



## 2 Computer-generated global map of valley networks on Mars

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5 [1] The presence of valley networks (VN) on Mars suggests that early Mars was warmer  
6 and wetter than present. However, detailed geomorphic analyses of individual networks  
7 have not led to a consensus regarding their origin. An additional line of evidence can  
8 be provided by the global pattern of dissection on Mars, but the currently available global  
9 map of VN, compiled from Viking images, is incomplete and outdated. We created an  
10 updated map of VN by using a computer algorithm that parses topographic data and  
11 recognizes valleys by their morphologic signature. This computer-generated map was  
12 visually inspected and edited to produce the final updated map of VN. The new map  
13 shows increase in total VN length by a factor of 2.3. A global map of dissection density,  
14  $D$ , derived from the new VN map, shows that the most highly dissected region forms a  
15 belt located between the equator and mid-southern latitudes. The most prominent regions  
16 of high values of  $D$  are the northern Terra Cimmeria and the Margaritifer Terra where  $D$   
17 reaches the value of  $0.12 \text{ km}^{-1}$  over extended areas. The average value of  $D$  is  $0.062 \text{ km}^{-1}$ ,  
18 only 2.6 lower than the terrestrial value of  $D$  as measured in the same fashion. These  
19 relatively high values of dissection density over extensive regions of the planet point  
20 toward precipitation-fed runoff erosion as the mechanism of valley formation. Assuming  
21 warm and wet early Mars, peculiarity of the global pattern of dissection is interpreted in  
22 the terms of climate controlling factors influenced by topographic dichotomy.

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### 26 1. Introduction

27 [2] Martian valley networks (hereafter referred to as VN),  
28 discovered in 1971 by the Mariner 9 spacecraft, are geo-  
29 morphic features on Mars exhibiting some visual resem-  
30 blance to terrestrial river systems. The VN are present  
31 mostly in the Martian highlands and date mostly to the  
32 Noachian [Fassett *et al.*, 2008; Hoke and Hynes, 2008].  
33 They point to a possibility that Noachian Mars was warmer  
34 and wetter than present-day Mars, perhaps capable of  
35 maintaining precipitation and some form of life. A geomor-  
36 phic analysis of VN should, in principle, be able to establish  
37 a predominant style of erosion leading to their formation.  
38 Finding features indicative of runoff erosion signals precip-  
39 itation and a warmer climate whereas features indicative of  
40 erosion by groundwater sapping allow for cold and dry  
41 conditions during the formation of VN. Alas, the geomor-  
42 phic evidence is inconclusive and the origin of VN remains  
43 a topic of active research.

44 [3] The VN are not as mature and integrated as typical  
45 terrestrial analogues. Many VN exhibit widely spaced  
46 tributaries with alcove-like terminations [Sharp and Malin,  
47 1975; Pieri, 1980], constant valley width downstream  
48 [Goldspiel *et al.*, 1993], short, stubby tributaries [Baker

and Partridge, 1986], flat longitudinal profiles [Aharonson 49  
*et al.*, 2002], and U-shaped cross sections [Carr, 1995]. On 50  
Earth such features are attributed to erosion by groundwater 51  
sapping. Groundwater sapping, which rarely dominates on 52  
Earth, generates landforms quite different from landforms 53  
formed by rainfall-fed runoff erosion [Laity and Malin, 54  
1985; Howard, 1988; Luo *et al.*, 1997]. The formation of 55  
VN by sapping could happen in cold and dry conditions, 56  
providing that groundwater was recharged by some mech- 57  
anism such as hydrothermal circulation [Gulick and Baker, 58  
1990; Tanaka *et al.*, 1998]. 59

[4] On the other hand, runoff origin of VN is supported 60  
by branching, dendritic patterns, origin near dividing ridges 61  
[Milton, 1973; Irwin and Howard, 2002; Craddock and 62  
Howard, 2002; Hynes and Phillips, 2003; Stepinski and 63  
Collier, 2004], consistency with crater degradation [Craddock 64  
and Maxwell, 1993; Craddock *et al.*, 1997; Forsberg-Taylor *et al.*, 65  
2004], and significant erosion [Goldspiel and Squyres, 66  
1991; Grant, 2000; Gulick, 2001]. In addition, surface runoff 67  
is necessary to carry away the debris generated by potential 68  
groundwater sapping [Craddock and Howard, 2002; Lamb 69  
*et al.*, 2006]. Runoff origin of VN points to precipitation on 70  
early Mars. 71

[5] A common feature of all aforementioned studies is 72  
their reliance on detailed examination of selected valley 73  
segments or individual networks. However, these detailed 74  
studies must be supplemented by an analysis of an overall 75  
dissection pattern in order to formulate a viable hypothesis 76  
of VN origin. Pattern analysis requires a global, sufficiently 77

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78 detailed map of VN. The existing global map was drawn by  
79 Carr [Carr, 1995; Carr and Chuang, 1997] using Viking-  
80 based images covering  $\pm 65^\circ$  of latitude. The Carr map  
81 contains  $\sim 348,000$  km of valleys; taken globally it depicts  
82 an immature drainage heavily concentrated in the southern  
83 highlands. The newer, different types of global mosaics of  
84 Mars, such as the Mars Orbiter Laser Altimeter (MOLA)  
85 Mission Experiment Gridded Data Records (MEGDR)  
86 [Smith et al., 2003], or THEMIS mosaic [Christiansen,  
87 2004] reveal existence of many valleys not depicted in Carr's  
88 map [Hynek and Phillips, 2003; Mangold et al., 2004;  
89 Stepinski and Collier, 2004; Ansan and Mangold, 2006].  
90 Therefore, construction of a new, more comprehensive  
91 global map of VN is in order. The absence of such map is  
92 due to the high cost of its acquisitions by standard means of  
93 visual interpretation of images, although there is an effort  
94 underway [Hynek et al., 2008] to construct such a map.

95 [6] Recent advances in machine extraction of geomorphic  
96 features from topographic data [Molloy and Stepinski, 2007]  
97 make it feasible to acquire the VN map automatically - by  
98 computer parsing of the MEGDR. In this paper we use such  
99 automated approach to construct a planet-wide map of  
100 VN that would reveal more accurately the global pattern of  
101 dissection on Mars. The resultant map shows much more  
102 dense dissection than was depicted by [Carr and Chuang,  
103 1997]. This result adds to the growing body of evidence that  
104 points toward precipitation-fed runoff erosion as an ultimate  
105 origin of VN, and thus toward the warm and wet climate on  
106 early Mars. We propose to explain the different patterns of  
107 dissection on early Mars and Earth in terms of different  
108 global topographies of the two planets rather than in terms  
109 of the overall availability of liquid water.

110 [7] The rest of the paper is organized as follows: section 2  
111 describes methodology and results of mapping the valley  
112 networks. Section 3 discusses the methodology for deriving  
113 a global map of dissection density from the mapped VN and  
114 zonal statistics of the dissection density. Finally, in section 4  
115 we discuss our results, assess the utility of the new maps,  
116 and point to directions of future development.

## 117 2. Mapping Valley Networks

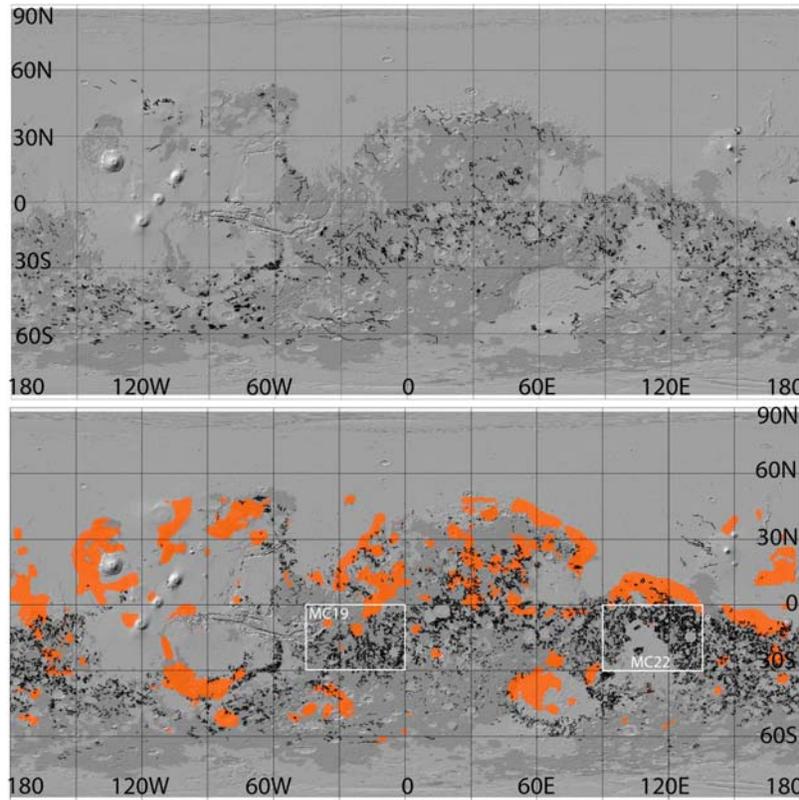
### 118 2.1. Mapping Methodology

119 [8] In principle, the THEMIS daytime IR data set of  
120 images represents the best quality global representation of  
121 Martian surface, but lack of reliable autodetection methods  
122 of VN from images prevents its utilization for our purposes.  
123 On the other hand, topographic data provide more direct  
124 information on landscape dissection and a suitable input for  
125 automapping of VN. The Martian global topographic data  
126 set (digital elevation model or DEM) is available as the  
127 MEGDR [Smith et al., 2003]. We use the MEGDR DEM  
128 with spatial resolution of 128 pixels/degree or about 463 m/  
129 pixel at the equator. Note that MEGDR is an interpolated  
130 DEM, as the cross-track spacing of the original MOLA  
131 measurements is variable and  $\sim 1000$  m at the equator  
132 [Neumann et al., 2001] so some of the grid cells lack actual  
133 measurement values. This limitation on the quality of the  
134 available data means that some valleys or portions of the  
135 valleys, located in places where the MEGDR is smudged  
136 due to lack of data, may not be properly mapped by our  
137 algorithm.

[9] The core technology underpinning our mapping method 138  
has been developed by Molloy and Stepinski [2007]. It 139  
detects incisions directly from terrain morphology, making 140  
it possible to accurately map regions with highly variable 141  
drainage density. The valleys are mapped only where they 142  
are "seen" by an algorithm without resorting to any particular 143  
model of dissection. The emphasis on mapping rather than 144  
modeling the location of VN is what distinguishes our 145  
method from previous applications of computer algorithms 146  
to delineate the VN. Modeling the drainage network from a 147  
DEM by assuming a uniform and constant rainfall and a 148  
particular criterion for surface channelization is routinely 149  
performed in terrestrial hydrology [O'Callaghan and Mark, 150  
1984; Tarboton et al., 1991, 1992; Montgomery and Dietrich, 151  
1992; Peckham, 1995]. These modeling methods were also 152  
applied to individual VN on Mars [Stepinski et al., 2002, 153  
2004; Caldarelli et al., 2004; Mest and Crown, 2008] in 154  
order to compare manually mapped networks with model 155  
predictions. However, the model-based automatic delineation 156  
of drainage fails to provide an accurate global map of 157  
VN because the underlying assumptions of the model, while 158  
apt to terrestrial conditions, are apparently unsuitable for 159  
Mars. 160

[10] The basic idea behind our method is to flag land- 161  
forms having a U-shaped morphology that is characteristic 162  
of the valleys. Using terrain morphology to extract valleys 163  
and drainage networks had been proposed in the past 164  
[Peucker and Douglas, 1975; Band, 1986; Howard, 1994; 165  
Tarboton and Ames, 2001], however, these early attempts 166  
had not led to robust mapping algorithms as they were 167  
unable to overcome a key technical challenge of minimizing 168  
the noise in calculation of a terrain parameter used to flag 169  
the valleys. The algorithm developed by Molloy and Stepinski 170  
[2007] calculates the topographic planar curvature (TPC) 171  
analytically for each pixel in the DEM using a polynomial 172  
approximation to the local patch of the surface. This method 173  
minimizes noise in the values of the TPC. The TPC 174  
influences the divergence/convergence of potential flow. 175  
The positive values of the TPC flag segments of terrain 176  
where flow converges (we refer to them as U-shaped terrain), 177  
whereas the negative values of the TPC flag segments of 178  
terrain where flow diverges. Thus, the valleys are identified 179  
as segments with relatively large positive values of the TPC 180  
( $TPC > 0.003 \text{ m}^{-1}$  has been used in the present calcula- 181  
tions). Note that TPC is a local measure, so valleys wider 182  
than  $\sim 10$  pixels ( $\sim 5$  km) may not register as U-shaped 183  
features. Molloy and Stepinski [2007] have designed a 184  
number of image processing transformations that turn these 185  
segments into an actual map of VN. 186

[11] Details of the accuracy assessment of the automap- 187  
ping algorithm were given by Molloy and Stepinski [2007]. 188  
Here we briefly summarize the result from that study. The 189  
accuracy was assessed based on eight test sites having areas 190  
of the order of  $10^5 \text{ km}^2$  each. For these sites manual 191  
delineations of VN (based on the THEMIS daytime IR 192  
images with the resolution of 100 m/pixel) were compared 193  
with the results of automapping using quantitative measures 194  
pertaining to the overall length of the valleys, size of the 195  
watershed, and the physical "closeness" of the two net- 196  
works. Overall, they found a good agreement between 197  
manual and automatic mapping. The total length of auto- 198  
mapped valleys is less than the total length of drawn 199



**Figure 1.** Visual comparison of valley networks mapping: (top) the Carr map and (bottom) the map generated by our algorithm. Dark gray background indicates extent of the Noachian terrain. Regions where algorithm has identified large number of false valleys are shown in orange. Two quadrangles previously mapped using our algorithm are outlined in white.

200 valleys; however this difference is attributed to the different  
 201 resolutions of THEMIS and MEGDR data sets. The water-  
 202 sheds of automapped VN are larger than the watersheds of  
 203 drawn VN because automapped networks benefit from  
 204 topographic information and conform better to the underly-  
 205 ing landscape. Finally, the automapped and drawn networks  
 206 are “close” to each other as measured by a statistics of  
 207 differences between pixel-to-valley drainage lengths calcu-  
 208 lated using the two networks.

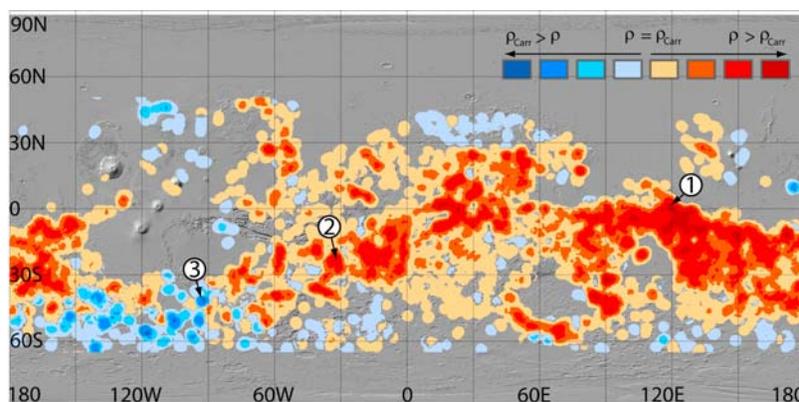
209 [12] The automapping algorithm has also been tested for  
 210 mapping VN over the entire quadrangles. Application of  
 211 this algorithm to the Mare Tyrrenum (MC22) quadrangle  
 212 [Luo and Stepinski, 2006], yielded 172,396 km of U-shaped  
 213 features, of which 142,885 km were visually confirmed to  
 214 be actual VN. For comparison, only 25,447 km of VN are  
 215 present in the Carr map of this quadrangle. Similarly,  
 216 applying our algorithm to the Margaritifer Sinus (MC19)  
 217 quadrangle [Stepinski et al., 2007] yielded 115,429 km of  
 218 VN versus 33,949 km of VN present in the Carr map of the  
 219 same region. The locations of these quadrangles are indi-  
 220 cated on Figure 1.

221 [13] We have set to automap the VN located between  
 222 50°N and 70°S on Mars. In order to obtain a global map we  
 223 subdivided the entire MEGDR into overlapping tiles having  
 224 size of 10° by 10° each. This mosaicing is necessitated by  
 225 the size of the complete MEGDR-based DEM (46080 ×  
 226 22528 pixels) that is too large to be processed in its entirety.  
 227 The overlapping buffer (with a width of 100 km) between  
 228 adjacent tiles prevents mapping incomplete valley frag-

ments. The valleys are identified and measured at each tile  
 229 separately and the results from individual tiles are concat-  
 230 enated into a single database from which duplicate valleys,  
 231 which are present due to buffering, are eliminated before the  
 232 final map is created. 233

## 2.2. Mapping Results 234

[14] The resulting (unverified) map was placed into an  
 235 ArcGIS shapefile that contained ~375,550 individual valley  
 236 segments. A segment is an unbifurcated line on a map; it  
 237 corresponds to a single “link” [Rodriguez-Iturbe and  
 238 Rinaldo, 1997] in a drainage network. For comparison,  
 239 the Carr map contains 11,336 segments. In order to facilitate  
 240 future analyses a database of actual networks instead of  
 241 segments is more desirable. A network is a set of connected  
 242 segments organized in a hierarchical system starting from  
 243 the outlet and bifurcating in the upstream direction. In order  
 244 to create a database of VN we parsed the set of 375,550  
 245 individual segments to determine which segments are asso-  
 246 ciated with each other and thus can be combined into a  
 247 network. The result of this calculation is another ArcGIS  
 248 shapefile, visually identical to the original shapefile of seg-  
 249 ments, but containing 178,012 “netlets.” Due to the limited  
 250 quality of the topographic data and the degraded character of  
 251 many VN our parsing algorithm is unable to collect all  
 252 segments constituting a complete network (as interpreted  
 253 by an analyst). The best it can do is to group segments that  
 254 connect to each other. We refer to these agglomerates as  
 255 netlets. Depending on the state of preservation of a given  
 256



**Figure 2.** Quantitative comparison of valley networks mapping using the difference,  $\rho - \rho_{\text{Carr}}$ , of line densities calculated for each map. Areas chosen for in-depth comparison are labeled 1 to 3.

257 VN and the quality of the MEGDR in its location, it may take  
 258 anywhere from a single netlet to several netlets to represent  
 259 an actual network. The database of netlets (an attribute table  
 260 associated with the netlets shapefile) lists numerical attrib-  
 261 utes associated with each netlet: ID number, outlet coordi-  
 262 nates, location (bounding box), the underlying geologic  
 263 units, the number of segments in the netlet, netlets's order  
 264 and magnitude [Rodriguez-Iturbe and Rinaldo, 1997], total  
 265 length of all tributaries in the netlet, the length of the main  
 266 valley, the elevation of the outlet and source of the main  
 267 valley, sinuosity, slope and aspect of the main valley, and  
 268 the area of a drainage basin associated with the netlet.

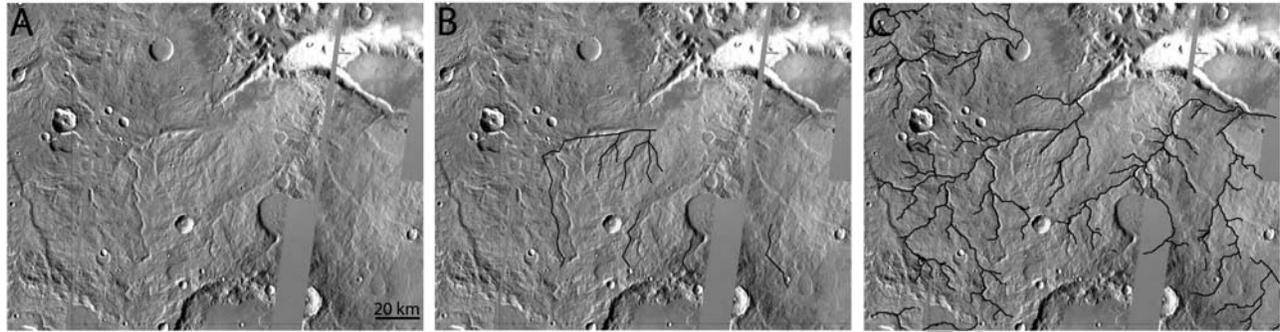
269 [15] Visual inspection of the uncorrected map of netlets  
 270 reveals that our automapping procedure yields a lot of false  
 271 positives (objects that are not really valleys), especially  
 272 outside of the Noachian surface. This is expected because  
 273 the algorithm identifies all features having a U-shaped  
 274 morphology. In the Noachian a great majority of such  
 275 features are indeed valleys, but other features on Mars  
 276 surface, such as, for example, grabens, also may be recog-  
 277 nized as having a U-shaped morphology. In order to eliminate  
 278 false negatives we visually inspected and edited the original,  
 279 unverified map. The result is a corrected map containing  
 280 24,941 netlets. Thus, 86% of netlets are eliminated by visual  
 281 inspection, many of which are small single segment netlets  
 282 resulting from imperfection in the data. Eliminated netlets  
 283 (false positives) constitute 76% of the total length of the  
 284 features originally mapped as "valleys." On Figure 1  
 285 (bottom) we indicated the regions on Mars where the  
 286 algorithm has mapped a large number of false positives.  
 287 These include areas located along the dichotomy boundary,  
 288 around the Olympus Mons, Tharsis bulge, Alba Patera, at  
 289 the floors of big impact craters such as Hellas and Argyre,  
 290 all characterized by tectonic and/or volcanic activities. The  
 291 total length of the valleys in the corrected database is  
 292 807,000 km, approximately 2.3 times the total length of  
 293 valleys in the Carr map. Figure 1 shows a visual comparison  
 294 between the Carr map and our automapped, corrected map.  
 295 Our algorithm finds VN in the same regions (mostly  
 296 Noachian) where they have been mapped by Carr, but it  
 297 finds more of them.

298 [16] Figure 1 offers only a qualitative comparison be-  
 299 tween the two maps. In order to perform a quantitative  
 300 comparison we calculated the line density,  $\rho$ , of valleys as

shown on the two maps. Line density at a given location is  
 the density of linear features (valleys) in the neighborhood  
 of that location, i.e., the total length of valleys divided by  
 the area of the neighborhood. We have used a grid of  
 sampling locations with 4 km spacing. For each location,  
 a circle centered on that location and having a radius of  
 150 km was used as a neighborhood. The result of the line  
 density calculation is a raster where each pixel has a value  
 of  $\rho$  calculated on the basis of its neighborhood. As applied  
 to VN, the concept of line density is very similar to the  
 concept of drainage density. However, whereas drainage  
 density pertains to hydrology, the line density pertains to  
 graphics. In calculating  $\rho$  the neighborhood over which  
 lines are accumulated is chosen arbitrarily and its shape and  
 size are fixed throughout the region. In calculating drainage  
 density the neighborhood is a drainage basin, its shape and  
 size are dictated by the topography and changes throughout  
 the region. Thus, values of  $\rho$  cannot serve as estimates of  
 drainage density and we utilize them only for assessing the  
 difference between the two maps.

[17] Figure 2 shows a map of  $\rho - \rho_{\text{Carr}}$ . The regions  
 shown in warm colors (yellow-to-brown) are the areas  
 where our corrected map depicts more valleys (by length)  
 than the Carr map. The regions shown in cool colors (light  
 blue-to-dark blue) are the areas where Carr map depicts  
 more valleys. The new map depicts more valleys almost  
 everywhere. The most notable exceptions are Aonia Terra  
 and Terra Sirenum located at high southern latitudes, and  
 Alba Patera and the northernmost extent of Arabia Terra  
 located in the mid northern latitude. We have chosen three  
 specific close-up sites, annotated 1, 2, and 3 on Figure 2,  
 to demonstrate the differences between the two maps. The first  
 two sites demonstrate locations where our algorithm has  
 mapped significantly more valleys than are present in the  
 Carr map. The third site focuses on a location where our  
 algorithm extracted less valleys than are present in the Carr  
 map.

[18] Site 1 (Figure 3) is located in the Terra Cimmeria and  
 centered on 121.21°E and 1.22°N; the background image is  
 the THEMIS daytime IR mosaic. The Carr map (Figure 3b)  
 registers a single prominent network and another less  
 prominent channel. Note that the Carr map is not correctly  
 registered with the MDIM 2.1 standard resulting in a  
 misalignment between the map of the network and the



**Figure 3.** Close-up of the VN mapping in a site labeled 1 in Figure 2. The site is located in the Terra Cimmeria and centered on 121.21°E and 1.22°N. (a) Image of the site, (b) Carr map of VN, and (c) the autogenerated map of VN. The background image is the THEMIS daytime IR mosaic.

345 image. Our map (Figure 3c) registers 5.8 more valleys (by  
 346 length) in the same site than the Carr map does. Site 2  
 347 (Figure 4) is located on the southwestern flank of the Ladon  
 348 basin and centered on 32.22°W and 20.41°S; the site  
 349 includes the Arda Valles. The Carr map (Figure 4b) shows  
 350 the Arda Valles, but it does not show an extensive system of  
 351 valleys located to the east and southeast of the Arda Valles  
 352 and clearly visible in the THEMIS image. Again, our  
 353 autogenerated map (Figure 4c) registers most valleys seen  
 354 in the THEMIS image, 2.6 times more (by length) than are  
 355 present on the Carr map.

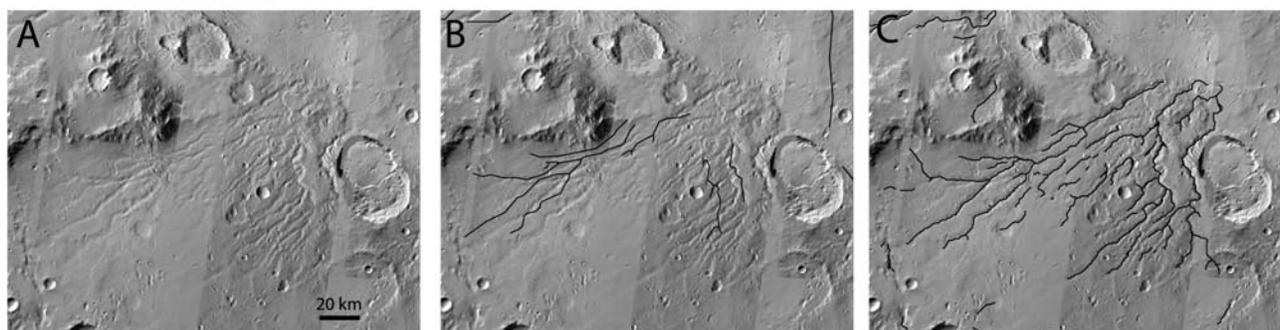
356 [19] Finally, Site 3 (Figure 5) is centered on 92.41°W and  
 357 41.84°S and shows the Warrego Valles: arguably the best  
 358 known example of VN on Mars. Figures 5a and 5b show  
 359 THEMIS daytime IR image and shaded relief derived from  
 360 the MEGDR, respectively. Also shown in Figure 5b are  
 361 MOLA tracks. Although the network is clearly visible in the  
 362 image, it is much less prominent in the topography as  
 363 depicted by the MEGDR. As it happens the quality of the  
 364 MEGDR in the location of the Warrego Valles is rather  
 365 poor; MOLA tracks show gaps several kilometers wide in  
 366 some places (Figure 5b). The lack of data results in a  
 367 presence of many smudges obscuring the network. Conse-  
 368 quently, the Carr map (Figure 5c) registers 2.3 times more  
 369 valleys (by length) than our algorithm does (Figure 5d).  
 370 Moreover, during the process of manual editing, we have  
 371 apparently overcorrected the original automap (Figure 5d)  
 372 resulting in further reduction of channels (Figure 5e). The

Warrego Valles site represents a rare example of location  
 373 where our autogenerated map shows less valleys than the  
 374 Carr map. Such sites are associated with the locations where  
 375 MEGDR quality is below average. There is also a possibility  
 376 that we have inadvertently deleted some actual valleys during  
 377 the manual edit of an uncorrected automap. 378  
 379

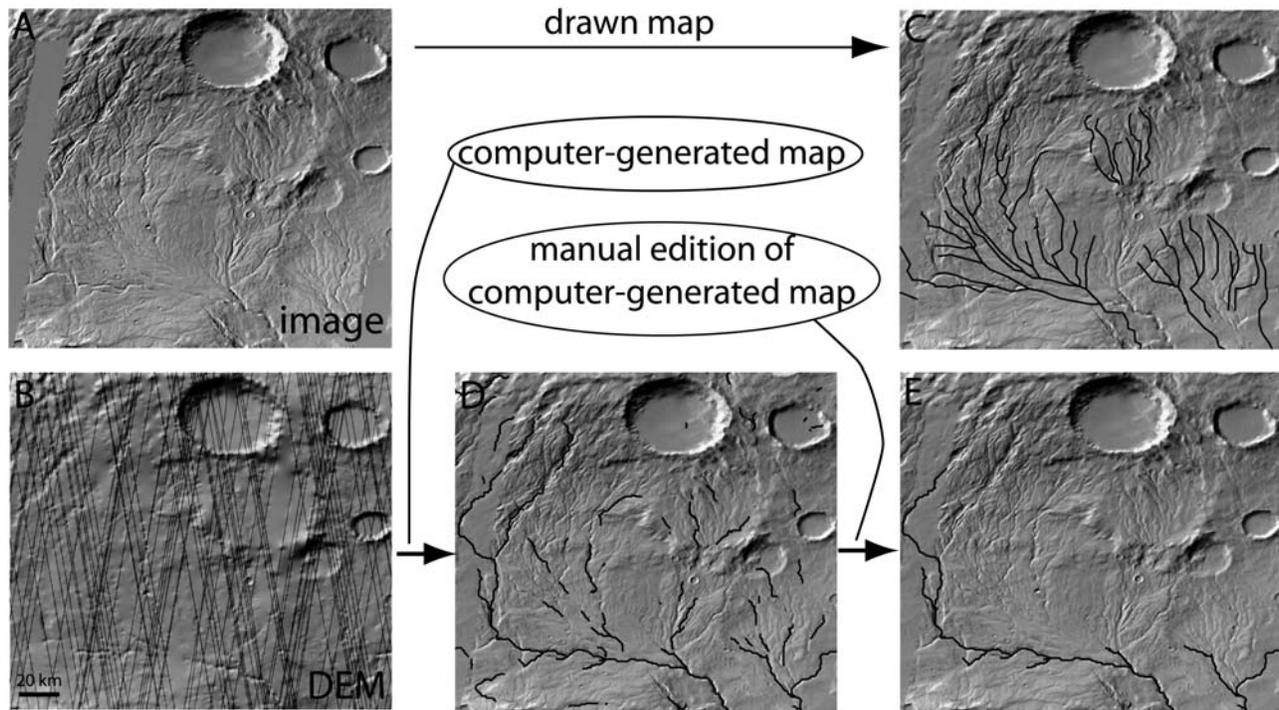
### 3. Mapping Dissection Density 380

#### 3.1. Methodology 381

[20] The map of VN depicts the spatial distribution of  
 382 dissection on Mars and provides a convenient tool for visual  
 383 and descriptive analysis of the dissection pattern. However,  
 384 the map of VN is not well suited for a quantitative analysis  
 385 geared toward determining factors responsible for the  
 386 observed pattern of dissection. In order to utilize the  
 387 information contained in the map of VN for quantitative,  
 388 computer-based analyses we derive a raster map of dissec-  
 389 tion density. Dissection density,  $D$ , is conventionally  
 390 defined as the total length of valleys per unit area. When the  
 391 area is well defined, for example, a geological unit or  
 392 drainage basin, calculation of  $D$  from a map of VN is  
 393 straightforward. However, often such spatial units are not  
 394 readily available. Moreover, even when the spatial units are  
 395 defined, conventional calculation does not facilitate map-  
 396 ping variations in  $D$  within each individual, possibly much  
 397 extended, unit. These problems are addressed by utilizing a  
 398 generalized measure of  $D$  that is defined for every pixel in a  
 399



**Figure 4.** Close-up of the VN mapping in the Arda Valles site labeled 2 in Figure 2. The Arda Valles site is centered on 32.22°W and 21.41°S. (a) Image of the site, (b) Carr map of VN, and (c) the autogenerated map of VN. The background image is the THEMIS daytime IR mosaic.



**Figure 5.** Close-up of the VN mapping in the Warrego Valles site labeled 3 in Figure 2. The Warrego Valles site is centered on  $92.41^{\circ}\text{W}$  and  $41.84^{\circ}\text{S}$ . (a) THEMIS daytime IR mosaic image of the site, (b) shaded relief based on the MEGDR, along with MOLA tracks, (c) Carr map of VN, (d) uncorrected, autogenerated map of VN, and (e) corrected, autogenerated map of VN.

400 DEM and does not require predefined spatial units for its  
401 calculation. We refer to such measure as the “continuous”  
402 dissection density.

403 [21] The continuous dissection density is defined [Tucker  
404 et al., 2001] in terms of a local (pixel-based) and easily  
405 calculated variable,  $L(x, y)$ : the downslope distance to the  
406 nearest valley from a given pixel  $(x, y)$ . The mean value  $\langle L \rangle$ ,  
407 calculated over a specified neighborhood of  $(x, y)$  is  
408 physically related to dissection density at the point  $(x, y)$ .

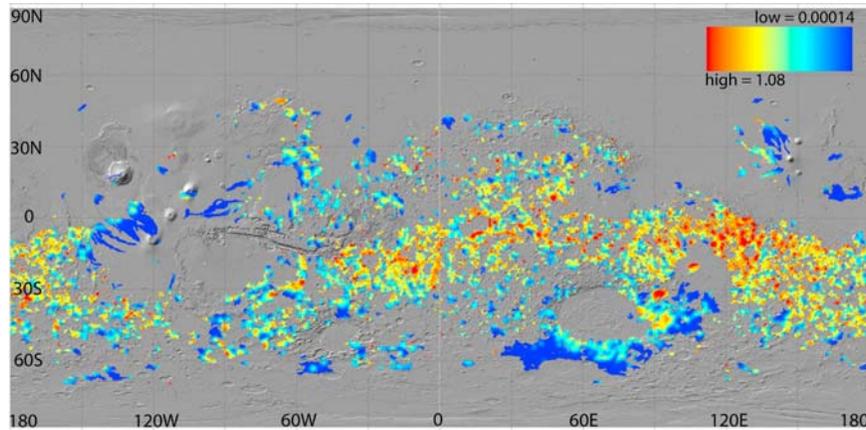
$$D(x, y) = \frac{1}{2\langle L \rangle} \quad (1)$$

411 [22] We use a correlation length of  $L(x, y)$ ,  $\Lambda$ , as a size of  
412 the neighborhood over which the average  $\langle L \rangle$  is calculated.  
413  $\Lambda$ , read from the covariance function of  $L(x, y)$ , is the  
414 distance beyond which  $L$  is not autocorrelated. This allows  
415 for calculating the value of  $D$  for every pixel in the DEM,  
416 for which a drainage path to a channel exists, resulting in  
417 construction of a continuous raster map of dissection  
418 density. Note there is a difference between the continuous  
419 dissection density and the line density used in section 2.2.  
420 Although both measures are pixel-based,  $D$  utilizes local  
421 topography and the estimated size of an average drainage  
422 basin ( $\Lambda$ ), while  $\rho$  ignores topography and is based on an  
423 arbitrary length scale. Thus, whereas the values of  $D$  are  
424 reasonable estimates of actual drainage density, the values  
425 of  $\rho$  are not. In section 2.2 we were forced to use  $\rho$  because  
426 the valleys, as depicted on the Carr map, do not conform to  
427 topography so calculating  $D$  from the Carr map is not  
428 possible.

### 3.2. Results

429 [23] We used the MEGDR together with the corrected  
430 automap of VN to calculate  $L(x, y)$ . The  $D(x, y)$  raster was  
431 calculated using correlation length of about 150 pixels.  
432 Figure 6 shows the map of  $D(x, y)$ , the blue-to-red gradient  
433 depicts values of  $D(x, y)$  from its minimum of  $0.00014 \text{ km}^{-1}$   
434 to its maximum of  $1.08 \text{ km}^{-1}$ . The effective range of  $D$   
435 values is  $0-0.12 \text{ km}^{-1}$ , the few values larger than  $0.12 \text{ km}^{-1}$   
436 are outliers caused by computational artifacts. The mean  
437 value of  $D$  is  $0.062 \text{ km}^{-1}$  and the value of standard  
438 deviation is  $0.041 \text{ km}^{-1}$ . The areas not covered by the color  
439 gradient are places where  $D$  is not defined. Comparing the  
440 map of  $D$  (Figure 6) with the map of VN (Figure 1, bottom),  
441 it is clear that the two maps convey the same information  
442 about the spatial distribution of dissection, but the map of  $D$   
443 does it more effectively by readily identifying locations with  
444 different levels of dissection.  
445

446 [24] According to the map on Figure 6 the distribution of  $D$   
447 is rather homogeneous on large spatial scale; the dissected  
448 region forms a belt roughly located between the equator  
449 and mid southern latitudes. At the same time the distribu-  
450 tion of  $D$  is quite inhomogeneous on smaller spatial scales;  
451 although the mean value of  $D$  is high throughout the  
452 aforementioned belt, there is a strong local variability. The  
453 most prominent region of concentrated areas of high dis-  
454 section density is located in the northern Terra Cimmeria,  
455 around and northwest of the Herschel Crater and extending  
456 up to the dichotomy boundary. Another region of high  
457 dissection density is located in the Margaritifer Terra. These  
458 two regions are located in quadrangles MC22 and MC19;  
459 these are the quadrangles that we automapped for VN [Luo



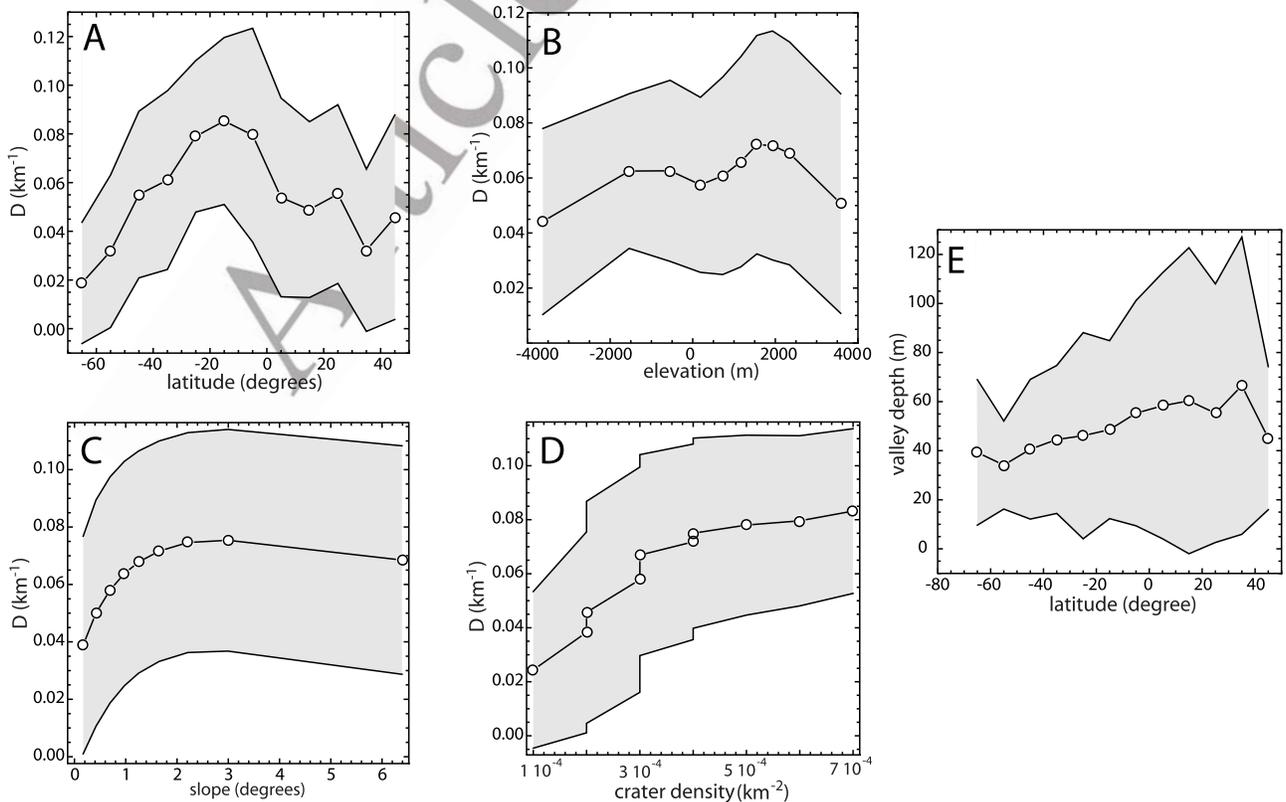
**Figure 6.** Global map of dissection density,  $D$ , on Mars. The blue-to-red gradient corresponds to low-to-high values of  $D$ , and gray indicates areas for which  $D$  is undefined.

460 and Stepinski, 2006; Stepinski et al., 2007] before attempt-  
 461 ing to generate a global map.

462 **3.3. Statistical Analysis**

463 [25] We take advantage of the fact that  $D(x, y)$  is a raster  
 464 of the same dimensions as the MEGDR and calculate zonal  
 465 statistics of dissection density with respect to several vari-  
 466 ables. The zonal statistics is a tool that allows us to calculate  
 467 statistics (such as the mean, standard deviation etc.) of  $D$  for  
 468 a number of predefined spatial zones. A zone is a set of  
 469 pixels having a common attribute.

[26] First, we calculate zonal statistics of  $D$  with respect  
 to the latitude. The latitudes covered by our map ( $50^\circ\text{N}$  to  
 $70^\circ\text{S}$ ) are classified into 12 equidistant zones. For each zone  
 the values of mean and standard deviation of  $D$  are  
 calculated using only pixels located with this zone. The  
 result of this calculation is shown on Figure 7a. The zonal  
 statistics of  $D$  with respect to latitude encapsulates what is  
 already seen on the map of  $D$  (Figure 6); regions located  
 between the equator and  $\sim 40^\circ\text{S}$  have on average highest  
 density of dissection.



**Figure 7.** Zonal statistics of dissection density  $D$  with respect to (a) latitude, (b) elevation, (c) slope, and (d) crater density. (e) Zonal statistics of valley depths with respect to latitude. Circles indicate mean values for each zone. The regions spanning the values of average  $\pm$  standard deviation are shown in gray.

[27] Second, we calculate zonal statistics of  $D$  with respect to the elevation. To derive elevation zones we use only the MEGDR pixels for which  $D$  is defined (only pixels represented by a color in Figure 6). These pixels are classified into 10 zones in such a fashion that each zone contains approximately the same number of pixels (equal area classification). The result of zonal statistics of  $D$  with respect to elevation is shown on Figure 7b. It shows that, on average, highest dissection density occurs in the regions located at elevations between 1000 m and 3000 m; these range of elevations coincides with the prominent belt of high values of  $D$  clearly seen on Figure 6 with a notable exception of the Margaritifer Terra which is located at lower elevations corresponding to a secondary hump in Figure 7b. Similarly, we have calculated zonal statistics of  $D$  with respect to the local slope (Figure 7c). The steepest-descent-based slope values are calculated from the MEGDR only at the pixels for which  $D$  is defined. These pixels are classified into 9 zones using equal area classification. The result of zonal statistics of  $D$  with respect to slope is shown on Figure 7c; on average, steeper locations are more dissected.

[28] Finally, we calculate zonal statistics of  $D$  with respect to crater density. The crater density is derived from the Catalog of Large Martian Impact Craters [Barlow, 1988] using a circular moving window of radius 150 km. A given pixel is assigned a value of crater density corresponding to a ratio of the total count of craters in a circular window to the area of the window. The pixels for which  $D$  is defined are classified into 10 zones using equal area classification. Figure 7d shows the results of zonal statistics of  $D$  with respect to crater density; it shows that dissection density and crater density are positively correlated.

[29] In addition, we have calculated zonal statistics of estimated valley depths with respect to latitude. The valley depths are estimated from DEM using a procedure described in [Howard et al., 2005] and [Barnhart et al., 2009]. Elevation values are collected from within a 3 km radius of a focus pixel identified as belonging to a valley. The 75th percentile elevation is assumed to capture the elevation of surrounding surface; valley depth is estimated as a difference between the elevation of surrounding surface and the elevation at the focus pixel. This procedure is repeated for all pixels delineating VN. Figure 7e shows that, on average, the valleys are shallowest in the southernmost locations and their depth systematically increases northward.

## 4. Discussion

[30] We demonstrated that semiautomapping of VN on a global scale is a feasible technique that offers speed and objectivity. Using our mapping technique we produced a significant update to Carr's global map of VN. All products associated with our map, uncorrected segments, uncorrected netlets, and corrected netlets, as well as the global map of continuous dissection density are available from the authors upon request.

### 4.1. Interpretation of Mapping Results

[31] Our map shows that the dissection is most prominent in a wave-like region (see Figure 6) located in the Noachian surface and extending roughly from the equator to mid southern latitudes. The crest of this "wave" is located at

60°E to 120°E, and the trough of the "wave" is located at 90°W to 150°W. The northern limit of the region of intensive dissection is mostly controlled by the geology; it approximately coincides with the northern extent of the Noachian surface. However, the southern limit of the high dissection region cannot be explained by the global geology and constitutes the first major feature of the global distribution of  $D$  that needs explanation. The large scale of the global map (Figure 6) makes it difficult to observe spatial distribution of  $D$  on smaller length scales. Zooming into the smaller spatial scales reveals a patchy character of dissection; local highs, where the values of  $D$  are as high as  $0.12 \text{ km}^{-1}$ , are interweaved with local lows, where the values of  $D$  are much lower or the surface lacks dissection altogether. This patchy distribution of  $D$  cannot be attributed exclusively to covering by subsequent processes, and appears to be a salient characteristic of an original emplacement of valleys; it constitutes the second major feature of the global distribution of  $D$  that requires explanation.

[32] These two major features of the global distribution of dissection were already noticeable in the Carr map; the major new insight from our updated global map of VN is the value of dissection density that is higher than was reported by Carr. [Carr and Chuang, 1997] found dissection density on Mars (as measured on  $\sim 1 \text{ km}$  length scale) to be an order of magnitude lower than on Earth. The disparity of the values of  $D$  on the two planets provided a major argument against runoff erosion as the leading mechanism of the VN formation. However, the mean value of  $D$ , as calculated from our map, is  $0.062 \text{ km}^{-1}$ . Our preliminary calculation for Earth using the same methodology and topographic data of similar resolution yielded a mean value of  $D$  for terrestrial landmass of  $0.16 \text{ km}^{-1}$  or only 2.6 higher than on Mars. By using topographic data instead of images, and by resorting to (semi) automatic mapping technique, we were able to map significantly more valleys thus reducing the disparity between the values of dissection density (as measured on  $\sim 1 \text{ km}$  length scale) on the two planets. (It is important to recall that terrestrial values of  $D$ , as measured on  $\sim 30 \text{ m}$  length scale are  $1\text{--}100 \text{ km}^{-1}$ , however, small valleys are not expected to survive on Mars, so dissection densities on the two planets need to be compared on a larger spatial scale.) Given the relatively high values of  $D$  over the large portion of Noachian surface, it is now difficult to argue against runoff erosion as the major mechanism of VN formation. This conclusion, arrived on the basis of examining the global pattern of dissection, supplements similar conclusions arrived on the basis of detailed examinations of individual valleys (see section 1).

[33] The zonal statistics (see section 3.3) provides few clues into the origin of VN. Zonal statistics of  $D$  with respect to latitude and elevation basically points to geographical location of VN. The zonal statistics of  $D$  with respect to the local slope shows increased dissection of steeper terrain, a feature consistent with runoff erosion process [Montgomery and Dietrich, 1989]. Statistics of  $D$  with respect to crater density points to potential correlation between cratering and dissection. However, it is more likely that crater density serves as yet another proxy for the geographical location of the highly dissected region and less likely that it indicates any direct casual relation between cratering and dissection. The region of high crater density on Mars [Stepinski and

602 Urbach, 2009] has many of the same geographical limits as  
 603 the region of intensive dissection. Finally, the zonal statis-  
 604 tics of valley depth with respect to latitude shows shallow-  
 605 ing toward the south pole. Thus, not only the number of  
 606 valleys decreases drastically southward of  $\sim 30\text{--}40^\circ\text{S}$ , but  
 607 their depths decrease as well. This is in agreement with the  
 608 results of similar calculations by [Williams and Phillips,  
 609 2001]. This trend could be a primordial feature related to the  
 610 mechanism of valleys emplacement (see discussion below),  
 611 the effect of subsequent surface modification, or the com-  
 612 bination of both. Indeed, the terrain located southward of  
 613  $\sim 30\text{--}40^\circ\text{S}$  is known for its “softened” appearance. This  
 614 attribute of mid-to-high-latitude terrain is connected to  
 615 surface modification due to either mantling [Soderblom *et*  
 616 *al.*, 1973] or viscous relaxation due to presence of ground  
 617 ice [Parmentier and Head, 1981]. Terrain softening may  
 618 result in postformation shallowing of valleys located  
 619 southward of  $\sim 30\text{--}40^\circ\text{S}$ .

620 [34] The full account of implications the global pattern of  
 621 dissection has on the origin of VN is beyond the scope of  
 622 this paper. However, we offer a preliminary discussion of a  
 623 particular scenario that may, conceivably, explain the major  
 624 features of this pattern. This scenario is based on an  
 625 exploratory discussion of climatic controlling factors on  
 626 ancient Mars [Richardson and Soto, 2008]. We focus on  
 627 explaining why the pattern of dissection on Mars is different  
 628 from the terrestrial pattern, even assuming a best case  
 629 scenario of an early Mars that is warm enough to support  
 630 liquid water on its surface. The fundamental reason is the  
 631 existence of the topographic dichotomy on Mars. Given the  
 632 particular global topography of Mars, the water will accu-  
 633 mulate in the topographic lows of the northern plains to  
 634 form an “ocean.” The confinement of a major water  
 635 reservoir to northern plains restricts the transfer of water  
 636 vapor to equatorial regions and the southern hemisphere  
 637 where the VN are found. Global atmospheric circulation  
 638 will carry moisture from northern midlatitudes southward  
 639 resulting in precipitation when the air mass rises orograph-  
 640 ically upon encountering the dichotomy boundary. In fact,  
 641 some of the heavily dissected regions are found where the  
 642 topographic dichotomy is the most pronounced (see Figure 6).  
 643 The supply of moisture to the regions located farther from  
 644 the ocean would be limited. Diffusive atmospheric motions  
 645 will deliver occasional storms to these regions, but in  
 646 quantities much smaller than could be expected from the  
 647 global inventory of water. Thus, in this scenario the equa-  
 648 torial and southern regions of Mars would experience an  
 649 arid, “continental” climate not because of the lack of liquid  
 650 water on the surface of the planet, but because of the  
 651 physical separation of the water reservoir and the landmass.  
 652 Such a scenario explains the relatively low dissection  
 653 density and the existence of the southern limit to the  
 654 presence of valley networks. The southernmost regions,  
 655 located farthest from the water reservoir would get no  
 656 rainfall and would develop no valleys. It also offers an  
 657 explanation to shallowing of VN from (roughly) the location  
 658 of topographic dichotomy, where largest rainfall rates result  
 659 in more dissection, southward, where smaller rainfall rates  
 660 result in less dissection. This hypothesis requires further  
 661 scrutiny; in particular, it needs to be investigated whether  
 662 potential existence of large “lakes” in the Hellas and Argyre  
 663 basins would alter our reasoning. It is unclear how to

explain the patchy character of dissection within a frame- 664  
 work of this scenario. 665

#### 4.2. Automapping Technique: Issues and Future 666 Research 667

[35] Mapping any features by means of computer parsing 668  
 of digital data is a challenging problem because even the 669  
 best algorithm lacks human comprehension of the problem 670  
 and is unable to make a determination on a basis of a 671  
 broader context. Our algorithm, in its present form, cannot 672  
 distinguish between valleys and other landscape features 673  
 having a U-shaped morphology. It is also not designed to 674  
 extrapolate the valleys (as a human mapper would frequently 675  
 do) to the parts of the landscape where topographic data is of 676  
 inferior quality or lacking altogether. In this paper the first 677  
 shortcoming is solved by means of visual inspection of the 678  
 initial, uncorrected map produced by the algorithm. Thus, 679  
 strictly speaking our mapping technique is semiautomated 680  
 because the final map of valley networks requires verifica- 681  
 tion from additional imagery data. Although such a semi- 682  
 automatic technique is faster and more objective than purely 683  
 manual mapping, the ultimate goal is to develop a fully 684  
 automatic technique of mapping VN. Our future research 685  
 will focus on the development of an algorithm, based on a 686  
 concept of machine learning [Witten and Frank, 2005], that 687  
 can automatically eliminate false positives. In machine 688  
 learning the features that are known a priori to be VN are 689  
 used to “train” an algorithm and to build a mathematical 690  
 model of a valley network against which all “candidate 691  
 networks” are tested for either inclusion or elimination. 692  
 This would allow us to do away with manual editing. The 693  
 second shortcoming is not addressed in this paper; the 694  
 valleys are mapped where they are “seen” (by the algorithm) 695  
 in the data and no attempt is made to extrapolate the netlets 696  
 so they form more integrated networks. Although models of 697  
 drainage can be used for such extrapolation (see discussion 698  
 in section 1) it is prudent to deliver a model-free map and to 699  
 leave the modeling to the user of the map in cases it is 700  
 deemed helpful or necessary. 701

[36] Is it “cost effective” to develop a technique for 702  
 automapping of VN? The up-front cost of developing an 703  
 automapping algorithm is high but this expense is recuper- 704  
 ated by a low operating cost of actual mapping. On the other 705  
 hand, manual mapping requires little up-front investment 706  
 but it imposes large operating cost. Because there is no 707  
 evidence that valleys extend to ever smaller spatial scales, 708  
 we don’t expect automapping of VN to offer the same 709  
 “economy of scale” advantage as, for example, the auto- 710  
 mapping of craters which can be used to map “millions” of 711  
 small craters. Nevertheless, we submit that developing a 712  
 means for autoanalysis of planetary data is an important task 713  
 because, in general, purely manual analysis is not efficient 714  
 enough for analyzing the deluge of new data. Our effort to 715  
 automap VN develops techniques that can be used for 716  
 mapping other landscape features, too. In addition, auto- 717  
 mapped VN, while sometimes lacking the integration of 718  
 manually mapped VN, possess some other characteristics 719  
 that are absent from manual maps. For example, automapped 720  
 VN are guaranteed to conform fully to topography, whereas 721  
 manually mapped VN are not. Automapping is also an 722  
 objective and reproducible procedure. Finally, the method 723  
 developed for automapping VN can be applied [Luo and 724

725 *Stepinski, 2008*] to map the drainage on Earth where  
 726 automation can be used to map even the smallest valleys.  
 727 This provides a new functionality for terrestrial geomor-  
 728 phology because conventional methods for mapping dissec-  
 729 tion are model-based and fail in the regions where model  
 730 assumptions are not fulfilled, like, for example, in regions  
 731 where dissection density changes abruptly [*Luo and Stepinski,*  
 732 *2008*].  
 733

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