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Title: Tracking geomorphic signatures of watershed suburbanization with multitemporal LiDAR

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Abstract: Urban development practices redistribute surface materials through filling, grading, and terracing, causing drastic changes to the geomorphic organization of the landscape. Many studies document the hydrologic, biologic, or geomorphic consequences of urbanization using space-for-time comparisons of disparate urban and rural landscapes. However, no previous studies have documented geomorphic changes from development using multiple dates of high-resolution topographic data at the watershed scale. This study utilized a time series of five sequential Light Detection and Ranging (LiDAR) derived digital elevation models (DEMs) to track watershed geomorphic changes within two watersheds throughout development (2002-2008) and across multiple spatial scales (0.01-1 km²). Development-induced changes were compared against an undeveloped forested watershed during the same time period. Changes in elevations, slopes, hypsometry, and surface flow pathways were tracked throughout the development process to assess watershed geomorphic alterations. Results suggest that development produced an increase in sharp topographic breaks between relatively flat surfaces and steep slopes, replacing smoothly varying hillslopes and leading to greater variation in slopes. Examinations of flowpath distributions highlight systematic modifications that favor rapid convergence in unchanneled upland areas. Evidence of channel additions in the form of engineered surface conduits is apparent in comparisons of pre- and post-development stream maps. These results suggest that topographic modification, in addition to impervious surfaces, contributes to altered hydrologic dynamics observed in urban systems. This work highlights important considerations for the use of repeat LiDAR flights in analyzing watershed change through time. Novel methods introduced here may allow improved understanding and targeted mitigation of the processes driving geomorphic changes during development and help guide future research directions for development-based watershed studies.

GEOMORPHOLOGY JOURNAL EDITORIAL REVIEW

All editorial changes and suggestions were accepted and can be seen in the “LiDAR timeseries paper Final 032714 track changes.docx” document.

Highlights (for review)

- Sequential LiDAR DEMs used to track geomorphic changes during urbanization process
- Developing watersheds compared against undeveloped forested reference watershed
- Smoothly varying hillslopes replaced by sharp topographic breaks
- Evidence of channel burial coupled with additions of engineered runoff conduits
- Increased watershed connectivity independent of impervious surface cover and pipes

1 **Tracking geomorphic signatures of watershed suburbanization with multitemporal LiDAR**

2

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26 **Abstract**

27

28 Urban development practices redistribute surface materials through filling, grading, and
29 terracing, causing drastic changes to the geomorphic organization of the landscape. Many studies
30 document the hydrologic, biologic, or geomorphic consequences of urbanization using space-for-
31 time comparisons of disparate urban and rural landscapes. However, no previous studies have
32 documented geomorphic changes from development using multiple dates of high-resolution
33 topographic data at the watershed scale. This study utilized a time series of five sequential Light
34 Detection and Ranging (LiDAR) derived digital elevation models (DEMs) to track watershed
35 geomorphic changes within two watersheds throughout development (2002-2008) and across
36 multiple spatial scales (0.01-1 km²). Development-induced changes were compared against an
37 undeveloped forested watershed during the same time period. Changes in elevations, slopes,
38 hypsometry, and surface flow pathways were tracked throughout the development process to
39 assess watershed geomorphic alterations. Results suggest that development produced an increase
40 in sharp topographic breaks between relatively flat surfaces and steep slopes, replacing smoothly
41 varying hillslopes and leading to greater variation in slopes. Examinations of flowpath
42 distributions highlight systematic modifications that favor rapid convergence in unchanneled
43 upland areas. Evidence of channel additions in the form of engineered surface conduits is
44 apparent in comparisons of pre- and post-development stream maps. These results suggest that
45 topographic modification, in addition to impervious surfaces, contributes to altered hydrologic
46 dynamics observed in urban systems. This work highlights important considerations for the use
47 of repeat LiDAR flights in analyzing watershed change through time. Novel methods introduced
48 here may allow improved understanding and targeted mitigation of the processes driving

49 geomorphic changes during development and help guide future research directions for
50 development-based watershed studies.

51

52 *Keywords:* LiDAR time series; urbanization; land cover change; digital elevation models;
53 anthropogenic geomorphology; watershed

54

55 **1. Introduction**

56

57 At least one-third of the Earth's continental surface has undergone some form of anthropogenic
58 geomorphological activity (Rózsa, 2010). Of these activities, construction and urban
59 development stand out as ongoing and expanding causes of geomorphic change characterized by
60 rapid, complex, and multiscale processes. Throughout urban development, extensive landscape
61 grading, leveling, and terracing can change the fundamental organization of topography within a
62 watershed (Tenenbaum et al., 2006; Csima, 2010). The current ability to track, quantify, and
63 predict development-induced geomorphic changes is hindered by insufficient data and reliance
64 on methodologies developed for traditional geomorphic studies in natural settings (Djokic and
65 Maidment, 1991; Haff, 2003; Gironás et al., 2010; Rózsa, 2010; Szabo, 2010). High-resolution
66 topographic data, such as Light Detection and Ranging (LiDAR), is required to resolve surface
67 flowpaths and visualize roads and other fine-scale features in urban landscapes (Djokic and
68 Maidment, 1991; Tenenbaum et al., 2006; Hunter et al., 2008; Gironás et al., 2010). Sequential
69 topographic data sets are needed to track and quantify the multiscale (Bochis and Pitt, 2005;
70 Tenenbaum et al., 2006; Bochis, 2007) and temporally variable (MDE, 2000; Dietz and Clausen,
71 2008) topographic changes throughout development. No studies have utilized a high-resolution

72 digital elevation model (DEM) time series to track geomorphic changes throughout the
73 development process.

74

75 The topographic footprint of development is visually striking on the ground and from the air,
76 highlighting large-scale landscaping carried out during initial development phases (Csima,
77 2010). Final smoothing and regrading occurs at finer scales that are less immediately apparent in
78 topographic data, but are nonetheless important for understanding surficial drainage and altered
79 geomorphic characteristics (MDE, 2000; Tenenbaum et al., 2006). How best to quantify such
80 topographic changes is unclear in terms of selecting geomorphic variables and spatial scales that
81 most effectively capture observed changes (Haff, 2003; Rózsa, 2010). Topographic variables
82 such as slope, curvature, and their distribution relative to watershed area play important roles in
83 sediment transport and erosion dynamics (Moore et al., 1991; Montgomery and Foufoula-
84 Georgiou, 1993; James et al., 2007), channel formation (Band, 1993; Montgomery and Foufoula-
85 Georgiou, 1993), and surface water – groundwater exchanges (Beven and Kirkby, 1979; Moore
86 et al., 1991; Tenenbaum et al., 2006). However, topographic controls on hydrologic and
87 geomorphic processes may be altered in urban landscapes because of infrastructure (e.g., roads,
88 pipes) (Djokic and Maidment, 1991; Tenenbaum et al., 2006; Gironás et al., 2009; Rózsa, 2010;
89 Choi, 2012).

90

91 Five sequential LiDAR-derived DEMs spanning the development of two small, historically
92 agricultural watersheds were obtained to track geomorphic changes throughout development.
93 The goal of this study is to understand how and where geomorphic changes manifest within these
94 watersheds and to understand their implications for watershed geomorphic and hydrologic

95 processes prior to accounting for influences of stormwater infrastructure. The DEM coverage of
96 a third, forested reference watershed was used to track temporal variation not attributable to
97 development. This work will introduce a number of novel methods for tracking temporal
98 geomorphic changes using sequential LiDAR DEMs and will highlight unique geomorphic
99 signatures of the development process. Results of this study will help guide ongoing research
100 efforts that focus on the coupled influence of topography and infrastructure on watershed
101 function.

102

103 **2. Site description**

104

105 For this study, three watersheds in Montgomery County, Maryland, within the Piedmont
106 Physiographic Province were examined. All three watersheds are within the Mt. Airy Uplands
107 District, characterized by siltstones and quartzite with underlying crystalline bedrock consisting
108 of a phyllite/slate unit, with average annual precipitation of 106.4 cm (Reger and Cleaves, 2008;
109 Hogan et al., ~~2013~~2014). Soper Branch (forested control) drains an area of ~ 3.4 km² with an
110 overall mean gradient of 13%. Land cover is predominantly classified as deciduous forest (~
111 85%) and small percentages of low-density housing and agriculture (6 and 9%, respectively)
112 (Hogan et al., ~~2013~~2014; Dr. J.V. Loperfido, USGS, oral communication, July 2013). Tributaries
113 104 and 109 (T104 and T109, urbanizing) drain areas of ~ 1.2 and 0.9 km² with mean gradients
114 of 11 and 8%, respectively. Tributaries 104 and 109 are historically agricultural watersheds that
115 underwent extensive suburban development from 2003 to the present. Tributary 104 land use
116 shifted from 41% agriculture, 0.3% barren, 42% forest, and 17% urban in 2002 to 2%
117 agriculture, 15% barren, 19% forest, and 64% urban in 2008. Across the same time period, T109

118 changed from 64% agriculture, 1% barren, 25% forest, and 10% urban to 45% agriculture, 13%
119 barren, 25% forest, and 17% urban (Hogan et al., ~~2013~~2014). Both T104 and T109 were used in
120 analyses of development-induced geomorphic changes. However, development in T109 was not
121 complete at the time of the last LiDAR flight, so the final T109 DEM represents a different stage
122 of development than T104.

123

124 Tributaries 104 and 109 fall within the Clarksburg Special Protection Area (CSPA, established in
125 1994) and are subject to development guidelines and restrictions (Maryland DEP, 2013). Broadly
126 speaking, the Special Protection Area designation identifies areas with high quality natural
127 resources and requires ongoing and future development projects to implement best available
128 water quality and quantity protection measures. Water quality and quantity measures often
129 exceed minimum local and state regulations and include extensive Best Management Practices
130 (BMPs) implementation during (sediment and erosion controls; S&EC) and after development
131 (stormwater management; SWM) (Hogan et al., ~~2013~~2014). Each watershed has extensive storm
132 sewer (SS) infrastructure in addition to noted BMP and SWM structures. While SS infrastructure
133 likely plays a key role in dictating watershed hydrology, the purpose of this study is to document
134 geomorphic changes and their hydrologic implications independent of SS influence. Ongoing
135 research will incorporate SS infrastructure to distinguish its effects from geomorphic impacts and
136 better account for watershed hydrologic dynamics.

137

138 **3. Data description and methods**

139

140 Five sequential LiDAR-derived 1-m bare-earth point clouds were collected at semiannual
141 intervals from 2002 to 2008 to track development-induced topographic changes in T104 and
142 T109 and background topographic changes in Soper Branch. Detailed LiDAR vendor,
143 instrumentation, and accuracy information are provided by Jarnagin (2010). Building removal
144 and bare-earth point cloud filtering was performed by each LiDAR vendor utilizing proprietary
145 software, summarized in Table 1. Differences in bare-earth filtering algorithms used by different
146 vendors may have caused interpolated topography within filtered building footprints to vary
147 between dates. Therefore, topographic patterns within building footprints may reflect filtering
148 artifacts and should be interpreted with a degree of uncertainty.

149

150 *Insert Table 1 near here.*

151

152 LiDAR was collected as part of a larger monitoring effort to track watershed changes throughout
153 development and to document development effects on local resources through time (Jarnagin,
154 2008). Detailed comparisons between LiDAR spot elevations and field-surveyed elevations
155 across a range of slope, elevation, and land use classes are reported by Gardina (2008) and
156 Jarnagin (2010). One-meter resolution DEMs were interpolated from bare-earth point clouds
157 using the natural neighbor interpolation algorithm (Sibson, 1981). The DEMs were then
158 coarsened to 2-m horizontal resolution to smooth and reduce noise in the interpolated
159 topography. All subsequent spatial analyses were performed using the interpolated 2-m DEMs
160 in ArcMap 9.3¹ (~~Esri~~ESRI, Redlands, CA).

161

¹ Any use of trade, firm, or product names is for identification purposes only and does not imply endorsement by the U.S. Government.

162 3.1. Elevation change

163

164 Differencing sequential DEMs has been used to estimate sediment budgets, to identify nutrient
165 and sediment sources and sinks (Thoma et al., 2005), and to track gully evolution through time
166 (Perroy et al., 2010). Raw difference calculations, however, do not incorporate any estimate of
167 the error inherent across different DEMs or LiDAR vendors (e.g., they imply that all observed
168 elevation differences are real). Further, LiDAR precision has been shown to vary across land
169 cover classes—with forest cover exhibiting the lowest precision (Gardina, 2008; Jarnagin, 2010).
170 Therefore, comparisons of absolute elevation change estimates across differing land cover
171 classes may be misleading.

172

173 To correct for the inherent variability in elevation estimates through time, DEM differences were
174 expressed as standard deviations (SD) through time calculated for each pixel in each watershed
175 as

$$SD = \sqrt{\frac{1}{N} \sum_{i=year}^N (x_i - \mu)^2} \quad (1)$$

176

177 where N is the number of DEM years (5), x_i is the elevation for a given pixel x in year i , and μ is
178 the mean elevation of pixel x across the five DEM years. The distribution of elevation SDs in
179 Soper Branch was then used to quantify background elevation variability in T104 and T109. The
180 SD distribution in Soper Branch was filtered to exclude all known areas with recent
181 anthropogenic activities or structures to assure the distribution of SDs were indicative of the
182 background signal. Any remaining non-zero SD in Soper Branch was assumed to be attributable

183 to either natural causes or uncertainty in the LiDAR data. Because of the decreased elevation
184 precision observed within forested pixels (e.g., Gardina, 2008; Jarnagin, 2010), the use of Soper
185 Branch to quantify background LiDAR variability was a conservative baseline for detecting
186 change in nonforest areas. To detect elevation changes associated with the development process,
187 resulting SD maps were classified based on the distribution of SDs in Soper Branch, with an
188 additional class added to distinguish SD values outside of the range observed in Soper Branch.
189 Classes represent SD values less than or equal to the mean, within three standard deviations of
190 the mean, and SD values greater than three standard deviations from the mean. Any elevation
191 change above the maximum SD observed in Soper Branch is assumed to be attributable to the
192 development process.

193

194 *3.2. Slope change*

195

196 The distribution of local slopes from before and after development in T104 were quantified and
197 classified to ranges relevant to landscape development decisions described by Csima (2010;
198 Table 2). Each slope class in the pre-development landscape provides an indication of the
199 relative development costs and earth-moving practices used across the watershed. A
200 representative cross section was extracted for pre- (2002) and post- (2008) development years to
201 highlight changes in the juxtaposition of high and low sloping areas and to differentiate spatial
202 patterns associated with modified and unmodified areas. More broadly, changes in local terrain
203 created by reallocating and relocating slopes were mapped using areal standard deviations of
204 slope within a 10 x 10 pixel square neighborhood of every focal pixel x_i . Comparison of maps

205 resulting from pre- and post-development DEMs further highlighted changes in the variability of
206 local landscapes associated with development practices.

207

208 *Insert Table 2 near here.*

209

210 *3.3. Hypsometric curves*

211

212 Hypsometric curves represent the proportion of a watershed's area (a/A) that is above a given
213 elevation (h/H), providing a generalized approximation of hillslope shape within the watershed.

214 Traditionally, hypsometric curves have been used in discussions of landscape evolution across

215 broad spatial extents (e.g., Strahler, 1957). Other studies have found hypsometric curve shape

216 indicative of dominant erosional mechanisms (Harlin, 1980; Luo, 2000), hydrologic response

217 (Luo and Harlin, 2003), and infiltration dynamics (Vivoni et al., 2008). Hypsometric curves were

218 generated for each DEM year based on first-order catchment and full watershed scales to

219 compare and contrast hillslope topographic alterations at increasingly broader spatial extents.

220 Consistent boundaries were used for each year to assure that any hypsometric variability through

221 time was attributable to topographic changes and not to boundary shifts caused by topographic

222 modification along drainage divides. Hypsometric curves for each year were compared to assess

223 hillslope level changes through time. A map of net fill and excavation ($DEM_{2008} - DEM_{2002}$) was

224 also developed to help interpret hypsometric trends.

225

226 *3.4. Drainage network change*

227

228 Topographically based drainage network delineation methods are widely used across a range of
229 land covers and climates. The most common approaches rely on threshold relationships of
230 derivative watershed properties (e.g., drainage area, slope ~~—~~—area relationship) to define
231 upslope channel extents (e.g., O’Callaghan and Mark, 1984; Montgomery and Foufoula-
232 Georgiou, 1993). Other techniques utilize local topography to extract convergent drainage
233 features and stream heads directly (e.g., plan/profile curvature); but have seen much less use ~~due~~
234 ~~to~~because of higher topographic data resolution required for feature detection (Tribe, 1992;
235 Band, 1993; Tarboton and Ames, 2001; Lindsay, 2006). A number of threshold-type approaches
236 have been shown to provide reasonable approximations of field-surveyed channel networks
237 (extent and complexity) (Heine et al., 2004; James et al., 2007; James and Hunt, 2010).
238 However, threshold-based methods are insufficient for determining the location and occurrence
239 of artificial urban conduits (e.g., swales, gutters) whose location is based on design
240 specifications, not physical processes (Gironás et al., 2010; Jankowsky et al., 2012).
241
242 Topographic openness (a measure of tangential curvature; see Yokoyama et al., 2002) was
243 employed to identify surface networks in the pre-development terrain of T104 and then applied
244 to the post-development terrain to compare surface network changes. Topographic openness is
245 an angular relation of horizontal distance to vertical relief, calculated from above
246 (positive/zenith) and below (negative/nadir) a topographic surface (DEM). For an angle ~~less~~
247 ~~than~~≤ 90°, openness is equivalent to the internal angle of an inverted cone, its apex at the focal
248 DEM pixel, constrained by neighboring elevations within a specified radial distance.
249 Topographic openness is more robust in identifying surface convexities and concavities than
250 commonly used profile and plan curvature (Yokoyama et al., 2002) and has been successfully

251 applied to LiDAR to identify convergent topography (Molloy and Stepinski, 2007; Sofia et al.,
252 2011). Topographic openness was calculated using an Arc Macro Language (AML) script
253 (written by M.E. Baker, [UMBC-2005](#), personal communication, [2005](#)) that reproduces the
254 methods detailed in Yokoyama et al. (2002). The AML extracts DEM values for all relevant cells
255 in a neighborhood set for each focal pixel x_i (defined by stepwise increments of one pixel width
256 in each of eight azimuth directions until a specified search radius is reached; here we used 100
257 m) in a landscape. For each increment, the horizontal distance from and elevation above the
258 focal pixel is calculated, and an arctan function converts opposite:adjacent ratios to angular
259 degrees. The AML tracks a running maximum and minimum angle across each radial increment
260 and all smaller increments for all eight directions, converts the resulting maxima and minima to
261 zenith and nadir angles, and computes the mean of each across all eight directions. Thus the
262 final openness values represent the averaged openness measure for all eight azimuths across the
263 specified search radius.

264

265 Difference DEMs and ortho-imagery were used to identify detention basins and other potential
266 barriers to surface flow. Pre- and post-development DEMs were filled to overcome internal
267 drainages (i.e., “pits”) and then subtracted from unfilled DEMs to assess filling extents. Paths
268 were carved from detention basin low points to the next downslope pixel of equal or lesser
269 elevation outside of the filled zone. Where possible, carved pathways followed detention basin
270 outlets identified from areal and ground surveys. No attempt was made to account for flowpaths
271 within internal drainage basins or subsurface drainage infrastructure. Straight-line pathways were
272 enforced for all unknown detention basin outlets. [Flow directions were calculated for each year](#)
273 [from the pit-resolved DEMs by enforcing drainage from each DEM pixel to the adjacent or](#)

274 | ~~diagonal pixel with the greatest drop in relief (see O'Callaghan and Mark, 1984). D8 flow~~
275 | ~~directions were calculated for each year using the pit-resolved DEMs (e.g., O'Callaghan and~~
276 | ~~Mark, 1984). Resulting flow direction surfaces were used to generate drainage networks and~~
277 | ~~other derivative surfaces from which drainage networks and other derivative surfaces were~~
278 | ~~generated.~~

279

280 | To create a stream map, ortho-imagery and hillshaded DEMs were overlain with openness grids
281 | to determine the critical openness for identifying surface depressions in the pre-development
282 | T104 DEM. A negative openness (nadir) angle of ~~91.5° degrees~~ was determined to sufficiently
283 | capture all visible surface depressions. All pixels at or above ~~91.5° degrees~~ were identified, and
284 | then connected using a D8-based accumulation operation to enforce downslope continuity from
285 | each depression (Tarboton and Baker, 2008). The pre-development stream map (as defined by
286 | downslope accumulation of depression pixels) was then tested for agreement with the constant
287 | drop law, which states that the mean elevation drop across channels within a given Strahler order
288 | should not be significantly different from the mean drop in the next higher order (Tarboton et al.,
289 | 1991). Incremental accumulation thresholds of depression pixels were used to prune the pre-
290 | development stream map until it satisfied the constant drop law. The pre-development
291 | accumulation threshold was then applied to the post-development depression accumulation map
292 | to define the post-development stream map. Pre- and post-development stream maps were
293 | overlain and compared to assess network changes. Field visits were also conducted to classify
294 | channels as either natural (defined banks, sorted bed-load, evidence of ongoing or recent flow) or
295 | artificial (e.g., swales, outflow pipes, rip rap).

296

297 **4. Results**

298

299 Hillshades of each DEM year highlight the spatial pattern of surface modifications incrementally
300 throughout development ([Figure-Fig. 1](#)). Initial large-scale resurfacing throughout the watershed
301 is clear in 2004 and 2006 hillshades, with urban or suburban infrastructure distinguishable in the
302 terrain. Fine-scale grading and smoothing that is much less visually apparent in the LiDAR
303 characterize late-stage development in 2006 and 2008. Valley burial is evident in the western
304 portion of the watershed (see red arrow in [Figure-Fig. 1](#)), replaced by a smooth upland housing
305 cluster and roadway. Mainstem valley form remained relatively constant throughout the time
306 series, with little to no visual evidence of lateral channel movement.

307

308 *Insert Figure 1 near here.*

309

310 *4.1. Elevation standard deviations*

311

312 Pixelate elevation standard deviations (SD) calculated across the five DEM years were
313 approximately normally distributed in Soper Branch with a mean of 0.063 m and a standard
314 deviation of 0.034 m, ranging from 0.002-to 0.870 m. Areas with temporal SDs greater than or
315 equal to three standard deviations of the watershed mean (≥ 0.163 m) were focused within
316 riparian zones at or near stream banks and near the basin outlet within a relatively flat and wide
317 floodplain area ([Figure-Fig. 2](#)). Jarnagin (2010) reported similar LiDAR errors within densely
318 vegetated riparian zones in comparisons between ground-truth and LiDAR spot elevations
319 (absolute mean difference of 0.216 m with a standard deviation of 0.723 m and a maximum

320 | difference of 1.180 m). Moderately high SDs (0.063—0.163 m) are also apparent on northeast-
321 | (NE) facing hillslopes. The temporal variation observed on ~~NE-NE~~-facing slopes was greater
322 | than systematic errors reported for ground survey and LiDAR-derived elevations, which
323 | exhibited a mean absolute difference of ~ 0.05 m (Gardina, 2008).

324

325 | Elevation changes in T104 and T109 classified to the distribution of temporal SDs in Soper
326 | Branch highlight large areas exhibiting substantial elevation changes greater than the variability
327 | observed in forested controls. White areas in T104 and T109 signify temporal elevation SD
328 | values that are above the maximum observed in Soper Branch (> 0.870 m), encompassing
329 | ~~approximately ~~~ 24.4% and 9.9% of total watershed area, respectively. Distinct patches of high
330 | elevation variability are primarily located in development parcels evident in the post-
331 | development hillshades (2008 panel in ~~Figure-Fig.~~ 1). High temporal variability (0.163-0.870 m)
332 | is also apparent along the mainstem channel of T104, but is less evident in T109. Relatively low
333 | temporal SD values in the southern undeveloped portion of T109 and within riparian zones in
334 | T104 reflect magnitudes of temporal variability similar to those observed in Soper Branch. Such
335 | heterogeneous distribution of elevation change is consistent with the observation that earth-
336 | moving practices were not spatially uniform across developing parcels.

337

338 | *Insert Figure 2 here.*

339

340 | *4.2. Slopes*

341

342 The footprint of development is clearly distinguishable from unmodified hillslopes in a visual
343 comparison of pre- and post-development slope maps (~~Figure Fig.~~ 3). Development created
344 abrupt slope changes in the landscape, replacing relatively smooth topographic profiles (i.e., gray
345 boxes in inset). Moderate to high slope classes ($> 12\%$) were present along roads, housing
346 parcels, and stormwater management features, despite the use of bare-earth LiDAR with
347 buildings removed by filtering. Valley infill is apparent in the slope transect between points *a*
348 and *b* at ~ 180 m with no evidence of the original valley structure in the post-development
349 topography. High-frequency variation between low slope and high slope features is evident in the
350 slope transect between 420 and 580 m across a developed subdivision.

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351

352 *Insert Figure 3 here.*

353

354 Areal slope standard deviations calculated using a 10 x 10 pixel moving window show a marked
355 increase in the spatial variability of local slope values in post-development topography (~~Figure~~
356 ~~Fig.~~ 4). The overall distribution of variation in local slopes exhibited a positive shift from the
357 pre- (mean areal slope standard deviation of $2.35 \pm 2.26\%$) to post-development terrain (mean
358 areal slope standard deviation of $4.69 \pm 3.54\%$). Highly variable slope zones in 2002 were
359 limited to riparian corridors and pre-existing development plots in the northern section of T104.
360 By contrast, post-development slopes exhibit greater local variation (i.e., ~~both~~-steeper and flatter
361 terrain in close proximity) associated with the grading of subdivisions and road embankments.

362

363 *Insert Figure 4 here.*

364

365 4.3. Hypsometry

366

367 Despite substantial topographic changes across the watershed visible in hillshades and slope
368 maps representing different stages of development, hypsometric changes ~~due to~~towing to
369 development only manifest when observed across small extents with relatively uniform, high
370 magnitude surface changes (~~Figure-Fig.~~Fig. 5). Distinctly terraced hypsometric curves appear in the
371 end-development phase topography, mirroring abrupt slope changes along roadways and
372 surrounding housing parcels (see insets A and C in ~~Figure-Fig.~~Fig. 5). Valley infill in subwatershed
373 A (see red arrow in ~~Figure-Fig.~~Fig. 1) created a sharp elevation gradient to the remaining riparian
374 lowlands, reflected in the 2007 and 2008 hypsometric curves. Terracing and leveling of upslope
375 transportation corridors also contribute to the terraced hypsometric form. Hypsometric trends of
376 the larger subwatershed B; and at the full watershed scale (D) did not exhibit detectable temporal
377 changes.

378

379 *Insert Figure 5 here.*

380

381 4.4. Drainage network changes

382

383 Substantial redistribution of overland flowpaths is evident in comparisons of pre- and post-
384 development flow direction grids (~~Figure-Fig.~~Fig. 6). As with slope comparisons, the most drastic
385 changes in flow directions are focused along roadways and within housing parcels. Small patches
386 of uniform flow directions mirroring housing and road footprints have replaced large, contiguous
387 patches of homogenous flow directions (i.e., parallel flow lines) on upland hillslopes. Fine-scale

388 variation in flow directions within relatively minor elevation change zones (see elevation change
389 inset in [Figure-Fig. 6](#)) are coupled with more dramatic modifications within drainage features
390 and near infrastructure. As a result of these alterations to the flow field, flow accumulation
391 distributions exhibited a negative shift, with accumulation quartiles decreasing from 4, 11, and
392 29 pixels in 2002 to 1, 4, and 12 pixels in 2008. Distributional shifts reflect drainage dissection
393 and the fragmented flow field apparent in [Figure-Fig. 6](#) with small, locally convergent pathways
394 replacing large, contiguous, parallel flowpaths.

395

396 *Insert figure 6 here.*

397

398 A depression pixel (openness $\geq 91.5^\circ$) accumulation threshold of three pixels was found to
399 satisfy the constant drop criterion in the pre-development terrain of T104. The three-pixel
400 threshold was then applied to the post-development terrain to evaluate surface network changes.
401 Comparisons of pre- and post-development drainage network structure in T104 indicate a net
402 gain in stream length of 4.25 km (~ 52% increase) throughout the watershed ([Figure-Fig. 7](#)).
403 Additions were common along transportation infrastructure as swales or gutters typical in this
404 watershed. Upslope extensions of existing channels also occurred extending beyond pre-existing
405 piped infrastructure at the apparent head of pre-development drainage lines ([inset A in Figure](#)
406 [Fig. 7A](#)). Valley infill ([see subwatershed A in Figure-see Fig. 6A](#)) caused the loss of a tributary;
407 but was replaced by an artificial conduit that paralleled a nearby road ([inset B in Figure-Fig. 7B](#)).
408 Similar conduits are apparent on the eastern side of the watershed as well ([inset C in Figure-Fig.](#)
409 [7C](#)). No substantial changes were apparent in the mainstem channel as suggested by the near 1:1
410 overlap in pre- and post-development networks.

411

412 *Insert Figure 7 here.*

413

414 **5. Discussion**

415

416 The growing footprint of development in T104 is clearly visible in the DEM hillshade time
417 series, with distinct early- and late-stage development phases operating at unique spatial scales
418 (~~Figure-Fig.~~ 1). Large-scale cutting and filling to support major infrastructure characterizes early
419 development, evident in the 2004 and 2006 hillshades. In contrast, fine-scale leveling and
420 grading of housing parcels characterize late-stage development (MDE, 2000). Fine-scale ~~earth~~
421 ~~movingearth-moving~~ practices are less immediately apparent in hillshades, but are detectable in
422 derivative data_sets, as this study has shown. Results show that geomorphic changes associated
423 with urban development are significant and non-uniform, and could have implications for
424 management and conservation efforts.

425

426 Initial analysis of background elevation changes in Soper Branch revealed non-zero standard
427 deviations focused on northeast- (NE) facing hillslopes, and within riparian and floodplain areas
428 (~~Figure-Fig.~~ 2). Increased variation on NE hillslopes may reflect flight paths taken during LiDAR
429 collection, or possible aspect differences in vegetation and soils. Slightly lower systematic errors
430 reported in Gardina (2008) were attributed in part to methods used to derive bare-earth point
431 clouds, and in part, to LiDAR system error. ~~It is likely that~~ The NE artifact likely is a
432 combination of vegetation and LiDAR system errors. Riparian and floodplain variability in
433 Soper Branch can be attributed to LiDAR errors reported by Jarnagin (2010). In comparisons

434 | between LiDAR elevations and ~~field-field~~-surveyed spot elevations, Jarnagin observed greater
435 | LiDAR error in densely vegetated riparian zones and on steep slopes common around incised
436 | channels. Distinguishing temporal variability attributable to artifacts in the LiDAR time series
437 | was necessary before reaching any quantitative conclusions about topographic changes resulting
438 | from development.

439

440 | Elevation standard deviations classified to distinguish background and development-induced
441 | changes clearly show that earth moving in T104 and T109 was nonetheless substantially greater
442 | than background geomorphic changes observed in Soper Branch. Magnitudes similar to temporal
443 | standard deviations found in Soper Branch were observed in T104 across pre-existing housing
444 | parcels and riparian zones left untouched to comply with riparian protection policies. Somewhat
445 | higher variability observed along the mainstem channel in T104 may indicate subtle channel
446 | shifts and bank erosion throughout the time period despite attempts to insulate the channel from
447 | the hydrologic and geomorphic effects of development. ~~It is also possible that~~ some of the
448 | observed elevation changes may possibly reflect artifacts produced by removing buildings from
449 | the raw LiDAR point clouds. Because of the proprietary nature of the LiDAR processing, ~~it was~~
450 | ~~not possible to evaluate~~ the bare-earth detection algorithms used by the data providers was
451 | not possible for this study.

452

453 | Surface changes throughout development in T104 and T109 altered slopes to support
454 | infrastructure, enforce drainage, and promote infiltration. Comparison of pre- and post-
455 | development slope maps in T104 shows a marked increase in high slope classes with
456 | development as well as reallocation across space (~~Figure Fig. 3~~). High slope classes focused

457 along transportation corridors and surrounding detention basins are likely to promote surface
458 drainage. Slopes leading toward streams in the pre-development map appear to steepen with
459 development as well, likely reflecting upland filling and terracing. Stream burial is a frequently
460 cited phenomenon in urban watersheds (Leopold et al., 2005; Elmore and Kaushal, 2008; Roy et
461 al., 2009; Doyle and Bernhardt, 2011) and is apparent on the western side of T104 (~180 m in
462 transect inset of ~~Figure-Fig.~~ 3). However, ~~it is unclear~~ without infrastructure design documents it
463 is difficult to determine whether the stream is piped under its original flowpath, or if it has been
464 redirected to flow alongside the nearby ~~main~~ road running north to south.

465
466 Regular shifts between high and low slopes within subdivisions likely reflect housing parcels
467 (low slopes) separated by graded lawns and swales (higher slopes) to promote drainage (~~Figure~~
468 ~~Fig.~~ 4). Low slopes within building footprints reflect the building removal algorithm applied to
469 LiDAR point clouds to extract bare-earth points used in this study to create DEMs. The spatial
470 distribution of high and low slopes in part controls the spatial distribution of runoff generation
471 and infiltration zones (Beven and Kirkby, 1979; Tenenbaum et al., 2006). Shifts in the spatial
472 pattern and magnitude of slopes may contribute to altered watershed processes including runoff
473 generation, nutrient and sediment transport, and surface ~~water-groundwater-water – groundwater~~
474 exchanges.

475
476 Hypsometric trends within small subwatersheds of T104 appear to mirror aforementioned
477 steepening of valley walls with development (~~Figure-Fig.~~ 5). Upland filling and grading further
478 exacerbates the disparity in elevations between the ~~now-now~~-developed uplands and the riparian
479 lowlands. Redistribution of flowpaths at and around valley infills likely altered surface and

480 subsurface water exchanges, thus changing the functional connectivity of upland and lowland
481 areas. Dynamics are further complicated by stormwater management infrastructure, such as
482 detention ponds and infiltration trenches, typically located between upland (development plots)
483 and lowland (riparian zone) areas to capture and treat overland runoff before it enters local
484 streams (Montgomery County Planning Department, 1994). Thus, riparian areas may no longer
485 be functionally connected to upland areas during rain events unless stormwater management
486 outfalls are triggered. Further research is required to investigate the nature of the altered ~~surface-~~
487 ~~groundwater-surface - groundwater~~ connections. Lack of temporal trends in the larger
488 subwatershed topography or throughout the entire watershed result from spatial averaging, and
489 may indicate a detection limit for tracking parcel or block-level topographic change with
490 hypsometry.

491

492 Development caused a substantial reconfiguration of overland flowpaths from large contiguous
493 areas of uniform directions to small, spatially fragmented patches (~~Figure-Fig.~~ 6). The non-
494 random distribution of smaller patches mirrors housing footprints, transportation corridors, and
495 stormwater management infrastructure, and reflects landscape-engineering practices that
496 promote efficient collection and drainage of overland runoff (~~Figure-Fig.~~ 7). Adjacent pixels with
497 uniform directions indicate parallel flowpaths, which will not converge until a pixel with a
498 differing direction is encountered downslope (O'Callaghan and Mark, 1984). By introducing
499 heterogeneity into the flow direction surface, development has created both more dissected
500 subbasins and increased opportunities for small flowpaths to converge. Increased convergence in
501 upland areas causes overland flow to reach moderately low accumulations more often and very
502 high accumulations more rarely than in the pre-development terrain. Decreased median and ~~first~~

503 | first-quartile values between 2002 and 2008 reflect the dominance of low accumulation pathways
504 | with development. Such alterations are likely the cumulative result of many site-specific efforts
505 | to increase local water conveyance, but their cumulative effect should also have implications for
506 | patterns of soil moisture, variable source area, erosion, infiltration, and recharge (Beven and
507 | Kirkby, 1979; Moore et al., 1991; Tenenbaum et al., 2006). To our knowledge, a geomorphic
508 | driver of increased upland convergence in developed terrain has yet to be described in the
509 | literature. Typically, pipes and impervious surfaces are thought to be the causal agents of altered
510 | runoff, nutrient, and sediment dynamics in urban watersheds (e.g., Paul and Meyer, 2001; Dietz
511 | and Clausen, 2008; Roy et al., 2008). However, our results indicate that the geomorphic
512 | modifications that occur with development are substantial and could also contribute to altered
513 | watershed functions.

514

515 | Comparisons of pre- and post-development drainage networks reveal channel additions through
516 | engineered surface conduits (e.g., swales, culverts); and losses ~~due to~~caused by valley infill or
517 | subsurface routing (~~Figure~~Fig. 7). Despite noted variability in elevations along stream banks,
518 | only minor shifts were apparent in the mainstem of T104. Results of this study also raise
519 | questions about what constitutes a channel in an urban landscape. Traditional channel
520 | definitions, characterized by defined banks and sorted bedloads (Montgomery and Dietrich,
521 | 1989; Heine et al., 2004) exclude engineered surface conduits that become active during
522 | precipitation events (Doyle and Bernhardt, 2011). Excluding these artificial conduits and
523 | subsurface stormwater pipes may under-represent the surficial connectivity of urban watersheds.
524 | Without infrastructure design documents, the true drainage connectivity of the watershed is
525 | unclear. Ongoing work will explore methods for incorporating subsurface drainage features into

526 T104 network representations (e.g., Gironás et al., 2009; Choi, 2012; Jankowsky et al., 2012)
527 and determine how best to incorporate BMP infrastructure into predicted network
528 representations. Nevertheless, this study indicates an overall increase in surface connectivity
529 with development independent of changes in impervious surface cover or subsurface stormwater
530 infrastructure. Increased connectivity would result in faster and larger storm peaks if left
531 unchecked by stormwater management infrastructure. Therefore, considering geomorphic
532 changes in addition to impervious surface cover impacts is important for remediation efforts in
533 urban areas.

534

535 **6. Conclusions**

536

537 This study has shown the utility of sequential LiDAR-derived DEMs for tracking and
538 quantifying the geomorphic changes associated with development. Using a forested watershed as
539 a reference, background topographic variability was quantified and differentiated from
540 development-induced topographic change. Utilizing first- and second-order topographic
541 derivatives, this study demonstrated that development generates increasingly disjointed hillslopes
542 characterized by abrupt slope changes separating flat upland areas. Temporal variation in
543 elevations, slopes, and hypsometric curves likely contributes to altered watershed functions, but
544 further research is required to understand their importance relative to stormwater infrastructure.
545 Substantial modification of overland flowpaths resulted in subtle changes to delineated network
546 structure, highlighting additions and losses to the pre-development network. This study's ability
547 to detect natural and manmade ephemeral surface conduits provides a more complete accounting
548 of watershed hydrologic connectivity. Better accounting of upslope ephemeral channel

549 extensions may help in designating areas for protection in future development projects. The
550 geomorphic signal of increased landscape convergence and drainage density indicates more rapid
551 runoff generation and conveyance independent of impervious surface cover or storm sewers. A
552 better understanding of how and where geomorphic changes occur throughout development may
553 enable a better understanding of pollutant mobilization and retention processes. Ongoing work
554 seeks to incorporate known subsurface pipe and BMP structures into surface network
555 representations to understand watershed hydrologic response dynamics.

556

557 Results presented here raise important considerations for temporal topographic studies. Standard
558 topographic methods developed for use in larger, mostly non-urban watersheds (e.g., DEM
559 differencing, hypsometry) have limited applicability for tracking urbanization unless summarized
560 across a relatively small extent. Additional research is needed to gain a better understanding and
561 [to](#) develop novel methods for tracking and quantifying urban topographic modifications.

562

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564

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576

577 **8. References**

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739

740 **9. List of Figures and Tables**

741

742 Fig. 1. Sequential light detection and ranging (LiDAR) derived digital elevation models (DEMs)
743 spanning the development process in T104 show the topographic footprint of development
744 increasing through time. Boxes highlight topographic differences between early- and late-stage
745 development practices. The red arrow denotes a valley that was buried during development.

746

747 Fig. 2. Elevation change through time quantified as temporal standard deviations (SD) calculated
748 for each DEM pixel. Map values are classified to the distribution of elevation variation observed
749 in Soper Branch (forested control). Black and dark gray areas represent values below and within
750 three standard deviations of the areal mean in Soper Branch, respectively. Light gray areas

751 represent temporal variation in elevation greater than three standard deviations above the areal
752 mean in Soper Branch; whereas white areas are greater than the maximum, and both indicate
753 areas of substantial development.

754

755 Fig. 3. Comparisons of pre- (2002) and post- (2008) development slope distributions in T104
756 classified to ranges relevant to development practices (see Table 2 after Csimá, 2010). Inset
757 shows a cross-sectional profile of slope along the dashed red line between points *a* and *c*.
758 Substantial differences in the variability of local slopes are evident in the profile between
759 developed (shaded zone) and undeveloped (unshaded) parcels. High slopes ring roads, swales,
760 and detention features and manifest as high magnitude variation across the lateral transect.

761

762 Fig. 4. Areal slope standard deviations calculated using a 10x10 pixel moving window for pre-
763 and post-development T104 slope maps. As in pixelate slope maps, highly variable slope zones
764 are concentrated along road corridors and around detention ponds. Nevertheless, a substantial
765 increase in the local variability of slope is apparent in 2008, with high standard deviation values
766 widely distributed across the watershed.

767

768 Fig. 5. Hypsometric variability through time exhibits distinctly different patterns across first-
769 order subwatersheds in T104 (A, B, and C). Hypsometry of relatively small subwatersheds that
770 underwent substantial resurfacing (A ~ 0.09 km² and C ~ 0.03 km²) shifted from smooth (black
771 curves) to terraced curve forms (gray curves) reflecting grading of developed parcels. However,
772 extensive resurfacing was not apparent in hypsometric trends of larger subwatersheds (B ~ 0.43
773 km²) or at the full watershed scale (D ~ 1.18 km²).

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Fig. 6. Extensive topographic modification (e.g., C) associated with grading throughout development (2002-2008) causes substantial variation in overland flowpath directions and elevation change (D) in T104. Comparisons between pre- and post-development years show that fine-scale flow direction changes at the hillslope scale are coupled with the designed orientation and gradients of road networks (A and B, with each color representing a unique direction of overland flow). Large patches of a single color indicate parallel flow lines down a hillslope, whereas fragmentation of the 2002 map in 2008 necessarily indicates redirection and convergence of previously unconnected contributing areas.

Fig. 7. Comparison of pre- (blue lines) and post- (red dashed lines) development channel network delineations suggest increased network density with development in T104. Added channels include upslope extensions beyond existing culverts (A) and vegetated swales parallel to roadways (C). Comparisons to pre-development topography show a minor tributary that has been buried during development (B).

Table 1-
 Building removal and bare-earth point cloud processing information for each LiDAR year provided by S.T. Jarnagin (2013, personal communication, 2013).^{4a}

LiDAR Year	Instrument	Vendor	Building Removal Software	Bare-earth Filtering
2002	Optech ALTM-2025	Airborne 1	Optech's REALM 2.27	Terrascan (running on Microstation)
2004	Optech ALTM-2033	Laser Mapping Specialists Inc.	Applanix POSPROC & Optech's REALM (versions not specified)	Spectra's Terramodel
2006	Optech ALTM-3100	Canaan Valley Institute	Optech's REALM (version not specified)	Microstation 8 with Terrascan
2007	Optech ALTM-3100	Canaan Valley Institute	Optech's REALM (version not specified)	Microstation 8 with Terrascan
2008	Leica ALS-50	Sanborn	Applanix POSPROC 4.3	Leica ALS post-processing and Terrasolid Terrascan

790 ⁴Any use of trade, firm, or product names is for identification purposes only and does not
 791 imply endorsement by the U.S. Government.

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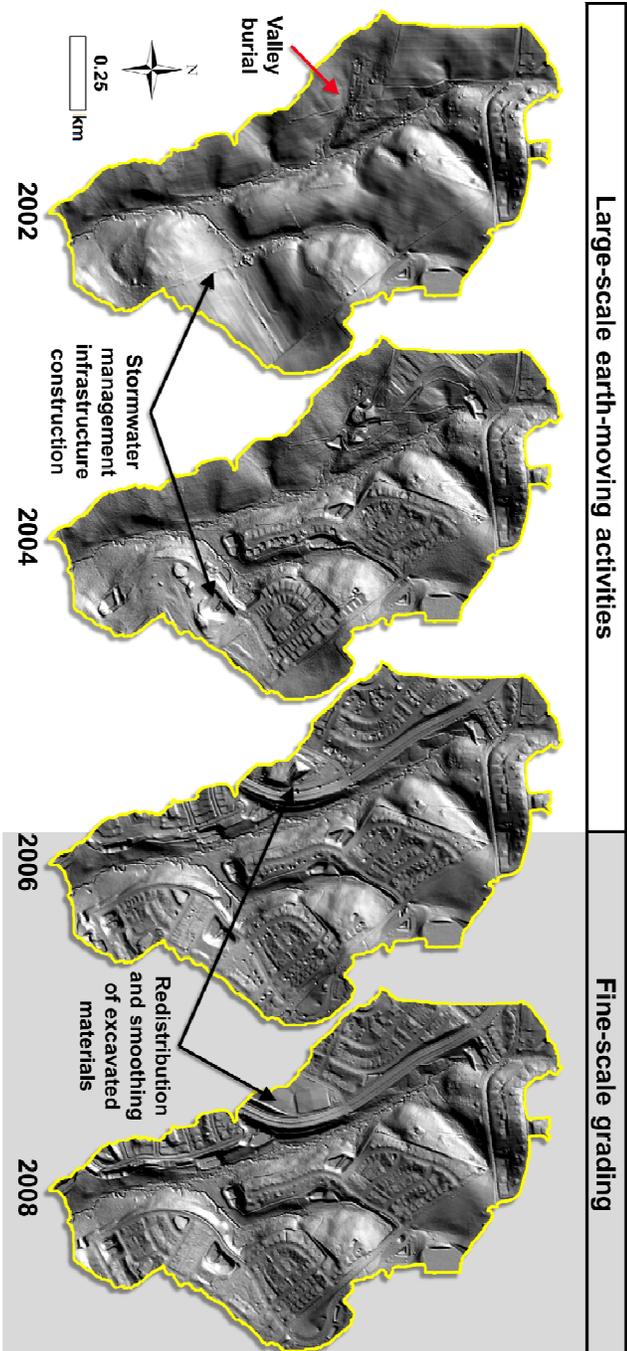
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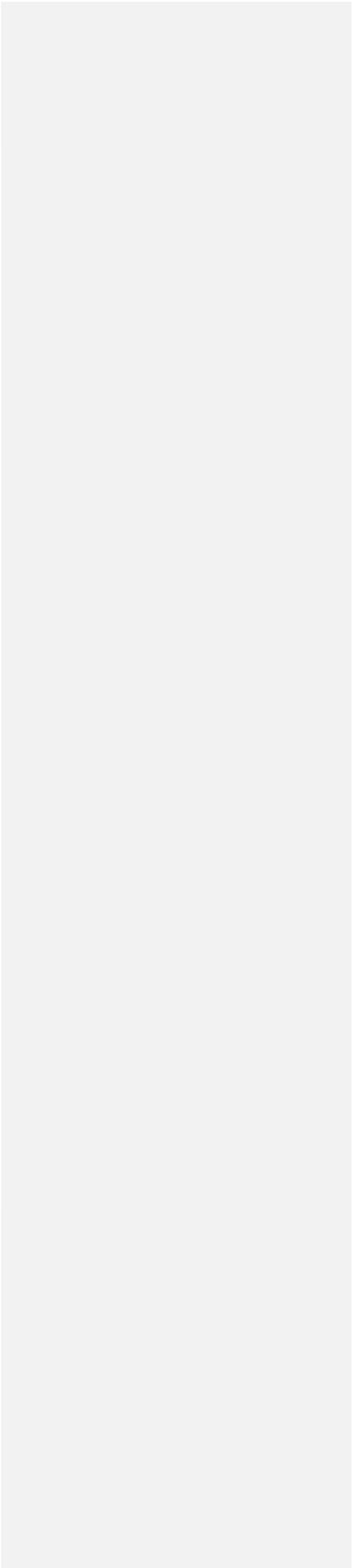
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Table 2-
Slope classes and associated economic costs and considerations for urban development (adapted from Csimá, 2010)-

Angle of slope	Development potential and the required landscaping
Up to 5%	Easy and economic development potential. In general, terracing not necessary; landscaping exclusively restricts to drainage. Relief does not pose a limit either to build-up density or building size.
5-12%	Increased development costs. Landscaping is inevitable; development is only possible with terracing and slope leveling. Somewhat limited development.
12-25%	Development potential at significant cost and labor expenditure, only with terraces and supporting walls provided. Major topographic transformation; relief fundamentally controls the type of development to be applied.
25-35%	Limited potential for urban development. Low building density with small-sized buildings.
Above 35%	Unsuitable for urban development.

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Soper Branch

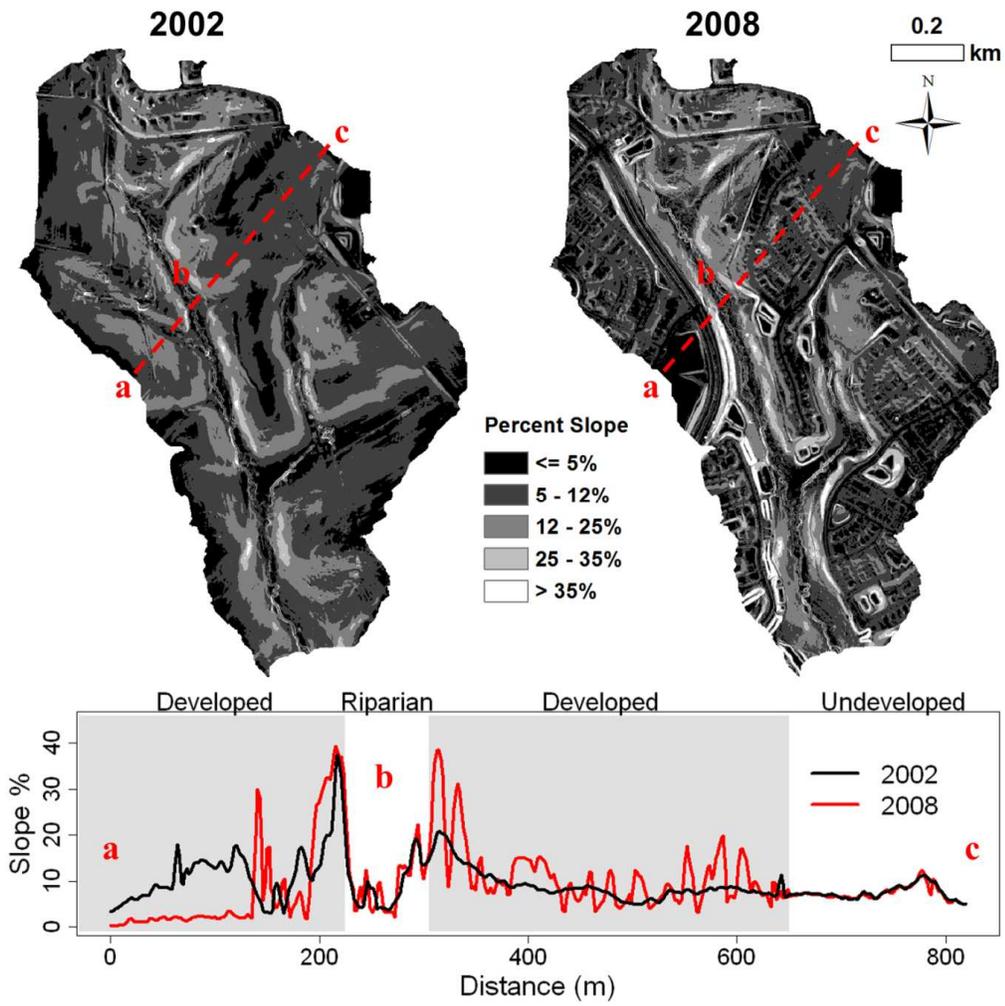


Tributary 104

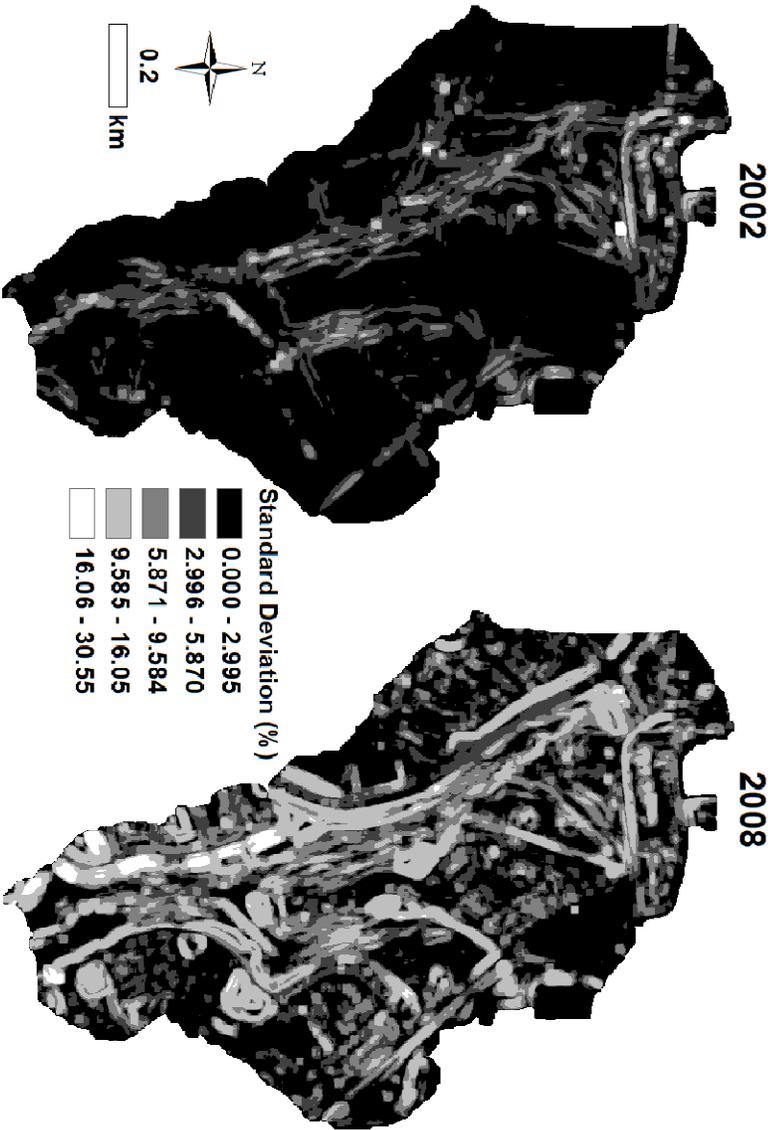


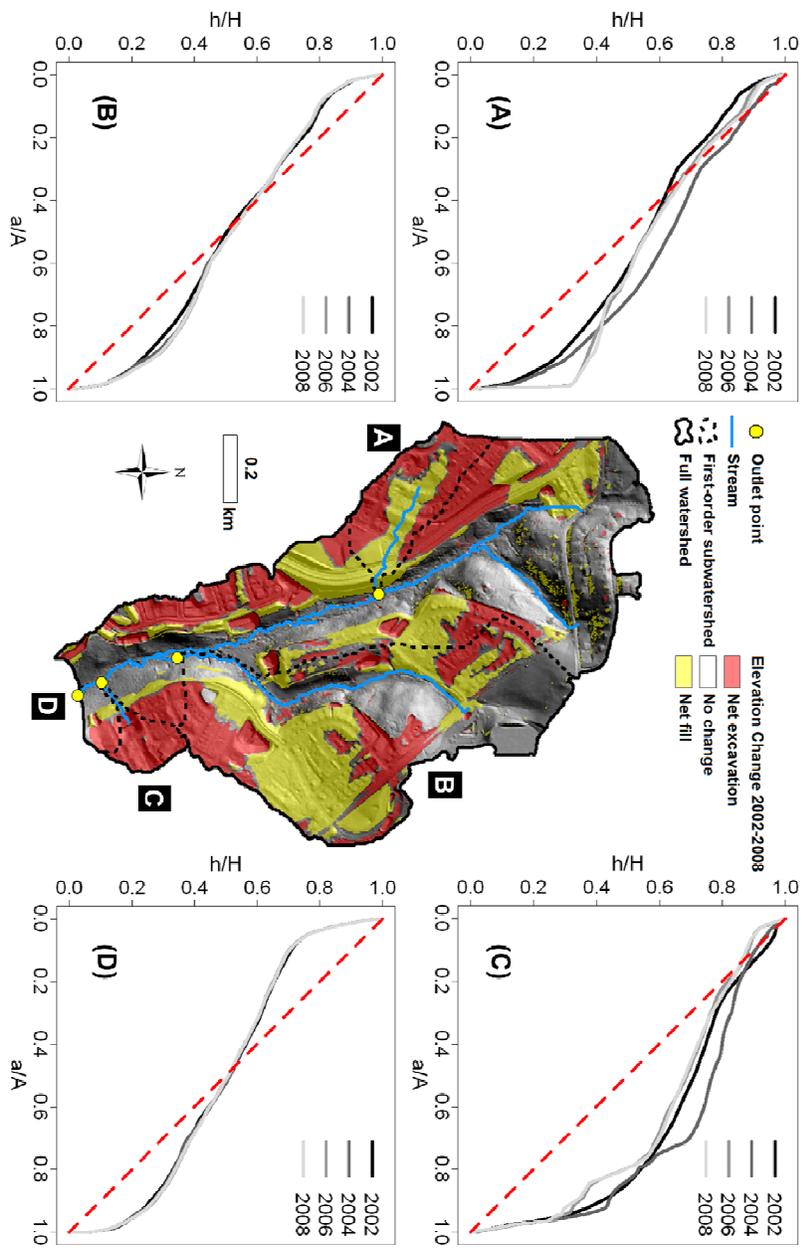
Tributary 109

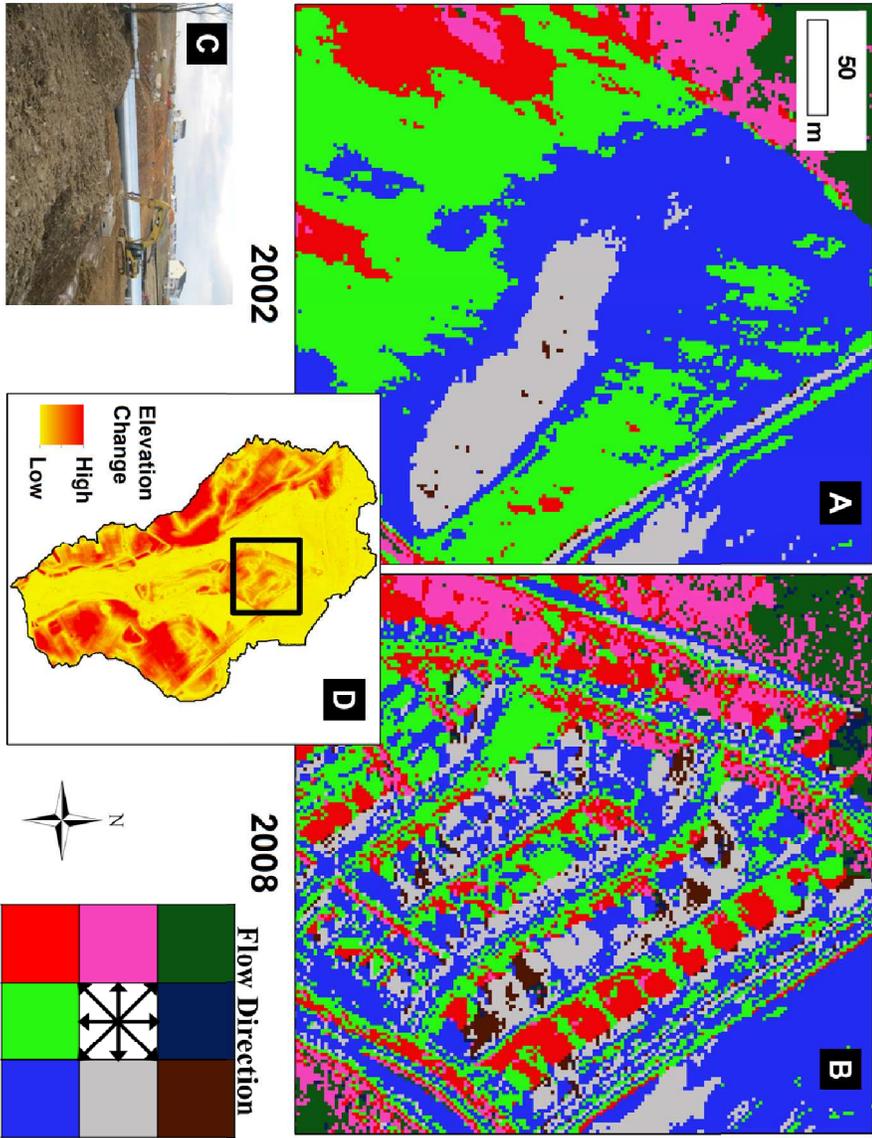


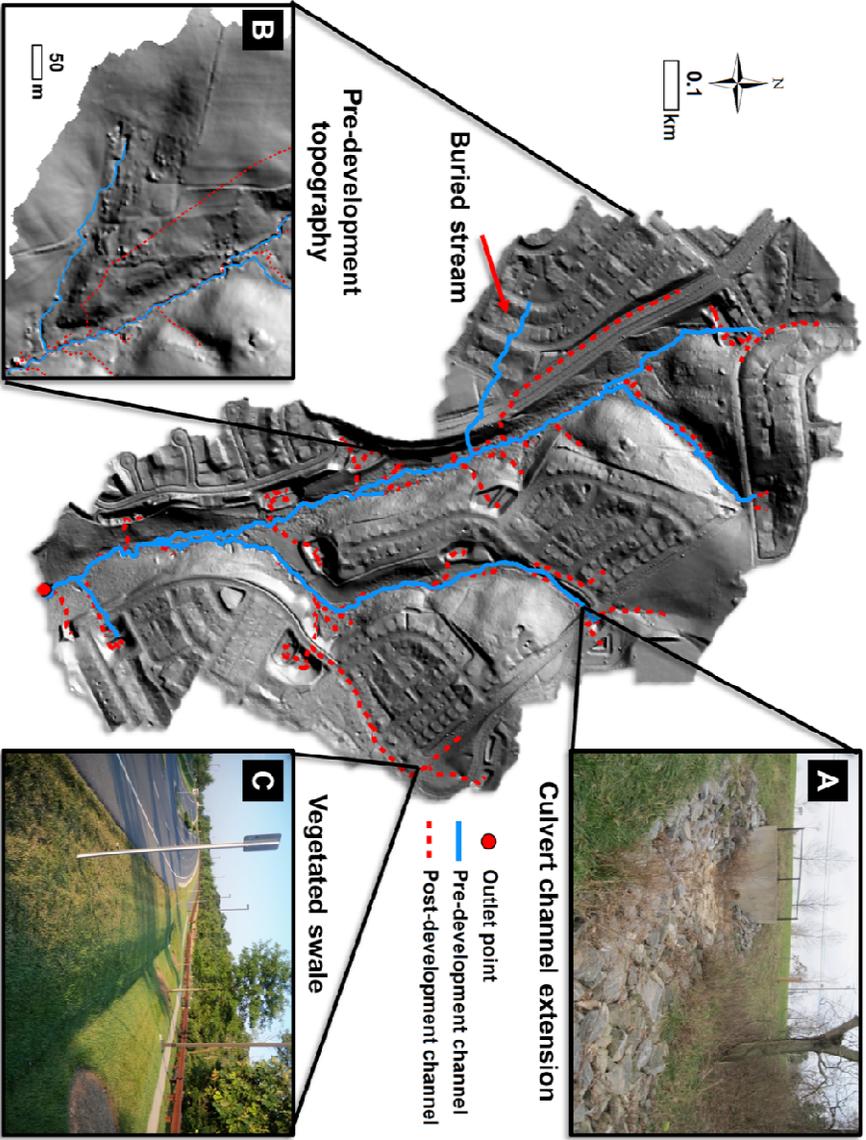


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1 **Tracking geomorphic signatures of watershed suburbanization with multitemporal LiDAR**

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25

26 **Abstract**

27

28 Urban development practices redistribute surface materials through filling, grading, and
29 terracing, causing drastic changes to the geomorphic organization of the landscape. Many studies
30 document the hydrologic, biologic, or geomorphic consequences of urbanization using space-for-
31 time comparisons of disparate urban and rural landscapes. However, no previous studies have
32 documented geomorphic changes from development using multiple dates of high-resolution
33 topographic data at the watershed scale. This study utilized a time series of five sequential Light
34 Detection and Ranging (LiDAR) derived digital elevation models (DEMs) to track watershed
35 geomorphic changes within two watersheds throughout development (2002-2008) and across
36 multiple spatial scales (0.01-1 km²). Development-induced changes were compared against an
37 undeveloped forested watershed during the same time period. Changes in elevations, slopes,
38 hypsometry, and surface flow pathways were tracked throughout the development process to
39 assess watershed geomorphic alterations. Results suggest that development produced an increase
40 in sharp topographic breaks between relatively flat surfaces and steep slopes, replacing smoothly
41 varying hillslopes and leading to greater variation in slopes. Examinations of flowpath
42 distributions highlight systematic modifications that favor rapid convergence in unchanneled
43 upland areas. Evidence of channel additions in the form of engineered surface conduits is
44 apparent in comparisons of pre- and post-development stream maps. These results suggest that
45 topographic modification, in addition to impervious surfaces, contributes to altered hydrologic
46 dynamics observed in urban systems. This work highlights important considerations for the use
47 of repeat LiDAR flights in analyzing watershed change through time. Novel methods introduced
48 here may allow improved understanding and targeted mitigation of the processes driving

49 geomorphic changes during development and help guide future research directions for
50 development-based watershed studies.

51

52 *Keywords:* LiDAR time series; urbanization; land cover change; digital elevation models;
53 anthropogenic geomorphology; watershed

54

55 **1. Introduction**

56

57 At least one-third of the Earth's continental surface has undergone some form of anthropogenic
58 geomorphological activity (Rózsa, 2010). Of these activities, construction and urban
59 development stand out as ongoing and expanding causes of geomorphic change characterized by
60 rapid, complex, and multiscalar processes. Throughout urban development, extensive landscape
61 grading, leveling, and terracing can change the fundamental organization of topography within a
62 watershed (Tenenbaum et al., 2006; Csima, 2010). The current ability to track, quantify, and
63 predict development-induced geomorphic changes is hindered by insufficient data and reliance
64 on methodologies developed for traditional geomorphic studies in natural settings (Djokic and
65 Maidment, 1991; Haff, 2003; Gironás et al., 2010; Rózsa, 2010; Szabo, 2010). High-resolution
66 topographic data, such as Light Detection and Ranging (LiDAR), is required to resolve surface
67 flowpaths and visualize roads and other fine-scale features in urban landscapes (Djokic and
68 Maidment, 1991; Tenenbaum et al., 2006; Hunter et al., 2008; Gironás et al., 2010). Sequential
69 topographic data sets are needed to track and quantify the multiscalar (Bochis and Pitt, 2005;
70 Tenenbaum et al., 2006; Bochis, 2007) and temporally variable (MDE, 2000; Dietz and Clausen,
71 2008) topographic changes throughout development. No studies have utilized a high-resolution

72 digital elevation model (DEM) time series to track geomorphic changes throughout the
73 development process.

74

75 The topographic footprint of development is visually striking on the ground and from the air,
76 highlighting large-scale landscaping carried out during initial development phases (Csima,
77 2010). Final smoothing and regrading occurs at finer scales that are less immediately apparent in
78 topographic data, but are nonetheless important for understanding surficial drainage and altered
79 geomorphic characteristics (MDE, 2000; Tenenbaum et al., 2006). How best to quantify such
80 topographic changes is unclear in terms of selecting geomorphic variables and spatial scales that
81 most effectively capture observed changes (Haff, 2003; Rózsa, 2010). Topographic variables
82 such as slope, curvature, and their distribution relative to watershed area play important roles in
83 sediment transport and erosion dynamics (Moore et al., 1991; Montgomery and Foufoula-
84 Georgiou, 1993; James et al., 2007), channel formation (Band, 1993; Montgomery and Foufoula-
85 Georgiou, 1993), and surface water – groundwater exchanges (Beven and Kirkby, 1979; Moore
86 et al., 1991; Tenenbaum et al., 2006). However, topographic controls on hydrologic and
87 geomorphic processes may be altered in urban landscapes because of infrastructure (e.g., roads,
88 pipes) (Djokic and Maidment, 1991; Tenenbaum et al., 2006; Gironás et al., 2009; Rózsa, 2010;
89 Choi, 2012).

90

91 Five sequential LiDAR-derived DEMs spanning the development of two small, historically
92 agricultural watersheds were obtained to track geomorphic changes throughout development.

93 The goal of this study is to understand how and where geomorphic changes manifest within these
94 watersheds and to understand their implications for watershed geomorphic and hydrologic

95 processes prior to accounting for influences of stormwater infrastructure. The DEM coverage of
96 a third, forested reference watershed was used to track temporal variation not attributable to
97 development. This work will introduce a number of novel methods for tracking temporal
98 geomorphic changes using sequential LiDAR DEMs and will highlight unique geomorphic
99 signatures of the development process. Results of this study will help guide ongoing research
100 efforts that focus on the coupled influence of topography and infrastructure on watershed
101 function.

102

103 **2. Site description**

104

105 For this study, three watersheds in Montgomery County, Maryland, within the Piedmont
106 Physiographic Province were examined. All three watersheds are within the Mt. Airy Uplands
107 District, characterized by siltstones and quartzite with underlying crystalline bedrock consisting
108 of a phyllite/slate unit, with average annual precipitation of 106.4 cm (Reger and Cleaves, 2008;
109 Hogan et al., 2014). Soper Branch (forested control) drains an area of $\sim 3.4 \text{ km}^2$ with an overall
110 mean gradient of 13%. Land cover is predominantly classified as deciduous forest ($\sim 85\%$) and
111 small percentages of low-density housing and agriculture (6 and 9%, respectively) (Hogan et al.,
112 2014; Dr. J.V. Loperfido, USGS, oral communication, July 2013). Tributaries 104 and 109
113 (T104 and T109, urbanizing) drain areas of ~ 1.2 and 0.9 km^2 with mean gradients of 11 and 8%,
114 respectively. Tributaries 104 and 109 are historically agricultural watersheds that underwent
115 extensive suburban development from 2003 to the present. Tributary 104 land use shifted from
116 41% agriculture, 0.3% barren, 42% forest, and 17% urban in 2002 to 2% agriculture, 15%
117 barren, 19% forest, and 64% urban in 2008. Across the same time period, T109 changed from

118 64% agriculture, 1% barren, 25% forest, and 10% urban to 45% agriculture, 13% barren, 25%
119 forest, and 17% urban (Hogan et al., 2014). Both T104 and T109 were used in analyses of
120 development-induced geomorphic changes. However, development in T109 was not complete at
121 the time of the last LiDAR flight, so the final T109 DEM represents a different stage of
122 development than T104.

123
124 Tributaries 104 and 109 fall within the Clarksburg Special Protection Area (CSPA, established in
125 1994) and are subject to development guidelines and restrictions (Maryland DEP, 2013). Broadly
126 speaking, the Special Protection Area designation identifies areas with high quality natural
127 resources and requires ongoing and future development projects to implement best available
128 water quality and quantity protection measures. Water quality and quantity measures often
129 exceed minimum local and state regulations and include extensive Best Management Practices
130 (BMPs) implementation during (sediment and erosion controls; S&EC) and after development
131 (stormwater management; SWM) (Hogan et al., 2014). Each watershed has extensive storm
132 sewer (SS) infrastructure in addition to noted BMP and SWM structures. While SS infrastructure
133 likely plays a key role in dictating watershed hydrology, the purpose of this study is to document
134 geomorphic changes and their hydrologic implications independent of SS influence. Ongoing
135 research will incorporate SS infrastructure to distinguish its effects from geomorphic impacts and
136 better account for watershed hydrologic dynamics.

137

138 **3. Data description and methods**

139

140 Five sequential LiDAR-derived 1-m bare-earth point clouds were collected at semiannual
141 intervals from 2002 to 2008 to track development-induced topographic changes in T104 and
142 T109 and background topographic changes in Soper Branch. Detailed LiDAR vendor,
143 instrumentation, and accuracy information are provided by Jarnagin (2010). Building removal
144 and bare-earth point cloud filtering was performed by each LiDAR vendor utilizing proprietary
145 software, summarized in Table 1. Differences in bare-earth filtering algorithms used by different
146 vendors may have caused interpolated topography within filtered building footprints to vary
147 between dates. Therefore, topographic patterns within building footprints may reflect filtering
148 artifacts and should be interpreted with a degree of uncertainty.

149

150 *Insert Table 1 near here.*

151

152 LiDAR was collected as part of a larger monitoring effort to track watershed changes throughout
153 development and to document development effects on local resources through time (Jarnagin,
154 2008). Detailed comparisons between LiDAR spot elevations and field-surveyed elevations
155 across a range of slope, elevation, and land use classes are reported by Gardina (2008) and
156 Jarnagin (2010). One-meter resolution DEMs were interpolated from bare-earth point clouds
157 using the natural neighbor interpolation algorithm (Sibson, 1981). The DEMs were then
158 coarsened to 2-m horizontal resolution to smooth and reduce noise in the interpolated
159 topography. All subsequent spatial analyses were performed using the interpolated 2-m DEMs
160 in ArcMap 9.3¹ (ESRI, Redlands, CA).

161

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162 3.1. Elevation change

163

164 Differencing sequential DEMs has been used to estimate sediment budgets, to identify nutrient
165 and sediment sources and sinks (Thoma et al., 2005), and to track gully evolution through time
166 (Perroy et al., 2010). Raw difference calculations, however, do not incorporate any estimate of
167 the error inherent across different DEMs or LiDAR vendors (e.g., they imply that all observed
168 elevation differences are real). Further, LiDAR precision has been shown to vary across land
169 cover classes—with forest cover exhibiting the lowest precision (Gardina, 2008; Jarnagin, 2010).
170 Therefore, comparisons of absolute elevation change estimates across differing land cover
171 classes may be misleading.

172

173 To correct for the inherent variability in elevation estimates through time, DEM differences were
174 expressed as standard deviations (SD) through time calculated for each pixel in each watershed
175 as

$$SD = \sqrt{\frac{1}{N} \sum_{i=year}^N (x_i - \mu)^2} \quad (1)$$

176

177 where N is the number of DEM years (5), x_i is the elevation for a given pixel x in year i , and μ is
178 the mean elevation of pixel x across the five DEM years. The distribution of elevation SDs in
179 Soper Branch was then used to quantify background elevation variability in T104 and T109. The
180 SD distribution in Soper Branch was filtered to exclude all known areas with recent
181 anthropogenic activities or structures to assure the distribution of SDs were indicative of the
182 background signal. Any remaining non-zero SD in Soper Branch was assumed to be attributable

183 to either natural causes or uncertainty in the LiDAR data. Because of the decreased elevation
184 precision observed within forested pixels (e.g., Gardina, 2008; Jarnagin, 2010), the use of Soper
185 Branch to quantify background LiDAR variability was a conservative baseline for detecting
186 change in nonforest areas. To detect elevation changes associated with the development process,
187 resulting SD maps were classified based on the distribution of SDs in Soper Branch, with an
188 additional class added to distinguish SD values outside of the range observed in Soper Branch.
189 Classes represent SD values less than or equal to the mean, within three standard deviations of
190 the mean, and SD values greater than three standard deviations from the mean. Any elevation
191 change above the maximum SD observed in Soper Branch is assumed to be attributable to the
192 development process.

193

194 *3.2. Slope change*

195

196 The distribution of local slopes from before and after development in T104 were quantified and
197 classified to ranges relevant to landscape development decisions described by Csima (2010;
198 Table 2). Each slope class in the pre-development landscape provides an indication of the
199 relative development costs and earth-moving practices used across the watershed. A
200 representative cross section was extracted for pre- (2002) and post- (2008) development years to
201 highlight changes in the juxtaposition of high and low sloping areas and to differentiate spatial
202 patterns associated with modified and unmodified areas. More broadly, changes in local terrain
203 created by reallocating and relocating slopes were mapped using areal standard deviations of
204 slope within a 10 x 10 pixel square neighborhood of every focal pixel x_i . Comparison of maps

205 resulting from pre- and post-development DEMs further highlighted changes in the variability of
206 local landscapes associated with development practices.

207

208 *Insert Table 2 near here.*

209

210 *3.3. Hypsometric curves*

211

212 Hypsometric curves represent the proportion of a watershed's area (a/A) that is above a given
213 elevation (h/H), providing a generalized approximation of hillslope shape within the watershed.

214 Traditionally, hypsometric curves have been used in discussions of landscape evolution across

215 broad spatial extents (e.g., Strahler, 1957). Other studies have found hypsometric curve shape

216 indicative of dominant erosional mechanisms (Harlin, 1980; Luo, 2000), hydrologic response

217 (Luo and Harlin, 2003), and infiltration dynamics (Vivoni et al., 2008). Hypsometric curves were

218 generated for each DEM year based on first-order catchment and full watershed scales to

219 compare and contrast hillslope topographic alterations at increasingly broader spatial extents.

220 Consistent boundaries were used for each year to assure that any hypsometric variability through

221 time was attributable to topographic changes and not to boundary shifts caused by topographic

222 modification along drainage divides. Hypsometric curves for each year were compared to assess

223 hillslope level changes through time. A map of net fill and excavation ($DEM_{2008} - DEM_{2002}$) was

224 also developed to help interpret hypsometric trends.

225

226 *3.4. Drainage network change*

227

228 Topographically based drainage network delineation methods are widely used across a range of
229 land covers and climates. The most common approaches rely on threshold relationships of
230 derivative watershed properties (e.g., drainage area, slope – area relationship) to define upslope
231 channel extents (e.g., O’Callaghan and Mark, 1984; Montgomery and Foufoula-Georgiou, 1993).
232 Other techniques utilize local topography to extract convergent drainage features and stream
233 heads directly (e.g., plan/profile curvature) but have seen much less use because of higher
234 topographic data resolution required for feature detection (Tribe, 1992; Band, 1993; Tarboton
235 and Ames, 2001; Lindsay, 2006). A number of threshold-type approaches have been shown to
236 provide reasonable approximations of field-surveyed channel networks (extent and complexity)
237 (Heine et al., 2004; James et al., 2007; James and Hunt, 2010). However, threshold-based
238 methods are insufficient for determining the location and occurrence of artificial urban conduits
239 (e.g., swales, gutters) whose location is based on design specifications, not physical processes
240 (Gironás et al., 2010; Jankowsky et al., 2012).

241
242 Topographic openness (a measure of tangential curvature; see Yokoyama et al., 2002) was
243 employed to identify surface networks in the pre-development terrain of T104 and then applied
244 to the post-development terrain to compare surface network changes. Topographic openness is
245 an angular relation of horizontal distance to vertical relief, calculated from above
246 (positive/zenith) and below (negative/nadir) a topographic surface (DEM). For an angle $< 90^\circ$,
247 openness is equivalent to the internal angle of an inverted cone, its apex at the focal DEM pixel,
248 constrained by neighboring elevations within a specified radial distance. Topographic openness
249 is more robust in identifying surface convexities and concavities than commonly used profile and
250 plan curvature (Yokoyama et al., 2002) and has been successfully applied to LiDAR to identify

251 convergent topography (Molloy and Stepinski, 2007; Sofia et al., 2011). Topographic openness
252 was calculated using an Arc Macro Language (AML) script (written by M.E. Baker, UMBC,
253 personal communication, 2005) that reproduces the methods detailed in Yokoyama et al. (2002).
254 The AML extracts DEM values for all relevant cells in a neighborhood set for each focal pixel x_i
255 (defined by stepwise increments of one pixel width in each of eight azimuth directions until a
256 specified search radius is reached; here we used 100 m) in a landscape. For each increment, the
257 horizontal distance from and elevation above the focal pixel is calculated, and an arctan function
258 converts opposite:adjacent ratios to angular degrees. The AML tracks a running maximum and
259 minimum angle across each radial increment and all smaller increments for all eight directions,
260 converts the resulting maxima and minima to zenith and nadir angles, and computes the mean of
261 each across all eight directions. Thus the final openness values represent the averaged openness
262 measure for all eight azimuths across the specified search radius.

263

264 Difference DEMs and orthoimagery were used to identify detention basins and other potential
265 barriers to surface flow. Pre- and post-development DEMs were filled to overcome internal
266 drainages (i.e., 'pits') and then subtracted from unfilled DEMs to assess filling extents. Paths
267 were carved from detention basin low points to the next downslope pixel of equal or lesser
268 elevation outside of the filled zone. Where possible, carved pathways followed detention basin
269 outlets identified from areal and ground surveys. No attempt was made to account for flowpaths
270 within internal drainage basins or subsurface drainage infrastructure. Straight-line pathways were
271 enforced for all unknown detention basin outlets. Flow directions were calculated for each year
272 from the pit-resolved DEMs by enforcing drainage from each DEM pixel to the adjacent or

273 diagonal pixel with the greatest drop in relief (see O'Callaghan and Mark, 1984). Resulting flow
274 direction surfaces were used to generate drainage networks and other derivative surfaces.

275

276 To create a stream map, orthoimagery and hillshaded DEMs were overlain with openness grids
277 to determine the critical openness for identifying surface depressions in the pre-development
278 T104 DEM. A negative openness (nadir) angle of 91.5° was determined to sufficiently capture
279 all visible surface depressions. All pixels at or above 91.5° were identified, and then connected
280 using a D8-based accumulation operation to enforce downslope continuity from each depression
281 (Tarboton and Baker, 2008). The pre-development stream map (as defined by downslope
282 accumulation of depression pixels) was then tested for agreement with the constant drop law,
283 which states that the mean elevation drop across channels within a given Strahler order should
284 not be significantly different from the mean drop in the next higher order (Tarboton et al., 1991).
285 Incremental accumulation thresholds of depression pixels were used to prune the pre-
286 development stream map until it satisfied the constant drop law. The pre-development
287 accumulation threshold was then applied to the post-development depression accumulation map
288 to define the post-development stream map. Pre- and post-development stream maps were
289 overlain and compared to assess network changes. Field visits were also conducted to classify
290 channels as either natural (defined banks, sorted bedload, evidence of ongoing or recent flow) or
291 artificial (e.g., swales, outflow pipes, rip rap).

292

293 **4. Results**

294

295 Hillshades of each DEM year highlight the spatial pattern of surface modifications incrementally
296 throughout development (Fig. 1). Initial large-scale resurfacing throughout the watershed is clear
297 in 2004 and 2006 hillshades, with urban or suburban infrastructure distinguishable in the terrain.
298 Fine-scale grading and smoothing that is much less visually apparent in the LiDAR characterize
299 late-stage development in 2006 and 2008. Valley burial is evident in the western portion of the
300 watershed (see red arrow in Fig. 1), replaced by a smooth upland housing cluster and roadway.
301 Mainstem valley form remained relatively constant throughout the time series, with little to no
302 visual evidence of lateral channel movement.

303

304 *Insert Figure 1 near here.*

305

306 *4.1. Elevation standard deviations*

307

308 Pixelate elevation standard deviations (SD) calculated across the five DEM years were
309 approximately normally distributed in Soper Branch with a mean of 0.063 m and a standard
310 deviation of 0.034 m, ranging from 0.002 to 0.870 m. Areas with temporal SDs greater than or
311 equal to three standard deviations of the watershed mean (≥ 0.163 m) were focused within
312 riparian zones at or near stream banks and near the basin outlet within a relatively flat and wide
313 floodplain area (Fig. 2). Jarnagin (2010) reported similar LiDAR errors within densely vegetated
314 riparian zones in comparisons between ground-truth and LiDAR spot elevations (absolute mean
315 difference of 0.216 m with a standard deviation of 0.723 m and a maximum difference of 1.180
316 m). Moderately high SDs (0.063-0.163 m) are also apparent on northeast- (NE) facing hillslopes.
317 The temporal variation observed on NE-facing slopes was greater than systematic errors reported

318 for ground survey and LiDAR-derived elevations, which exhibited a mean absolute difference of
319 ~ 0.05 m (Gardina, 2008).

320

321 Elevation changes in T104 and T109 classified to the distribution of temporal SDs in Soper
322 Branch highlight large areas exhibiting substantial elevation changes greater than the variability
323 observed in forested controls. White areas in T104 and T109 signify temporal elevation SD
324 values that are above the maximum observed in Soper Branch (> 0.870 m), encompassing ~
325 24.4% and 9.9% of total watershed area, respectively. Distinct patches of high elevation
326 variability are primarily located in development parcels evident in the post-development
327 hillshades (2008 panel in Fig. 1). High temporal variability (0.163-0.870 m) is also apparent
328 along the mainstem channel of T104 but is less evident in T109. Relatively low temporal SD
329 values in the southern undeveloped portion of T109 and within riparian zones in T104 reflect
330 magnitudes of temporal variability similar to those observed in Soper Branch. Such
331 heterogeneous distribution of elevation change is consistent with the observation that earth-
332 moving practices were not spatially uniform across developing parcels.

333

334 *Insert Figure 2 here.*

335

336 *4.2. Slopes*

337

338 The footprint of development is clearly distinguishable from unmodified hillslopes in a visual
339 comparison of pre- and post-development slope maps (Fig. 3). Development created abrupt slope
340 changes in the landscape, replacing relatively smooth topographic profiles (i.e., gray boxes in

341 inset). Moderate to high slope classes ($> 12\%$) were present along roads, housing parcels, and
342 stormwater management features, despite the use of bare-earth LiDAR with buildings removed
343 by filtering. Valley infill is apparent in the slope transect between points *a* and *b* at ~ 180 m with
344 no evidence of the original valley structure in the post-development topography. High-frequency
345 variation between low slope and high slope features is evident in the slope transect between 420
346 and 580 m across a developed subdivision.

347

348 *Insert Figure 3 here.*

349

350 Areal slope standard deviations calculated using a 10 x 10 pixel moving window show a marked
351 increase in the spatial variability of local slope values in post-development topography (Fig. 4).
352 The overall distribution of variation in local slopes exhibited a positive shift from the pre- (mean
353 areal slope standard deviation of $2.35 \pm 2.26\%$) to post-development terrain (mean areal slope
354 standard deviation of $4.69 \pm 3.54\%$). Highly variable slope zones in 2002 were limited to riparian
355 corridors and preexisting development plots in the northern section of T104. By contrast, post-
356 development slopes exhibit greater local variation (i.e., steeper and flatter terrain in close
357 proximity) associated with the grading of subdivisions and road embankments.

358

359 *Insert Figure 4 here.*

360

361 *4.3. Hypsometry*

362

363 Despite substantial topographic changes across the watershed visible in hillshades and slope
364 maps representing different stages of development, hypsometric changes owing to development
365 only manifest when observed across small extents with relatively uniform, high magnitude
366 surface changes (Fig. 5). Distinctly terraced hypsometric curves appear in the end-development
367 phase topography, mirroring abrupt slope changes along roadways and surrounding housing
368 parcels (see insets A and C in Fig. 5). Valley infill in subwatershed A (see red arrow in Fig. 1)
369 created a sharp elevation gradient to the remaining riparian lowlands, reflected in the 2007 and
370 2008 hypsometric curves. Terracing and leveling of upslope transportation corridors also
371 contribute to the terraced hypsometric form. Hypsometric trends of the larger subwatershed B
372 and at the full watershed scale (D) did not exhibit detectable temporal changes.

373

374 *Insert Figure 5 here.*

375

376 *4.4. Drainage network changes*

377

378 Substantial redistribution of overland flowpaths is evident in comparisons of pre- and post-
379 development flow direction grids (Fig. 6). As with slope comparisons, the most drastic changes
380 in flow directions are focused along roadways and within housing parcels. Small patches of
381 uniform flow directions mirroring housing and road footprints have replaced large, contiguous
382 patches of homogenous flow directions (i.e., parallel flow lines) on upland hillslopes. Fine-scale
383 variation in flow directions within relatively minor elevation change zones (see elevation change
384 inset in Fig. 6) are coupled with more dramatic modifications within drainage features and near
385 infrastructure. As a result of these alterations to the flow field, flow accumulation distributions

386 exhibited a negative shift, with accumulation quartiles decreasing from 4, 11, and 29 pixels in
387 2002 to 1, 4, and 12 pixels in 2008. Distributional shifts reflect drainage dissection and the
388 fragmented flow field apparent in Fig. 6 with small, locally convergent pathways replacing large,
389 contiguous, parallel flowpaths.

390

391 *Insert figure 6 here.*

392

393 A depression pixel (openness $\geq 91.5^\circ$) accumulation threshold of three pixels was found to
394 satisfy the constant drop criterion in the pre-development terrain of T104. The three-pixel
395 threshold was then applied to the post-development terrain to evaluate surface network changes.
396 Comparisons of pre- and post-development drainage network structure in T104 indicate a net
397 gain in stream length of 4.25 km (~ 52% increase) throughout the watershed (Fig. 7). Additions
398 were common along transportation infrastructure as swales or gutters typical in this watershed.
399 Upslope extensions of existing channels also occurred extending beyond preexisting piped
400 infrastructure at the apparent head of pre-development drainage lines (Fig. 7A). Valley infill (see
401 Fig. 6A) caused the loss of a tributary but was replaced by an artificial conduit that paralleled a
402 nearby road (Fig. 7B). Similar conduits are apparent on the eastern side of the watershed as well
403 (Fig. 7C). No substantial changes were apparent in the mainstem channel as suggested by the
404 near 1:1 overlap in pre- and post-development networks.

405

406 *Insert Figure 7 here.*

407

408 **5. Discussion**

409

410 The growing footprint of development in T104 is clearly visible in the DEM hillshade time
411 series, with distinct early- and late-stage development phases operating at unique spatial scales
412 (Fig. 1). Large-scale cutting and filling to support major infrastructure characterizes early
413 development, evident in the 2004 and 2006 hillshades. In contrast, fine-scale leveling and
414 grading of housing parcels characterize late-stage development (MDE, 2000). Fine-scale earth-
415 moving practices are less immediately apparent in hillshades but are detectable in derivative data
416 sets, as this study has shown. Results show that geomorphic changes associated with urban
417 development are significant and nonuniform and could have implications for management and
418 conservation efforts.

419

420 Initial analysis of background elevation changes in Soper Branch revealed nonzero standard
421 deviations focused on northeast- (NE) facing hillslopes and within riparian and floodplain areas
422 (Fig. 2). Increased variation on NE hillslopes may reflect flight paths taken during LiDAR
423 collection or possible aspect differences in vegetation and soils. Slightly lower systematic errors
424 reported in Gardina (2008) were attributed in part to methods used to derive bare-earth point
425 clouds and, in part, to LiDAR system error. The NE artifact likely is a combination of vegetation
426 and LiDAR system errors. Riparian and floodplain variability in Soper Branch can be attributed
427 to LiDAR errors reported by Jarnagin (2010). In comparisons between LiDAR elevations and
428 field-surveyed spot elevations, Jarnagin observed greater LiDAR error in densely vegetated
429 riparian zones and on steep slopes common around incised channels. Distinguishing temporal
430 variability attributable to artifacts in the LiDAR time series was necessary before reaching any
431 quantitative conclusions about topographic changes resulting from development.

432

433 Elevation standard deviations classified to distinguish background and development-induced
434 changes clearly show that earth moving in T104 and T109 was nonetheless substantially greater
435 than background geomorphic changes observed in Soper Branch. Magnitudes similar to temporal
436 standard deviations found in Soper Branch were observed in T104 across preexisting housing
437 parcels and riparian zones left untouched to comply with riparian protection policies. Somewhat
438 higher variability observed along the mainstem channel in T104 may indicate subtle channel
439 shifts and bank erosion throughout the time period despite attempts to insulate the channel from
440 the hydrologic and geomorphic effects of development. Also, some of the observed elevation
441 changes may possibly reflect artifacts produced by removing buildings from the raw LiDAR
442 point clouds. Because of the proprietary nature of the LiDAR processing, evaluating the bare-
443 earth detection algorithms used by the data providers was not possible for this study.

444

445 Surface changes throughout development in T104 and T109 altered slopes to support
446 infrastructure, enforce drainage, and promote infiltration. Comparison of pre- and post-
447 development slope maps in T104 shows a marked increase in high slope classes with
448 development as well as reallocation across space (Fig. 3). High slope classes focused along
449 transportation corridors and surrounding detention basins are likely to promote surface drainage.
450 Slopes leading toward streams in the pre-development map appear to steepen with development
451 as well, likely reflecting upland filling and terracing. Stream burial is a frequently cited
452 phenomenon in urban watersheds (Leopold et al., 2005; Elmore and Kaushal, 2008; Roy et al.,
453 2009; Doyle and Bernhardt, 2011) and is apparent on the western side of T104 (~ 180 m in
454 transect inset of Fig. 3). However, without infrastructure design documents it is difficult to

455 determine whether the stream is piped under its original flowpath or if it has been redirected to
456 flow alongside the nearby road running north to south.

457

458 Regular shifts between high and low slopes within subdivisions likely reflect housing parcels
459 (low slopes) separated by graded lawns and swales (higher slopes) to promote drainage (Fig. 4).
460 Low slopes within building footprints reflect the building removal algorithm applied to LiDAR
461 point clouds to extract bare-earth points used in this study to create DEMs. The spatial
462 distribution of high and low slopes in part controls the spatial distribution of runoff generation
463 and infiltration zones (Beven and Kirkby, 1979; Tenenbaum et al., 2006). Shifts in the spatial
464 pattern and magnitude of slopes may contribute to altered watershed processes including runoff
465 generation, nutrient and sediment transport, and surface water – groundwater exchanges.

466

467 Hypsometric trends within small subwatersheds of T104 appear to mirror aforementioned
468 steepening of valley walls with development (Fig. 5). Upland filling and grading further
469 exacerbates the disparity in elevations between the now-developed uplands and the riparian
470 lowlands. Redistribution of flowpaths at and around valley infills likely altered surface and
471 subsurface water exchanges, thus changing the functional connectivity of upland and lowland
472 areas. Dynamics are further complicated by stormwater management infrastructure, such as
473 detention ponds and infiltration trenches, typically located between upland (development plots)
474 and lowland (riparian zone) areas to capture and treat overland runoff before it enters local
475 streams (Montgomery County Planning Department, 1994). Thus, riparian areas may no longer
476 be functionally connected to upland areas during rain events unless stormwater management
477 outfalls are triggered. Further research is required to investigate the nature of the altered surface

478 – groundwater connections. Lack of temporal trends in the larger subwatershed topography or
479 throughout the entire watershed result from spatial averaging and may indicate a detection limit
480 for tracking parcel or block-level topographic change with hypsometry.

481

482 Development caused a substantial reconfiguration of overland flowpaths from large contiguous
483 areas of uniform directions to small, spatially fragmented patches (Fig. 6). The nonrandom
484 distribution of smaller patches mirrors housing footprints, transportation corridors, and
485 stormwater management infrastructure and reflects landscape-engineering practices that promote
486 efficient collection and drainage of overland runoff (Fig. 7). Adjacent pixels with uniform
487 directions indicate parallel flowpaths, which will not converge until a pixel with a differing
488 direction is encountered downslope (O'Callaghan and Mark, 1984). By introducing heterogeneity
489 into the flow direction surface, development has created both more dissected subbasins and
490 increased opportunities for small flowpaths to converge. Increased convergence in upland areas
491 causes overland flow to reach moderately low accumulations more often and very high
492 accumulations more rarely than in the pre-development terrain. Decreased median and first-
493 quartile values between 2002 and 2008 reflect the dominance of low accumulation pathways
494 with development. Such alterations are likely the cumulative result of many site-specific efforts
495 to increase local water conveyance, but their cumulative effect should also have implications for
496 patterns of soil moisture, variable source area, erosion, infiltration, and recharge (Beven and
497 Kirkby, 1979; Moore et al., 1991; Tenenbaum et al., 2006). To our knowledge, a geomorphic
498 driver of increased upland convergence in developed terrain has yet to be described in the
499 literature. Typically, pipes and impervious surfaces are thought to be the causal agents of altered
500 runoff, nutrient, and sediment dynamics in urban watersheds (e.g., Paul and Meyer, 2001; Dietz

501 and Clausen, 2008; Roy et al., 2008). However, our results indicate that the geomorphic
502 modifications that occur with development are substantial and could also contribute to altered
503 watershed functions.

504

505 Comparisons of pre- and post-development drainage networks reveal channel additions through
506 engineered surface conduits (e.g., swales, culverts) and losses caused by valley infill or
507 subsurface routing (Fig. 7). Despite noted variability in elevations along stream banks, only
508 minor shifts were apparent in the mainstem of T104. Results of this study also raise questions
509 about what constitutes a channel in an urban landscape. Traditional channel definitions,
510 characterized by defined banks and sorted bedloads (Montgomery and Dietrich, 1989; Heine et
511 al., 2004) exclude engineered surface conduits that become active during precipitation events
512 (Doyle and Bernhardt, 2011). Excluding these artificial conduits and subsurface stormwater
513 pipes may underrepresent the surficial connectivity of urban watersheds. Without infrastructure
514 design documents, the true drainage connectivity of the watershed is unclear. Ongoing work will
515 explore methods for incorporating subsurface drainage features into T104 network
516 representations (e.g., Gironás et al., 2009; Choi, 2012; Jankowsky et al., 2012) and determine
517 how best to incorporate BMP infrastructure into predicted network representations. Nevertheless,
518 this study indicates an overall increase in surface connectivity with development independent of
519 changes in impervious surface cover or subsurface stormwater infrastructure. Increased
520 connectivity would result in faster and larger storm peaks if left unchecked by stormwater
521 management infrastructure. Therefore, considering geomorphic changes in addition to
522 impervious surface cover impacts is important for remediation efforts in urban areas.

523

524 **6. Conclusions**

525

526 This study has shown the utility of sequential LiDAR-derived DEMs for tracking and
527 quantifying the geomorphic changes associated with development. Using a forested watershed as
528 a reference, background topographic variability was quantified and differentiated from
529 development-induced topographic change. Utilizing first- and second-order topographic
530 derivatives, this study demonstrated that development generates increasingly disjointed hillslopes
531 characterized by abrupt slope changes separating flat upland areas. Temporal variation in
532 elevations, slopes, and hypsometric curves likely contributes to altered watershed functions, but
533 further research is required to understand their importance relative to stormwater infrastructure.
534 Substantial modification of overland flowpaths resulted in subtle changes to delineated network
535 structure, highlighting additions and losses to the pre-development network. This study's ability
536 to detect natural and manmade ephemeral surface conduits provides a more complete accounting
537 of watershed hydrologic connectivity. Better accounting of upslope ephemeral channel
538 extensions may help in designating areas for protection in future development projects. The
539 geomorphic signal of increased landscape convergence and drainage density indicates more rapid
540 runoff generation and conveyance independent of impervious surface cover or storm sewers. A
541 better understanding of how and where geomorphic changes occur throughout development may
542 enable a better understanding of pollutant mobilization and retention processes. Ongoing work
543 seeks to incorporate known subsurface pipe and BMP structures into surface network
544 representations to understand watershed hydrologic response dynamics.

545

546 Results presented here raise important considerations for temporal topographic studies. Standard
547 topographic methods developed for use in larger, mostly nonurban watersheds (e.g., DEM
548 differencing, hypsometry) have limited applicability for tracking urbanization unless summarized
549 across a relatively small extent. Additional research is needed to gain a better understanding and
550 to develop novel methods for tracking and quantifying urban topographic modifications.

551

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553

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565

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709

710 **List of Figures**

711

712 Fig. 1. Sequential light detection and ranging (LiDAR) derived digital elevation models (DEMs)
713 spanning the development process in T104 show the topographic footprint of development
714 increasing through time. Boxes highlight topographic differences between early- and late-stage
715 development practices. The red arrow denotes a valley that was buried during development.

716

717 Fig. 2. Elevation change through time quantified as temporal standard deviations (SD) calculated
718 for each DEM pixel. Map values are classified to the distribution of elevation variation observed
719 in Soper Branch (forested control). Black and dark gray areas represent values below and within
720 three standard deviations of the areal mean in Soper Branch, respectively. Light gray areas
721 represent temporal variation in elevation greater than three standard deviations above the areal
722 mean in Soper Branch; whereas white areas are greater than the maximum, and both indicate
723 areas of substantial development.

724

725 Fig. 3. Comparisons of pre- (2002) and post- (2008) development slope distributions in T104
726 classified to ranges relevant to development practices (see Table 2 after Csima, 2010). Inset

727 shows a cross-sectional profile of slope along the dashed red line between points *a* and *c*.
728 Substantial differences in the variability of local slopes are evident in the profile between
729 developed (shaded zone) and undeveloped (unshaded) parcels. High slopes ring roads, swales,
730 and detention features and manifest as high magnitude variation across the lateral transect.

731

732 Fig. 4. Areal slope standard deviations calculated using a 10x10 pixel moving window for pre-
733 and post-development T104 slope maps. As in pixelate slope maps, highly variable slope zones
734 are concentrated along road corridors and around detention ponds. Nevertheless, a substantial
735 increase in the local variability of slope is apparent in 2008, with high standard deviation values
736 widely distributed across the watershed.

737

738 Fig. 5. Hypsometric variability through time exhibits distinctly different patterns across first-
739 order subwatersheds in T104 (A, B, and C). Hypsometry of relatively small subwatersheds that
740 underwent substantial resurfacing (A ~ 0.09 km² and C ~ 0.03 km²) shifted from smooth (black
741 curves) to terraced curve forms (gray curves) reflecting grading of developed parcels. However,
742 extensive resurfacing was not apparent in hypsometric trends of larger subwatersheds (B ~ 0.43
743 km²) or at the full watershed scale (D ~ 1.18 km²).

744

745 Fig. 6. Extensive topographic modification (e.g., C) associated with grading throughout
746 development (2002-2008) causes substantial variation in overland flowpath directions and
747 elevation change (D) in T104. Comparisons between pre- and post-development years show that
748 fine-scale flow direction changes at the hillslope scale are coupled with the designed orientation
749 and gradients of road networks (A and B, with each color representing a unique direction of

750 overland flow). Large patches of a single color indicate parallel flow lines down a hillslope,
751 whereas fragmentation of the 2002 map in 2008 necessarily indicates redirection and
752 convergence of previously unconnected contributing areas.

753

754 Fig. 7. Comparison of pre- (blue lines) and post- (red dashed lines) development channel
755 network delineations suggest increased network density with development in T104. Added
756 channels include upslope extensions beyond existing culverts (A) and vegetated swales parallel
757 to roadways (C). Comparisons to pre-development topography show a minor tributary that has
758 been buried during development (B).

759

Table 1

Building removal and bare-earth point cloud processing information for each LiDAR year provided by S.T. Jarnagin (personal communication, 2013).^a

LiDAR year	Instrument	Vendor	Building removal software	Bare-earth filtering
2002	Optech ALTM-2025	Airborne 1	Optech's REALM 2.27	Terrascan (running on Microstation)
2004	Optech ALTM-2033	Laser Mapping Specialists Inc.	Applanix POSPROC & Optech's REALM (versions not specified)	Spectra's Terramodel
2006	Optech ALTM-3100	Canaan Valley Institute	Optech's REALM (version not specified)	Microstation 8 with Terrascan
2007	Optech ALTM-3100	Canaan Valley Institute	Optech's REALM (version not specified)	Microstation 8 with Terrascan
2008	Leica ALS-50	Sanborn	Applanix POSPROC 4.3	Leica ALS post-processing and Terrasolid Terrascan

760 ^aAny use of trade, firm, or product names is for identification purposes only and does not imply
761 endorsement by the U.S. Government.

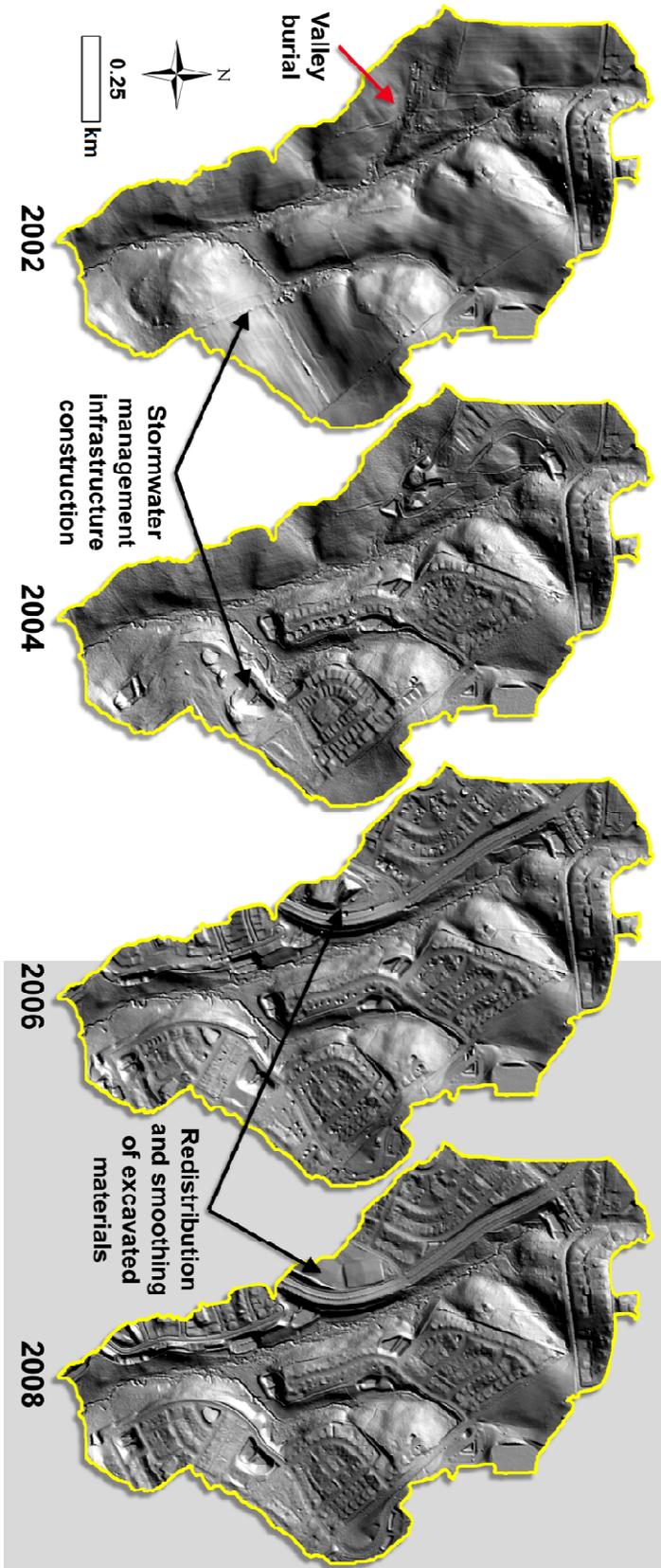
Table 2
 Slope classes and associated economic costs and considerations for urban development (adapted from Csimá, 2010)

Angle of slope	Development potential and the required landscaping
Up to 5%	Easy and economic development potential. In general, terracing not necessary; landscaping exclusively restricts to drainage. Relief does not pose a limit either to buildup density or building size.
5-12%	Increased development costs. Landscaping is inevitable; development is only possible with terracing and slope leveling. Somewhat limited development.
12-25%	Development potential at significant cost and labor expenditure, only with terraces and supporting walls provided. Major topographic transformation; relief fundamentally controls the type of development to be applied.
25-35%	Limited potential for urban development. Low building density with small-sized buildings.
Above 35%	Unsuitable for urban development.

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Large-scale earth-moving activities

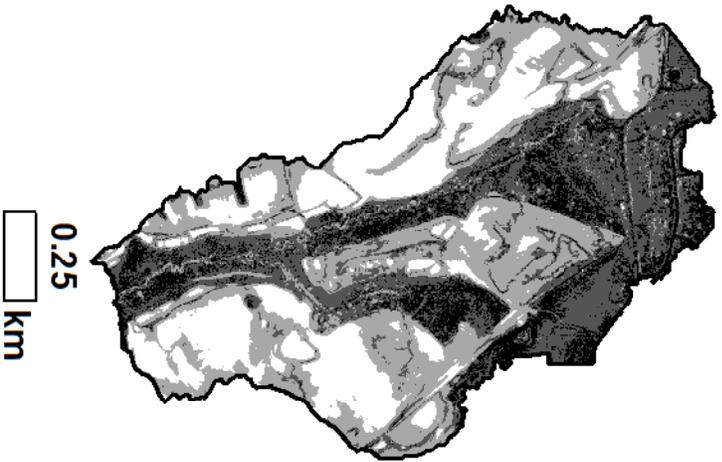
Fine-scale grading



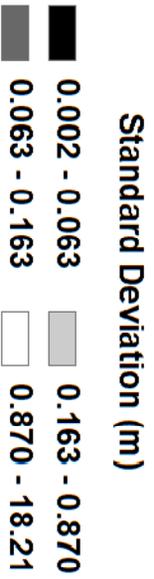
Soper Branch

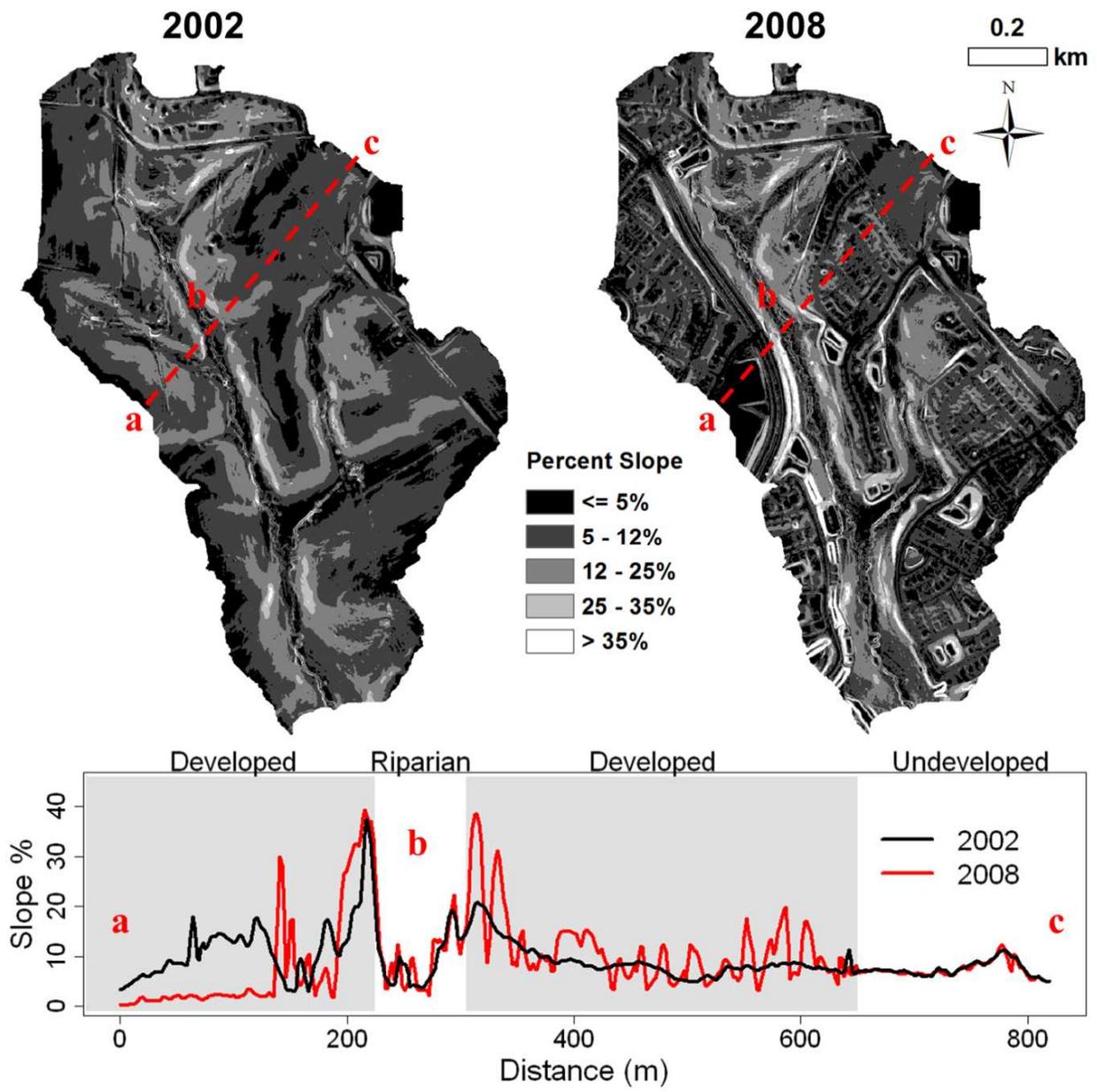


Tributary 104

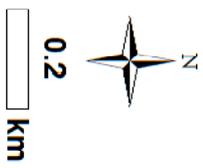
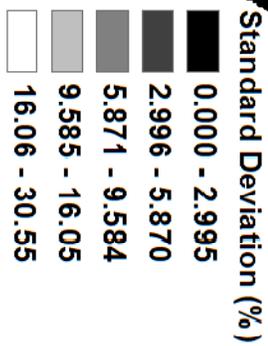
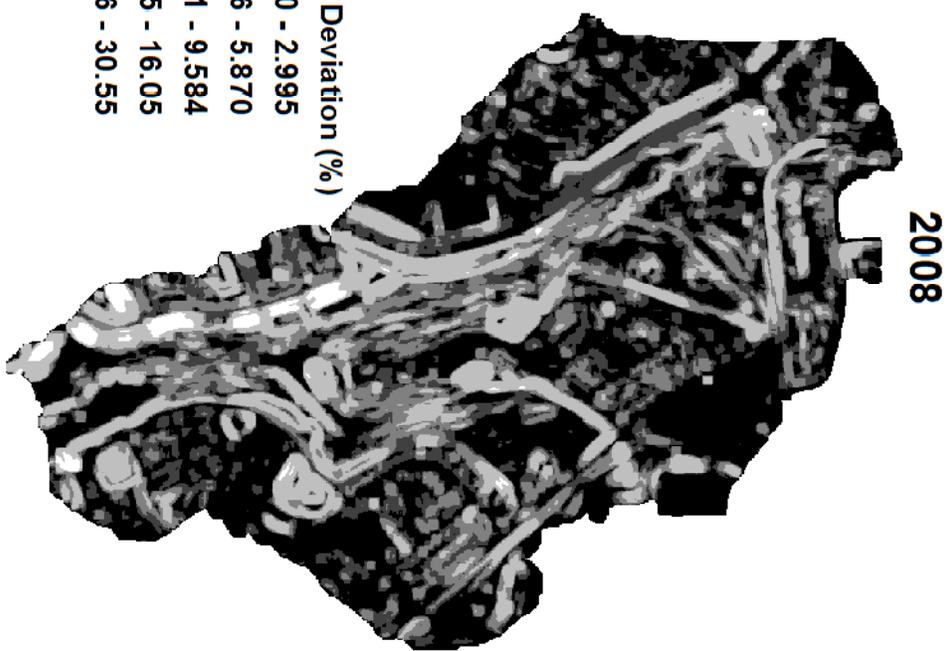
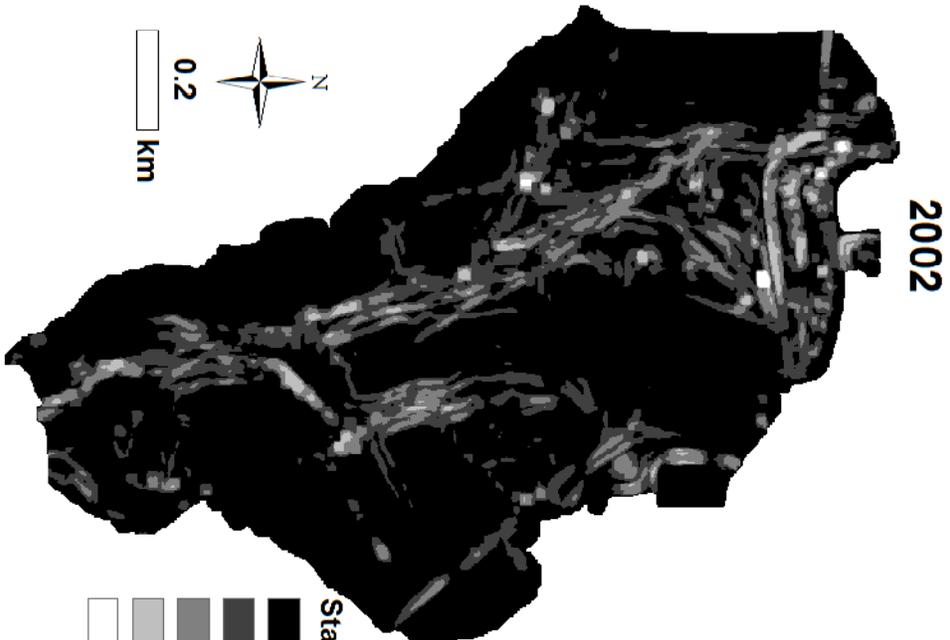


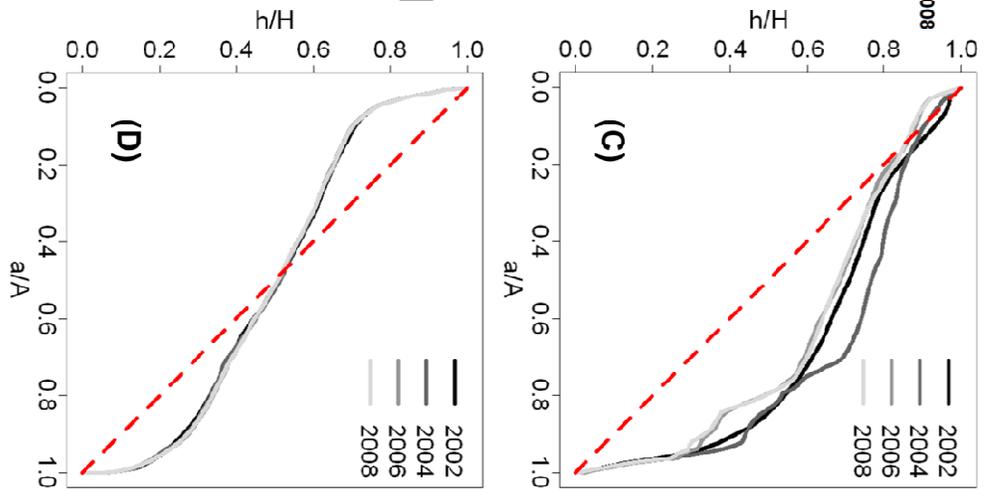
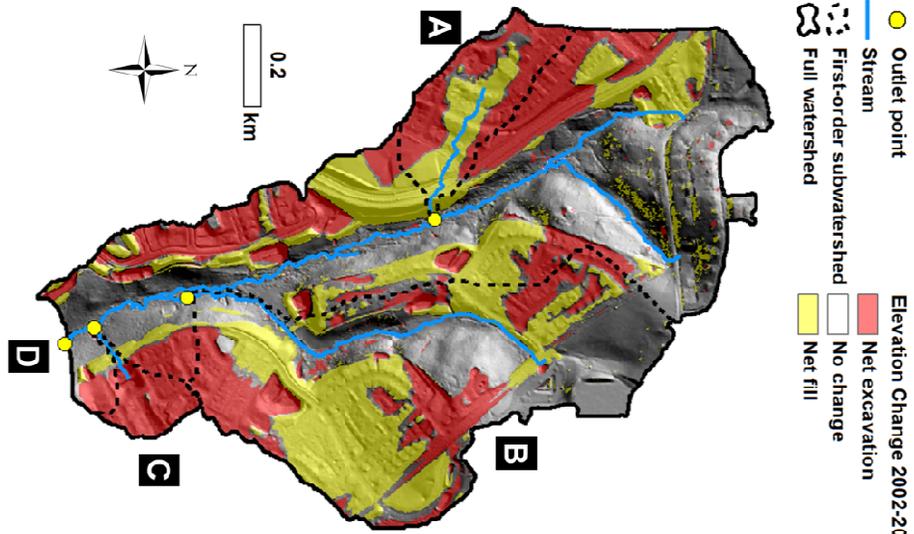
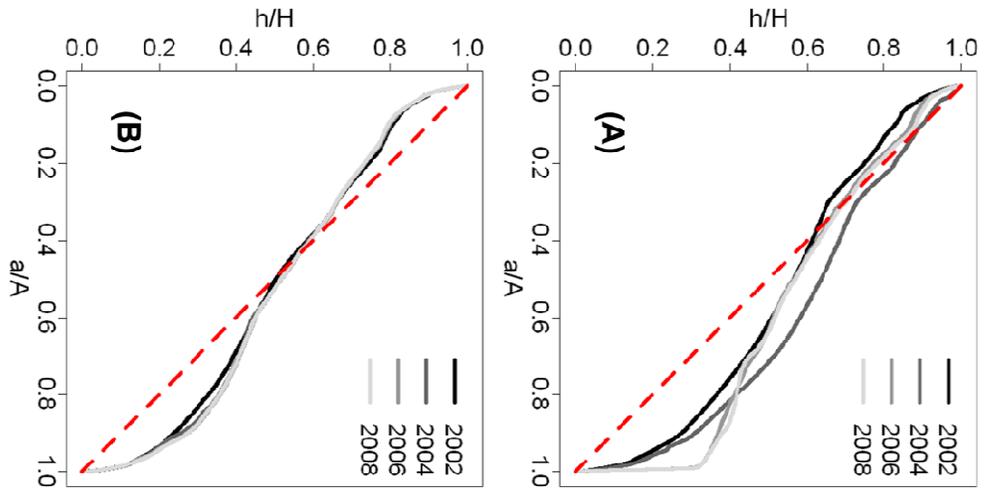
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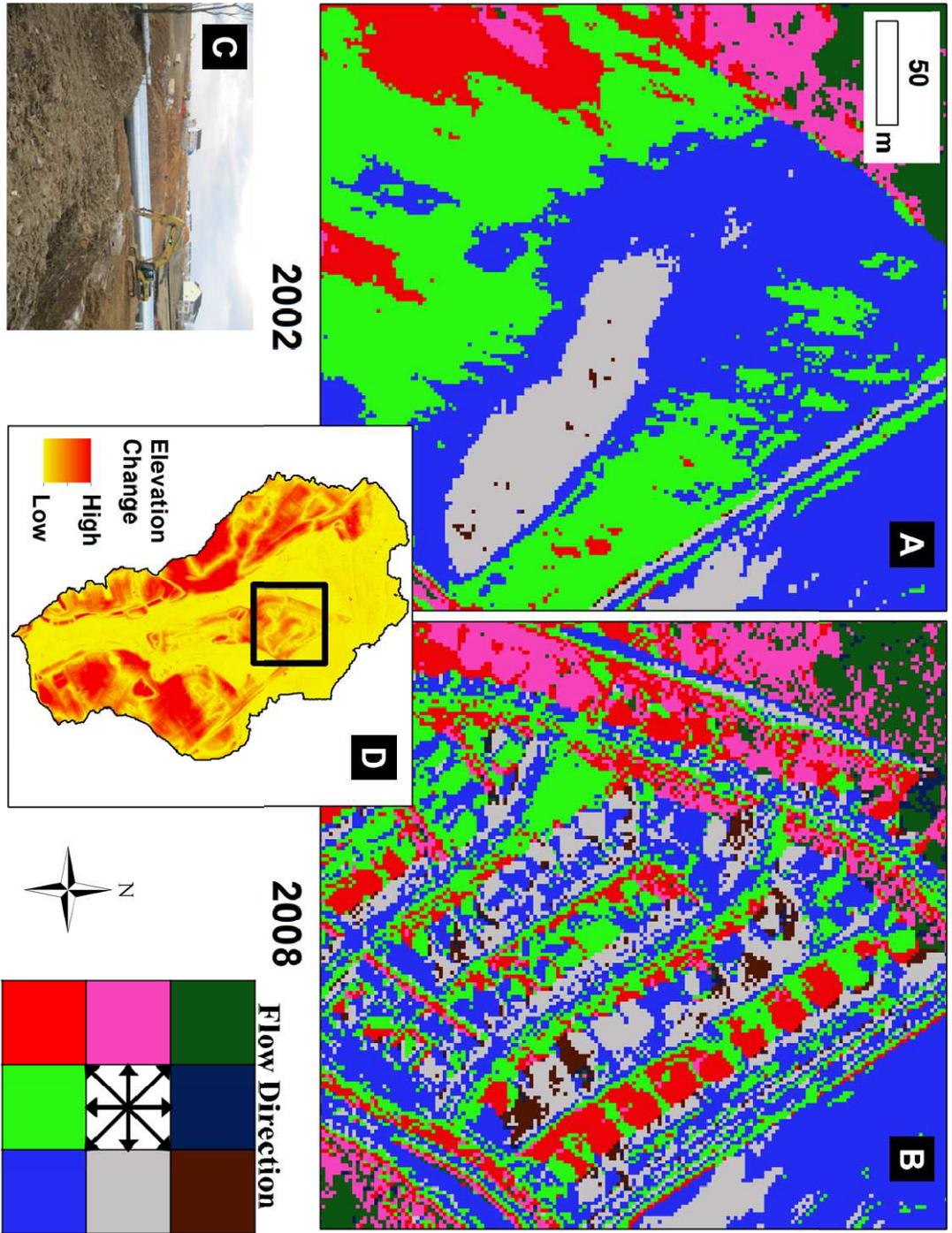




765







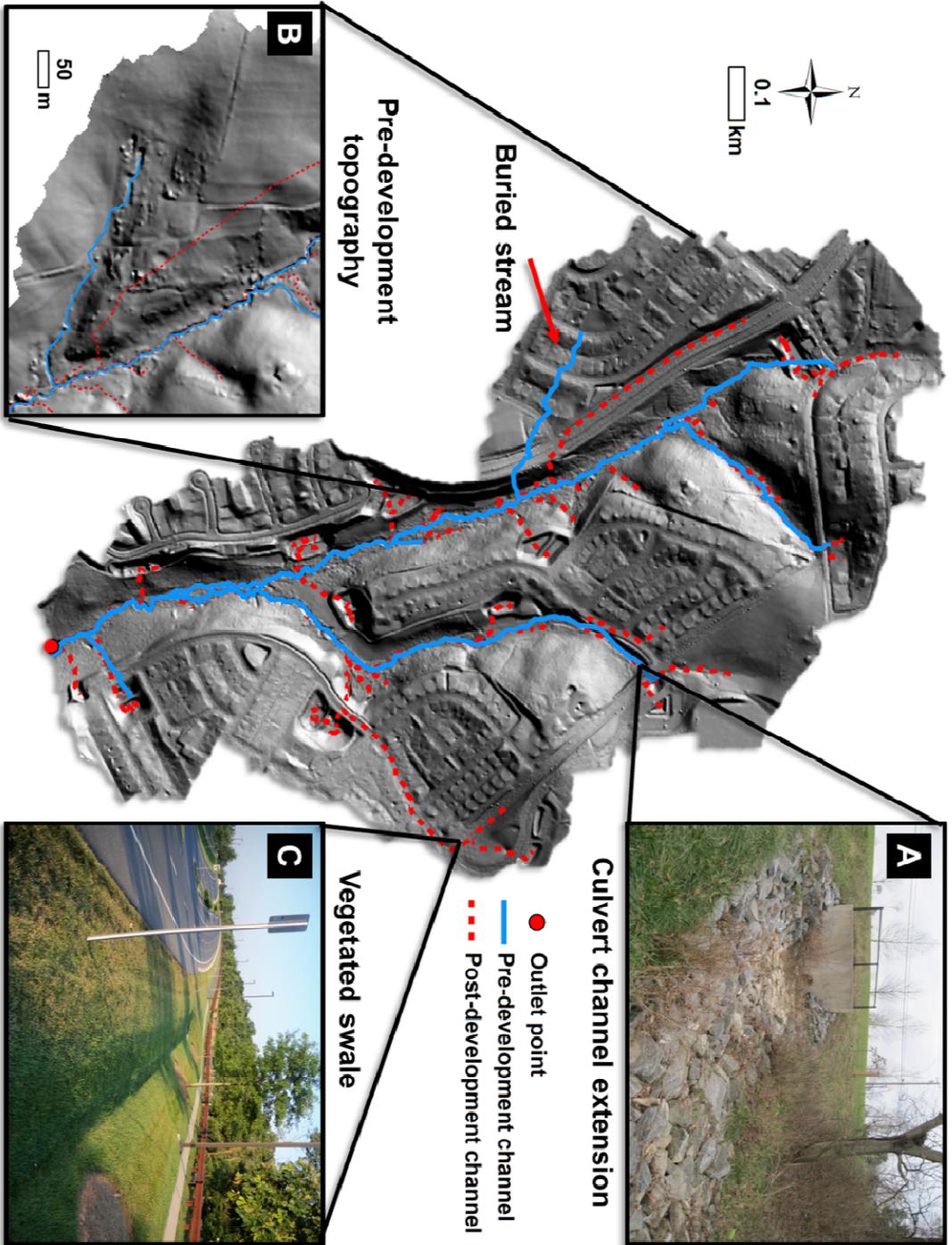


Table 1

Building removal and bare-earth point cloud processing information for each LiDAR year provided by S.T. Jarnagin (personal communication, 2013).^a

LiDAR year	Instrument	Vendor	Building removal software	Bare-earth filtering
2002	Optech ALTM-2025	Airborne 1	Optech's REALM 2.27	Terrascan (running on Microstation)
2004	Optech ALTM-2033	Laser Mapping Specialists Inc.	Applanix POSPROC & Optech's REALM (versions not specified)	Spectra's Terramodel
2006	Optech ALTM-3100	Canaan Valley Institute	Optech's REALM (version not specified)	Microstation 8 with Terrascan
2007	Optech ALTM-3100	Canaan Valley Institute	Optech's REALM (version not specified)	Microstation 8 with Terrascan
2008	Leica ALS-50	Sanborn	Applanix POSPROC 4.3	Leica ALS post-processing and Terrasolid Terrascan

^aAny use of trade, firm, or product names is for identification purposes only and does not imply endorsement by the U.S. Government.

Table 2
Slope classes and associated economic costs and considerations for urban development (adapted from Csimá, 2010)

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Figure 1
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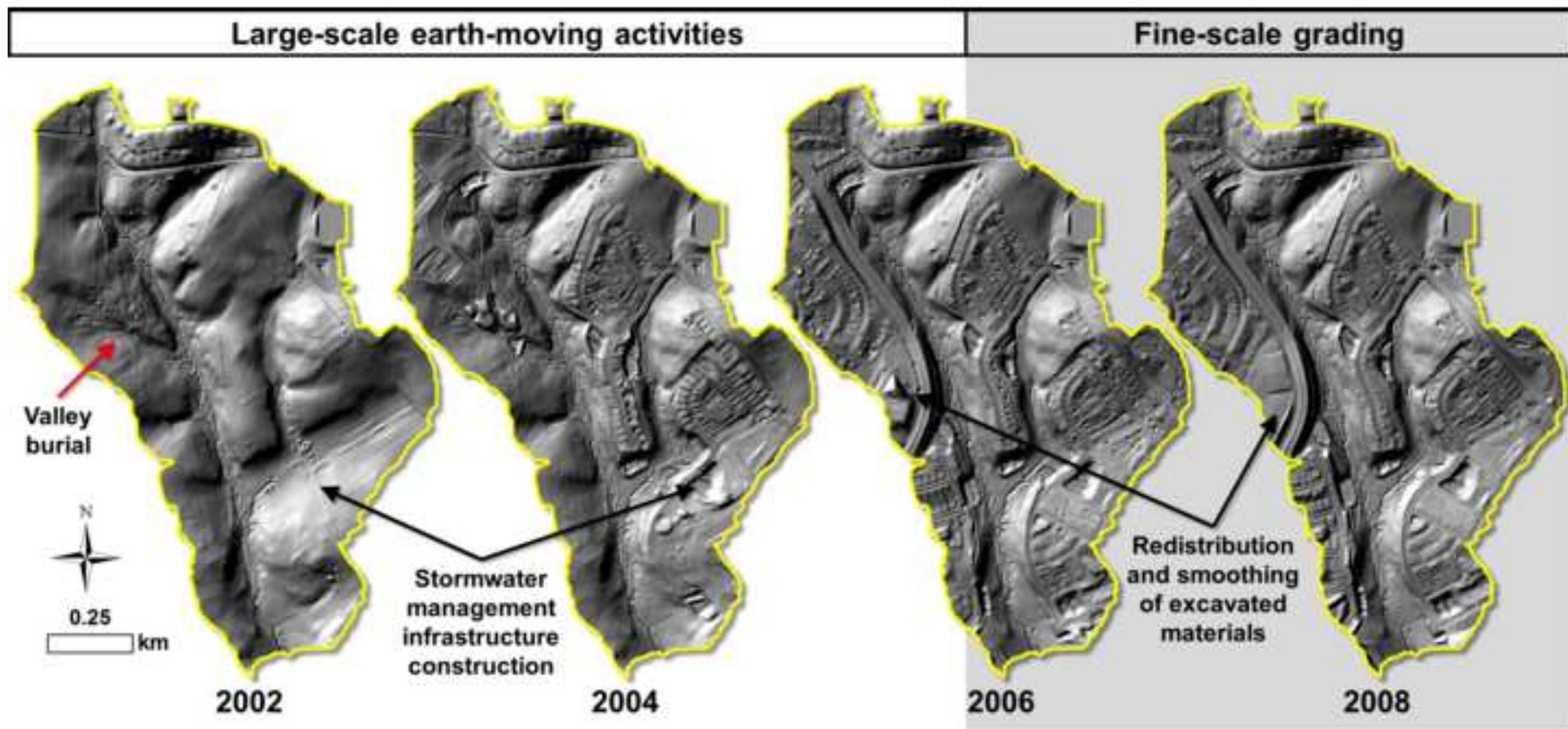


Figure 2
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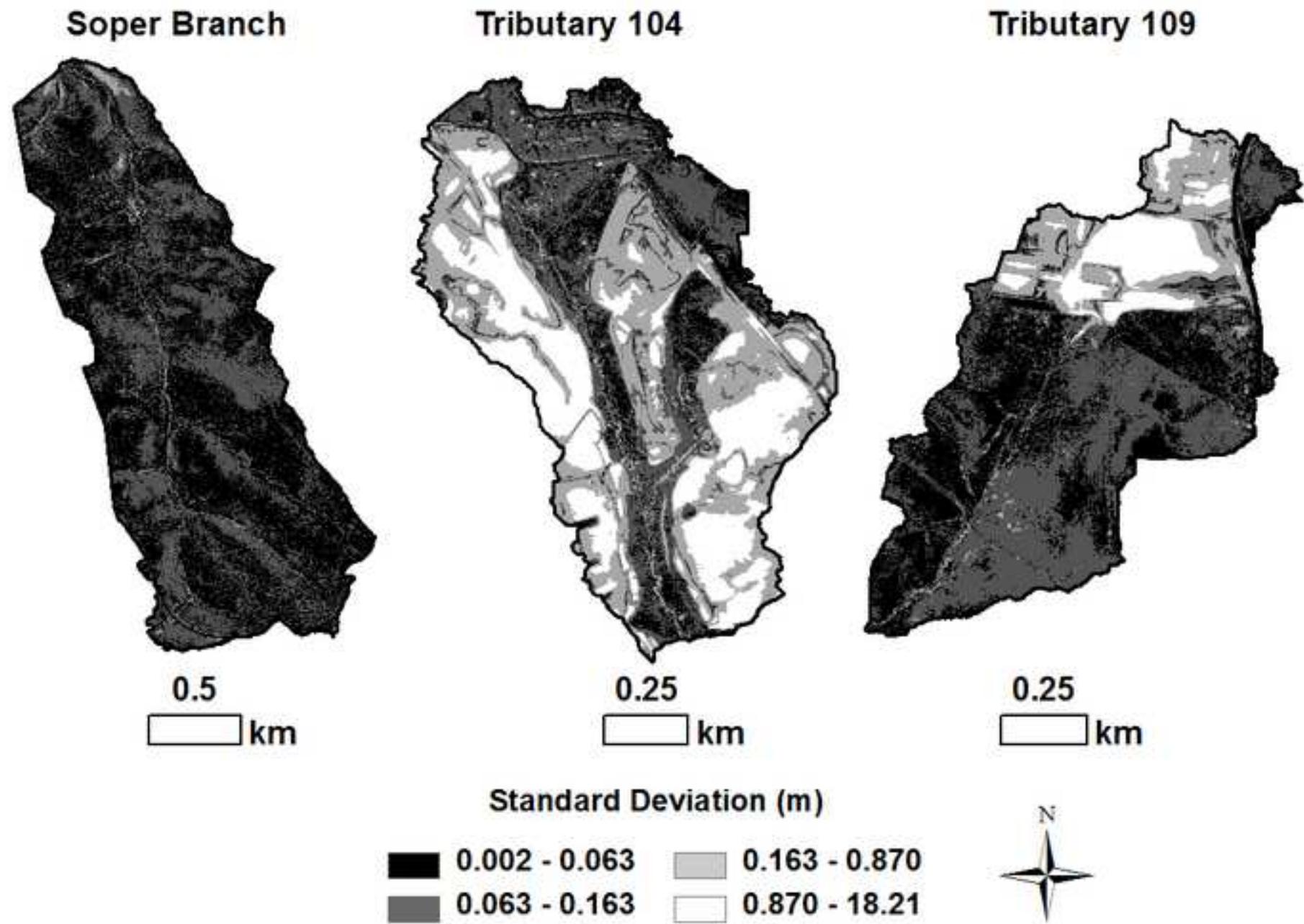


Figure 3
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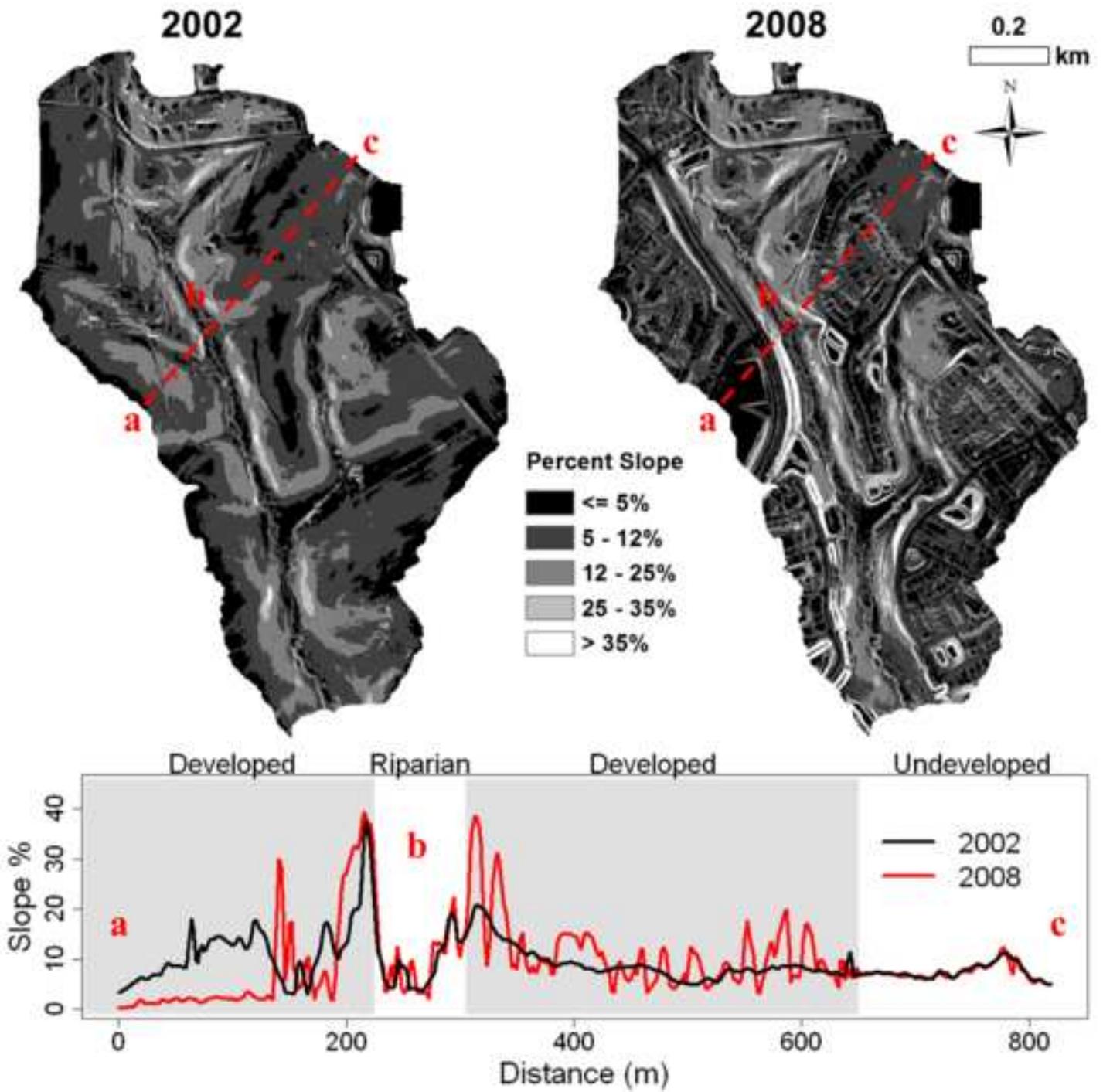


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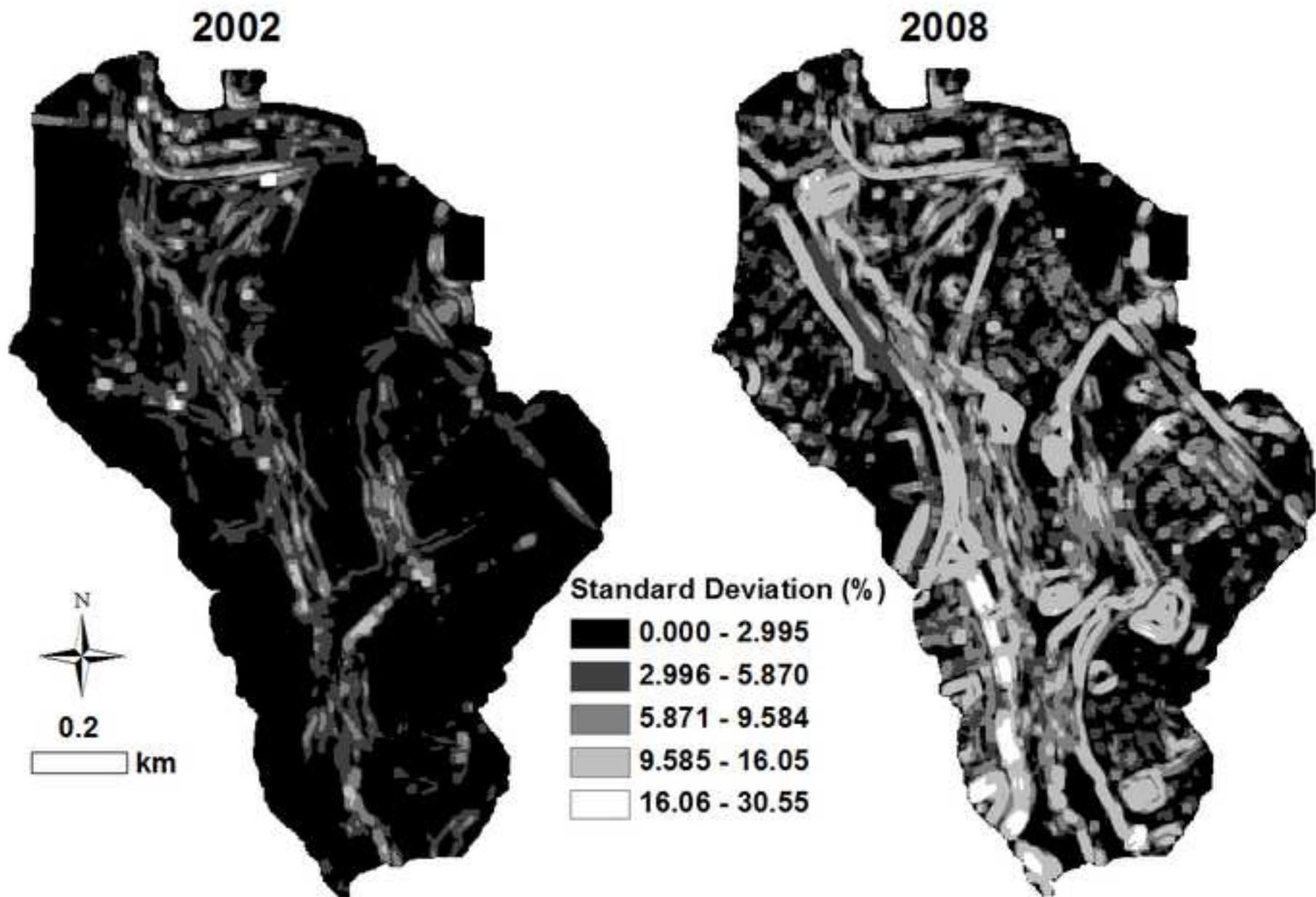


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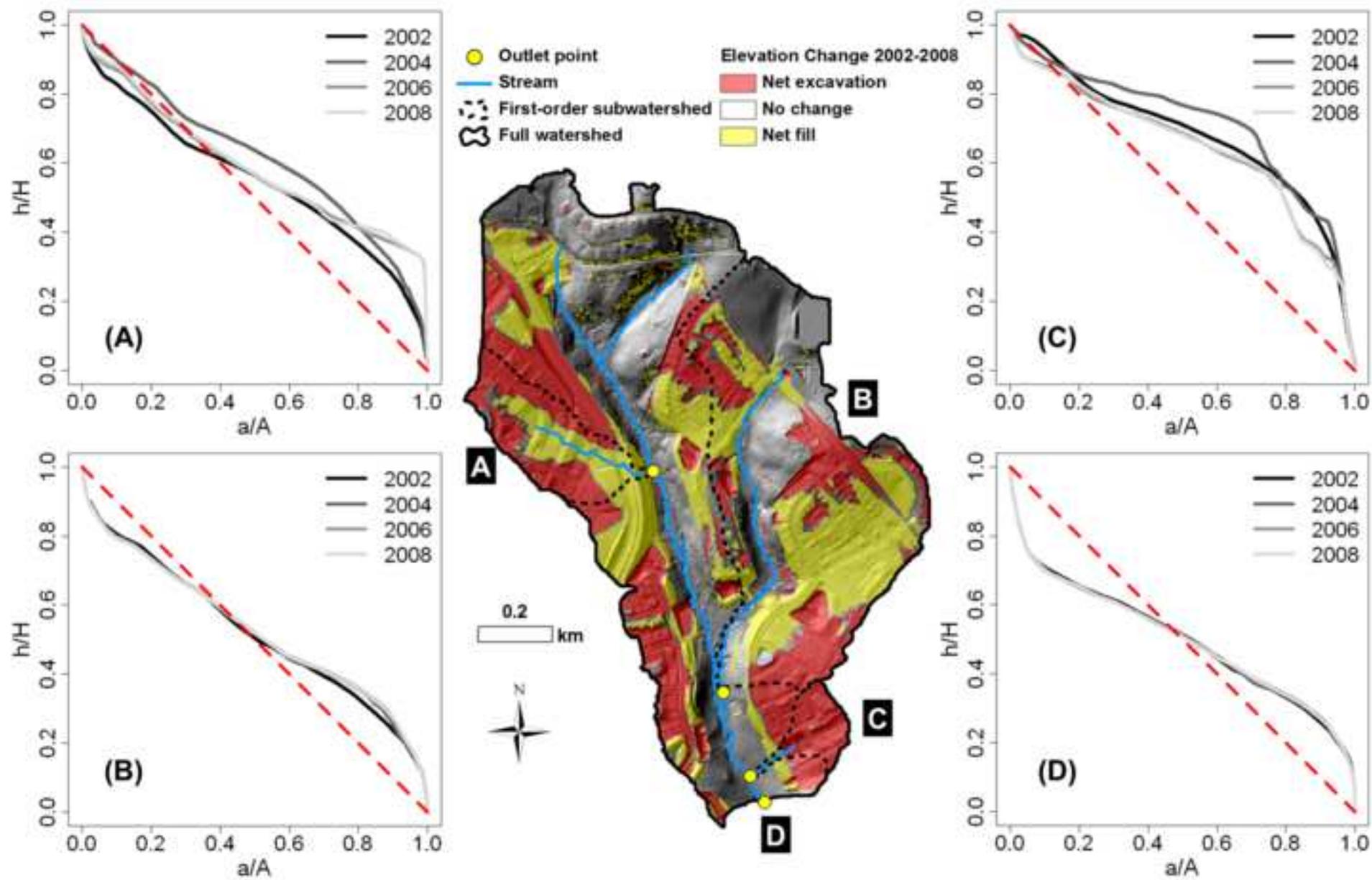


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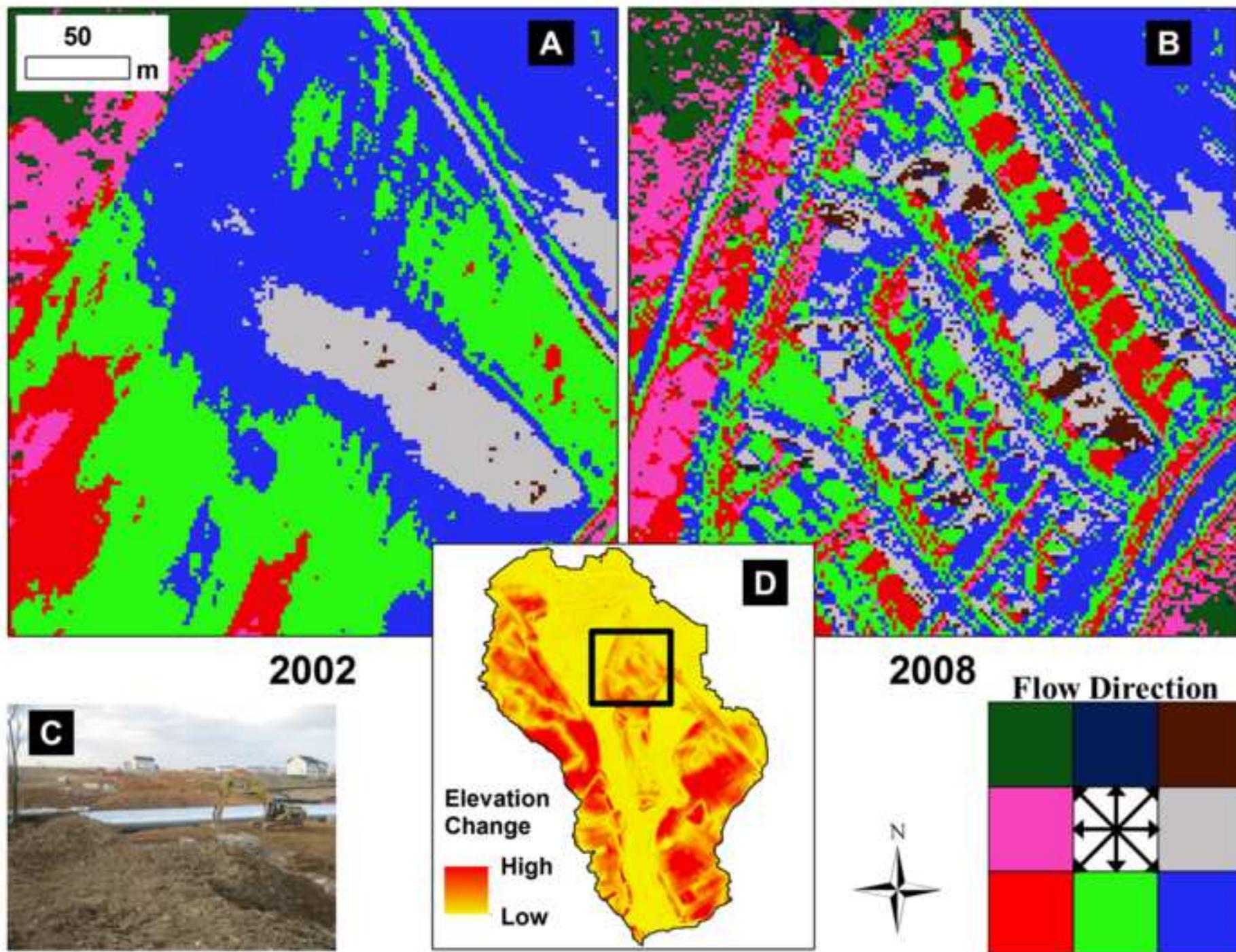


Figure 7
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