Original software publication

Bedforms-ATM, an open source software to analyze the scale-based hierarchies and dimensionality of natural bed forms

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A B S T R A C T

Bedforms-ATM (Bed forms analysis toolkit for multiscale modeling) is a software designed to hierarchize and quantify the dimensionality of natural bed forms fields. It comprises four modular applications, namely: (1) wavelet analysis, (2) Hovmöller analysis, (3) multiscale discrimination, and (4) three-dimensionality analysis. Bedforms-ATM also provides insights on bed form systems dynamics and their interrelationship with the surrounding hydrodynamic characteristics. The software structure encourages its expandability via the collaboration from the community of users. Both fluvial and synthetic bed form data accompany Bedforms-ATM.

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Code metadata

Current code version
v1.1

Permanent link to code/repository used of this code version
https://sourceforge.net/projects/bedforms-atm/
https://github.com/ElsevierSoftwareX/SOFTX_2018_28

Legal Code License
GNU GPL v3 or later

git

Code versioning system used
MATLAB

Compilation requirements, operating environments & dependencies
Microsoft Windows, Unix

If available Link to developer documentation/manual
https://sourceforge.net/projects/bedforms-atm/
https://github.com/rgutierrezl/BedformsATM
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1. Introduction

Although the need of discriminating bed forms scales (e.g., bars, dunes and ripples) that are commonly present in unidirectional, oscillatory and combined flows has been highlighted by many researchers [e.g. 1–3], no standard nomenclature and procedure exist to systematically identify the scale and magnitude of such sedimentary features [3]. As a result, the discrimination between ripples and dunes is taken as obvious and many authors provide limited or no explanation regarding the criteria they followed for their discrimination in field and laboratory studies [1]. Reference [3] introduced a spectral-structural technique that allowed for a quantitative characterization of three bed form hierarchies (Fig. 1-a and 1-b) even for curved bed form fields (Fig. 1-c). However, it is not able to differentiate between two dimensional (2D) and three dimensional (3D) bed form features. The structural component of the technique presented by [3] obeys the scale-based classification scheme proposed by [4]. It defines ripples as features having wavelengths less than 0.6 m; dunes as features characterized by wavelength thresholds of 5, 10, and 100 m determining small, medium and large dunes; and bars as features having wavelengths higher than 100 m. The spectral component of the
technique proposed by [3] is based on the coupled application of robust spline filters and the one dimensional continuous wavelet transforms frame developed by [5].

Bed forms in sedimentary environments are predominantly 3D in planform [3]. This characteristic induces marked differences in local turbulence, drag coefficient, and dispersal patterns of sediments [6–8]. Therefore, a complete bed forms description requires distinguishing between 2D and 3D bed forms. To this end, [9] proposed a quantitative continuum metric, named 3-dimensionality index (3DI), which varies between 0 (strictly 2D) and 1 (strictly 3D). 3DI (cf. Fig. 1) quantifies the normalized autocorrelation between the fluctuations of a bed form field $\eta(x, y)$ respect to the local cross-sectional mean $\bar{\eta}(y)$, and the fluctuations of the local stream-wise mean $\bar{\eta}_x(x)$ respect to the bed form field mean $\bar{\eta}_{xy}$ (Eqs. (1)-(4)). Thus, a strictly 2D bed form has a flat crest perpendicular to the flow, and the more irregular the cross-sectional profiles are, the greater the value of 3DI.

Most computer programs designed to analyze bed forms fields (e.g., Bed form tracking tool by [2]) are not currently publicly available or are discontinued. Herein we present Bedforms-ATM (Bed forms analysis toolkit for multiscale modeling), a MATLAB open source software that performs the variability analysis of bed form fields by quantitatively describing both their hierarchies and dimensionality.

This paper presents the technical aspects of the development and capabilities of Bedforms-ATM and is organized as follows: Section 2 describes the data processing methods used in this study, Section 3 provides a functional description of Bedforms-ATM, Section 4 focuses on its implementation, Section 5 presents case histories of its application on bed forms and synthetic bed forms, and Section 6 discusses on its potential impact. Finally, Section 7 outlines the main conclusions of this study.

2. Background

Fig. 1 shows an approximately equally spaced (i.e., $\Delta x \approx$ constant) elevation bed form profile $\eta(x, y)$, also represented herein by the vector $\eta^{\text{vi}}(x)$. The software performs the wavelet analysis (Application 1) of each bed form profile in $\eta(x, y)$ by adapting the one dimensional continuous wavelet analysis package developed by [5]. The continuous wavelet transform of $\eta^{\text{vi}}(x)$ is mathematically defined as the convolution of it with a continuous wavelet function. By performing a Monte Carlo analysis, [5] estimated empirical factors that allow for determining both the local wavelet power spectrum (i.e., the modulus of the wavelet transform) and its cone of influence (i.e., the divide that discriminates reliable from spurious local power spectrum results). This is done for continuous wavelet functions such as Morlet, Paul, Marr and the derivative of the Gaussian, at a given confidence level. Thus, the spatial distribution of the wavelengths along $\eta^{\text{vi}}(x)$ is obtained. Subsequently, the wavelet spectrum provides information about the dominant constituent wavelengths of $\eta^{\text{vi}}(x)$. The reader is kindly referred to [5] and [3] for technical details of the application of wavelet transforms on bed forms variability analysis.

Application 2 (power Hovmöller analysis) is an extension of Application 1 [5]. Basically, the Power Hovmöller analysis allows for locating bed form features in $\eta^{\text{vi}}(x)$ that lie in a specific wave-length interval by passing the bed form profile with a continuous-wavelet-based high-pass filter and low-pass filter, which are defined by such interval. Thus, by applying that procedure to each bed form profile in $\eta(x, y)$, the locations of such features are mapped on the bed form field domain [3].

Application 3 (scale-based discrimination) allows for obtaining three hierarchies from $\eta^{\text{vi}}(x)$ by setting two wavelength thresholds, $\theta_{1,3}$ and $\theta_{2,3}$. The smallest scales are comprised in Hierarchy 1, which are determined following the structural criteria presented by [3]. For instance, if Hierarchy 1 corresponds to ripples, $\theta_{1,3} = 0.6$ m and if Hierarchy 1 corresponds to small dunes, $\theta_{1,3} = 5$ m. The largest scales are comprised in Hierarchy 3, and are retrieved from the wavelet spectrum according to a threshold $\theta_{3}$ typically set $> 100$ m. Hierarchy 2 is determined by subtracting scales in Hierarchy 1 and Hierarchy 3 from $\eta^{\text{vi}}(x)$. A given situation could present more than three scales of interest, i.e ripples, small dunes, intermediate dunes, large dunes, and bars. In such cases, the described procedure would first discriminate ripples, dunes and bars as hierarchies 1, 2 and 3, respectively. Further, the discrimination procedure can be applied again on the intermediate scales in Hierarchy 2, distinguishing small, intermediate and large dunes.

It is important to note that the methods used for Applications 3 are optimized versions of those proposed by [3] in the following aspects: learning curves to allow for an efficient bias–variance trade–off at each hierarchy are used, and the bias–variance error is distributed among them, and thus the discriminated hierarchies are denoised at a better extent. Likewise, when the bed form data is provided as a water depth field (e.g. $h(x, y)$ in Fig. 1), the application verifies the expected weak correlation between $h_1(x, y)$, namely a Hierarchy 1 field, and the mean water depth (e.g. $\bar{h}_y(y)$) in Fig. 1) by using the proxy parameter named scale-variance-ratio (SVR$_{3}$) proposed by [3] representing the ratio between the standard deviation of Hierarchy 1 and that of Hierarchy 2. The reader is kindly referred to [3] and the flow diagrams in the User Manual of Bedforms-ATM for details on the methodology of Applications 1, 2, and 3.

The three-dimensionality analysis (Application 4) is performed by using 3DI definition proposed by [9] (Eqs. (1)-(4)).

$$\bar{\eta}_y(x) = \frac{\sum_{i=1}^{n} \eta(x, y)}{n}$$

(1)

$$\bar{\eta}_x(y) = \frac{\sum_{i=1}^{m} \eta(x, y)}{m}$$

(2)

$$\bar{\eta}_{xy} = \frac{\sum_{i=1}^{n} \sum_{j=1}^{m} \eta(x, y)}{m \times n}$$

(3)

$$R = \frac{1}{m \times n} \sum_{i=1}^{n} \sum_{j=1}^{m} \left( \frac{\eta(x, y) - \bar{\eta}_y(y)}{\sigma_{x}} \right) \left( \frac{\eta(x, y) - \bar{\eta}_{xy}}{\sigma_{xy}} \right)$$

(4)

where $\sigma_{x}(y)$ and $\sigma_{xy}$ are the local stream-wise standard deviation and the bed form field standard deviation, respectively, for $i = 1, 2, \ldots, m$ and $j = 1, 2, \ldots, n$. Thus, 3DI ($T_b$) is defined by Eq. (5).

$$T_b = 1 - R^2$$

(5)

Subsequently, the average 3DI of a resolution cell, which is limited by nodes $\eta(x, y)$ and $\eta(x+1, y+1)$ (Figs. 1-a and 1-b), is obtained by estimating $T_b$ over a spatial window of arbitrary fixed length $L_x \times Ly$ being centered in such resolution cell. Finally, contours of $T_b$ over the whole bed form field are obtained. It is important to note that the aforementioned process excludes the resolution cells of the bed form field periphery; although, the algorithm minimizes the number of such cells.

3. Software framework

3.1. Software architecture

Bedforms-ATM applications are built to run in successive sequence. This stems from the fact that some of Application 2 input data is Application 1’s output data, both Application 3 and Application 4 use Application 1’s output, and Application 4 inherits input
Fig. 1. Schematic of bed form fields discretization. (a) Rectangular field $\eta(x, y)$ represented by a structured mesh having approximately constant resolution cell size of $\Delta x = x_{i+1} - x_i$ and $\Delta y = y_{j+1} - y_j$ for $(i,j) = (1, \ldots, m; 1, \ldots, n)$, where $\Delta x$ not necessarily equals $\Delta y$. (b) schematic of the 3-hierarchy scheme for bed form discrimination proposed by [3] (when $j = J$, bed form profile $\eta(x_i, y_J) \in \eta(x_i, y_j)$ is also represented by vector $\eta^J(x_i)$ in this paper). By following the nomenclature proposed by [3], vector-wise Hierarchies 1, 2, and 3 from such bed form profile are $\eta_{y, J}^1(x_i)$, $\eta_{y, J}^2(x_i)$, and $\eta_{y, J}^3(x_i)$, respectively. This representation also applies to bed form fields $h(x, y)$ defined by water depth measurements. (c) Curved bed form field which is analyzed by transforming rectangular coordinates into streamwise $s$ and normal $n$ local coordinates, such that all bed form profiles have a constant sinuosity. Thus, resolution cells are trapezoidal, where $\Delta s_j = s_{i+1} - s_i$ and $\Delta n = n_{i+1} - n_i$ is constant, provided that $\Delta s_j = s_{i+1} - s_{i+1}$, and thereby, all bed form profiles have the same sample size $m$. Synthetic bed form data for both rectangular and curved fields are provided as accompanying information of this paper (see Appendix A, Supplementary File 1, Section 2).

3.2. Software functionalities

Application 1’s interface (Appendix A, Supplementary File 1, Section 1) lists all the steps to create, select input data, preview the input data, perform a run time test (for large data), and run the application. It also highlights the steps that have been successfully completed. We foresee that many potential users of Bedforms-ATM may not be familiar with the wavelet analysis theoretical background; therefore, Application 1’s interface lists all the necessary parameters to be set for successfully performing the wavelet analysis of bed forms profiles (advanced users can however edit them to obtain parameter specific estimations.) Thus, it guides the user to the appropriate selection of both the significance level and the frequency increment for the wavelet analysis; likewise, it allows for the use of two sets of wavelet functions, namely: (1) the derivatives of the Gaussian, among which the second derivative, i.e. the Ricker wavelet function or so called Mexican hat wavelet function, has been widely used in the analysis of geophysical signals [3]; and (2) the Morlet function which is most suited for analyzing bed forms variability [3]. If the latter wavelet function is selected, the interface guides the user to the proper selection of the Morlet’s nondimensional frequency parameter, $k_0$. Additionally, it allows the user to set the output settings (e.g., figures title, signal name, unit, sample name, and sample unit). Finally, the user can select the file format (e.g., MATLAB .fig, .jpg, .tiff, or .PDF) in which the application’s output will be saved. When the run time data from Application 1 and 3 (see Appendix A, Supplementary File 1, Section 1). However, all of these applications are presented in a single MATLAB platform.

The input data is collected from the Application 1’s interface, and in order to avoid data inconsistencies, it is collected as a set of streamwise equally spaced vectors representing bed form profiles from either linear or curved elevation or water depth bed form fields, provided that all vectors are sampled at the same sampling period, approximately (Bedforms-ATM verifies that not a single sample period is higher than four times the average sample period, and that the number of sample period outliers is less than 5% of the vector length. If these conditions are not met, the program is prevented to run.) These vectors can be described by either relative (i.e., local streamwise coordinate and elevation or water depth) or absolute coordinates (i.e., Easting, Northing, and elevation or water depth) being saved in .TXT or .MAT file extensions. Additionally, the input data can be provided as a cell array .MAT file containing a bed form profile vector in each cell.
test indicates that it will take a long period of time, the application
suggests the user to print the output to a MAT file and provides a
function to print it later, which results in a reduced computational
time.

Application 2 performs a scale-averaging of the wavelet power
spectra at multiple locations (Fig. 5 in Appendix A, Supplementary
File 1), and thus, the spatial variability of a field of bed form tran-
sects can be assessed. This application estimates the appropriate
wavelength intervals to perform the power Hovmöller analysis, which
is presented in its interface, and thereby the analysis of
low density scale populations or non-existent scales are avoided.
Lastly, Application 2’s interface provides the user with two alter-
natives, namely: (1) to export the output in Tecplot® format, or (2)
to print the output to file in jpg, tiff or PDF formats. Since the code
is open, advanced users can modify it to include other formats such
as eps format.

Application 3 presents a plot of the structural classification
scheme suggested by [4], and once the input files are selected
(i.e., Application 1’s output), the application depicts the scales over
the aforementioned plot and suggests the user the three scales
of analysis, namely: Hierarchy 1, Hierarchy 2, and Hierarchy 3.
Additionally, when the whole analysis is performed over a water
depth field, this application allows the user for selecting the cross
correlation analysis between SVR1,2 and SVR2,3, and the mean
water depth at each bed form profile as presented in Fig. 2. This
operation represents a physically based assessment of the discrimi-
nation output. Finally, the application prints the output to file in
both PDF and MAT file extensions. Additionally, it can be exported
to a Tecplot® input format file.

Application 4 allows the user to select: (1) the analysis field,
namely: the bed form field \( \eta(x_i, y_j) \), or one of the hierarchies being
obtained from Application 3; and (2) a sub-domain where the
analysis will be performed. The application suggests the window
size over which \( T_p \) is integrated for the user-selected hierarchy as
well as bed form field. Additionally, the user can choose printing
the output in PDF, MAT or Tecplot® file formats.

4. Implementation

The program along with the corresponding user manual, and
test data from the Paraná River (150 bed form profiles) are available
for free at https://sourceforge.net/projects/bedforms-atm/
and can be easily downloaded and installed. Additionally, the same
information can be obtained from https://github.com/r Gutierrez
BedformsATM. Synthetic data from both curved and rectangular
plots to test the program are provided by [10]. Please refer to
Appendix A, Supplementary File 1, Section 2 for technical details
on the aforementioned data. An explicative video is available at
www.pucp.edu.pe/trbV8 and can also be found at Appendix A,
Supplementary File 2. Bedsforms-ATM is divided into four inter-
facing. To call them, the user should execute the following scripts
from the MATLAB console:

- BedformsATM_App1_WaveletAnalysis.m
- BedformsATM_App2_HovmollerAnalysis.m
- BedformsATM_App3_ScaleBasedDiscrimination.m
- BedformsATM_App4_3DAnalysis.m

5. Illustrative examples

Bedforms-ATM was tested by using field data from the Paraná
River from [11], and synthetic data from [10]. The application
of the software on the former data effectively provides information
of the spatial distribution of the bed forms wavelength spectrum (Application 1 and Application 2) as shown in Appendix A, Supple-
mentary File 1, Section 3.

Fig. 2 presents the output of Application 3 for the Paraná River.
Fig. 2-a shows the centered first hierarchy (i.e., small dunes, \( h_{1,3} \),
and Fig. 2-b presents the centered second hierarchy (i.e., medium
and large dunes, \( h_{2,3} \)) which shows less noise when compared
to the output presented by [3] as the algorithm distributes the
resulting bias–variance error of the method. Similarly, Fig. 2-c
shows the third hierarchy (i.e., bars, \( h_{3,3} \)) exhibiting a manifestly
less noisy output than that presented by [3].

The scale-variance-ratio between the first and second hierarchy
(SVR1,2, Fig. 2-d) confirms the expected weak correlation with
water depth, which is explained by the fact that \( h_{1,3} \) is mainly
induced by local turbulence. On the contrary, SVR2,3 (Fig. 2-d)
shows a strong correlation with water depth, due to the fact that
larger bed form scales are controlled by hydraulic conditions in
fluvial environments [3].

Fig. 3-a shows Application 4 output for the Paraná River when
the analysis is performed over \( h(x_i, y_j) \), in which the effective area
of the output is smaller than that of \( h(x_i, y_j) \). Different integration
window sizes are used in the analysis, and consequently different
degrees of smooth outputs are obtained. This stems on the fact
that, similarly to the application of Parzen windows, the bigger the
window the smoother the output. Thus, no single window width is
ideal overall because some sizes allow to gain information in some
areas, whereas losing in others [12]. The user can visually select
the output he/she considers the best based on his/her interest as
shown in Fig. 3-b, and thus a localized analysis can be performed
(Fig. 3-c).

Three-dimensional analysis on the Paraná River bed forms indicates
that high values of \( T_p \), namely marked 3D bed forms, are
mostly concentrated at the deeper areas, possibly because three di-
dimensional bed forms control at some extent the Reynolds stresses,
the drag coefficients, and the dispersal patterns of sediments [7].

6. Impact

Even though meteorological models have been usually tested
with observation using objective scientifically rigorous methods,
many morphological models have been evaluated by means of the
modeler’s subjective judgment, or by mean squared score
(i.e. the Brier skill score) in which the initial bed is chosen as refer-
ence [13,14]. In this contribution, the efficiency of Application 3 is
analyzed and presented on the basis of objective statistics that are
commonly used to evaluate meteorological models. It was success-
fully applied on both synthetic and fluvial bed forms, [15] reported
the effectiveness of Bedforms-ATM to discriminate seabed form
features from the Gräubd tidal channel inlet in Denmark, which
were previously studied by [16]. Thus, in the light of the results
(see Appendix A, Supplementary File 1, Section 4), by coupling the
capabilities of wavelet and Hovmöller analysis and 3DI, Bedsforms-
ATM provides an effective frame to analyze the variability of bed
forms features from different sedimentary environments. It was
coded following the best practices of object-oriented and user-
friendly software, which allows for analyzing bed form data sets in
a single platform without the need of passing from one software
to another. In the instances that users are not familiar with the
wavelet analysis, all the applications guide them to select the
appropriate parameters to provide reasonable results. Likewise,
the graphical outputs of the software can be exported as vector
graphics in publication-quality resolution.

We designed Bedforms-ATM as a modular and straightforward
software, which allows for expanding the software.

We expect and encourage new applications to be developed by the
users’ community. In this context, any feedback (e.g., bug reports,
suggestions for further development, and pieces of contributed
code or even ready plug-ins) benefits the users community [17].

The usability of the software was tested by releasing a beta
version in .p MATLAB files, which unlike .m format files, cannot be
Fig. 2. Bed form fields from the Paraná River, Argentina (flowing from left to right). (a) Hierarchy 1: centered small dunes ($h_{1,3}$), (b) Hierarchy 2: centered medium and large dunes ($h_{2,3}$), (c) Hierarchy 3: bars ($h_{3,3}$), (d) $SVR_{1,2}$, and (e) $SVR_{2,3}$. In all cases, coordinates are in meters.

Fig. 3. Bedforms-ATM Application 4 output for (a) $h(x_i,y_j)$ from the Paraná River in the entire domain, and a sub-domain selected in (b) and its corresponding output (c). In all cases, the contours represent the 3DL $T_b$, such that the higher the value of $T_b$, the higher the three dimensionality.

edited. An error report form was also provided. Few reports were received, and the present version, corrects the reported errors.

Finally, the results presented herein provide valuable information for both scientific and practical applications. For the former application, for instance, when coupled to river morphodynamic models, it could enhance our understanding of the interrelationship between the three hierarchies discriminated herein. For the latter application, it could potentially be used to estimate the input of each hierarchy in the global roughness of a reach or perform a bathymetry reduction which are relevant for flood and navigability modeling.

7. Conclusions

Most of the computer programs aimed to analyze bed forms in both fluvial and marine environments are not public available or are discontinued. Bedforms Analysis Toolkit for Multidimensional Modeling (Bedforms-ATM) potentially fills this niche, and, in the light of our results from applying the software on fluvial and synthetic bed form fields, we believe it has the potential to be used as an initial standard frame towards bed forms dimensional analysis and scale-based discrimination, and provide valuable insights on the understanding of bed form systems dynamics and
their interrelationship with the surrounding hydrodynamic characteristics. Bedforms-ATM is presented herein as an open source code software whose structure has been designed to encourage its expandability via the collaboration from the users community. The software will greatly benefit from future MATLAB code for parallel computing and GPU which unfortunately are not provided in this version.

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Appendix A. Supplementary data

Supplementary material related to this article can be found online at https://doi.org/10.1016/j.softx.2018.06.001.

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