques, unifying their particular advantages. This is referred to as the Hydrological Signatures Classification (HSC) method. The extraction of HRSS from multispectral/high-resolution RS maps using SPOT-5 satellite data for the regional case of the State of Jalisco in Mexico is reported to probe the efficiency of the developed technique in hydrological resources management applications.

## **1. INTRODUCTION**

Considerable progress has been made generally in the application of remote sensing techniques to both research and operational problems for urban planning and natural resource management. Modern applied theory of image processing for urban planning and hydrological resources management is now a mature and well developed research field, presented and detailed in many works ([1] thru [4] are only some indicative examples). Although the existing theory offers a manifold of statistical and descriptive regularization techniques to tackle with the particular environmental monitoring problems, in many applications areas there still remain some unresolved crucial theoretical and data processing problems.

One of those unsolved problems is particularly related to the extraction of hydrological characteristics (e.g., water, humid and dry content) for applications in resource management (modeling and planning).

The development of a tool for the supervised segmentation and classification of the hydrological remote sensing signatures (HRSS) from remote sensing (RS) imagery unifies the statistical strategies based on weighted order statistics (WOS) and minimum distance to means (MDM) applications. This is addressed as the Hydrological Signatures Classification (HSC) method.

## 2. WEIGHTED ORDER STATISTICS

The WOS method is considered as a generalization of the median filter [1], and is characterized by a weight vector and a threshold value. The order statistics (OS) filtering methodology [2] shifts a  $n \times n$  window **W** (with cardinality  $n \times n$ , i.e.,  $|\mathbf{W}| = n \times n$ ) over an input remote sensing (RS) image frame and, at each position of the frame, takes the  $n \times n$  inputs ( $w_{11}, w_{12}, ..., w_{ij}, ..., w_{nn}$ ) under  $\mathbf{W}_{ij}$  and then outputs the *r*-th element of the sorted input.

The WOS method is a generalization of the OS filter that is characterized by a weight vector  $\mathbf{Y}_{ij} = (v_1, v_2, ..., v_{n \times n})$  of  $n \times n$  positive weight thresholds w,  $0 \le w \le 255$  (gray-level threshold). To compute the output of the filter, each input wis duplicated to the number of corresponding weight v, then they are sorted and the *w*-th order element (median) is chosen as the output. This is expressed as

$$WOS_{ij} = median(Y_{ij}), \qquad (1)$$

where **WOS**<sub>*ij*</sub> is the weighted order of the (i, j)-th pixel of the image. The decision rule for classification based on the WOS filter determines that, based on the a priori information for class segmentation (number of HRSS to be classified and their respective thresholds), the WOS value for each image pixel is compared with the a priori thresholds (gray-level) and classified according to the most proximal value.

## **3. MINIMUM DISTANCE TO MEANS**

The MDM decision rule is computationally simple and can result in classification accuracy comparable to other more computationally intensive algorithms [3]. It is characterized by the mean values of the HRSS classes and the Euclidean distances based on the Pythagorean Theorem. An important aspect of this method is that it is applied to the multispectral RS imagery. The a priori information for class segmentation (number of HRSS to be classified and their respective mean values) conform the means matrix **E** ( $c \times b$  size) that contains the mean values  $\mu_{cb}$ : ( $0 \le \mu_{cb} \le 255$ , gray-level) of the HRSS classes for every RS band. Here, c is the number of HRSS classes to be classified, and b is the number of spectral bands contained in the RS imagery. The input is defined by the vector  $\Pi_{ij}$ , which contains the (i, j)-th image pixel values  $\pi_{ijb}$  employed for every spectral band.

To compute the output of the classifier, the distance between each input  $\Pi_{ij}$  and the means matrix **E** is calculated using the Euclidean distance based on the Pythagorean Theorem. This is expressed by Eq. 2, where  $\mathbf{D}_{ij,c}$  is a vector ordered by multi-index (ij,c) of  $c \times I$  size that contains the distances between the (i, j)-th image pixel value and the *c*class value for each band *b*.

$$\mathbf{D}_{ij,c} = \sqrt{\left(\pi_{ij,1} - \mu_{c1}\right)^2 + \dots + \left(\pi_{ij,b} - \mu_{cb}\right)^2}$$
(2)

The decision rule for classification based on the MDM filter determines that, based on the a priori information for class segmentation (number of HRSS to classify and their respective mean values), each image pixel generates an ordered distance vector  $\mathbf{D}_{ij}$ , and therefore, the pixel is classified according to the minimum value on the vector.

#### 4. MULTISPECTRAL HRSS

Both WOS and MDM techniques provide a high level of HRSS segmentation and classification. Nevertheless, to ensure an accurate high-resolution process, the fusion of both algorithms is performed as a systematical tool for supervised HRSS segmentation and classification of RS scenes via combining the WOS and MDM techniques.

The developed technique is referred to as the Hydrological Signatures Classification (HSC) method [4]. The detailed stages of the computational algorithm of the HSC method for HRSS classification of the RS scenes is described as follows

- 1. Set the number of HRSS classes and their respective threshold mean values for each band in the multispectral RS image. This is the a priori information.
- 2. Separate the multispectral RS image in its respective bands (*b*).
- 3. Apply the WOS method (1) to each band *b* present on the RS image. With this, a number *b* of WOS classified HRSS images are obtained.
- 4. Apply the MDM method (2) using the WOS classified HRSS images as the bands *b*.

The classification obtained with the MDM method is the desired HSC electronic HRSS map extracted from the RS image.

#### **5. SPOT-5 IMAGERY**

SPOT Image is the worldwide distributor of geographic information products and services derived from the SPOT Earth observation satellites. A SPOT satellite image is a view of the Earth seen through one of the satellite's highresolution imaging instruments. The technical characteristics of each instrument determine the resolution and spectral mode of the image. The acquired image is then processed to suit users' requirements in terms of geographic information. It is delivered in a standard format able to be integrated directly in current geographic information software packages [5].

The image used for this paper was obtained from SPOT-5 and corresponds to "La Vega" dam in the state of Jalisco Mexico. It was provided by the Mexican NAVY (SEMAR Secretaría de Marina/Armada de México) under the ERMEXS program (Estación de Recepción México de la Constelación SPOT) [6]. The spatial resolution is 10m (spectral mode Hi) and the spectral resolution corresponds to 3 bands:

- B1: green band  $(0.50 0.59 \ \mu m)$ .
- B2: red band (0.61 0.68 μm).
- B3: near infrared band  $(0.78 0.89 \,\mu\text{m})$ .

## 6. HRSS SIMULATION EXPERIMENT

In the reported results, a HRSS electronic map is extracted from the multispectral high-resolution RS image. Both, the WOS and MDM methodologies employs a decision rule used for the classification process, nevertheless, some pixels from the original RS scene may have particular characteristics that causes a uncertainty in the decision rule (e.g., for the WOS method, the median of the weighted vector for a particular pixel is exactly between the values of two classes; for the MDM method, the distance to two or more classes are the same). In this case, the decision rule considers those pixels as unclassified zones.

Figure 1 show the original multispectral high-resolution (1024x1024-pixel) RS scene.

Figures 2, 3 and 4 show the different bands of the original image as described in section 5.

Figure 5 show the HRSS maps obtained applying the fusion of the WOS and MDM approaches via the developed HSC method.



Figure 1 – Multispectral RS scene (courtesy of ERMEXS).



Figure 3 – B2 band (red).

Three level HRSS are selected for this particular simulation process, moreover, unclassified zones must be also considered (2-bit classification) described as

- Light-gray regions represents the HRSS that relate to the wet zones of the RS image.



Figure 2 – B1 band (green).

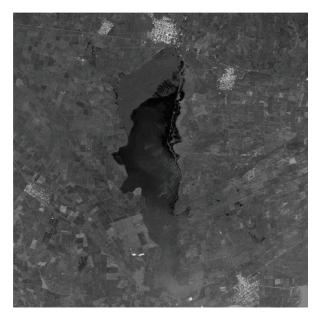


Figure 4 – B3 band (near infrared).

Black regions represents the HRSS that relate to the humid zones of the RS image.

- Dark-gray regions represents the HRSS that relate to the dry zones of the RS image.

□ – White regions represent the unclassified zones of the HRSS map.

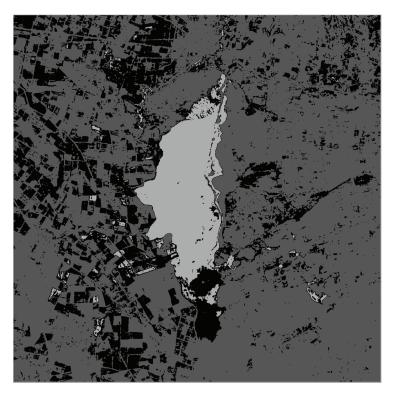


Figure 5 – HRSS map extracted using the HSC method.

# 6. CONCLUDING REMARKS

The developed HSC method that fuses the WOS and MDM classifiers provides the high resolution HRSS electronic map with a high-accurate classification and less unclassified zones than those WOS and MDM without fusion. This is achieved because the unified HSC classifier uses more detailed robust a priori information from the original RS scene (provided by separate supervised classifiers). The resulting HRSS map ensures better results in the classification achieved with the developed HSC method [4].

The developed HSC method for HRSS extraction can be applied to several RS images from a particular geographical region obtained in different moments of time to generate a HRSS atlas of environmental electronic maps. This process is a powerful tool for hydrological resource management applications [4].

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