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Unlocking Flow: An Investigation of topographical controls on Canadian river runoff

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42 **Unlocking Flow: An Investigation of topographical**
43 **controls on Canadian river runoff**

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Abstract

Digital Elevation Models (DEMs) are a crucial tool for watershed analysis, offering valuable insights into landscape-scale hydrology. Traditional watershed delineations are derived by filling a DEM to force flow paths through topographic depressions, thus creating a continuous drainage network throughout the domain. However, this approach is challenged in landscapes characterized by abundant real-world depression storage, internally drained lake basins, and intermittently blocked stream channels such as the Canadian Shield (CS). The CS landscape is influenced by “fill-and-spill” surface water hydrology, with runoff flow paths controlled by bedrock sills and rocky cascades that are overtopped when water levels are high but cease flowing when water levels are low. To better represent this intermittently connected drainage pattern, we apply an alternative, less aggressive DEM filling model (Fill-Spill-Merge or FSM) to a continental-scale DEM (MERIT) for all of Canada. To ensure adequate filling of DEM errors while also preserving real-world topographic depressions, we propose a climatic method to initialize a key FSM parameter (“runoff depth”) that optimizes observed discharges from 1690 Environment and Climate Change Canada (ECCC) active gauging stations with climate model data. Our application of FSM to all 1690 gauges identifies 255 topographical control points that can influence the actively flowing area of the watershed. Extending the approach to ungauged areas of the CS identifies 583 control points. This research demonstrates how CS watersheds are affected by fill-and-spill hydrology, emphasizing the critical role of accurate terrain modeling for delineating surface water flow paths in depressional landscapes.

Keywords: Watershed delineation, surface water hydrology, terrain analysis, Canadian Shield

Introduction

Extraction of drainage patterns from Digital Elevation Models (DEMs) constitutes a fundamental step in watershed analysis, providing invaluable insights into landscape hydrology[1–4]. Traditional watershed analysis methods have been established for this purpose, relying on algorithms designed to fill topographic depressions to create continuous drainage networks[5,6]. These methods have earned widespread recognition for their effectiveness in characterizing watershed flow paths and runoff behavior [7–10]. The fundamental principle behind these techniques involves treating depressions (also called “pits”) in the DEM as artifacts, which are systematically filled to facilitate unimpeded water flow, but also eliminating any

74 water storage in these depressions. This approach effectively removes all topographic barriers to flow, ensuring a consistent
75 flow direction for each cell in the derived raster and a coherent, well-integrated drainage pattern[11–14].

76 Traditional drainage pattern analysis has known challenges in landscapes having abundant depression storage (i.e.
77 internally drained basins, swales, wetlands, and lakes). These depressions are real-world topographical elements on the
78 landscape that interrupt the regional drainage pattern and should not be removed as artifacts. There are many such places in the
79 world [15–17]. One such place is the Canadian Shield (CS), dominated by glaciated bedrock with numerous topographical
80 depressions (Figure 1). Previous studies have identified “fill-and-spill” hydrological processes on the CS[18,19], which
81 intermittently connect topographic depressions on the landscape causing watershed source areas to expand or contract [4, 8–
82 10]. During wet conditions, topographic depressions and/or lake basins fill and overtop bedrock sills and/or ephemeral
83 channels, connecting to a down-elevation surface drainage network. This effect can increase downstream river discharge
84 [20,21], recharge other lake basins, and mobilize the transport of sediment and nutrients across the landscape [22–25].
85 Controlling bedrock sills, low-gradient ephemeral channels, and waterfalls (rocky cascades) thus create topographical gateways
86 or “control points” that shut off upstream drainage areas during low flow [26] but switch on during high flow. Depending on
87 regional water balance, these control points can release or withhold flow to a downstream drainage network, producing erratic
88 behavior in river hydrographs. Such control points are not identified with traditional DEM filling approaches, because they
89 overfill most depressions to remove all topographic barriers to flow.

90 To mitigate DEM over-filling in depressional and/or poorly drained landscapes, new alternative algorithms that
91 preserve topographic depressions and associated fill-and-spill processes have been developed. The contour tree method [27]
92 constructs depression hierarchies by connecting local minima and maxima of a DEM-generated contour map. While accurate
93 and straightforward, it is computationally intensive so is suboptimal for large geographical areas. A rainfall-based sequential
94 depression-filling algorithm [28] fills low-lying areas sequentially based on their retention capacity and contributing area.
95 However, since rainfall often occurs simultaneously over large geographical areas, this algorithm is not suitable for large study
96 domains. The Fill-Spill-Merge (FSM) algorithm by Barnes et al [29], however, uses a “depression hierarchy” concept to enable
97 high computational efficiency. Larger depressions can contain smaller ones, forming a hierarchy. FSM leverages this hierarchy
98 to fill, overtop, and merge depressions at low computational cost as it does not treat each cell in the DEM to identify and process
99 depressions. Instead, it utilizes the information about the hierarchical relationships between depressions to make decisions
100 about filling and merging. This approach significantly speeds up the process of handling depressions in DEMs and unlike

101 traditional methods, fills a DEM more realistically by preserving many real-world depressions. This, in turn, should enable
102 detection of topographical control points likely to influence watershed source area(s) and drainage pattern.

103 Here we proposed a climatic method to estimate a key parameter for ensuring the right filling of the DEM which
104 removes DEM error but preserves the depressional storage. The key parameter in runoff depth is the sum of minimum filling
105 depth and inundation depth. That parameter was used as the starting runoff depth value to identify those control points. Manual
106 checking and validation were done to filter out artificial control points. After this analysis, we produced a novel dataset of 255
107 control points which can be cascades, lake overflows, or drypaths. These control points can impact the environment by affecting
108 the flow of water, sediment, and nutrients. Mapping them is crucial for understanding and mitigating the environmental impact
109 of human activities, such as dam construction or land development.



Figure 1: Aerial photo of CS region. This photo was taken at NASA AVoBE field campaign 2017 captured by Sarah Cooley

Materials and Methods

Data Used

The Multi-Error-Removed Improved-Terrain (MERIT) DEM was used for digital terrain analysis of Canada's drainage pattern. The MERIT DEM was developed by removing multiple error components (absolute bias, stripe noise, speckle noise, and tree height bias) from two existing satellite-derived DEMs (SRTM3 v2.1 and AW3D-30m v1) [30]. It represents terrain elevations at 3 arc sec resolution (~90m at the equator) and covers land areas between 90N-60S, referenced to EGM96 geoid. The absolute vertical accuracy of this DEM is less than 2m and horizontal accuracy is less than 17m. MERIT DEM's removal of noise and other errors from its underlying satellite data products from the make it well-suited for drainage network extractions[31,32].

To estimate an initial value for a critical FSM parameter ("runoff depth", see section Methodology) we compared Coupled Model Intercomparison Project Phase 6 (CMIP6) [33] Precipitation (P) – Evapotranspiration (ET) data with Environment and Climate Change Canada (ECCC) Water Survey of Canada (WSC) observations of river discharge for 1,690 active gauging stations throughout Canada. CMIP6 datasets (<https://cds.climate.copernicus.eu/cdsapp#!/dataset/projections-cmip6?tab=form>) were used for this purpose instead of climate reanalysis or weather station datasets to enable future projections and uniform coverage over low-population areas, respectively. The 30-year mean of annual hydrologic year (1992 - 2022) runoff (P-ET) from the historical CMIP6 CanESM5 model for Canada was intersected with the watershed areas of all 1,690 active river gauges to retrieve 30-year mean CMIP6 modeled discharge for each basin. Similarly, Monthly discharge observations for all 1,690 active river gauges for 30 water year were downloaded from the Hydrometric Database (HYDAT), available from the ECCC WSC National Water Data Archive (<https://www.canada.ca/en/environment-climate-change/services/water-overview/quantity/monitoring/survey/data-products-services/national-archive-hydat.html>). Yearly average was calculated from those discharge data. The watershed area was also obtained from HYDAT as well. The information about those basins, their name, area, station ID can be found on the supplementary files.

Potential control points identified from FSM terrain analysis were validated visually through inspection of high-resolution map and satellite image products. An initial visual assessment for each control point was conducted using ArcGIS base map, a high-resolution cartographic and satellite image product derived by Esri using DigitalGlobe, GeoEye, i-cubed,

135 USDA FSA, USGS, AEX, Getmapping, Aerogrid, IGN, IGP, swisstopo, and the GIS User Community. If visual inspection of
136 ArcGIS base map was inclusive, a second visual validation was performed using multi-temporal Planet smallsat imagery
137 (<https://www.planet.com/products/planet-imagery/>). Planet provides a Data API that allows users to search for specific imagery
138 based on metadata criteria such as cloud cover percentage. Planet scenes acquiring during high flow and low flow conditions
139 were used to determine if a DEM-flagged location appeared to be a lake basin sill, river cascade, intermittent stream, or similar
140 topographic impediment surface water flow.

141 **Methods**

142 **Fill-Spill-Merge algorithm**

143 This study applies the recently developed Fill-Spill-Merge (FSM) algorithm [34] to delineate Canadian surface
144 drainage patterns from the MERIT DEM. FSM employs a two-step process as follows. First, the algorithm identifies all DEM
145 depressions by finding lower cells than their neighboring cells. It then creates a depression hierarchy, a tree-like structure
146 representing all depressions in the DEM. Conceptually, an ocean is the tree's root, and individual depressions the tree's leaves.
147 Next, the algorithm applies a user-specified runoff depth parameter (e.g., 0.5 m) of water uniformly across the DEM and routes
148 it through the depression hierarchy, filling and merging them according to the underlying topography (Figure 2). Starting from
149 the tree's root, the algorithm checks if each node is a depression. If yes, it fills it with water; if not, it moves up to the next
150 node. Since water is distributed through the depression hierarchy rather than each individual DEM cell, the FSM algorithm can
151 handle the large MERIT DEM at a continental scale.

152 Choice of the FSM runoff depth parameter – in essence a small, uniform DEM filling depth that is then redistributed
153 across the landscape according to topography - is a key decision [35]. From this single parameter the algorithm calculates the
154 total volume of water that will be used to (lightly) fill DEM depressions. The amount of water diverted into each will depend
155 on the shape of the depression and the area of its local catchment. To initialize our study, MERIT DEM was filled using FSM
156 with various 10 runoff depths ranging from 0.2 to 2 m, producing 10 corresponding flow direction and flow accumulation maps
157 for Canada.

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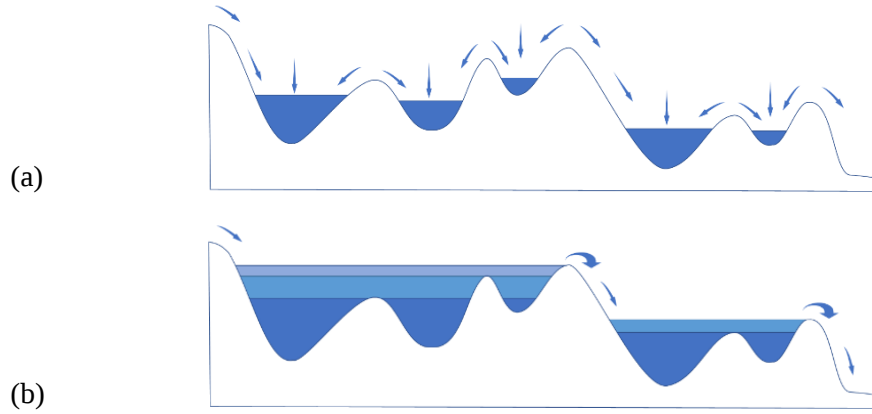
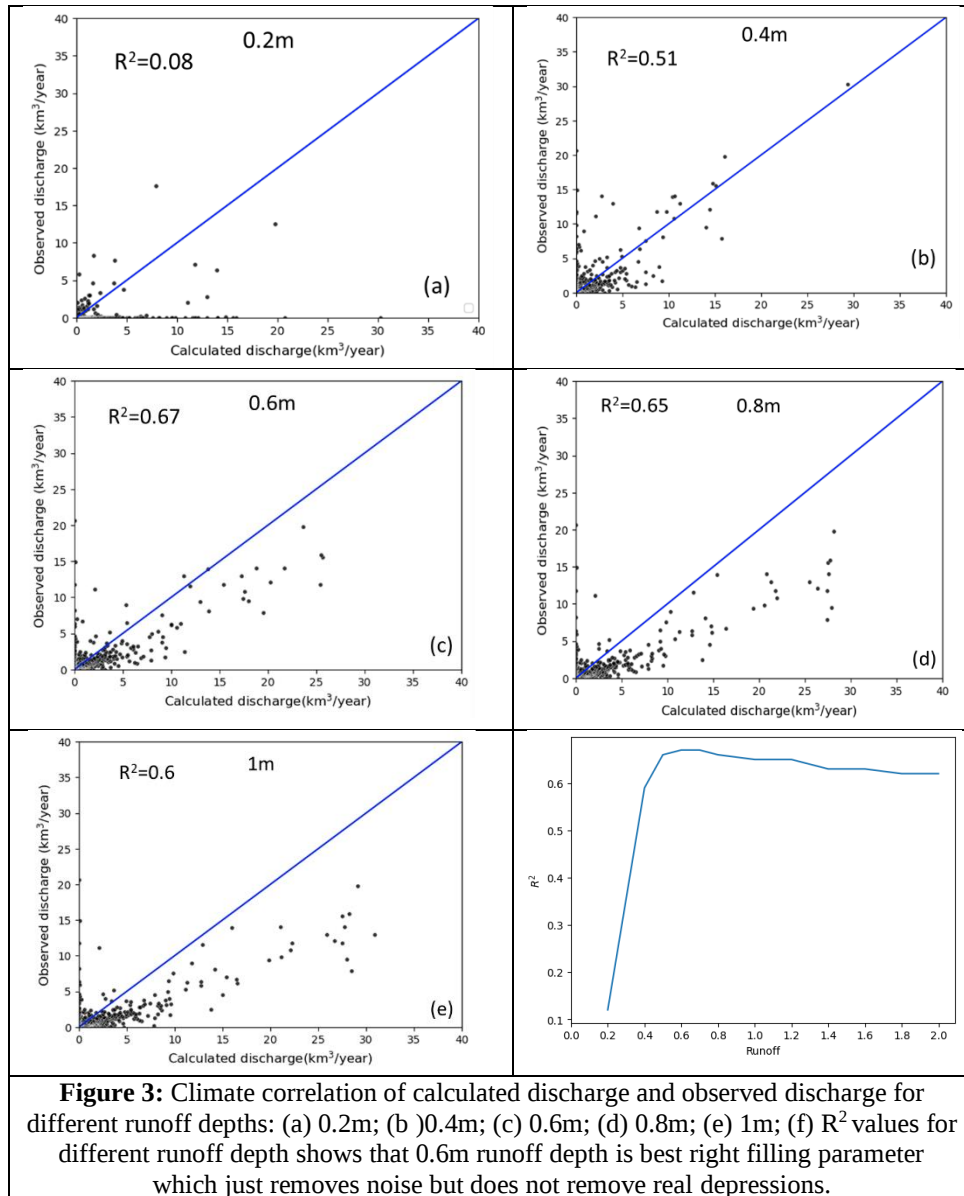


Figure 2: FSM process showing filling, merging, and overflowing process in DEM.

Empirical method for runoff depth

To determine an optimal minimum runoff depth parameter – large enough to smooth and fill most DEM artifacts, but small enough to preserve real-world topographic depressions – we correlated climate model P-ET with observed mean annual discharge for all 1,1690 WSC gauging stations assuming a range of plausible runoff depth parameter values (0.2 – 2.0 m). A smaller runoff depth yields a smaller, more fragmented drainage pattern while a higher runoff depth yields a large, more integrated drainage pattern. In principle, the runoff depth value that best correlates P-ET with observed discharge is optimal because its produce the most physically realistic watershed (under the assumptions that 1) mean annual P-ET $\sim$ mean annual discharge; and 2) both climate model P-ET and gauging station observations are reasonably accurate). For each iteration of the runoff depth parameter, the upstream watersheds for all 1,690 gauges were delineated and their respective mean annual runoff correlated with corresponding CMIP6 P-ET for 30 water years (1992-2022). Maximum correlation between these two datasets ($R^2 = 0.67$) was observed for a runoff depth of 0.6 m (Figure 3). Therefore, 0.6 m was selected as an initial minimum runoff depth parameter for topographic control-point detection.



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173 Detection of topographic control points

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Flow control points be any topographic control or obstruction which causes a sharp increase (decrease) in upstream drainage area when overtopped (exposed) by high (low) water. Their overall effect is to connect to a downstream flow network during wet (high water level) conditions and detach during dry (low water level) conditions. Waterfalls, rapids, dry overspills, lake sills, dams, and road embankments are all examples of potential control points. In this study, potential control points throughout Canada with identified using the FSM algorithm, by applying the FSM algorithm to MERIT DEM with an initial runoff depth of 0.6m and progressively increasing it to 2.0 m. For each 0.2 m increase in runoff depth, grid cells experiencing

180 a +20% and/or +5,000 km² increase in upstream watershed area were considered sufficiently “abrupt” to be flagged as a
 181 potential control point. In total, some 2519 of locations satisfying these criteria were identified within the known watersheds
 182 of the 1,690 WSC gauges across Canada. Control points outside of the ECCC watershed were also detected by finding the
 183 location of new flow path formation all over CS.

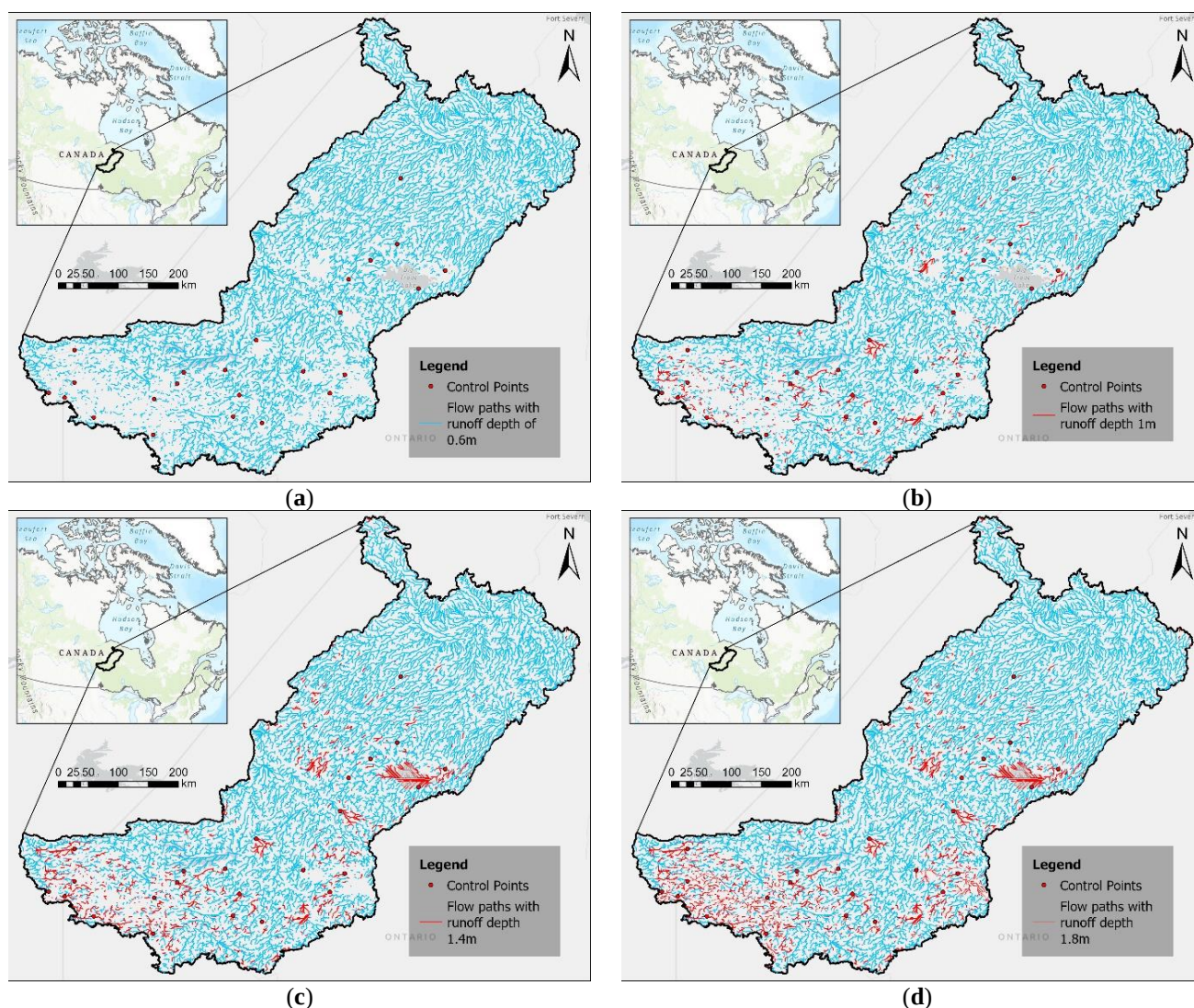


Figure 4: Increasing drainage pattern and control points in Severn River watershed of CS on increasing runoff depth: (a) 0.6m; (b) 1m; (c) 1.4m (d) 1.8m.

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185 Validation

186 Validation of the 362 identified potential control points was performed visually using ArcGIS base map imagery and
 187 Planet imagery. Most of the identified control points (~91%) were located in rocky cascades, lake sills, moist swales,

188 intermittent streams, dams, road embankments, beaver dams and other visually evident features of intermittent and/or controlled
 189 water flow (Figure 7). An initial visual inspection of each location was done in ArcGIS basemap. Based on this inspection
 190 each point was either assigned to a respective descriptive category (Rocky Cascade, Dry path, Sills, dam, beaver dam, farm
 191 land, bridge, narrow stream total = 342 locations). Others were narrow rivers and streams erroneously detected as control
 192 points owing to the limitations of the 90m resolution of MERIT DEM and were removed from the dataset (total = 20 locations).
 193 Control points that could not be positively in ArcGIS basemap were verified using multi-temporal Planet imagery, by
 194 examining images during both high- and low-flow conditions. A total of 362 control points were detected between 0.6m and
 195 0.8m runoff depth.

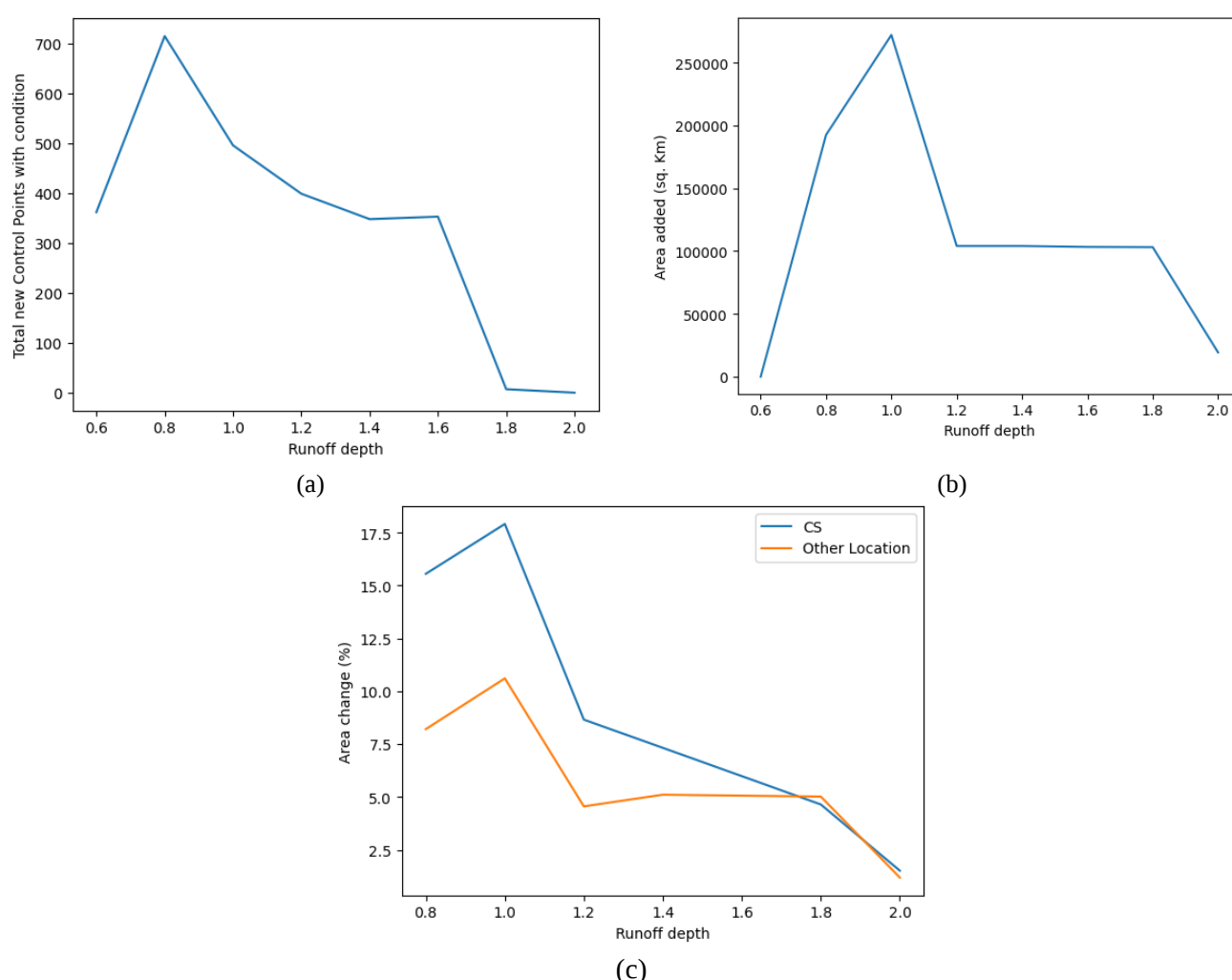


Figure 5: Number of control points detected and watershed area increase with progressive increase of runoff depth starting from 0.6m up to 2m: (a) New control points formed in various runoff depths when we applied the condition of more than 20% or 5000 sq. km; (b) The percentage of watershed area increasing with increasing runoff depth for different places of Canada shows there is a huge increase in runoff depth in CS (20 % in 0.8-1.0m) and less in other parts of Canada; (c) watershed area added with increasing runoff depth.

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197 **Results**

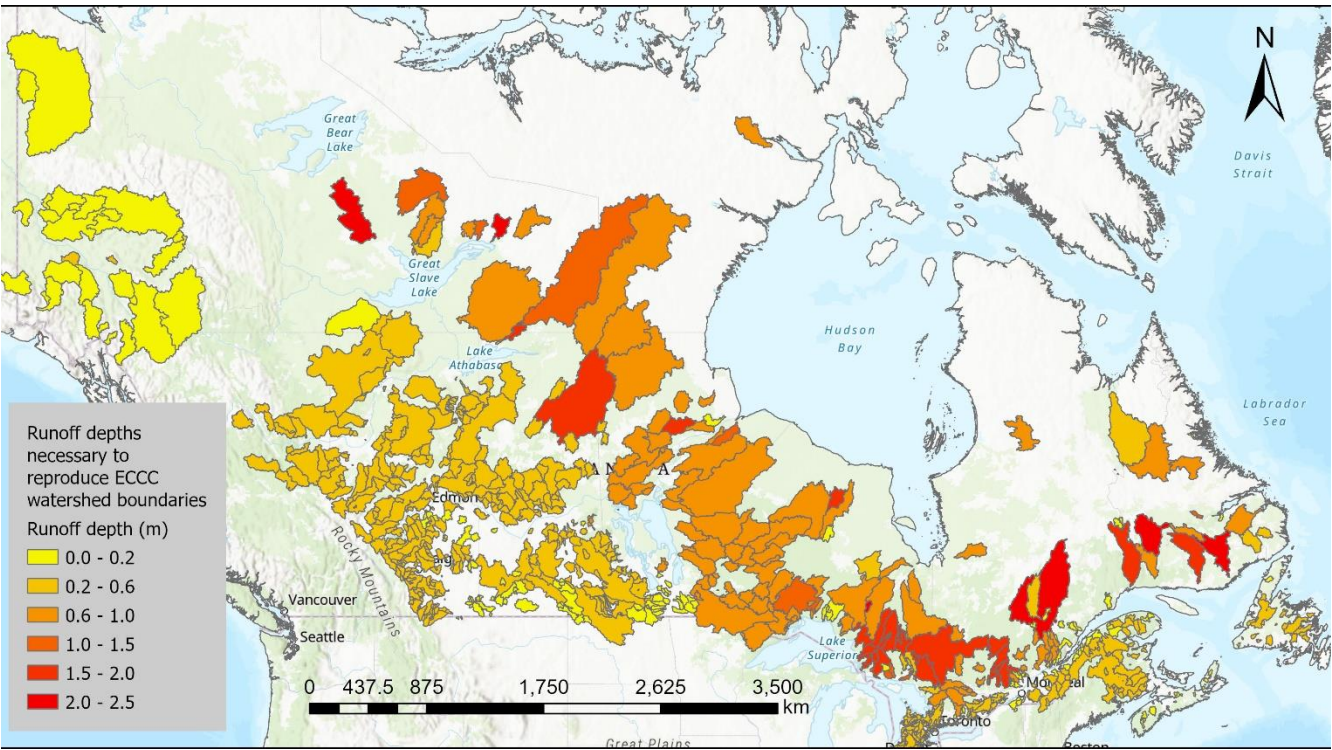


Figure 6: Different ECCC watershed has different depressional storage. This figure shows the runoff depth required to match up with the ECCC watershed area. From this, we can see CS needs more runoff depth because of fill-and-spill hydrology.

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199 Progressive DEM filling with FSM identifies As anticipated derived drainage patterns expand as MERIT DEM is
200 progressively infilled with runoff depths above 0.6 m (Figure 4), and sometimes abruptly in fill-and-spill landscapes (e.g. the
201 Severn River watershed on the Canadian Shield, Figure 4). Control points influencing at least 20% or 5,000 km² of upstream
202 watershed area were mostly found within 0.6m – 1.0m. Above 1.0 m, the number of new control points encountered decreases
203 (Figure 5). Watershed area increases largely track the number control points (Figure 5b). Watershed area increases are nearly
204 20% for the CS, roughly double other parts of Canada (~10%). Most of these increases occur within the first 1 m of inundation,
205 gradually declining out to 2 m (Figure 5c). As expected, the total depression storage varies strongly across Canada, with the
206 greatest filling required in the CS (Figure 6, Figure 5c). This affirms the need to use a range of runoff depths to identify control
207 points, as well as greatest depression storage within the CS physiographic region.

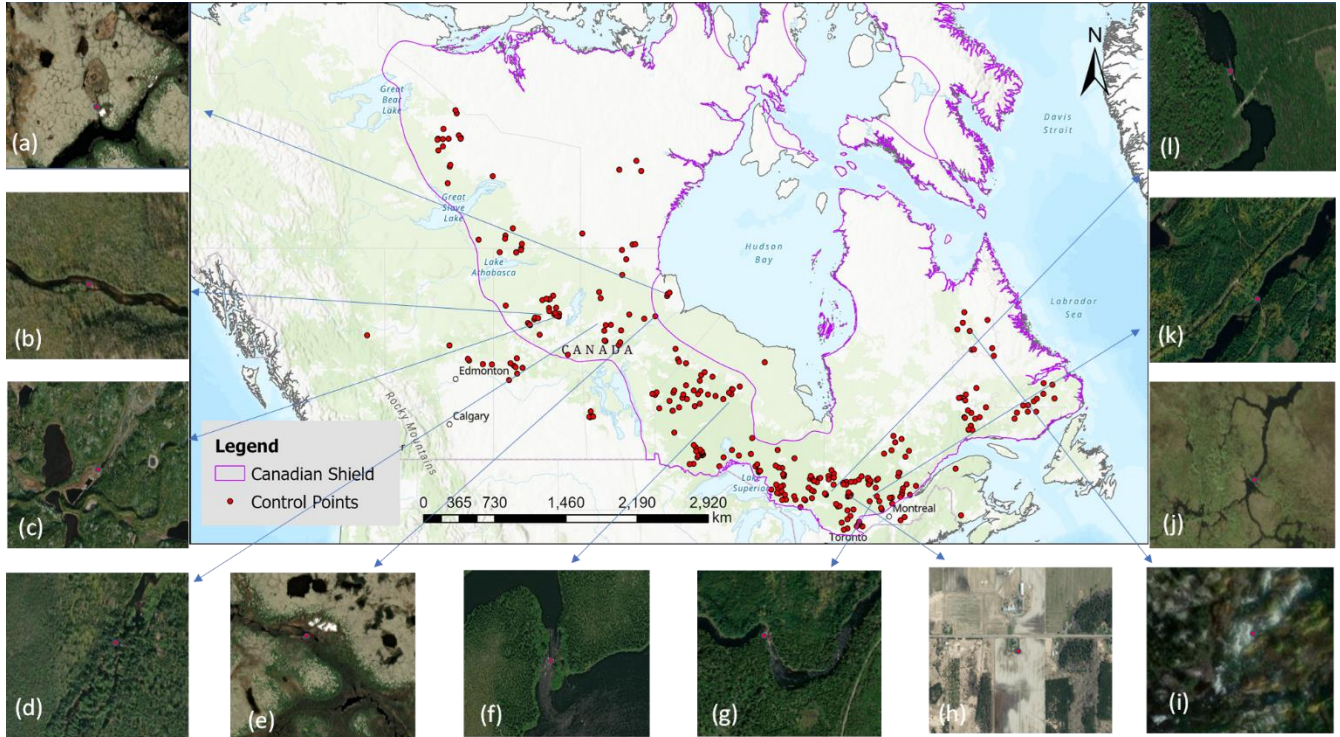


Figure 7: 326 control points of Canada detected using our algorithm when filling with increasing fill with 0.2 m starting from 0.6m: (a) A dry River; (b) Sill; (c) Dry Stream; (d) Sill; (e) River; (f) Cascade; (g) Cascade; (h) farm; (i) Cascade; (j) Beaver Dam; (k) Beaver Dam; (l) Bridge;

Out of 362 major control points identified throughout Canada, 290 (~80%) are in the CS and most of the rest are in topographically similar terrains resembling the CS (Figure 6). From the visual inspection of ArcGIS Basemap and satellite imagery, we removed those points that were in the river without any obstruction. We also removed Dams, Bridges, Beaver Dams, Farms, and Human intervention from the datasets. These types of obstruction are shown in Figure 5. After removing those we had 255 control points. Among the remaining control points, 147 are cascades (~56%), and 57 are protruded bedrock sills (29%) but not cascade. 32 are dry paths (12.54%) and the remaining are some points between lakes and intermittent streams.

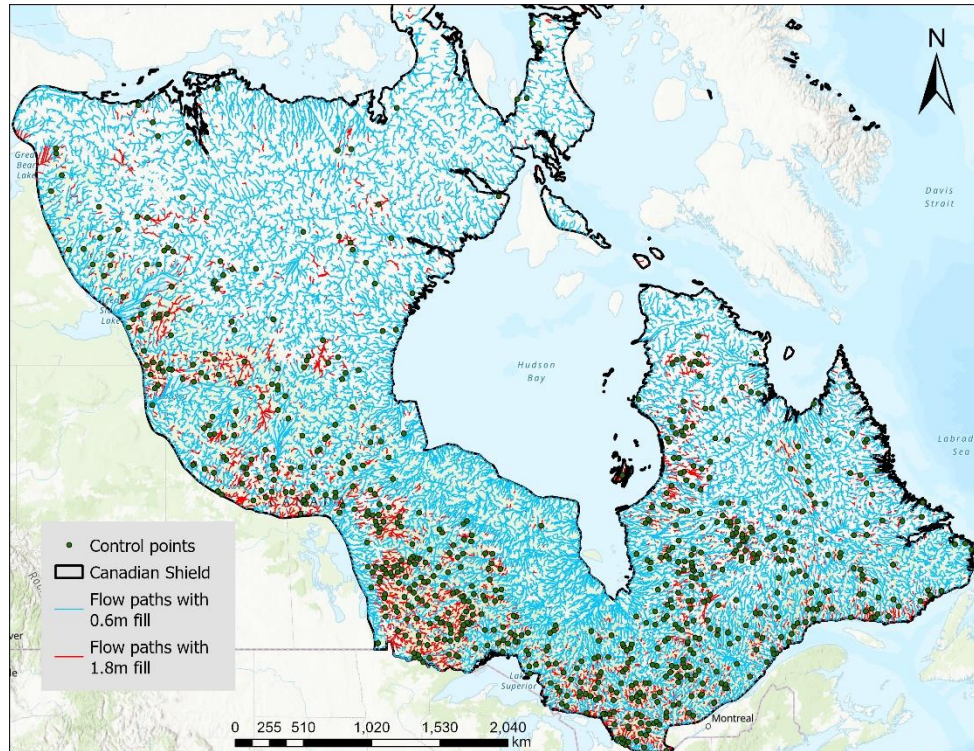


Figure 8: Control points in CS with flow path at 0.6m fill in blue and increased flow path in 1.8 m fill in blue. This shows fill-and-spill hydrology of CS

The CS is characterized by fill-and-spill hydrology, especially in Northwest Territories, Manitoba, and Saskatchewan (Figure 7). A total of 21% of ECCC gauged watersheds contain control points influencing at least 20% or 5000 sq. km particularly Geiki River, Seal River, Kazan River, and Tazin River. By applying FSM to the entire CS domain, we identify 763 control points (including 362 lying within ECCC watersheds) (Figure 8). These control points mostly lie in the rivers (e.g. rocky cascades). There are also numerous beaver dams, dams, and bridges that do not necessarily restrict streamflow on the CS.

Discussion and Conclusion

Depressional landscapes contain internally drained basins and topographic depressions that store water and fragment regional drainage patterns. Traditional DEM-based watershed analysis approaches commonly override these features by overfilling DEMs to create continuous, well-integrated drainage patterns that may or may not reflect reality. FSM terrain modeling offers a novel approach for detecting topographic controls on regional and continental-scale drainage patterns, including bedrock sills, rocky cascades, and intermittent flow paths that can switch on or off depending on water levels. By

229 applying FSM to the 90m MERIT DEM for all of Canada, we find that depressional landscapes are common in certain areas,
230 most notably in the Canadian Shield and some topographically similar areas in northern Alberta, Manitoba, and Saskatchewan.
231 This broad-scale analysis thus affirms previous studies reporting fill-and-spill hydrology in the CS [18,19,36–39], as well as
232 heterogeneity in depression storage[40] and the importance of hydrological connectivity to runoff processes [41–43].

233 A classical challenge in watershed analysis is determination of a DEM filling parameter[44,45]. This filling parameter
234 should sufficiently large to reduce DEM artifacts, pits, and error but not so large as to remove real-world topographic
235 depressions. Our pairing of climate data with gauging station discharge records (Figure 3) offers a novel, physically-based
236 rationale for estimating a plausible initial value of this critical parameter. Depending on project aims, modest increases (or
237 decreases) in this parameter are likely owing to variable depression storage for different physiographic regions of Canada
238 (Figure 6). Greatest filling is necessary in the CS, where depression storage of currently gauged watersheds is here estimated
239 to be as high as 2.0m.

240 Our application of FSM across 1,690 gauged Canadian watersheds identifies 255 control points with the potential to
241 influence the actively flowing area of the watershed. An estimated 323 of 1623 (~12%) of gauged watersheds contain one or
242 more control points influencing at least 20% and/or 5,000 km² of the affected area. Such variable source areas may explain
243 irregular river hydrograph behavior and/or unusual flood risk statistics downstream. Control points may also contribute to
244 variable sediment and/or geochemical loads downstream. Based on our first analysis here, we suggest the Gieki, Snare, Kazan,
245 and Seal rivers may be particularly affected as 20-25% of their watersheds are potentially impacted by flow control points.

246 Most of the flow control points identified in this study are activated with sub-meter increases in regional water balance
247 (e.g., 0.6-0.8m, Figure 4). A total of 91 percent of (235 out of 255) control points identified in our dataset appear to be active
248 or occasionally active today, even at runoff depths as high as 1.8m.

249 A primary limitation of our work is the 90m spatial resolution of MERIT DEM. Smaller channels, narrow connections
250 between lakes and other fine-scale features are not well-resolved at this resolution so our analysis is necessarily
251 regional/continental in scale. Similarly, topographical slope breaks in continuous river channels sometimes erroneously flagged
252 as control points in MERIT DEM, because the channel was not resolved. While computational expense currently prohibits
253 analysis of higher-resolution DEMs (e.g. ArcticDEM) at continental scales, future studies should consider using ArcticDEM
254 or other high-resolution DEMs for local or regional studies.

255 In conclusion, we find that FSM improves drainage pattern delineations of Canadian watersheds, by striking an
256 optimal balance between terrain smoothing vs. information loss in DEM analysis. Through a continental-scale analysis of the
257 MERIT DEM, we identify at least 255 control points in currently gauged Canadian watersheds, and another 508 control points
258 in other, ungauged areas of the Canadian Shield. These natural features likely connect intermittently to downstream flow
259 networks, complicating their fluxes of water, sediment, and geochemical loads. While acknowledging the limitations of DEM
260 analysis, our research highlights the value of nuanced, physically-based terrain modeling for understanding the surface water
261 hydrology of poorly drained landscapes with abundant depression storage.

262 **Author Contributions**

263 **Conceptualization:** Nimisha Wagle; Laurence C. Smith

264 **Funding procurement:** Laurence C. Smith

265 **Data curation:** Nimisha Wagle

266 **Formal analysis:** Nimisha Wagle

267 **Investigation:** Nimisha Wagle, Laurence C. Smith

268 **Methodology:** Nimisha Wagle; Laurence C. Smith

269 **Software:** Nimisha Wagle

270 **Supervision:** Laurence C. Smith

271 **Validation:** Nimisha Wagle

272 **Writing – original draft:** Nimisha Wagle; Laurence C. Smith

273 **Writing – review & editing:** Nimisha Wagle, Laurence C. Smith

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277 **Data Availability Statement**

278 All data are available in USGS EarthExplore and EarthData. The data URL is provided in the Main text itself.

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281 journey.

282 **Conflicts of Interest**

283 The authors declare no conflict of interest.

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