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1	DOES FLUVIAL CHANNEL BELT CLUSTERING PREDICT NET SAND TO GROSS
2	ROCK VOLUME?
3	ARCHITECTURAL METRICS AND POINT PATTERN ANALYSIS OF A DIGITAL
4	OUTCROP MODEL
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14	ABSTRACT
15	Spatial point pattern analyses (PPAs) are used to quantify clustering, randomness, and
16	uniformity of the distribution of channel belts in fluvial strata. Point patterns may reflect end-
17	member fluvial architecture, e.g., uniform compensational stacking and avulsion-generated
18	clustering, that may change laterally, especially at greater scales. To investigate spatial and
19	temporal changes in fluvial systems, we performed PPA and architectural analyses on extensive
20	outcrops of the Cretaceous John Henry Member of the Straight Cliffs Formation in southern
21	Utah, USA. Digital outcrop models (DOMs) produced using unmanned aircraft system-based
22	stereophotogrammetry form the basis of detailed interpretations of a \sim 250 m-thick fluvial
23	succession over a total outcrop length of 4.5 km. The outcrops are oriented roughly

perpendicular to fluvial transport direction. This transverse cross-sectional exposure of the 24 25 fluvial system allows a study of the system's variation along depositional strike. We developed a 26 workflow that examines spatial point patterns using the quadrat method, and architectural 27 metrics such as net sand to gross rock volume (NTG), amalgamation index, and channel belt 28 width and thickness within moving windows. Quadrat cell sizes that are $\sim 50\%$ of the average 29 channel belt width-to-thickness ratio (16:1 aspect ratio) provide an optimized scale to investigate 30 laterally elongate distributions of fluvial channel belt centroids. Large-scale quadrat point 31 patterns were recognized using an array of 4 quadrat cells, each with 237x greater area than the 32 median channel belt. Large-scale point patterns and NTG correlate negatively, which is a result 33 of using centroid-based PPA on a dataset with disparately-sized channel belts. Small-scale 34 quadrat point patterns were recognized using an array of 16 quadrat cells, each with 21x greater 35 area than the median channel belt. Small-scale point patterns and NTG correlate positively, and 36 match previously observed stratigraphic trends in the fluvial John Henry Member, suggesting 37 that these are regional trends. There are deviations from these trends in architectural statistics 38 over small distances (100s of meters) which are interpreted to reflect autogenic avulsion 39 processes. Small-scale autogenic processes result in architecture that is difficult to correlate 40 between 1D datasets, for example when characterizing a reservoir using well logs. We show that 41 1D NTG provides the most accurate prediction for surrounding 2D architecture.

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INTRODUCTION

44 The architecture of fluvial deposits records complex interactions between allogenic and
45 autogenic processes. Quantified parameters that are commonly used to help define stratigraphic
46 architecture in fluvial deposits include net sand to gross rock volume (NTG; 'N' in Lake and

47	Jensen 1991), channel belt amalgamation (Allen 1979), and channel belt width and thickness
48	(Gibling 2006), as well as paleomorphodynamic reconstructions based on grain size and
49	estimates of bankfull depth (Paola and Mohrig 1996; Hajek and Heller 2012; Benhallam et al.
50	2016). Sandstone body density and interconnectedness (essentially NTG and amalgamation,
51	respectively) were the key subjects of early modeling work by Leeder (1978), Allen (1978), and
52	Bridge and Leeder (1979), known collectively as the LAB models (Paola 2000). These two
53	architectural characteristics of fluvial strata strongly impact reservoir quality by determining the
54	volume of connected porous rock capable of hosting hydrocarbons or water (Larue and Hovadik
55	2006). Models of fluvial architecture are frequently 2D or 3D while subsurface resource
56	investigations often rely primarily on 1D borehole data. Detailed digital outcrop datasets allow
57	us to connect 1D data to 2D outcrop architectural data and expand to 3D architectural prediction.
58	Sequence stratigraphic models link fluvial architecture with different stages of a relative
59	sea level curve, notably predicting valley incision during lowstand regression which can later be
60	filled by clustered channel belts (Posamentier and Vail 1988; Shanley and McCabe 1993; Wright
61	and Marriott 1993). These models broadly infer allogenic controls of climate, tectonics, and
62	eustasy based on fluvial architecture. Subsequent work building on the LAB models
63	demonstrated the importance of autogenic processes such as avulsion in the development of
64	fluvial architecture (Heller and Paola 1996; Hajek et al. 2010; Hajek and Straub 2017).
65	Additionally, models of distributive fluvial systems (DFSs) (Weissmann et al. 2010), suggest
66	predictable trends in NTG, amalgamation, and channel belt size, dependent on position relative
67	to the DFS axis and its evolution over time (Owen et al. 2015).
68	Point pattern analysis (PPA) is a class of methods used to categorize spatial point patterns

69 and classify points as randomly, clustered, or uniformly spaced (Cressie 1993). PPA has been

70 applied to test for clustering in fluvial architecture and to compare outcrops to conceptual and 71 physical models (Hajek et al. 2010; Hofmann et al. 2011; Flood and Hampson 2015; Chamberlin 72 et al. 2016; Benhallam et al. 2016). Results from PPA quantify the spatial arrangement of points 73 in 2D space (Cressie 1993). PPA is typically applied to fluvial deposits by treating each channel 74 belt as a single centroid point and then analyzing their distribution over the study area (Hajek et 75 al. 2010). Reducing complex channel belts to their centroid points is necessary using current 76 PPA methods, but is an oversimplification and therefore potentially limits the usefulness of PPA 77 (Hajek et al. 2010; Benhallam et al. 2016). For example, an asymmetric channel belt may have a 78 centroid point that does not serve as an accurate center of mass for that sand body, and thus 79 could not be used to describe NTG. However, the specific consequences of simplifying channel 80 belts to centroid points are relatively unknown.

81 Clustering of channel belts has been attributed to allogenic controls, such as valley 82 incision caused by base-level fall (e.g., Shanley and McCabe 1993) as well as autogenic controls, like avulsion (Hajek et al. 2010; Hajek et al. 2012; Huling and Holbrook 2016). Uniformity of 83 84 channel belts and of channel belt clusters may reflect compensational stacking, whereby a river 85 preferentially occupies the lowest topography on the floodplain, avoiding previous channel belts 86 or channel belt clusters (Mohrig et al. 2000; Straub et al. 2009; Hofmann et al. 2011; Chamberlin 87 et al. 2016). Point patterns, and to a large extent architectural metrics, have previously been used 88 to highlight stratigraphic trends and their variation down-depositional-dip (Flood and Hampson 89 2015, Benhallam et al. 2016).

Leveraging a well-studied outcrop of the fluvial John Henry Member in southern Utah,
this study seeks to: 1) quantify lateral and vertical trends in channel belt NTG, degree of
amalgamation, width and thickness, and spatial point pattern using 4.5 km of depositional-strike-

93 oriented outcrop to compare to previously studied outcrops along depositional-dip, 2) quantify 94 the correlation between NTG and clustering, 3) describe specific characteristics of a fluvial 95 dataset that could lead to more reliable results when using centroid-based PPA, and 4) test the 96 limitations of 1D prediction by correlating 1D metrics to corresponding 2D architectural metrics. 97 The results of this study can be used to guide 2D and 3D facies modeling of fluvial strata, which 98 is highly useful in the exploration and development of hydrocarbons and water. This study adds 99 to a growing body of literature which utilizes DOMs to supplement traditional outcrop data.

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GEOLOGIC BACKGROUND

102 The Cretaceous John Henry Member of the Straight Cliffs Formation (southern Utah, 103 USA) serves as an excellent laboratory to evaluate the spatial trends in fluvial channel belt 104 architecture due to a long (~40 km) continuous depositional-strike-oriented outcrop exposure of 105 fluvial deposits along the south-western margin of the Kaiparowits Plateau, known as the 106 Cockscomb (Fig. 1). Numerous studies provide excellent sedimentologic and stratigraphic 107 context for these strata and their down-depositional-dip paralic and marine equivalents (Peterson 108 1969a, 1969b; Shanley and McCabe 1991, 1993, 1995; Allen and Johnson 2010a, 2010b, 2011; 109 Gallin et al. 2010; Szwarc et al. 2015; Benhallam et al. 2016; Johnson et al. 2016; Gooley et al. 110 2016; Mulhern and Johnson 2017).

The John Henry Member was deposited in a retroarc foreland basin from mid-Coniacian to early Campanian time. Accommodation was generated by the Sevier fold-thrust belt and dynamic topography (Liu and Nummedal 2004; Painter and Carrapa 2013), with sediment supplied from the Mogollon Highlands to the south and the Cordilleran magmatic arc to the southwest, in addition to the orogenic belt to the west (Szwarc et al. 2015; Primm et al. 2017).

116	Down-depositional-dip trends in fluvial architecture have been studied in detail by
117	documenting stratigraphic trends in fluvial style and analyzing channel belt point patterns from
118	Rock House Cove to Bull Canyon (~20 km distance; Fig. 1; Gooley et al. 2016; Benhallam et al.
119	2016). Gooley et al. (2016) divided the John Henry Member into 7 depositional units (DUs)
120	based on facies associations and fluvial architecture, and observed 2 consistent stratigraphic
121	trends. Trend 1 shows a decrease in grain size, channel-belt frequency, and width of channel
122	belts up through stratigraphy from DU-1 to DU-3, with a shift in channel geometry from braided
123	to meandering with tidal influence. Trend 2 shows an increase in grain size, channel-belt
124	frequency, and width of channel belts up through stratigraphy from DU-4 to DU-6, with a shift in
125	channel geometry from meandering to braided (Gooley et al. 2016). This previous work
126	establishes excellent context for a detailed, statistical analysis of fluvial architecture in the John
127	Henry Member, particularly looking for the first time at trends laterally in the fluvial system
128	(~N-S) rather than down-depositional-dip.
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130	DATASET AND METHODS
131	Data Collection, Model Generation and Interpretation
132	The dataset for this study consists of two digital outcrop models (DOMs) of the John
133	Henry Member, CC1 and CC5 (Fig. 1), and three measured stratigraphic sections (Fig. 1, CCX.
134	1-3). Lithology, grain size, sedimentary structures, bedding geometry, paleocurrent indicators,
135	and cross-bed thicknesses were recorded for each measured section (supplemental data in Koch
136	2018). The basal and upper contacts of the John Henry Member are well-defined, regional
137	correlation surfaces in this area (Figs. 2, 3; Primm et al. 2017; Gooley et al. 2016; Peterson
138	1969b).

139 An unmanned aircraft system, the DJI Phantom 3 Professional quadcopter, was used to 140 collect georeferenced photographs of each outcrop. CC1 is one continuous model created from 141 378 photographs (Fig. 2) and is approximately 1.4 km long and 275 meters in thickness. CC5 is 142 composed of two overlapping models made from 288 and 343 photographs (Fig. 3) and is 143 approximately 3.1 km long and 220 meters in thickness. Photos were collected by flying a grid 144 pattern across each outcrop, maintaining on average 70% overlap between adjacent photos. 145 DOMs were generated with Agisoft Photoscan Professional (version 1.2.5) using 146 Structure-from-Motion stereophotogrammetry, a method by which common points are identified 147 in images taken from different locations, and 3D locations of those points are triangulated 148 (Westoby et al. 2012). This process is repeated for every common pixel from each photo of the 149 dataset to generate a 3D point cloud. A tiled model is generated from the 3D point cloud which 150 incorporates pixels from the source photographs resulting in a photorealistic 3D surface. The 151 DOMs were georeferenced using remotely selected control points collected with a differential 152 GPS (Trimble GeoExplorer 6000) paired with a laser rangefinder (TRUPULSE 360°B). Rather 153 than using traditional ground control targets, features on the outcrop were targeted with the laser-154 dGPS pair and their position was then photographed and annotated. Detailed descriptions of the 155 entire DOM workflow can be found in Westoby et al. (2012). The average resolution of each 156 DOM is 3.5 cm/pixel, such that most decimeter-scale sedimentary features are readily 157 identifiable. 158 In the DOMs, 104 channel belts were interpreted in outcrop CC1, and 265 in outcrop 159 CC5 (Fig. 6). Channel belts were interpreted in Agisoft Photoscan (version 1.2.5) using the

161 surfaces were interpreted using the hierarchical methods of Miall (1988), where lower order

polyline tool by outlining prominent sandstone bodies in the model. Channel belt bounding

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surfaces are truncated by higher order surfaces. Channel belts are bound by 5th order surfaces: a 162 163 laterally persistent erosive surface on the base, and a contact with floodplain facies or the erosional base of an overlying channel belt on the top (Miall 1988). Amalgamated 5th order 164 surfaces between two distinct channel belts can be confused with 4th order surfaces, which result 165 166 from major shifts in channel or bar migration without the channel abandoning its position. Distinguishing 5th from 4th order surfaces in the DOMs posed a challenge, and in highly 167 168 amalgamated areas it is likely that some channel belt interpretations represent multiple amalgamated channel belts (Fig. 4). Where present, lower order surfaces (e.g., 2nd and 3rd order) 169 170 aided in distinguishing distinct packaging. Measured sections aided minimally in making surface 171 distinctions due to their limited intersection with the much larger model areas, however they 172 provided ground-truthing where possible.

173 The strata of outcrops CC1 and CC5 are dipping between 30-40° to the E-SE. In order to 174 correct polyline interpretations for their structural orientation, they were exported from Agisoft 175 as shapefiles, and imported into Schlumberger's Petrel (Version 2015). In Petrel, they were 176 translated and rotated until the bedding orientation was perpendicular to the horizon. The rotated 177 polylines were again exported as a shapefile and imported into ESRI ArcMap (Version 10.3) to 178 be projected in 2D, resulting in a clean cross-section of the polyline interpretations. The 179 polylines were then corrected for broad folding (maximum limb dip of 1.9°) along the outcrop in 180 order to make them as consistently horizontal as possible. Polylines were cut at the peak of a 181 fold, rotated until horizontal, and rejoined. To determine where bedding was truly horizontal we 182 followed the methods of Calvo and Ramos (2015), identifying flat channel belt tops as datums 183 for the section. After structural correction of folding and 2D projection, the DOMs still have 184 distortion related to the topography of the outcrop, such that some amount of error can be

expected in all calculations and analyses. Current software is not capable of manipulating DOMs
to fully compensate for the complex interaction of strata and topography. Polylines were
converted to closed polygons, and their centroid points were calculated in ArcGIS. The
structurally corrected channel belt polygons, including their centroids, the 2D area of each
polygon (Fig. 5A), contact polylines between amalgamated channel belts, and channel belt basal
polylines (Fig. 5B) were all exported as shapefiles from ArcMap, and imported into MATLAB
for analysis.

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Definition of Architectural Metrics

194 **Net Sand to Gross Rock Volume (NTG).**—NTG is calculated by:

$$NTG(2D) = \frac{\sum_{i=1}^{n} A_i}{A_T}$$
(1.1)

where A_i is the transverse cross-sectional area of channel belt *i* out of *n* channel belts, and A_T is the total area over which the analysis is performed, which in this study is a moving window (Fig. 5A). This analysis is performed in 2-dimensional (eq. 1.1), or 1D (eq. 1.2):

$$NTG(1D) = \frac{\sum_{i=1}^{n} Th_i}{Th_T}$$
(1.2)

where Th_i is the thickness of sandstone bed in a 1D section divided by the total thickness of the moving window, Th_T . *NTG* is presented in this study as a dimensionless fraction between 0 and 1, where 0 is 0% sandstone and 1 is 100% sandstone. This is a simplification of the true outcrop lithology because there are minor tabular sandstone beds in the overbank deposits, and there are minor mudstone beds within the channel deposits. At this scale of investigation, the vast majority of each channel fill is composed of sandstone (see measured sections in Figs. 2, 3).

206 Channel Belt Amalgamation Index.— An index to quantify the degree of amalgamation
207 of channel belts (after Funk et al. 2012; Peter et al. 2017):

$$AI(2D) = \frac{\sum_{j=1}^{m} LAmalgamated_j}{\sum_{i=1}^{n} LBase_i}$$
(1.3)

209 Where *LAmalgamated*_i is length of contact between two channel belts and is a pairwise 210 statistic performed over m pairs where m = n - 1 contacts, and when summed, equals the total 211 length of all amalgamation contacts (Fig. 5B). *LBase_i* is the length of the channel belt bases *i* 212 over *n* channel belts, and when summed, is the total length of all channel belt bases. AI, then, is 213 the ratio of the length of amalgamated contacts over the total length over which amalgamation 214 could occur, or the proportion to potential amalgamation. This analysis is performed in 2-215 dimensional (eq. 1.3), or 1D (eq. 1.4): $AI(1D) = \frac{m}{n}$ 216 (1.4)217 Where m is the number of amalgamation contacts and n is the total number of channel 218 belt bases encountered in the 1D section. AI is presented in this study as a dimensionless fraction 219 between 0 and 1. 220 Channel Belt Width and Thickness.—Channel belt width is the maximum horizontal 221 extent of the channel belt, and is measured by subtracting the minimum x-coordinate from the 222 maximum x-coordinate of each channel belt polygon. Channel belt thickness is calculated as the 223 maximum vertical distance between two points on the top and base of the polygon having equal 224 x-coordinates (Fig. 5C). Paleocurrent indicators collected for this study (n = 251) primarily from 225 trough cross-bed axes and dune accretion surfaces indicate an average paleocurrent direction of 226 109.2° , with a standard deviation of 2.1° (Fig. 1). The mean direction is nearly normal (89.2°) to 227 the average strike of the outcrop faces (020°) , indicating that apparent widths, used for the 228 entirety of this study, are close to true channel belt widths.

Vertical and Lateral Trends in Architectural Metrics

231	A 600m x 25m "moving window" is stepped over the data, in 50m lateral and 5 m
232	vertical increments to quantify spatial trends in architectural metrics (Fig. 6i). A 1D observation
233	was also collected from a vertical line positioned at the center of each moving window. For each
234	lateral moving window position, an R^2 value was calculated between the vertical trends of 1D
235	and 2D architectural metrics, to quantify how well a 1D (e.g., borehole) dataset characterizes the
236	architecture laterally for 300 m on either side. A separate moving window, 25 m thick and
237	spanning the entire width of the outcrop, was used to analyze gross vertical stratigraphic trends,
238	also moving vertically by 5 m increments (Fig. 6ii). For all metrics, channel belts were cropped
239	to the moving window, except for channel belt width calculations. For channel belt thickness
240	measurements, channel belts were cropped to the moving window laterally but not vertically.
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242	Point Pattern Analysis of Channel Belt Distribution
243	Point pattern analysis characterizes the distribution of a set of points over a finite 2D
244	area. Each method tests the hypothesis of complete spatial randomness (CSR; Cressie 1993)
245	which asserts that a random set of points in a finite 2D area will have a Poisson distribution, i.e.,
246	each point location is independent of all other points, and has equal probability of being at any
247	location in the study area. The centroid point of each channel belt polygon is used as the input
248	point to understand the distribution of channel belts across each outcrop. We use the quadrat
249	method in this study for its compatibility with a moving window analysis, which allows for
250	comparison to results from architectural data.
251	We use the methods of Benhallam et al. (2016), who presented the first application of the

253 study area (i.e., moving window area) with a grid of equally-sized rectangular cells, and counts 254 the number of points within each cell. The variance of points within each rectangle is then 255 divided by the mean to produce the cluster index. When variance and mean are equal (clustering 256 index = 1) points are distributed randomly, when variance is greater than the mean (clustering 257 index > 1) points are clustered, and when variance is less than the mean (clustering index < 1) 258 points are distributed uniformly (Fig. 7). A robust way to compare results to those that could 259 have been produced by a random process is to generate 99 Monte Carlo simulations where the 260 number of simulated points and the simulated 2D area are both equal to that of the measured 261 dataset. Using the same moving window and quadrat cell size, the quadrat analysis is performed 262 for each Monte Carlo simulation. The range of cluster indices produced by 99 simulations is 263 considered possible by random process, and is used to determine the statistical significance of a 264 cluster index produced by the measured data. Cluster indices from the actual data are considered 265 uniformly distributed if they are lower than the simulation envelope, random if they are within 266 the simulation envelope, and clustered if they are greater than the simulation envelope. Point 267 patterns of channel belts groups are anisotropic, loosely matching the elongate geometry of the 268 channel belts themselves (Fig. 6). To introduce anisotropy to the quadrat method, the cells of the 269 quadrat grid are manipulated to have the desired aspect ratio (Fig. 6iii, iv, v).

The quadrat method is performed within moving windows to analyze vertical and lateral trends. We used three different scales – small, medium, and large – of quadrat cells (Fig. 6iii, iv, and v respectively). Quadrat cells for all scales are arranged in two rows. Table 1 summarizes the various moving window and quadrat cell dimensions for each scale of analysis. Just as for the architectural analysis, the moving window begins at the base of the outcrop at the northern end. After each quadrat analysis, the window is stepped vertically by 5 m, until the top of the window

276	reaches the top of the outcrop. The window is then returned to the base of the outcrop, and steps
277	50 m laterally to the south. Vertical and lateral movement of the window continues in this pattern
278	until the window has traversed the entire outcrop. To assess the similarity or difference between
279	NTG and clustering we use Pearson's R^2 correlation coefficient. For each quadrat cell scale
280	described above - small, medium, and large - NTG was averaged over the entire quadrat cell
281	grid (i.e., moving window). The vertical moving window results of quadrat cluster index and
282	NTG at the first lateral moving window position are used to compute an R^2 value. This method is
283	repeated for each lateral moving window position, generating a series of R^2 values which
284	represent how well the two characteristics' vertical trends match with changing lateral position
285	across the outcrop.
286	
287	RESULTS
288	Architectural Trends
289	General Architectural Trends.—The John Henry Member has broadly similar
290	architectural trends at outcrops CC1 and CC5, with stratigraphic and lateral variability both
291	between and within outcrops (Figs. 6, 8). Previous work divided the John Henry Member at
292	Rock House Cove (6.5 km south of CC5) and Bull Canyon (17 km E-SE of CC5) into 7
293	depositional units (DUs) on the basis of architectural and sedimentological characteristics
294	(Gooley et al. 2016; Benhallam et al. 2016). Similar trends and units are present in the study
295	area, but are simplified into three main units (dashed lines in Fig. 6) which correspond to the
296	DUs of Gooley et al. (2016) as follows: lower = DUs 0-2, middle = DUs 3-4, and upper = DUs
297	5-6. In outcrops CC1 and CC5, the lower and upper units have greater average NTG and
298	amalgamation index, with wider and thicker channel belts. In contrast, the middle unit tends to

have lower average NTG and amalgamation index, with narrower channel belts, which isconsistent with the previous interpretations of Gooley et al. (2016).

Each outcrop was divided into lateral segments to facilitate the detailed description of lateral variability. CC1 is referred to in terms of northern and southern segments, and CC5 is referred to as northern, central, and southern segments (vertical hashed lines, Fig. 8). Moving window analyses highlight and quantify outcrop trends of NTG, amalgamation index, and channel belt width and thickness (Fig. 8A-H) whereas gross vertical stratigraphic analyses (Fig. 8I-P) show laterally-averaged trends in architectural metrics.

307 Net-to-Gross (NTG).—Window-averaged NTG for CC1 ranges from 0.04 to 0.76. The
308 CC1 gross vertical analysis (Fig. 8I) reveals two key trends; 1) an upward decrease from the
309 lower unit into the middle unit, and 2) an upward increase from the middle unit to the upper unit.
310 The CC1 moving window analysis (Fig. 8A) reveals minor lateral variation where the lower unit
311 decreases in NTG from north to south, and the upper unit increases in NTG from north to south
312 (Fig. 9).

313 Window-averaged NTG for CC5 ranges from 0.002 to 0.56, with less overall NTG than 314 CC1. The CC5 gross vertical analysis (Fig. 8J) reveals three key NTG trends; 1) a decrease from 315 the lower unit to the middle unit, 2) an intermediate peak within the middle unit, and 3) an 316 upward increase to the upper unit (Fig. 9). The CC5 moving window analysis (Fig. 8B) reveals 317 high lateral variability primarily in the middle and upper units. The middle unit has isolated 318 packages of high NTG in the central and southern segments. The upper unit has increasing NTG 319 from north to south. Thus, although overall NTG trends are similar across the study area, CC1 320 has less lateral variability than CC5.

Amalgamation Index.—For both outcrops, amalgamation index closely mimics the trends of NTG. Window-averaged amalgamation index for CC1 ranges from 0 to 0.65. The CC1 gross vertical analysis (Fig. 8K) reveals two key amalgamation index trends; 1) an upward decrease from the lower unit into the middle unit, and 2) and upward increase from the middle unit to the upper unit. The CC1 moving window analysis (Fig. 8C) reveals high lateral variability where amalgamation index in the lower unit decreases from north to south, and in the upper unit increases from north to south (Fig. 9).

328 Window-averaged amalgamation index for CC5 ranges from 0 to 0.65. The CC5 gross 329 vertical analysis (Fig. 8L) for amalgamation index is more sporadic and irregular than CC5 NTG, 330 however it loosely holds the same three trends; 1) a decrease from the lower unit to the middle 331 unit, 2) an intermediate peak within the middle unit, and 3) an upward increase to the upper unit 332 (Fig. 9). The CC5 moving window analysis (Fig. 8D) reveals high lateral variability in all units, 333 where amalgamation index increases from north to south. The lower and middle units contain 334 isolated packages of high amalgamation index, where the upper unit contains more laterally 335 continuous amalgamated packages.

336 **Channel Belt Thickness.**— Channel belt thicknesses for CC1 range from 1.2 m to 14.2 337 m, with a mean of 5.2 m (Fig. 10). The CC1 gross vertical analysis (Fig. 8M) reveals two key 338 channel belt thickness trends; 1) an upward decrease from the lower unit into the middle unit, 339 and 2) an upward increase from the middle unit to the upper unit. The CC1 moving window 340 analysis (Fig. 8E) reveals moderate lateral variation of channel belt thickness that mimics the 341 NTG and amalgamation index trends in the lower unit, where channel belt thickness decreases 342 from north to south. Channel belt thickness in the upper unit also decreases from north to south 343 (Fig. 9).

344 Channel belt thicknesses for CC5 range from 0.3 m to 13.1 m, with a mean of 4.2 m (Fig. 345 10). The CC5 gross vertical analysis (Fig. 8N) reveals three key channel belt thickness trends; 1) 346 a minor decrease from the lower unit to the middle unit, 2) a more significant upward increase to 347 the middle unit-upper unit contact, and 3) a decrease within the upper unit. This third trend is 348 unique to this analysis, and is caused by high lateral variability, as revealed in the CC5 moving 349 window analysis in all stratigraphic units (Fig. 8F). The lower unit channel belt thicknesses 350 decrease moderately from north to south. The middle unit channel belt thicknesses increase from 351 the northern to the central segment, then decrease from the central to the southern segment. The 352 upper unit has an inverse trend, where its channel belt thicknesses decrease from the northern to 353 the central segments, then increase from the central to the southern segments (Fig. 9).

354 **Channel Belt Width.**— Apparent channel belt widths for CC1 range from 15.6 m to 355 1340 m, with a mean of 191.5 m (Fig. 10). The CC1 gross vertical analysis (Fig. 8O) reveals two 356 key channel belt width trends; 1) an upward decrease from the lower unit into the middle unit, 357 and 2) and upward increase from the middle unit to the upper unit. The CC1 moving window 358 analysis (Fig. 8G) reveals moderate lateral variation that trends opposite the lateral variation of 359 NTG and amalgamation index, where the lower unit increases in average channel belt width 360 from north to south, and the upper unit decreases in average channel belt width from north to 361 south (Fig. 9).

Channel belt widths for CC5 range from 8.6 m to 2450.1 m, with a mean of 127.3 m (Fig. 10). The CC5 gross vertical analysis (Fig. 8P) reveals two key channel belt width trends; 1) a moderate decrease from the lower unit to the middle unit, and 2) a more significant upward increase to the upper unit. The CC5 moving window analysis (Fig. 8H) reveals moderate lateral variability primarily in the middle and upper units. The middle unit channel belt widths increase 367 from the northern to the central segments, then decrease from the central to the southern

segments. The upper unit has an inverse trend, where it decreases from the northern to the centralsegments, then increases from the central to the southern segments (Fig. 9).

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Point Pattern Analysis of Channel Belt Centroids

372 We analyzed small-, medium-, and large-scale point patterns of channel belt centroids for 373 outcrop CC5, and only medium-scale point patterns for CC1. Beginning with CC5, at small point 374 pattern scales with 16:1 aspect ratio quadrat cells (Fig. 6iii), this outcrop is dominated by 375 clustering in the lower and upper units, and by uniformity and randomness in the middle unit 376 (Fig. 11A). Small scale point patterns of CC5 have low lateral variability. The lower unit has the 377 most clustering in the southern segment, and the upper unit has the most clustering in the central 378 unit. At medium point pattern scales (Fig. 6iv), outcrop CC5 has high lateral variability with two 379 main categories of clustering; a) major clustering in the lower and middle units in the central and 380 southern segments, and b) very sparse clustering throughout the upper unit (Fig. 11C). 381 Elsewhere, the medium scale CC5 analysis is dominated by randomness with sparse uniformity. 382 At large point pattern scales with 16:1 aspect ratio quadrat cells (Fig. 6v), outcrop CC5 is 383 dominated by clustering in the lower unit and lower half of the middle unit, with very sparse 384 uniformity in the middle unit of the central and southern segments (Fig. 11B). Laterally, large-385 scale clustering has low variability, with a slight increase in abundance from north to south. 386 At medium point pattern scales with a low aspect ratio of 2.4:1, outcrop CC1 has high 387 lateral variability with two regions of high clustering; a) the middle to upper unit in the northern 388 segment, and b) the lower to middle unit in the southern segment (Fig. 11D). The CC1 quadrat

analysis also shows two regions of uniformity; a) the lower unit in the northern segment, and b)the middle unit in the southern segment.

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Correlation of Point Pattern Results to NTG

393 Correlations between quadrat point pattern results and NTG are summarized in Figure 394 11E-H. In outcrop CC5, small scale point patterns have a weak-to-moderate positive correlation to NTG, with a mean R^2 value of 0.32, and a standard deviation of 0.14 (Fig. 11E). In contrast, 395 396 large-scale point patterns have a moderate-to-strong negative correlation to NTG, with a mean R² 397 value of -0.61, and a standard deviation of 0.10 (Fig. 11F). The medium scale analyses for both 398 CC5 and CC1 have correlations with high lateral variability. Medium scale point patterns of 399 outcrop CC5 have a strong positive correlation to NTG in the north, and a moderate negative 400 correlation in the central and the southern portions of the outcrop (Fig. 11G). Medium scale point 401 patterns of outcrop CC1 have a strong negative correlation to NTG in the north, non-correlation 402 in the center of the outcrop, and a strong positive correlation in the south (Fig. 11H)

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404 Correlation of 1D and 2D Channel Belt NTG, Amalgamation Index, and Thickness. 405 Correlations between 1D and 2D channel belt NTG, amalgamation index, and thickness 406 are summarized in Figure 12. Channel belt width and point pattern cannot be measured in a 407 vertical 1D section so they were omitted for this analysis. For both outcrops, NTG has the strongest correlation between 1D and 2D, with a mean R² value of 0.93 for CC1, and 0.78 for 408 CC5. NTG also has the lowest standard deviation in R^2 values of each lateral window, at 0.02 for 409 410 CC1, and 0.09 for CC5. The metric with the second strongest correlation is different for each outcrop. For CC1, amalgamation index is second best ($R^2 = 0.68$), and for CC5 thickness is 411

second best ($R^2 = 0.54$). The least correlative features for each outcrop are thickness for CC1 (R^2 412 413 = 0.57), and amalgamation index for CC5 ($R^2 = 0.41$). 414 415 DISCUSSION 416 Methodology of Point Pattern Analysis 417 Point pattern analysis permits statistical characterization of feature clustering, and thus is 418 a potentially powerful tool for quantifying aspects of fluvial architecture (Hajek and Wolinsky 419 2012). PPA methods have been used to investigate avulsion behavior and characterize 420 stratigraphic patterns (Hajek et al. 2010; Flood and Hampson 2015; Chamberlin et al. 2016; 421 Benhallam et al. 2016). PPA methodology applied to fluvial deposits is still relatively new, and 422 only a few methods have been tested on a limited number of outcrops such as the Cretaceous 423 Ferris, Blackhawk, and Williams Fork Formations, as well as other outcrops of the John Henry 424 Member to the south and southeast of the Cockscomb (Hajek et al. 2010; Flood and Hampson 425 2015; Chamberlin et al. 2016; Benhallam et al. 2016). Outcrops CC1 and CC5 are laterally 426 extensive, permitting the analysis of point patterns and fluvial architecture at a variety of scales. 427 The results reveal dataset-size guidelines for any PPA method, provided that the data set is 428 similar in scale and channel belt density to CC1 and CC5. In total, CC1 and CC5 cover an area 4.5 km wide by 250 m tall, and have on average 4 channel belts per 100 m². Channel belt 429 430 centroids are spaced 12 m apart laterally on average. 431 To date, point pattern techniques used in fluvial stratigraphy have mostly been isotropic, 432 searching for point patterns occupying an equant space (Hajek et al. 2010; Chamberlin et al. 433 2016). In contrast, Flood and Hampson (2015) vertically exaggerated their channel belt point set 434 to incorporate anisotropy into the analysis, and Benhallam et al. (2016) used quadrat cells with

435 width-to-thickness ratios higher than 1:1. For the approach of Flood and Hampson, vertical 436 exaggeration by a factor of x results in an analysis capable of identifying point patterns with x:1 437 width-to-thickness ratio, which better matches the elongate channel belt geometry of most fluvial 438 deposits. Flood and Hampson (2015) used the ratio of mean channel belt width to mean channel 439 belt thickness as the vertical exaggeration factor. By qualitatively identifying clusters in the 440 present dataset, we determined that approximately half the ratio of mean channel belt width to 441 mean channel belt thickness (16:1 for the Cockscomb) should produce the most geologically 442 realistic results, i.e., this ratio highlights geologically relevant clusters rather than artifacts or 443 outlier signals. This aspect ratio was applied in the quadrat method following Benhallam et al. 444 (2016) by using cells of the desired aspect ratio. Further work is needed, particularly statistical 445 analysis using data from many different field examples, to empirically determine the most 446 common aspect ratios of channel belt clusters. Until such a dataset is compiled, workers applying 447 anisotropic point pattern techniques should perform sensitivity analysis on a range of aspect 448 ratios, as well as search for qualitative evidence of clusters within their dataset. 449 Scale flexibility is limited in the quadrat method, because the moving window must be 450 evenly divisible by the size of the quadrat cells in both dimensions, so tailoring the experiment to 451 the exact point pattern scale and aspect ratio can be inconvenient. However, the quadrat 452 method's requirement of fitting equally within the window boundaries means it requires no edge 453 correction. There is an inherent problem with edge effects in any PPA method, because all 454 outcrops have finite dimensions. For our purposes and for the present dataset, we argue that the 455 quadrat method is most appropriate. Our results indicate that a minimum total data frame width 456 (i.e., a moving window in this study) should be at least 500 m for the average lateral centroid 457 spacing of 12 m in the present dataset. This is the size necessary to have observed the small-scale

458 clusters of outcrop CC5. Edge effects will decrease as the data frame increases width beyond 500
459 m. In order to observe large-scale clusters accurately, the results indicate that a dataset with total
460 width >3 km is optimal.

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Correlation of Point Patterns and Net-to-Gross

How do fluvial architectural properties statistically relate to point patterns? Clustering is the presence of channel belts in close proximity to one another compared to surrounding channel deposits. Therefore, we focused on NTG to answer this question. Clustered channel deposits have been previously identified as having high relative NTG (Hofmann et al. 2011), increased channel belt width and thickness, and elevated amalgamation (Benhallam et al. 2016).

468 At small point pattern scales, clustering and NTG have a weak positive correlation (Fig. 469 11E). The stratigraphic trends of small scale clustering and NTG found in outcrop CC5 match 470 those found by Benhallam et al. (2016) at nearby outcrops of Rock House Cove and Bull Canyon 471 (Fig. 1). There is an upward decrease in clustering and NTG from the base to the middle of the 472 section, followed by an upward increase from the middle to the top of the section (Fig. 12 of 473 Benhallam et al. 2016). However, at medium and large scales, clustering and NTG within 474 outcrops CC5 and CC1 are mostly negatively correlated, or highly variable resulting in non-475 correlation on average (Fig. 11F-H). To explain these anti-correlative results, we consider two 476 possible interpretations: 1) the methodology is fundamentally flawed, because analyzing point 477 patterns using centroids cannot accurately describe the spatial arrangement of fluvial channel 478 belts of disparate geometry, and/or 2) the methods are reasonably accurate and this signal is true, 479 indicating that large-scale clusters tend to be composed of smaller channel belts resulting in a 480 lower total NTG than their larger, uniform and random counterparts.

481 Although these two interpretations may be compatible to some degree, we favor the 482 notion that the methodology for identifying large-scale point patterns has high error. Primarily, 483 this inaccuracy is due to the simplification of each channel belt to a single point for PPA, which 484 previous studies suggested was likely problematic (e.g., Hajek et al. 2010; Benhallam et al. 485 2016). However, in previous work there has been no quantification of the geometry (area, width, 486 thickness) of channel belts in a way that can be statistically compared to PPA results. The 487 apparent anti-correlation of large-scale clustering to NTG in outcrop CC5 provides quantitative 488 evidence that centroid-based PPA cannot accurately describe large-scale fluvial architecture for 489 the present dataset. The ability of centroid-based PPA to characterize architecture is influenced 490 by the variety of channel belt widths within the dataset, and the resolution of internal architecture 491 of large channel belts. Our interpretation of this PPA methodological pitfall could be tested in 492 future work by repeating the correlation exercise but using vertical rather than lateral moving 493 windows. Anti-correlation in a vertical moving window correlation would corroborate our 494 current hypothesis, whereas correlation would raise new questions. 495 Both CC5 and CC1 have significant stratigraphic variation in the width of channel belts 496 (Fig. 8G and H). The lower and middle units of CC5 result in large-scale clusters that are 497 composed of more numerous and narrower channel belts, resulting in lower total NTG (Figs. 8, 498 11). In contrast, the highest NTG values of outcrop CC5 are associated with a series of very wide 499 (>1 km: 1/3 total outcrop width) channel belts in the upper unit of the southern segment (Figs. 500 8B and H). In outcrops with disparate channel belt widths, the narrower belts produce higher 501 point density, whereas the wider, laterally amalgamated channel belts have fewer centroids, 502 resulting in lower point density. When the two distinct styles of channel belts are juxtaposed in 503 the quadrat method, the area with higher point density has a greater chance of having variance of

points-per-cell much greater than the mean, resulting in a high clustering index. This variety of channel belt widths occurring together is common in the John Henry Member, and many other fluvial successions worldwide (e.g., Robinson and McCabe 1997; Bridge et al. 2000; Rittersbacher et al. 2014; Flood and Hampson 2015). We predict that PPA and NTG proxy one another more accurately when the dataset has a low range of channel belt widths. Small-scale analysis shows a stronger positive correlation because a smaller moving window, especially in

510 the vertical dimension, incorporates a lower variety of channel belt geometries.

511 Wide channel belts were likely deposited via post-avulsion lateral migration of a river, 512 rather than purely aggradational filling of a very wide river (Gibling 2006). It is also likely that 513 many of these wide belts were formed by multiple generations of avulsion, reoccupation, and 514 lateral river migration (Leeder 1978; Bridge 1993; Larue and Hovadik 2006). From certain 515 portions of the CC1 and CC5 models with exceptionally high resolution, and from outcrop 516 analysis, detailed internal architecture exists that is not observable in the majority of each DOM 517 (Fig. 4). Bounding surfaces that indicate avulsion (5th-order) are often difficult to differentiate 518 from 4th-order surfaces which indicate a major change in the migration of a river or barform, 519 rather than an avulsion (surfaces labeled "4/5?" in Fig. 4; Miall 1988).

520 Further evidence for avulsion-generated sand bodies includes irregular 'sawtooth' channel 521 margins, and correlation of individual channel stories with floodplain horizons laterally 522 (Chamberlin and Hajek 2015). For very wide channel belts, both of these observations are 523 limited. Many of the channel margin contacts are sand-on-sand, so irregularity in the margin 524 geometry is difficult to detect. In these and many other fluvial outcrops, floodplain deposits are 525 poorly exposed, making their correlation to channel belt stories difficult. Without internal 526 architectural geometry, i.e., a clear 5th order surface indicating incision and establishment of a

527 new channel (Miall 1988), the location of the avulsion that initiated a wide channel belt and the 528 location of any avulsions that came later are both unclear. Because of this uncertainty, the 529 centroid point is a poor approximation of avulsion location for channel belts that are much wider 530 than estimated paleo-river width. The use of PPA to understand avulsion dynamics becomes 531 more reliable with better understanding of internal architecture of wider and more complex 532 channel belts. If internal architecture is not readily interpretable, PPA is more effective when 533 channel belt width is closer to estimated paleo-river width.

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Correlation of Architectural Metrics from 1D to 2D

536 Subsurface resource investigations rely heavily on 1D borehole data collected at multiple 537 locations across an area of interest. Correlation between 1D datasets can inform a 2D or 3D 538 interpretation of the subsurface architecture. The present outcrop dataset provides an excellent 539 opportunity to collect 1D measurements of NTG, channel belt amalgamation index, and channel 540 belt thickness in a hypothetical borehole. For each of these architectural metrics we determined 541 how similar the 1D dataset is to the 2D moving window extending 300 m away from the 542 hypothetical borehole in either direction. The results show that consistently for both outcrops 543 NTG has the highest predictivity from 1D to 2D compared to the other two metrics (Fig. 12). 544 Channel belt amalgamation index and thickness did not have a consistent pattern, varying in their 545 predictivity from 1D to 2D within and between outcrops CC1 and CC5. For example, 1D 546 measurements of amalgamation index in outcrop CC5 are negatively correlated to their 547 surrounding 2D area in the north, but in the south they are positively correlated. Over the scale of 548 100s to 1000s of meters along strike, channel belt amalgamation index and thickness are 549 unreliable 1D to 2D predictors, while NTG is very reliable.

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Controls on Sedimentation

552 Small-scale clustering, NTG, amalgamation index, and channel belt width and thickness 553 along the Cockscomb broadly follow two key stratigraphic trends; Trend 1 is a decrease in all 554 metrics from the lower unit to the middle unit, and Trend 2 is an increase in all metrics from the 555 middle unit to the upper unit. These trends are regionally persistent, spanning at least 15 km 556 south to Rock House Cove and 20 km southeast to Bull Canyon (Figs. 1, 9; Gooley et al. 2016; 557 Benhallam et al. 2016). The 600 m-wide moving window analysis reveals significant lateral 558 variation from these key trends, on the order of 100s to 1000 m. In outcrop CC5 there are 559 multiple isolated packages of higher NTG, amalgamation index, and channel belt width and 560 thickness in the middle unit, disrupting the broad up-section decreasing-to-increasing trends 561 (Figs. 8, 9). These trends are the result of clustered large channel belts, which stand out in the 562 relatively low-NTG middle unit (Fig. 6). We interpret these clusters to be formed by autogenic 563 processes because they contradict trends observed regionally, and they themselves do not form a 564 regionally persistent pattern (Hajek and Straub 2017). A possible autogenic mechanism for the 565 generation of these clusters is via preferred re-occupation of an abandoned channel by an 566 avulsing river (Mohrig et al. 2000; Jerolmack and Paola 2007; McHargue et al. 2011). 567 Autogenic processes are likely always important signals in a fluvial system (Budd et al. 568 2016; Paola 2016), but broad changes in fluvial style, such as those associated with Trends 1 and 569 2, may represent changing allogenic forces which influence autogenic processes. Gooley et al. 570 (2016) suggest that Trend 1 is caused by a primarily tectonically driven relative base-level fall,

571 followed by a time of high accommodation coeval to transgressive shoreline deposits down-

572 depositional-dip. Trend 2 records the autogenic progradation of a distributive fluvial system

573 influenced by tectonic subsidence and increased sediment supply (Gooley et al. 2016). The 574 results of this study corroborate the conclusions of Gooley et al. (2016) but do not allow for 575 further refinement of controls on sedimentation. An important caveat to the interpretation of 576 Trend 2 is that the general scale of DFSs is much greater than the scale of observation for this 577 study. Modern DFSs range in width from 10s to 100s of km (Weissmann et al., 2010; Hartley et 578 al., 2010). The Salt Wash DFS is estimated to be \sim 450 km wide based on maps from Owen et al. 579 (2015). Thus, the scale of observation for this study is significantly smaller than the full scale of 580 the DFS, such that only a portion of the DFS is observed (Primm et al. 2017). 581 582 583 **CONCLUSIONS** 584 This study adds to a growing body of literature that utilizes statistical methods to 585 understand the character and organization of fluvial channel deposits. We investigated point 586 patterns and architectural metrics in the John Henry Member, and the correlation between the 587 two at different scales. The studied outcrops are ideal for this type of analysis due to their scale 588 and orientation roughly perpendicular to paleoflow. The outcrops used in this work comprise a 589 total of 4.5 km of along-strike exposure, with an average stratigraphic thickness of 250 m, 590 containing a total of 369 interpreted channel belts. Point patterns and architectural metrics 591 broadly follow two key stratigraphic trends: Trend 1, an upward decrease from the base to the 592 middle of the section, and Trend 2, an upward increase from the middle to the top of the section. 593 There is divergence from these trends along depositional strike as architecture varies at small 594 scales (100s of meters) and large scales (kilometers). Small scale, and possibly larger scale, 595 variation is likely caused by autogenic dynamics of the fluvial system, such as avulsion.

596 Autogenic processes are thought to be influenced by allogenic forces, such as climate, tectonics,

eustasy, and dynamic mantle topography, resulting in the broad consistency of Trends 1 and 2.

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598 Subsurface resource investigations rely heavily on interpolation between 1D borehole 599 data points in order to interpret 2D and 3D architecture. Results from the 1D-2D correlation 600 show that 1D NTG is more useful than 1D amalgamation index or channel belt thickness for 601 predicting the surrounding 2D architecture. While amalgamation index and channel belt 602 thickness have a high positive 1D-2D correlation across some portions of the outcrop, NTG has a 603 consistently high positive correlation at all lateral positions of both outcrops.

604 Channel belt clustering predicts net-to-gross when channel belt architecture is uniform 605 laterally and vertically, a situation that is unlikely to be encountered at all scales of any fluvial 606 system. At small point pattern scales (quadrat cells 21x greater area than the median channel 607 belt), channel belt clustering and NTG are positively correlated and both broadly follow two key 608 trends: Trend 1, an upward decrease from the base to the middle of the section, and Trend 2, an 609 upward increase from the middle to the top of the section. These trends were also described by 610 previous workers as far as 15 km to the south, and 20 km to the southeast, along a depositional 611 dip profile (Gooley et al. 2016; Benhallam et al. 2016). In contrast, at medium and large point 612 pattern scales (quadrat cells respectively 35x and 237x greater area than the median channel 613 belt), clustering has a mostly negative correlation to NTG. This anti-correlation is the result of 614 the simplification of channel belts to centroid points in the PPA analysis for an outcrop with a 615 large range in channel belt sizes. In a dataset with high channel belt width variety, belts of 616 disparate width are each represented equally by a single centroid. When they are juxtaposed in 617 the same analysis, smaller, more abundant belts are more likely to result in clustering, which is 618 then negatively correlated to NTG. A centroid approximation for PPA analysis is more effective

619 when an outcrop is composed of uniformly-sized channel belts. For this same reason centroid

- 620 approximation introduces uncertainty to using PPA to interpret paleoavulsion. Using centroids in
- 621 avulsion analysis is more effective the closer channel belt width is to estimated paleoriver width,
- 622 and when there is high confidence in the internal architecture of highly amalgamated channel
- 623 belts.

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819

FIGURE CAPTIONS

820 Figure 1 - Modified from Gooley et al. (2016). Map of the Kaiparowits Plateau, highlighting 821 regional outcropping of the John Henry Member (gray), general paleogeography, field locations 822 of the current study (the Cockscomb, CC1 and CC5) and past studies (Rock House Cove and 823 Bull Canyon). Locations of measured sections are indicated by bold points, and areas collected 824 as digital outcrop models are indicated by labeled polygons in the 'Areas of Interest' expansion 825 of the map. Rose diagrams show paleocurrent indicators collected in this (The Cockscomb) and 826 previous studies (Rock House Cove and Bull Canyon, Gooley et al. 2016). Fm = Formation; 827 WIS = Western Interior Seaway. 828 829 Figure 2 - Digital outcrop model (DOM) of outcrop CC1 created using stereophotogrammetry. 830 The top image is the uninterpreted DOM, and the bottom image is with channel belt 831 interpretations (white polygons). North is to the left of the image. Interpretations span the entire 832 John Henry Member, bounded by the contacts with the underlying Smoky Hollow Member (red), 833 and the overlying Drip Tank Member (blue). Measured sections CCX.2 (left) and CCX.3 (right) 834 show facies and grain size (M = mud; S = sand, and G = gravel), and are correlated to the DOM 835 image with major formation contacts. Double-sided arrows indicate the distance each measured 836 section is away from the DOM edge. Note that the scale of the DOM and measured sections are 837 different.

838

Figure 3 – Digital outcrop models of outcrop CC5 created using stereophotogrammetry. The
outcrop is covered by a northern (A) and southern (B) DOM. The interpretations from both

841 DOMs were combined for data analysis. The top image of each DOM is uninterpreted, and the

842 bottom image is with channel belt interpretations (white polygons). North is to the left of the 843 image. Interpretations span the entire John Henry Member, bounded by the contact with the 844 underlying Smoky Hollow Member (red), and the overlying Drip Tank Member (blue). 845 Measured section CCX.1 shows facies and grain size, and is correlated to the DOM image with 846 major formation contacts. Note that the scale of the DOMs and measured section are different. 847 848 Figure 4 - Example of the bounding surface hierarchy method employed to interpret the 849 Cockscomb DOMs. Numbers indicate the bounding surface hierarchy level, following the 850 methods of Miall (1988), with question marks for surfaces of uncertain hierarchy. See text for 851 further description of this method. A) Uninterpreted DOM image of a channel belt within 852 outcrop CC5, which has higher resolution than the DOM on average. B) Detailed interpretation 853 of bounding surfaces visible in image A. C) Interpretation that would likely result from this same 854 outcrop at a lower resolution, such as the average resolution for the rest of this DOM. 855 856

Figure 5 - Illustrations of how NTG (A), amalgamation (B), and channel belt width and thickness
(C) are calculated within a moving window. See methods section for explanation of variables.

Figure 6 – Plots of channel belts (gray polygons) and their centroids (black points) for CC1 (A),

and CC5 (B). Red rectangles illustrate the 600 x 25 m moving window used to average

861 architectural metrics (i) and the moving window spanning the entire outcrop width to capture

gross vertical trends (ii). C) Channel belt centroid points of outcrop CC5 with blue quadrat cells

within a red moving window, illustrating one window at small scale (iii), medium scale (iv), and

large scale (v) point pattern analyses. The dimensions of each of these scales are listed Table 1.

865 Stratigraphic subdivisions of lower, middle, and upper units are denoted with dashed lines.

866 Vertical exaggeration for both outcrops is 2.5x.

867

868 Figure 7 – Modified from Figure 5 from Benhallam et al. (2016). A) Examples of uniform,

869 random, and clustered point patterns. B) Demonstration of the quadrat method performed on the

870 entire window, and the resulting relationship of variance to mean for each type of point pattern.

871

872 Figure 8 – A-H: Outcrop properties averaged over moving windows 600 m wide by 25 m thick, 873 moving vertically by 5 m increments, and laterally by 50 m increments, for outcrops CC1 (left) 874 and CC5 (right). I-P: Outcrop properties averaged over moving windows spanning the entire 875 outcrop width and 25 m thick, moving vertically by 5 m increments, for outcrops CC1 (left) and 876 CC5 (right). All properties are shaded from white (low) to black (high) with a linear gradient. 877 The minimum and maximum values for each plot are indicated inside a white and black box 878 respectively, and encompass the entire range for both outcrops to allow direct comparison. 879 Horizontal dashed lines indicate lower, middle, and upper stratigraphic units, and vertical dashed 880 lines indicate lateral segments.

881

Figure 9 – Trends of NTG, amalgamation, channel belt width and thickness, and point patterns averaged laterally by outcrop segments; northern (N), central (C), and southern (S). Vertical moving windows 25 m thick, moving in 5 m increments, were used for NTG, amalgamation, and channel belt width and thickness. Point patterns were assigned qualitatively based on the dominant moving window trends in Figure 11. Architectural properties increase from white to black, while point patterns are represented as either gray (uniform), white (random), or black

(clustered). Lower, middle, and upper stratigraphic units are noted by dashed lines and define thevertical scale.

890

891 Figure 10 - Histograms and vital statistics of channel belt (CB) widths (top) and thicknesses 892 (bottom) for CC1 (left) and CC5 (right). Arrows on plot 'CC5 CB Width' (upper right) are 893 pointing to columns with only one channel belt, for improved visibility. Widths and thicknesses 894 presented are the maximum width and thickness of each individual channel belt. For CC1 n =895 104, and for CC5 n = 265. 896 897 Figure 11 - Results of the quadrat analysis for both CC5 (A-C) and CC1 (D) at different point 898 pattern scales and aspect ratios. A) CC5 small scale point patterns using quadrat cells 241.5 m 899 wide by 15 m thick, with an aspect ratio of 16:1, and 1800 m wide by 30 m thick moving 900 windows. B) CC5 large scale point patterns using quadrat cells 805 m wide by 50 m thick, with 901 an aspect ratio of 16:1, and 1610 m wide by 100 m thick moving windows. C) CC5 medium 902 scale point patterns using quadrat cells 120 m wide by 50 m thick, with an aspect ratio of 2.4:1, 903 and 960 m wide by 100 m thick moving windows. C) CC1 medium scale point patterns using 904 quadrat cells 120 m wide by 50 m thick, with an aspect ratio of 2.4:1, and 480 m wide by 100 m

905 thick moving windows. E-H) Results of correlation between the corresponding cluster analysis

906 (A-D), and 2D NTG results using the same moving window size.

907

908 Figure 12 – Correlations of 1D-to-2D architectural metrics: NTG (left, A and B), amalgamation

909 (middle, C and D), and thickness (right, E and F), for both CC1 (top, A, C, and E) and CC5

910 (bottom, B, D, and F). The X-axis represents the lateral moving window position increasing from

- 911 north to south, and the Y-axis is the correlation coefficient between the 1D and 2D trends for
- 912 each metric. The inset box within each plot shows the mean, standard deviation, minimum, and
- 913 maximum R^2 value for its respective analysis.

		Moving Window Quadrat Cells						
Outcrop	Quadrat Scale	Width (m)	Thickness (m)	Width (m)	Thickness (m)	Number Cells	Aspect Ratio	Figure Reference
	Small	1800	30	241.5	15	16	16.1:1	Fig. 6iii
CC5	Medium	960	100	120	50	16	2.4:1	Fig. 6iv
	Large	1610	100	805	50	4	16.1:1	Fig. 6v
CC1	Medium	480	100	120	50	8	2 1.1	Half the width of
001	Medium	-00	100	120	50 8	2.4.1	grid in Fig. 6iv	





























