Statistical inevitability of Horton's laws and the apparent

randomness of stream channel networks



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ABSTRACT

The remarkably regular geometric relations observed in stream networks have been widely interpreted as evidence of a distinctive structure that reflects particular geomorphic processes. These relations have also been interpreted as evidence that stream networks are topologically random, formed by the laws of chance. Neither of these inferences is justified. The officited geometric properties are not specific to particular kinds of stream networks or to topologically random networks; instead, they describe virtually all possible networks. They therefore compel no particular conclusion about the origin or structure of stream networks.

OBSERVED REGULARITIES IN STREAM NETWORK STRUCTURE

Stream networks are important, both as controls on drainage basin hydrology (Kirkby, 1976) and as indicators of geologic processes. The patterns formed by stream channels are thought to reflect regional tectonics (Ollier, 1981; Cox, 1989; Burbank, 1992) and local geologic structure (Abrahams and Flint, 1983), as well as prevailing erosional mechanisms (Dunne, 1980) and climate (Gregory, 1976; Daniel, 1981).

The morphology of stream channel networks is often characterized in terms of Horton's (1945) "laws of drainage network composition" (Fig. 1). Horton's "law of stream numbers" states that N_{ω} , the number of streams of order ω , decreases geometrically with stream order:

$$\frac{N_{\omega}}{N_{\omega+1}} \approx R_{\rm B} \text{ or } N_{\omega} \approx R_{\rm B}^{\Omega-\omega}, \tag{1}$$

where Ω is the order of the network's main stream and $R_{\rm B}$ is the "bifurcation ratio." Horton's "law of stream lengths" holds that $L_{\rm w}$, the mean length of streams of each order, increases geometrically with stream order.

$$\frac{L_{\omega}}{L_{\omega-1}} \approx R_{\rm L} \text{ or } L_{\omega} \approx L_1 R_{\rm L}^{\omega-1}, \tag{2}$$

where L_1 is the mean length of the first-order tributaries and R_L is termed the "length ratio." The "law of stream areas," proposed by Schumm (1956) in the spirit of Horton, holds that drainage-basin area A_{ω} increases geometrically with stream order:

$$\frac{A_{\omega}}{A_{\omega-1}} \approx R_{A} \text{ or } A_{\omega} \approx A_{1} R_{A}^{\omega-1}, \tag{3}$$

where A_1 is the mean area draining into each first-order tributary and R_A is the "area ratio."

Today, many textbooks teach Horton's "laws" as central principles of drainage-basin structure (e.g., Chorley et al., 1984;

Selby, 1985; Press and Siever, 1986; Ritter, 1986; Skinner and Porter, 1987; Judson and Kaufman, 1990; McKnight, 1990; Bloom, 1991; Summerfield, 1991; Easterbrook, 1993). Dozens of studies on stream networks in diverse landscapes have confirmed that, as equations 1-3 predict, semilogarithmic plots of stream lengths, numbers, and areas vs. stream order (e.g., Fig. 1B) are nearly linear. Except in networks subject to structural controls, these studies have also shown that bifurcation, length, and area ratios are restricted to relatively narrow ranges; $R_{\rm B}$ generally varies between 3 and 5, with a modal value of 4, R_L usually ranges between 1.5 and 3, with a modal value of roughly 2, and RA typically ranges between 3 and 6 (Chorley, 1957; Smart, 1972; Abrahams,

Many geomorphic theories yield networks that satisfy Horton's laws and give values of $R_{\rm B}$, $R_{\rm L}$, and $R_{\rm A}$ that closely resemble those typically observed. Horton's

laws have been widely used as empirical tests of specific models of drainage network development (e.g., Roth et al., 1989; Rodriguez-Iturbe et al., 1992). They have also been used to test more general claims that stream networks form from successive generations of rills (Horton, 1945), that networks develop according to principles of maximum entropy or maximum efficiency (Leopold and Langbein, 1962; Woldenberg, 1969; Leopold, 1971; Rodriguez-Iturbe et al., 1992), and that networks are topologically random, developing largely by the laws of chance (Shreve, 1966, 1969, 1975).

NEED FOR A NULL HYPOTHESIS

Attempts to test geomorphic theories against Horton's laws have generally suffered from the lack of an appropriate null hypothesis. They have implicitly (and sometimes explicitly) assumed that Horton's laws, and the observed Horton ratios, characterize a distinctive type of network structure and thus represent a strict empirical test of whether the theories outlined above yield realistic networks. Here I demonstrate that this premise is false by showing that almost all possible networks fit Horton's laws and have Horton ratios similar to those observed in nature.

Directly enumerating all possible networks is impractical, because in networks

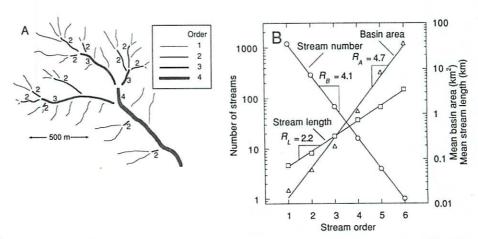


Figure 1. A: Exploded diagram of stream channel network in Marin County, California (Montgomery and Dietrich, 1989), illustrating application of conventional stream-ordering rules (Strahler, 1952): (1) all streams without tributaries are first order; (2) where two streams of order ω join, they both terminate and a stream of order $\omega+1$ begins, and (3) where two streams of unequal order meet, lower-order stream terminates, and higher-order stream continues through junction. B: Number (circles), mean length (squares), and mean drainage area (triangles) of streams in drainage network of Daddy's Creek, Tennessee (data of Morisawa, 1962), plotted as function of stream order, with bifurcation, length, and area ratios ($R_{\rm B}, R_{\rm L}$, and $R_{\rm A}$; see equations 1–3) calculated from slopes of regression lines.

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natural channel networks. Hortonian "laws" observed in studies of most all possible networks obey the same $R_{\rm L}$ plots have $r^2 \ge 0.80$. In other words, algression coefficients $r^2 \ge 0.98$, and 95% of straight; 96% of RB and RA plots have rethree ratios (as in Fig. 1B) are also very (Fig. 2). The Horton plots for each of the are close to those observed in nature $R_L \le 3, 3 \le R_A \le 6$), and the modal values

of theoretical models of network structure. tonian analysis, is an exceedingly weak test ditional stream-ordering protocols and Horstream networks, as viewed through the tralaws. Therefore, the observed regularity of tion of possible networks violates Horton's ries and models, because only a small fraccessfully predicted by many different theonot surprising that Horton's laws are sucsome subset of all possible networks. It is conclusion that natural stream networks are them, yielding only the singularly imprecise acteristics, Horton's ratios fail to identify If stream networks have distinctive char-

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randomness, as I show below. clearly violate the assumption of topological Horton's laws, but so do networks that logically random networks certainly obey data do not justify that interpretation. Topocause they are topologically random, but the stream networks obey Horton's laws be-1975). Today, it is generally believed that sample in Figure 2 (Shreve, 1966, 1969, pologically random, like the Monte Carlo works develop by chance and are in fact tostructural controls, natural channel netdifferent conclusion—that in the absence of have previously been used to draw a very Statistical methods like those used above

ranked set in half yields two topologically various topological criteria. Dividing a cally random) and rank them according to networks in Figure 2 (which are topologi-My approach is to take the Monte Carlo

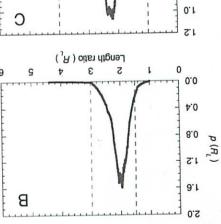
> Figure 1A to each network; tally the numchannels. I then apply the ordering rules in formed by a given number of first-order

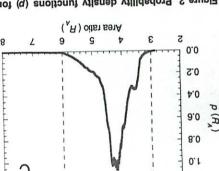
point corresponding to a short main stream accordance with typical practice, if a single Fig. 1B) for each Monte Carlo network (in R_L , and R_A by linear regression (as in each order's drainages, I can estimate RB, in each order, and the number of links in of streams in each order, the number of links detail below). Thus, by tallying the number (these assumptions are addressed in more ies, that make up each stream in a network the total number of links, including tributarorder, so area ratios can be estimated from tributing area per link does not change with streams. I also assume that the average conlinks), rather than physical lengths, of lengths (i.e., the number of component ratios can be estimated from the topological link lengths do not vary with order, so length Here, for simplicity, I assume that average ment (or "link") connecting adjacent nodes. and contributing area of each channel segpoth on network topology and on the length Stream lengths and drainage areas depend which the channels merge (Melton, 1959). property; it depends only on the sequence in streams of each order is a purely topological In a branching network, the number of furcation, length, and area ratios.

natural stream networks (3 \leq $R_{\rm B}$ \leq 5, 1.5 \leq fall within the ranges considered typical for ratios (RL), and 98% of all area ratios (RA) all bifurcation ratios (RB), 95% of all length first-order channels, $20 \le N_1 \le 1000$, 96% of works that can be formed by merging N1 In an unbiased sample of all possible netfrom R_L regressions).

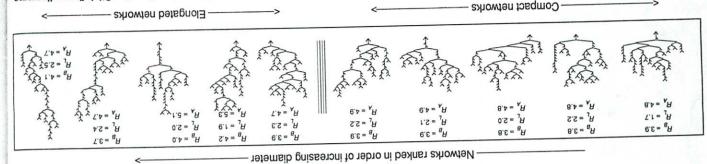
would exert undue leverage, it is omitted

of each order; and calculate the resulting bibers, lengths, and drainage areas of streams an unbiased sample of all possible networks Monte Carlo method (Shreve, 1974) to select very large (Shreve, 1966). Instead, I use a the number of possible configurations is with realistic numbers of first-order streams, Bifurcation ratio (RB) 5 0.0 2.0 4.0 9.0 8.0 0.1 1.2





served in natural stream networks (dashed ton ratios that fall within ranges typically obnetworks in this Monte Carlo sample have Horthat distributions are narrow and that almost all tributed between M₁ = 20 and M₁ = 1000. Note sample (see text) of 9800 networks evenly disbifurcation, length, and area ratios for unbiased Figure 2. Probability density functions (p) for



because they include only networks that are more elongated, or more compact, than median diameter. elongated and compact networks are not reflected consistently in bifurcation, length, and area ratios. Subsets shown are topologically nonrandom of network, and vertical position of individual links indicates number of junctions separating them from outlet. Note that obvious differences between describe network topology (i.e., order in which channels merge), but do not show size, shape, or orientation of channel links. Arrow represents outlet Figure 3. Topologically random set of ten networks formed from 60 first-order streams, ranked in order of increasing diameter. Stick-figure diagrams

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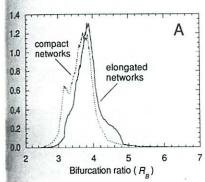
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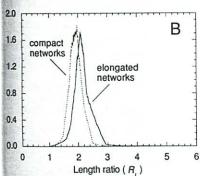
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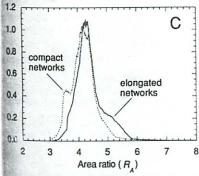
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random subsets of the original random to I then test whether the Horton ratios of ese topologically nonrandom networks deate significantly from those of the topologically random networks.

Figures 3 and 4, I divide the networks fording to diameter, which is the number links separating the outlet from its farthest st-order tributary (Smart, 1978). As Figure shows, networks with larger diameters are pologically longer, whereas networks with aller diameters are more compact. The onte Carlo sample shown in Figure 2 indesten networks for each number of first-der channels N_1 . For Figures 3 and 4, I may the ten networks for each N_1 in order of







gure 4. Distributions of bifurcation, length, id area ratios for topologically nonrandom in which each network is more elongated user diameter) or more compact (smaller diater) than median for each number of first-tar streams (see Fig. 3). Distributions for mapact and elongated networks are almost inserting in the long at the lo

increasing diameter, then divide the ranked set in half. The two resulting sets of networks are mutually exclusive and markedly nonrandom; one set contains only networks that are more elongated than the median diameter for their N_1 , and the other set includes only networks that are more compact than the median diameter.

Even though both the elongated networks and the compact networks clearly violate the premise of topological randomness, they both have distributions of bifurcation, length, and area ratios (Fig. 4) that are almost indistinguishable from those of topologically random networks (Fig. 2). The degree of similarity between these distributions can be quantified through the following thought experiment. Imagine one had a sample of networks drawn from either of the nonrandom distributions shown in Figure 4. How large a sample would be needed to detect a statistically significant difference between either of these distributions and the random set of networks shown in Figure 2? Table 1 shows that roughly $170 R_{\rm B}$ values, 60 $R_{\rm L}$ values, or 280 $R_{\rm A}$ values from elongated or compact networks would be needed to reject, at p = 0.05, the hypothesis that the networks were actually random. Horton's ratios cannot readily distinguish between random and nonrandom sets of networks. Therefore, the fact that the "random model" successfully predicts Horton's laws and the observed Horton ratios does not demonstrate that natural stream networks are actually random.

DISCUSSION

The argument presented above has two central points. First, almost all possible net-

works have Horton ratios similar to those observed in studies of natural channel networks, so those ratios are a very weak test of theories of stream network structure. Second, because both random and nonrandom sets of networks yield essentially the same Horton ratios, the ratios observed in nature do not show whether natural stream networks are random. The generality of these results depends on the following three questions.

First, do link lengths and contributing areas change with order, contrary to the assumptions used here? Some studies of natural channel networks indicate that links in first-order streams are somewhat longer, on average, than those of higher order streams, and others suggest that they are shorter (Shreve, 1969; Smart, 1972; Montgomery and Dietrich, 1989). First-order link lengths are sensitive to the criteria used to define where first-order channels begin, and there are very few published measurements of link contributing areas. However, plausible variations in link lengths and contributing areas are unlikely to affect the results reported here. If first-order links were 50% longer than higher order links, on average, as some field data suggest (Smart, 1972), the length ratio $(R_{\rm L})$ values reported above would be reduced by an average of only 0.1. Doubling the average contributing area to first-order links (Shreve, 1969) would decrease the reported area ratio (R_A) values by an average of only 0.2. The effects of plausible variations in link lengths and areas are small compared to the effects of the stream-ordering definitions, which geometrically compound the numbers of links in each order.

Second, do these results depend on the

TABLE 1. SAMPLE SIZE NEEDED TO DISTINGUISH NONBANDOM GROUPS OF NETWORKS

Networks	R_B	sst	R _L *	ss†	R_A	ss†
Random networks	3.8 ± 0.4		2.1 ± 0.4		4.3 ± 0.6	
Nonrandom subsets	selected accord	ding to:				
Diameter						
≥ median	3.9 ± 0.5	169	2.2 ± 0.4	6 1	4.4 ± 0.7	276
≤ median	3.7 ± 0.4	169	1.9 ± 0.3	6 1	4.2 ± 0.5	277
Width						
≥ median	3.6 ± 0.4	143	1.9 ± 0.3	77	4.2 ± 0.5	232
≤ median	3.9 ± 0.5	140	2.2 ± 0.4	75	4.5 ± 0.7	227
Mean source heigh	aht					
≥ median	3.9 ± 0.5	190	2.2 ± 0.4	64	4.4 ± 0.7	382
≤ median	3.7 ± 0.4	190	1.9 ± 0.3	6 4	$4.2\ \pm\ 0.5$	383
Mean source hei	ght / diameter					
≥ median	3.8 ± 0.4 1	2719	2.0 ± 0.3	4906	4.3 ± 0.6	3290
≤ median	3.8 ± 0.5	2506	2.1 ± 0.4	4855	4.3 ± 0.6	3262

*Means \pm standard deviations for Monte Carlo sample of 9800 networks evenly distributed from $N_1 = 20$ to $N_1 = 1000$, and for nonrandom subsets created by dividing random sample in half according to various topological criteria (see text).

†Approximate number of samples from nonrandom distribution that would be needed to reject (p = 0.05) hypothesis that samples were drawn from random networks, using Kolmogorov-Smirnov test for differences in distributions. Student's t-test is not used, because distributions are both skewed and leptokurtic.

number of first-order streams (N_1) in the networks that were analyzed? Because the likelihood of particular bifurcation, length, and area ratios varies with N_1 , the details of the distributions shown in Figures 2 and 4 will vary somewhat with the range of N_1 sampled. However, repeating this analysis with widely differing N_1 ranges $(20 \le N_1 \le 1000, 20 \le N_1 \le 200, 20 \le N_1 \le 500, 500 \le N_1 \le 1000$, and $20 \le N_1 \le 10000$) yields substantially the same results as reported above.

Third, are the results shown in Figure 4 sensitive to the particular topological criterion used to create the two nonrandom sets of networks? I have repeated the analysis described above using three other structurally important ranking criteria: (1) network width (Kirkby, 1976), the largest number of links equidistant from the network outlet, (2) mean source height (Jarvis and Werrity, 1975), the average distance separating firstorder links from the network outlet, and (3) mean source height divided by diameter, which expresses the degree to which the networks in Figure 3 appear top-heavy or bottom-heavy. In each case, the results agree with those reported above; the Horton ratios of the nonrandom networks are virtually indistinguishable from those of the random networks (Table 1).

These results do not show that natural channel networks are actually nonrandom; they simply show that Horton ratios are poor indicators of whether or not networks are random. Other measures have revealed systematic departures from topological randomness in natural stream networks, even where geologic controls are absent (Smart, 1978; Abrahams, 1984). The "random model" remains particularly useful as an explicit null hypothesis; because it is parsimonious, its premises can be stated concisely, and its implications can be calculated readily. However, failure to reject this null hypothesis does not indicate that networks are in fact random, if the measures used (such as Horton ratios) are not sensitive to randomness in network structure.

This analysis shows that $R_{\rm B}$, $R_{\rm L}$, and $R_{\rm A}$ are profoundly indifferent to network structure. Because the stream-ordering rules create a particular hierarchy of separate streams in channel networks, they constrain $R_{\rm B}, R_{\rm L}$, and $R_{\rm A}$ to relatively narrow ranges, thus enforcing substantial uniformity in the derived stream statistics whether or not there is structural uniformity in the underlying networks themselves. My results support earlier intuitive arguments that Horton's laws must be an artifact of stream-ordering methods (Bowden and Wallis, 1964; Milton, 1966; Smart, 1978), by showing that few possible networks lie outside the usual ranges of $R_{\rm B}$, $R_{\rm L}$, and $R_{\rm A}$, and by showing

that these ratios are insensitive to marked changes in network structure. For example, the Horton ratios of the networks shown in Figure 3 vary little and do not reflect, in a consistent way, the obvious differences between elongated and compact networks. These ratios vary over such a small range (Fig. 2, Table 1) and are so insensitive to pronounced differences in network configuration (Figs. 3 and 4, Table 1) that it is unsurprising that they are also insensitive to prevailing geologic and climatic conditions (Smart, 1978).

These results do not imply that stream networks lack distinctive geometric or topological characteristics. Rather, these results show that if stream networks have unique structural features, those characteristics are unlikely to be revealed in the numbers, lengths, and areas of streams defined through the conventional stream-ordering rules. Factors that regulate channel formation, such as competition for drainage area (Dunne, 1980; Abrahams, 1984) and erosional thresholds controlling channel incision (Montgomery and Dietrich, 1989), should give rise to distinctive network patterns. Devising morphometric techniques to detect the characteristic structure of natural channel networks and explaining that structure in mechanistic terms remain central problems in quantitative fluvial geomorphology.

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