

Quantitative analysis of randomness exhibited by river channels using chaos game technique: Mississippi, Amazon, Sava and Danube case studies

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Abstract. This paper presents a numerical evaluation of the randomness which can be observed in the geometry of major river channels. The method used is based upon that of generating a Sierpinski triangle via the chaos game technique, played with the sequence representing the river topography. The property of the Sierpinski triangle is that it can be constructed only by playing a chaos game with random values. Periodic or chaotic sequences always produce an incomplete triangle. The quantitative data about the scale of the random behaviour of the river channel pathway was evaluated by determination of the completeness of the triangle, generated on the basis of sequences representing the river channel, and measured by its fractal dimension. The results show that the most random behaviour is observed for the Danube River when sampled every 715 m. By comparing the maximum dimension of the obtained Sierpinski triangle with the gradient of the river we can see a strong correlation between a higher gradient corresponding to lower random behaviour. Another connection can be seen when comparing the length of the segment where the river shows the most random flow with the total length of the river. The shorter the river, the denser the sampling rate of observations has to be in order to obtain a maximum degree of randomness. From the comparison of natural rivers with the computer-generated pathways the most similar results have been produced by a complex superposition of different sine waves. By adding a small amount of noise to this function, the fractal dimensions of the generated complex curves are the most similar to the natural ones, but the general shape of the natural curve is more similar to the generated complex one without the noise.

1 Introduction

Fractal analysis has been in use for about thirty years and is particularly useful for describing non-Euclidean objects such as the patterns found in nature which are far more complex than simple objects such as lines, circles, ellipses, etc. (Mandelbrot, 1983; Feder, 1988). The fractal properties of river networks and channels were among the first natural objects to be analysed using fractal methods. Horton's (1945) well-known power law of bifurcation and stream-order length and Hack's (1957) power law of length and basin area had already pointed the way. Fractal methodologies for analysis of river behaviour have been applied to many aspects, such as describing the self-similarity properties of river networks (Tarboton, 1996) and quantifying the sinuosity of streams (Schuller et al., 2001), karst rivers (Kusumayudha et al., 2000) or fluvial topography (Veneziano and Niemann, 2000).

A fractal scaling system is determined by the fractal dimension, which provides insights regarding the origin and processes involved in forming the fractal objects, and therefore the understanding of the physical properties of geological materials (Turcotte, 1992; Dillon et al., 2001). This dimension can consist of non-integer values, as opposed to the integer values characteristic of Euclidean objects, such as 2-D surfaces or 3-D cubes. Rivers can be characterised by two fractal dimensions (Dodds and Rothman, 1999). The dimensions of individual streams represent measures of the stream's irregularity, or the extent of a stream's meandering and the dimensions of river networks describe the ability of a network to fill the plane. Both stream channels and river networks have been found to exhibit fractal structure (Tarboton et al., 1988; La Barbera and Rosso, 1989; Phillips, 1993; Veltri et al., 1996; De Bartolo et al., 2006 and many others). Fractal dimensions have been observed for several natural river networks and can vary widely (Schuller et al., 2001). Variations occur because natural stream networks are



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not strictly self-similar objects, but rather self-affine (Beauvais and Montgomery, 1997) and thus, empirically derived dimensions for these objects are only estimates of their true fractal dimension. The ideal surface with a fully developed river network should have a fractal dimension of 2, implying topological randomness. However, geological, topological and hydrological effects dictate that the fractal dimension of networks are always lower than 2 (La Barbera and Rosso, 1989). There are not many studies concerning individual river channels, as most studies have focused on drainage patterns (Tarboton et al., 1988; Nikora et al., 1993; Schuller et al., 2001; Angeles et al., 2004; De Bartolo et al., 2006 and others). The multifractal method has also been used for description of multichannel rivers (Veneziano and Niemann, 2000; De Bartolo et al., 2000; Gaudio et al., 2006 and others).

In this paper a novel approach has been used for the quantitative analysis of the behaviour of the river channel from its source to its mouth. It is based on creating a Sierpinski triangle (also known as Sierpinski gasket) via the chaos game, based on the data of the geometry of the channel. The *chaos game* is a method of generating deterministic objects from apparently random ones (Barnsley, 1988; Peitgen et al., 2004). It can be used to produce a visual representation of a sequence of random numbers (Mata-Toledo and Willis, 1997). It has been usefully applied to studies of DNA sequences (Jeffrey, 1992) and economics (Matsushita et al., 2007). The complete Sierpinski triangle via chaos game can be produced only with the help of the “perfect dice”, i.e. by an entirely random process. Every periodic sequence fails to reproduce a complete triangle and only some parts of it are created.

This paper presents an application of the chaos game algorithm to the characterisation of the randomness of river channel geometry. This approach concerns neither the fractal analysis of a single river channel nor the fractal analysis of river networks, as both have been already studied extensively by many other authors (see paragraph above). Here, the river is analysed on the basis of the representation of its pathway with the sequence of three classes which act as an input for the chaos game. Following this the fractal dimension of the outcome of the chaos game (a more or less complete Sierpinski triangle) is the measure of the randomness of the sequence representing the randomness of the river channel. When the river forms perfect meanders or when it flows straight, the obtained sequence would be very deterministic and only a limited part of the Sierpinski triangle would be created. Hypothetically when the river behaves completely unpredictably, as the perfect dice do, the sequence will be random and a complete triangle should appear. All intermediate possibilities produce more or less incomplete triangles. The higher the randomness of the river geometry, the higher the dimension of the triangle will be. Another issue to be mentioned is the sampling rate, which describes the length of the sequence which represents the river pathway. The denser

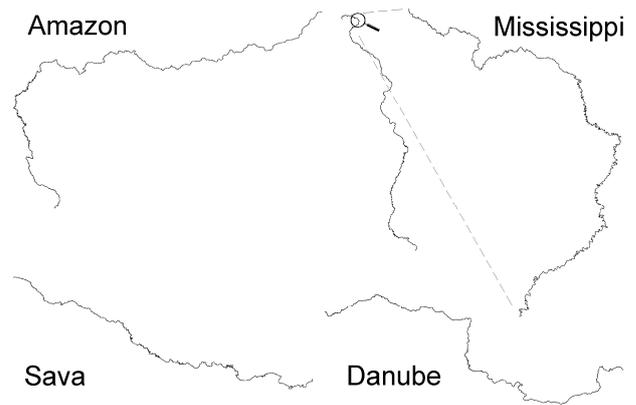


Fig. 1. Analysed rivers (scale varies, projection to Euclidean plane).

the sampling rate, the longer the sequence and the shorter the segment between two sequence points and vice versa.

2 Materials and methods

2.1 The choice of rivers and sources of data

Four rivers were analysed in this study – the Amazon, Mississippi, Danube and Sava (the largest tributary of the Danube) (Fig. 1). These watercourses were selected as their maps are easily accessible and they belong to different geologic and climatic environments. The river courses were digitised from various sources. The Amazon was digitised from the data on the Microsoft Encarta World Atlas web site¹; the Mississippi from the Google Earth web site²; the Danube from the topographic maps on a scale of 1:200 000 from the 3rd Military Mapping Survey of Austria-Hungary (between the years 1889–1915)³; and the Sava from the national maps of former Yugoslavia on a scale of 1:25 000 (raster TIF images). Despite the fact that we have used maps of different scales, all of them were sufficiently precise to present all of the river meanders and other necessary details. Moreover, different map scale does not affect the results because rivers were later in the process split into the 3-class sequence and this process is not affected by the map scale. Additionally, all of the rivers are very large so all of them were projected to the Euclidean plane. This had to be done as on the Euclidean plane all the geometrical calculations and manipulations are much easier to perform and such a projection should not much affect the obtained dimension of the Sierpinski triangle because the sequencing was done before the triangle was constructed. Chaos game technique is also not very sensitive to errors because of the iterative nature of the process, so the results quickly stabilise after approximately 10 steps.

¹ [http://encarta.msn.com/map_701510067/Amazon_\(river\).html](http://encarta.msn.com/map_701510067/Amazon_(river).html)

² <http://earth.google.com/>

³ <http://lazarus.elte.hu/hun/digkonyv/topo/3felmeres.htm>

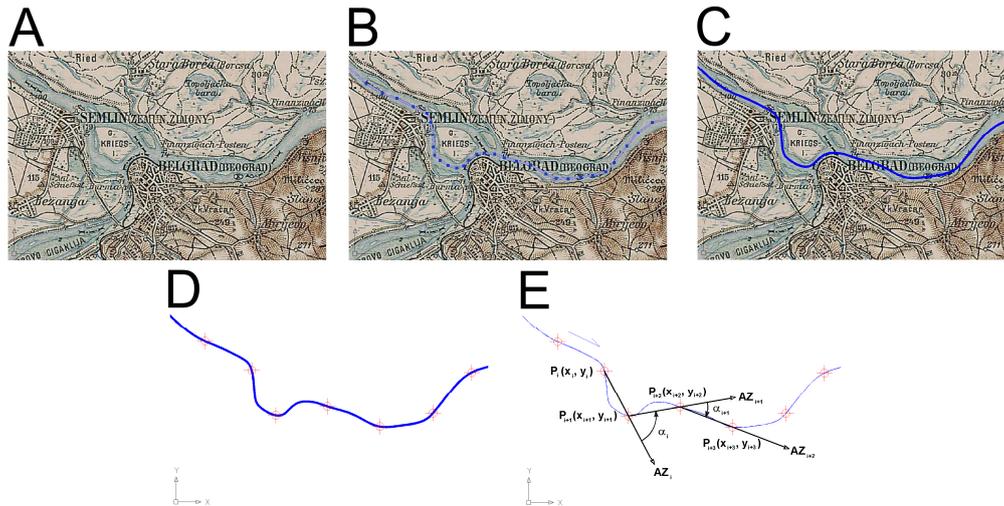


Fig. 2. The procedure for converting the river topography into a set of angles. (A) – raw tiff map; (B) – digitalisation using polyline method; (C) – conversion of the polyline into spline; (D) – division of spline into a defined set of segments – points (1000-segment Danube case); (E) – calculation of the set of angles.

Prior to sequencing, the rivers were digitised using a commercial software AutoCAD package (Autodesk, Inc.) with a single polyline which was later transformed to spline which better represents the course of water flow. In cases where the river behaves as a “braided river” (has several active channels) only the most active channel was digitised due to the limitations of our methodology. Although braided rivers can be analysed using special techniques (Sapozhnikov and Foufoula-Georgiou, 1996), the chaos game algorithm requires that the rivers are presented with only one channel, digitised as a continuous line (spline). Similarly, the pathways through the lakes were digitised as a spline (as a single channel) fitting from feeder to effluent locations.

Another issue has to be discussed prior to the generation of the sequence. Digitalisation started at the river mouth, going upstream. According to the Horton-Stahler stream ordering system, the longest lower-ordered channel was proceeded along.

2.2 Generation of a direction sequence

The main difficulty of this research was the transformation of the topography of the river channel into the three different classes needed to “play” the chaos game. The procedure was as follows. The river courses were digitised from the image files (Fig. 2A) using Autodesk AutoCAD software with the polyline method (Fig. 2B). The polylines were then smoothed into splines (Fig. 2C), to simulate the natural course as much as possible. A fixed number of points were drawn along the curves for each analysis. Their quantities were chosen exponentially, as a power of 2: 1000, 2000, 4000, 8000, 16 000 and 32 000 (Fig. 2D) and will be de-

scribed by the terms “sampling rate” or “segment length” in this article. The lower limit of 1000 points was regarded as the minimum requirement for a Sierpinski triangle image analysis. Raising the number of points above 32 000 was not reasonable because it would yield very monotonous sequences of classes. As a result, 24 arrays of points P_i with coordinates (x_i, y_i) were obtained for further calculations, where i_{max} varies from 1000 up to 32 000 (Fig. 2E). The final step of this procedure was to calculate the array of the angles $(\alpha_i, \text{Fig. 2E})$ which describe the topography of the river by using Eq. (1) and Eq. (2). Following this procedure, the two azimuths from the first point, $P_{i-1}(x_{i-1}, y_{i-1})$ to the second one $P_i(x_i, y_i)$, and from the second point to the third one $P_{i+1}(x_{i+1}, y_{i+1})$ have to be calculated by Eq. (1). That is how the two azimuths AZ_i and AZ_{i+1} have been obtained. The second step is the calculation of the angles (α_i) between these two azimuths using Eq. (2). The number of obtained angles is therefore equal to the number of points minus two.

$$AZ_i = \left\{ \begin{array}{l} (x_{i+1} - x_i) > 0 \Rightarrow AZ_i = 90^\circ - \frac{180^\circ}{\pi} \cdot \arctan\left(\frac{y_{i+1} - y_i}{x_{i+1} - x_i}\right) \\ (x_{i+1} - x_i) < 0 \Rightarrow AZ_i = 270^\circ - \frac{180^\circ}{\pi} \cdot \arctan\left(\frac{y_{i+1} - y_i}{x_{i+1} - x_i}\right) \\ x_{i+1} = x_i \Rightarrow AZ_i = \begin{cases} (y_{i+1} - y_i) < 0 \Rightarrow AZ_i = 180^\circ \\ (y_{i+1} - y_i) > 0 \Rightarrow AZ_i = 0^\circ \end{cases} \end{array} \right\} \quad (1)$$

$$\alpha_i = \left\{ \begin{array}{l} (AZ_{i+1} - AZ_i) < -180^\circ \Rightarrow \alpha_i = (AZ_{i+1} - AZ_i) + 360^\circ \\ (AZ_{i+1} - AZ_i) > 180^\circ \Rightarrow \alpha_i = (AZ_{i+1} - AZ_i) - 360^\circ \\ -180^\circ \leq (AZ_{i+1} - AZ_i) \leq 180^\circ \Rightarrow \alpha_i = AZ_{i+1} - AZ_i \end{array} \right\} \quad (2)$$

Finally, the array of the angles was classified into the three classes. If the direction was inclined to the left more than the critical angle α_L , the segment was assigned a value of L . A value of R was assigned if the direction to the right was greater than a critical angle α_R and the value F (forward) was allocated if it stayed within the limits of α_L and α_R .

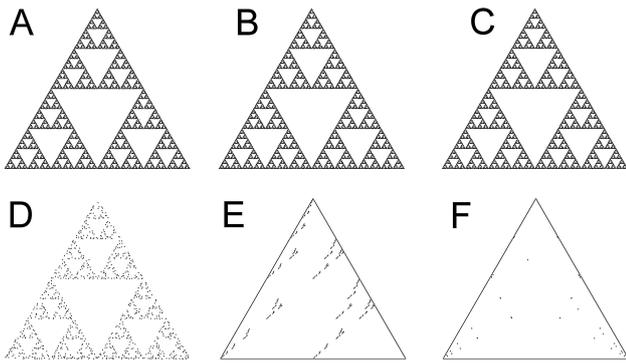


Fig. 3. Technique testing results for various pseudo-random and non-random generators: (A) – digits of π , (B) – digits of e , (C) – MS Excel RAND() function, (D) – LCG algorithm, (E) – bifurcation equation, (F) – sinus function. In cases D and F the size of the points has been enhanced due to visibility and in cases E and F the borders of the triangle have been included for easier visibility.

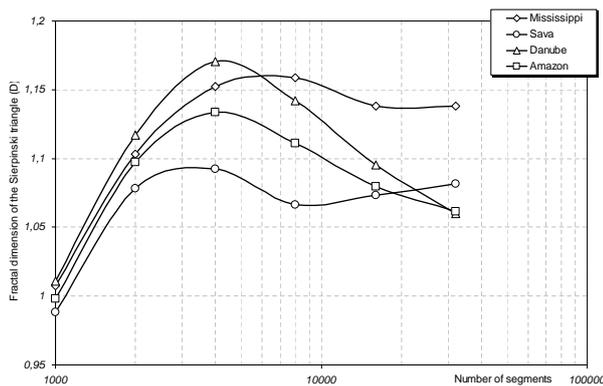


Fig. 4. The dimensions of the Sierpinski triangle, obtained via the chaos game procedure in relation to the number of segments.

The angles were determined by the percentiles of 33.33% and 66.67% of the complete array of angle deviation, to keep the directions uniformly distributed in all three classes and also to avoid the impact of angle choice on the results. For instance, with large critical angles, most of the directions would fall into the F group, and contrarily, with low critical angles, great deviations (mostly L and R) would appear. This was done to obtain the most complete triangle possible. A sequence with dominant L and R classes will yield a triangle with points concentrated only on one side of the triangle. On the other hand, a sequence with the F class dominant will result in a triangle with concentration of points close to the one corner. Finally, a sequence of letters was assigned to the river channel pathway, for instance LLRRRFFLRFLLL-RFRL, which was used as the input for the chaos game.

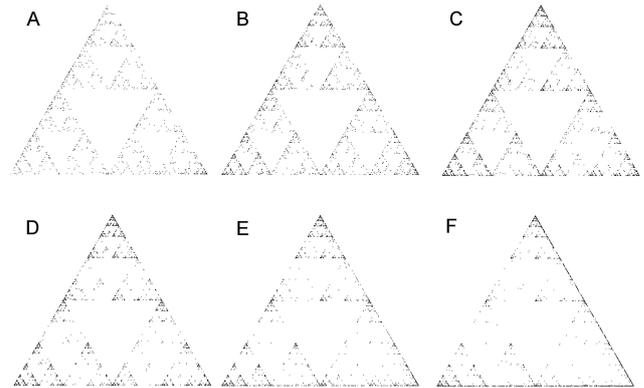


Fig. 5. Plot of chaos game results for the Amazon River, divided by different numbers of observations. (A) – 1000, (B) – 2000, (C) – 4000, (D) – 8000, (E) – 16000 and (F) – 32000 observations. Lower left vertex attracts flow directions to the left (L), lower right forward (F) and upper vertex directions to the right (R).

2.3 The analysis of chaotic behaviour by the chaos game technique

The sequences were analysed by the method of the chaos game (Barnsley 1988; Jeffrey, 1992; Peitgen et al., 2004). This technique is useful for the determination of random patterns in data. The original method relies on visual estimation and in this paper it has been updated to a quantitative approach by examining the fractal dimensions of different approximations of Sierpinski triangles. The “game”, played long enough with a random sequence, produces a complete Sierpinski triangle (Mandelbrot, 1983). If a non-random generating sequence is used, the Sierpinski triangle does not completely appear, as in the case of pseudo-random number generators with short periods, logistic equations or periodic functions (Figs. 3D, E and F). If the random numbers are not properly balanced (for example if some weight is placed on the one side of the dice), the triangle appears biased towards one vertex with greater probability.

The chaos game technique is described in detail in many references (Jeffrey, 1992; Mata-Toledo and Willis, 1997; Peitgen et al., 2004 etc.), so only a brief description is presented here. Points in the equilateral triangle are drawn by the iteration technique with a simple rule: one class is assigned to every corner of the triangle and the seed point is chosen randomly. The next point is drawn halfway between the previous point and the corner of the triangle, corresponding to the class in the sequence. In all cases in this research the seed point of the triangle has been chosen at the origin of the Euclidean coordinate system $P_0(0, 0)$.

Prior to analysing the river data, the method was tested to check if the triangle could be constructed by the appropriate formulas, random or pseudo-random generators and other functions. For each run, 65 536 data pairs were used to represent the triangle as faithfully as possible. In the

Table 1. Results for testing the technique with random and non-random number generators. D – fractal dimension of the constructed triangle; R^2 – correlation between dots on the log-log plot and the linear regression interpolation; $dif\%$ – difference between Sierpinski triangle mathematical dimension and D in percent. The mathematical dimension of the Sierpinski triangle is $D = \log(3)/\log(2) \cong 1.585$.

Generator	D	R^2	$dif\%$
pi	1.5567	0.9965	1.78
e	1.5563	0.9964	1.78
RAND()	1.5518	0.9963	1.78
LCG	1.0108	0.8796	36.2
bifurcation	1.0226	0.9807	35.5
sinus	0.4728	0.8125	75
chaos game, theoretically played with “perfect dice”	1.5850	1	0

literature is stated that 10 000 points are needed to construct a Sierpinski triangle (Peitgen et al., 2004). The sequences are:

- Digits of the constant π . The digits were grouped into packets of 8 decimals and divided by 10^8 . Thus, a number in the range of 0 to 1 was obtained. By the rules of the chaos game, the numbers were classified into three equal sized groups, each corresponding to the triangle vertices. A similar approach was applied to the following formulas.
- Digits of the constant e , grouped into 10 digits.
- Bifurcation equation. The sequence was obtained by the iterated equation: $a(n+1) = 3.999 \cdot a(n) \cdot (1 - a(n))$ (Peitgen et al., 2004) with the value a_1 between 0 and 1. This sequence exhibits chaotic behaviour as its period is infinite or nearly infinite and the sequence is very sensitive to the initial conditions.
- LCG: linear congruential generator generates pseudo-random sequences and has a general equation of the form $f(x+1) = (a \cdot f(x) + n) \pmod{m}$ (Mata-Toledo and Willis, 1997). It requires three parameters (a , n and m), and in this study the formula $a(n+1) = (600 \cdot a(n) + 500) \pmod{977}$ with the number 5 as a seed has been used. Using this array of parameters the function is periodical with a period of 976 iterations.
- RAND() random generator, as an internal function in MS Excel application. This generator is supposed to be superior to other pseudo-random number generators and

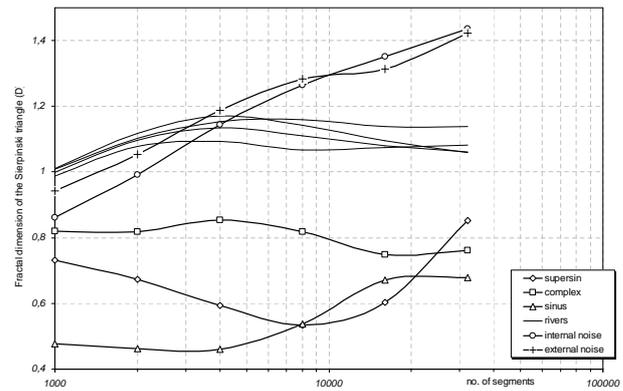


Fig. 6. The comparison of fractal dimensions of natural rivers and the artificially computer-generated curves.

is guaranteed not to repeat the sequence until the value of 10^{13} iterations (MS Excel help).

- Sinus function: $a(n) = \sin(n)$; $n \in \mathbb{N}$ where n is expressed in radians to avoid periodical sampling of the sinus function.

2.4 Determination of the fractal dimension of obtained Sierpinski triangles by box-counting method

The fractal dimensions of the Sierpinski triangles were determined by the box-counting method, which has the most applications in science (Peitgen et al., 2004) and is probably also the most commonly used, as the principle of its use is rather simple. The digitised map of an object (for instance, a river or fracture network) is covered by boxes of different side lengths s , and then the number of occupied boxes $N(s)$ is counted for each box size (Feder, 1988; Bonnet et al., 2001). The process is repeated by reducing the box sizes until the minimum size is reached. For fractal objects, the number of occupied boxes $N(s)$ follows the power-law relationship with the box size s Eq. (3).

$$N(s) \approx s^{-D} \tag{3}$$

The fractal dimension D is therefore calculated as the slope of linear regression best-fit line of log-log data, Eq. (4) (Foroutan-pour et al., 1999).

$$D = - \log N(s) / \log s \tag{4}$$

The coefficient of determination R^2 represents the value of how closely the linear regression fits to the plotted log-log data.

The box-counting method was used to measure the dimensions of the generated image approximations of the Sierpinski triangles and thus to determine the degree of random behaviour of the rivers. The computer code from BCFD program (Verbovšek, 2009) was used for calculation of fractal

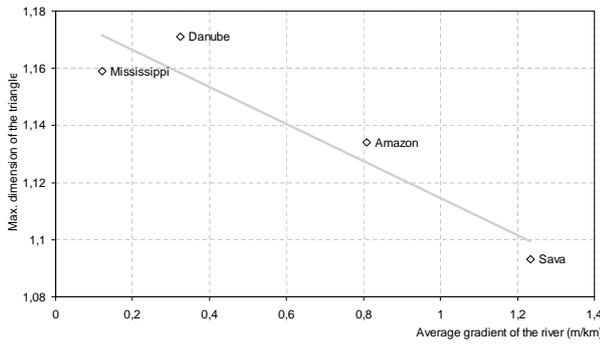


Fig. 7. Scaling between average gradient of the river, expressed in metres per kilometre with the maximum dimension of the triangle, acquired via the chaos game.

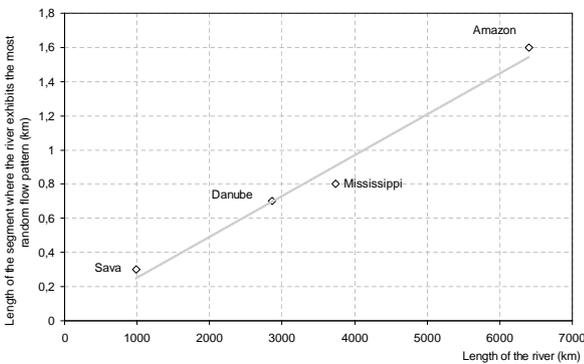


Fig. 8. Length of the segment where the river expresses the most random behaviour in relation to its total length.

dimensions. In this way, the generated image of the triangle was quantified and subjective visual judgments were avoided. All of the triangles were drawn as 1024 × 1024 black and white bitmap images prior to the box counting procedure. This resolution provides a satisfactory compromise between the resolution of the image and the number of pixels.

2.5 Computer generated lines

In order to evaluate our results, the same procedure was applied to three different mathematical functions which might represent the natural rivers. Since the angles are not preserved when the stretching of channel geometry in any of coordinate axes is performed, the precise functions will be presented. First one is a simple sinus function (Eq. 5). The second one, named as “supersin” is a superposition of four sine waves (Eq. 6) with different amplitudes and wavelengths. In all cases empirical coefficients are chosen in such way that the functions behave similarly to the natural rivers.

$$\sinus\left(\frac{x}{500}\right) = \sin\left(\frac{x}{100}\right) \quad (5)$$

$x=1\dots32\,000\text{rad}$

$$\begin{aligned} \text{supersin}\left(\frac{2\pi x}{360}\right) &= 0.7 \cdot \sin(x/50) + \sin(x/100) \\ &+ 1.4 \cdot \sin(x/250) + 2 \cdot \sin(x/1000) \end{aligned} \quad (6)$$

$x=1\dots30\,000\text{rad}$

The third function is a complex superposition of sine waves named “complex” (Eq. 7) where we tried to simulate the natural flow of the river as accurately as possible.

$$\begin{aligned} \text{complex}\left(\frac{2\pi x}{360}\right) &= 5 \cdot \sin(0.5x) + 0.1 \cdot e^{\frac{(x-150)^2}{5000}} \cdot \sin(7x) \\ &+ \frac{x \cdot 360 \cdot \sin(0.1x)}{2 \cdot \pi \cdot 1000} + 10 \cdot \sin(3\sqrt{x}) \end{aligned} \quad (7)$$

$x = 1\dots30\,000\text{rad}$

A complex function is made from four waves (in the same order as in the Eq. 7):

- a basic wave with constant wavelength and amplitude;
- a high frequency wave with a changing amplitude described by a gaussian function, which is set to be maximum in the first third, resembling small meanders in the upper flow of the river;
- a low frequency wave with an increasing amplitude to resemble long-wavelength meanders at the lower flow of the river;
- a steady amplitude but an increasing frequency wave to simulate flow without meanders, conditioned by the regional tectonic conditions in the upper part of the rivers.

Finally we added a low amount of noise to the complex function. Two different approaches have been used. *Internal noise* approach means that the noise has been added to the dependent variable and *external noise* approach means that the noise has been added to the functional result (Eq. 8). In both cases noise influenced approximately 1% of the fluctuations.

$$\begin{aligned} \text{internal}(x) &= \text{complex}(x + \text{noise}) \\ \text{external}(x) &= \text{complex}(x) + \text{noise} \end{aligned} \quad (8)$$

External noise can be explained geologically by the influence of external geological factors, such as geomorphology or tectonic control on a greater observation scale, with possible major faults dictating the directions and deviations of the river course while internal noise is explained by deviations inside the river channels, and these are obviously less important than the external deviations.

3 Results

3.1 Results of testing of the method used in the research

The results from preliminary testing (Table 1) show that the best random sequence is the sequence of decimal digits of the constant π . Also the randomness of the sequence of e

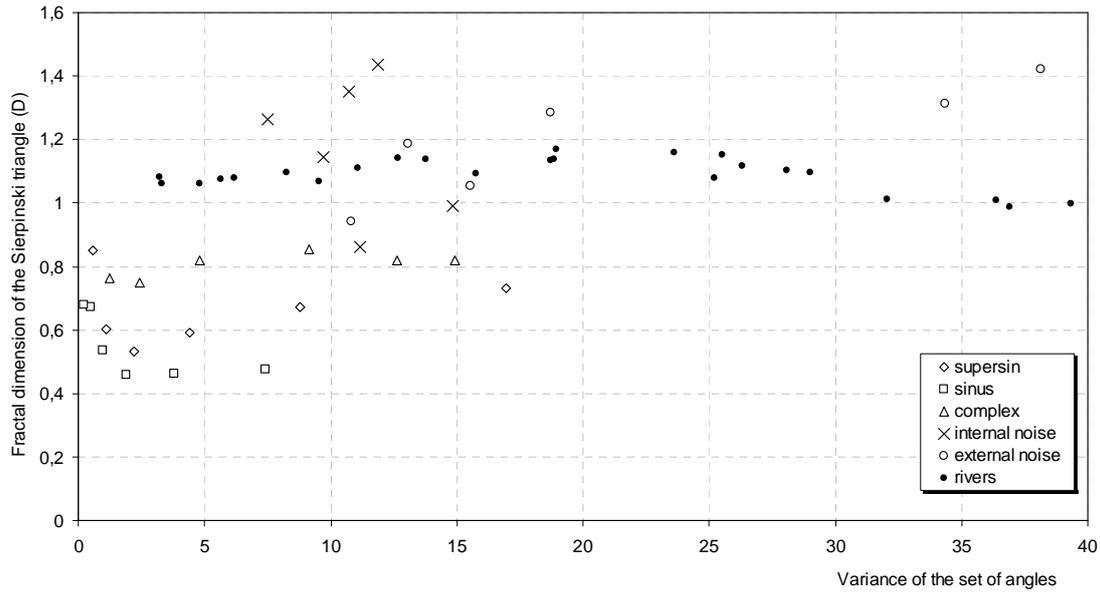


Fig. 9. The dimension of the triangle in relation to the variability of the angles.

Table 2. Results for the four analysed rivers. N – number of segments; AA – average angle (degrees); SD – standard deviation of the angles; D – fractal dimension of the “triangle”; R^2 – coefficient of determination.

Amazon River					Danube River				
N	AA	SD	D	R^2	N	AA	SD	D	R^2
1000	-0.073	39.35	0.9981	0.8901	1000	-0.090	32.07	1.0108	0.8891
2000	-0.017	28.99	1.0974	0.9243	2000	-0.049	26.32	1.1172	0.9231
4000	-0.004	18.73	1.1338	0.9399	4000	-0.029	18.95	1.1705	0.9453
8000	-0.001	11.10	1.1110	0.9459	8000	-0.016	12.68	1.1421	0.9480
16000	-0.001	6.20	1.0796	0.9423	16000	-0.008	8.25	1.0953	0.9476
32000	0.000	3.30	1.0615	0.9453	32000	-0.002	4.81	1.0598	0.9499
Mississippi River					Sava River				
N	AA	SD	D	R^2	N	AA	SD	D	R^2
1000	-0.230	36.40	1.0076	0.8951	1000	0.019	36.93	0.9881	0.8846
2000	-0.134	28.06	1.1031	0.9225	2000	0.003	25.20	1.0784	0.9177
4000	-0.054	25.52	1.1525	0.9512	4000	0.002	15.76	1.0925	0.936
8000	-0.031	23.63	1.1585	0.9477	8000	-0.003	9.52	1.0666	0.9314
16000	-0.015	18.25	1.1383	0.9503	16000	-0.001	5.65	1.0735	0.94
32000	-0.007	13.76	1.1379	0.9552	32000	0.000	3.24	1.0815	0.9462

digits and the RAND() function is satisfactory as the Sierpinski triangle was drawn accurately (Fig. 3A, B and C) and the obtained dimension is the greatest. The small difference between the dimensions of the triangle generated by decimal

numbers of π and the theoretical dimension is probably due to a systemic error (computing with finite accuracy, imaging and box counting with finite resolution, etc.). Examples of incomplete triangles, generated with the deterministic or

chaotic sequences, are presented in Fig. 3D, E and F where the triangle was incomplete or even indistinguishable. The conclusion is that the methods used for creating the triangle and calculating its fractal dimension are acceptable and that the systematic error encountered in using this specific methodology is less than 2%.

3.2 Chaos game representation of the analysed rivers

Since it has been established that the methodology yields accurate results, the sequence from the river directions was used to construct the Sierpinski triangle. As a result, the calculated dimension was used as a measure of how much a river channel flows randomly in its course. The more randomness it contains or, in other words, the more non-periodic behaviour it exhibits, the more faithfully the complete triangle is reproduced and the larger is the dimension D . Table 2 shows the results and also basic statistical parameters of the set of angles prior to 3-class transformation, while Fig. 4 represents the results graphically.

3.2.1 The Amazon

Obtained triangles for the Amazon River sequence are presented in Fig. 5. The largest dimension of the triangle is for a segment length of 4000 points (Table 2, Fig. 4), as the triangle was reproduced most closely in this case (Fig. 5C). With a total length of approximately 6400 km, this means that the Amazon River exhibits the greatest random behaviour at a sampling rate of every 1.6 km. For the largest amount of data (32 000), the points in the triangle are drawn mostly on its edges.

As the segment length increases (or number of points decreases), it is expected that the random behaviour is less pronounced on this scale, because the river channel topography on a large scales follows the macrotectonic environment and regional geological structures and is not very sensitive to local conditions, as is seen in denser observations. In addition, the meanders cannot be seen at this sampling rate. The results clearly confirm these expectations.

3.2.2 The Mississippi

Similar results were obtained for the Mississippi River, and the dimension is greatest for a number of segments between 4000 and 8000 (Fig. 4, Table 2). This value corresponds to a density of observation of between 0.93 and 0.47 km. Visual inspection of the triangles indicates that different behaviour than that of the Amazon River is observed for segments numbering 8000 and especially 16 000 and 32 000. Most of the points in the triangle lie between the lower left and upper vertices, indicating that river directions alternate mostly between left (L) and right (R) without intermediate F class. This indicates that most of the shapes of the river which enhance randomness are larger than 230 m. Smaller shapes are of a more deterministic nature. In comparison with the other

three triangles, the points in these images lie mostly inside the triangle, indicating more random behaviour. With 32 000 segments, almost all points lie on all three of the borders of the triangle. This is a consequence of the observation rate being too dense compared to the size of the meanders. However, in comparison with the Amazon River, where points lie only on two edges of the triangle (Fig. 5F), we can see in the case of the Mississippi that meanders are less regular (less sinusoidal) because they turn from left (L) to right (R) without an intermediate forward (F) flow direction. Regarding this, it is obvious that the standard deviation of the angles at a higher frequency of observation for the Mississippi is much greater than that for the Amazon River (Table 2).

3.2.3 The Danube

The results for the Danube are very comparable with those of the Amazon for all groups of segments (Table 2), except for the fact that the overall dimension of the triangle is greater. The dimension is greatest for the 4000 segments, which corresponds to a sampling rate of every 715 m. Up to 4000 segments, the points lie mostly inside the triangle and, for a higher number of segments, they concentrate on the two triangle edges, indicating more deterministic behaviour or a greater predictability of the meander shapes.

3.2.4 The Sava

Of all the analysed rivers, the Sava exhibits the highest triangle dimension of between 2000 and 4000 observation points. This corresponds to a sampling rate of approximately every 450 m. Another interesting fact that makes the Sava River different from the other three rivers is that when a denser sampling rate is used, the fractal dimension of the triangle slightly increases. At the river sequencing with 8000 observations the points on the triangle are concentrated on its edges on all levels. Contrary to this, when sampling with 16 000 and 32 000 sequence objects, the sampling rate became too dense and the sub-sequences representing particular parts of the meanders became very long. So we obtained sub-sequences of same classes and this resulted in the points on the triangle being concentrated only on its outer sides. This is the main reason why a smaller dimension of the triangle is observed.

3.3 Comparison with computer generated lines

Results for all rivers are represented and compared to computer generated functions with and without noise (Fig. 6). First, the classical sinus function, which is by definition completely deterministic and periodic, so the fractal dimensions of Sierpinski triangles are, as predicted, very low. More importantly, they are much lower than the ones of the natural rivers. The results from the supersin curve (Eq. 6) are similar to the former sinus function, and as the superposition

of several sinus functions is still periodical, the fractal dimensions are still low; below 0.8. Rather higher dimensions can be observed at the higher sampling frequency, at 32 000 points per complete length. Results for the complex function (Eq. 7) yielded the most similar results to natural rivers in the sense that their general shape is the most similar but lower dimensions have been observed. For these reasons, the complex function was used further to investigate the effect of noise on the results. The curves (triangle dimensions) with the added noise intersect curves of natural rivers. Deviations from the natural rivers occur at the higher sampling intervals, as the noise is more pronounced at smaller sampling intervals. Slightly better approximation is obtained with the external noise.

4 Discussion

The shape of the curves (Fig. 4), which represents the dependence of the dimension of the Sierpinski triangle on the density of the observation points, tells us how the random behaviour of the river topography changes in relation to the sampling pattern or, in other words, to the observation patterns. The Sava and the Mississippi Rivers have more or less regular meanders of various sizes and this is the reason why the dimension of the triangle does not drop significantly with an increase in the density of observation. In contrast, both the Danube and Amazon Rivers flow for a significant part of their total length in mountainous and hilly regions with a relatively small amount of discharge. When they reach the lowlands, the discharge is so great that it is not possible any longer to develop small meanders so only large ones are formed. The similarity between the Sava and Mississippi and between the Danube and Amazon Rivers in the sense of random behaviour in relation to sampling rate can be observed from Fig. 5. It is likely that the Sava and Mississippi have developed some features regarding the observation scale (irregular meanders of all sizes for example) while in the Danube and Amazon this is likely not the case.

Another interesting pattern can be discovered from plotting the maximum dimension of the triangle against the average gradient of the river, expressed in metres per kilometre. Figure 7 shows the dependence, expressed with a linear trend line. While with a larger gradient, the river has greater energy and small-scale parameters, as geological or vegetational parameters became less significant to the direction of water flow which has a high energy. The consequence is that the topography of a river channel becomes less random and more deterministic, controlled by large scale geological, tectonic and climatic settings. On the other hand, when the energy of the river is low, its flow is very sensitive to the initial condition in the sense that the shape of one meander has a major influence on the shape of the complete river flow downstream, and also the flow of the river is more influenced by small irregularities in the topography, vegetation, geology or other more local parameters. However, to prove the rela-

tion more of the same type of analysis from different world rivers would be needed.

Another conclusion can be made by examining the scatter plot of the length of the segment when the dimension of the triangle is greatest versus the total length of the river. The plot (Fig. 8) suggests a strong dependence between these two values. This can be explained by the hypothesis that every river develops a maximum degree of random behaviour for a certain observation density, and this density is strongly dependent on the total length of the river. The connection could be the consequence of the self-similar fractal properties of all rivers, since with the shorter rivers (Sava), the length of the segment of maximum chaotic behaviour is shortest, and vice-versa for the longest rivers (Amazon). The argument that the map scale influences this property is not valid because all of the river features were digitised and included. For example, for the Mississippi case Google earth satellite images allow us to digitise the river course with high precision. Despite that, the point representing Mississippi fits well into the regression and is not placed nearby Sava River point (Fig. 8) where the most accurate maps were used. Thus the size of the meanders is connected with the river length or, more possibly, with the river discharge at the mouth of the river, since the discharge is mostly also dependent on the river length. This might represent a problem for further investigations, namely a problem with applying the same technique to rivers in arid environments.

Intuitively the non-predictable behaviour of the river is expected to be connected with the intensity of the river meandering in its pathway or with the number and sharpness of the curves it makes. One should expect that sharper the curves are the less deterministic is the river behaviour and vice versa. As shown in Fig. 9 (rivers dots) this is not the case. The most random behaviour is observed at one specific interval of variance (or curvature of meanders) between 15 and 25. This is likely to be the consequence of some flow regime between regular sinusoidal meanders formed by low gradient rivers and the more or less straight flow of highly energetic waters. But, more probably, the low dimensions of Sierpinski triangles at high angle variances are a consequence of the fact that an insufficient number of points has been used for triangle construction in these cases.

When comparing these four rivers we can divide them into two groups. Firstly, the Danube and Amazon rivers that have meanders of mostly one size. They exhibit the most random behaviour on only one particular observation scale (Fig. 4). The second group of rivers, the Mississippi and the Sava, similarly exhibit one particular peak of random behaviour, but as the magnification increases the rate of random behaviour does not drop significantly (Fig. 4). A possible explanation is that there are two different flow regimes or two different meander sizes: smaller ones when the river has a small current intensity, especially in the upper parts of the flow, and larger ones when the discharge increases. This is likely to be connected with the formation of river

valleys, especially in the upper parts of the Danube and Amazon rivers, where the narrow width of the valleys does not allow rivers to develop meanders and the flow is mostly influenced by tectonic and other regional settings.

The last point discussed is how the size of the meanders, the observation rate and dimensions of the obtained Sierpinski triangle are interconnected. When the observation rate is infrequent, the meanders cannot be detected properly by the sequencing and periodic behaviour is exhibited in the sense that the flow can be described by a ...LRLR... sequence in the case of meandering and, contrarily, with a ...FFF... sequence in the case of a more or less straight flow. On the other hand, if the observation rate is too dense, the sequence also exhibits periodic behaviour in the sense that one meander is described with a long sequence of the same classes, such as ...FFFFFFFF RRRRRRRR FFFFFFFF LLLLLL... In such a case, the nondeterministic behaviour does not appear, as the points on the Sierpinski triangle are concentrated on the two edges (between F and R and F and L vertices). Only with a specific sampling density does the sequence exhibit such behaviour that the majority of the points lie inside the triangle, meaning that the sequence is more or less non-periodic or random. Therefore, we propose that the best measure of the randomness in the river topography is the determination of the maximum dimension of the Sierpinski triangle.

As a final comment, the authors would like to state that word “randomness” in the article is used to describe the sensitivity to a large number of factors influencing the river pathway, such as tectonic and geological settings, types of rock and sediment, vegetation, climatic regime, tributaries etc... We think that if the river exhibits more random behaviour its flow is more influenced by these factors and vice versa. So, random behaviour in this article should be recognised as random behaviour of dice which is, if viewed from a strictly theoretical point, actually deterministic but influenced by a vast number of factors which are all impossible to predict and thus almost impossible to simulate by a computer.

By comparing the fractal dimensions of the natural rivers with the computer generated rivers at different sampling intervals (Fig. 6), some differences are observed which can be explained geologically. Rivers can be best approximated by the complex function, and not by simple sinus function or a linear superposition of different sinus functions. The latter are in any way periodic and their fractal dimensions are much lower than the ones of natural rivers. The complex function includes the changes of the river course at short and long wavelengths, plus some greater deviations in the middle segment of river course (represented by the inclusion of the Gauss function). The natural rivers in the sense of proposed methodology are best approximated by the addition of noise, and the external noise fits the data slightly better than the internal, indicating greater external geological control (for example major faults) than the internal (inside the river channels). At the smaller sampling intervals (higher

sampling resolution) noise did not seem to have a major influence on the natural river pathway, contrary to the lower resolutions (large sampling intervals) where rivers’ “natural regional geological noise” outreach high resolution artificial noise added to the artificial curves.

5 Conclusions

All rivers exhibit some sort of random behaviour which is far greater than the “chaotic” behaviour of the periodic sinus function (Tables 1 and 2, Fig. 6). The dimensions of the generated triangles are even greater than some pseudo-random functions with a short period (LCG for example) or than a completely chaotic bifurcation function. A comparative study of the chaotic behaviour of four major rivers using the chaos game technique shows that the Danube River exhibits the greatest ratio of random behaviour when we sample our observations at every 715 km. Overall, the Mississippi River exhibits the largest proportion of random behaviour, which is not so strongly dependent on the observation frequency and is probably connected with low gradient.

Conclusions drawn from the comparison of natural rivers with the computer generated rivers show that the rivers can be best approximated by computer by the inclusion of random noise into the complex combinations of sinus functions. Any combination or superposition of sinus functions without the noise gives greater deviations with lower fractal dimensions of the Sierpinski triangles. Some noise obviously exists in nature, and this can be detected and quantified by our proposed chaos game method.

The method developed in this article can be further applied to the analysis of other natural geometric patterns, for example coastlines, mountain ridges or everywhere where the transformation of an object geometry into a three-class sequence can be applied. The method is not sensitive to the map scale, as prior to the usage of the chaos game technique sequencing has to take place and this is independent of linear map transformations. But one issue needs to be addressed – the low fractal dimension of the Sierpinski triangles at short sequences. The results at this scale are probably not accurate while too few points are present. Our estimation is that for accurate results at least 4000 points will be necessary.

The main limitation of the proposed method is that it can not evaluate multifractal properties as a measure of network or channel complexity (De Bartolo et al., 2006; Gaudio et al., 2006; Gangodagamage et al., 2007; Gupta et al., 2007 and others). In the case of braided rivers where the most active channel is undistinguishable, or in the case of several equally active channels, some approximations have to be made and the method should be accompanied by other methods.

This method can be further applied to more detailed studies of drainage basins where sequences of all rivers might be joined together. In this way we might study the random patterns in drainage basins. Moreover, the same method might

be applied to coastlines or lines of mountain ridges. We hope that the ideas presented in this paper will provoke some new research. In this paper it has only been possible to provide some remarks about how we should interpret the results. This is why it is crucial to study the results for several different rivers in different climatic and geological settings (and from such institutions which have access to detailed topographic maps) before drawing any final conclusions on how to appropriately interpret the results.

The supplementary files used in this research with examples and original data are available upon request.

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