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Bed Particle Displacements and Morphological Development in a Wandering Gravel-Bed River

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Abstract Bed particles were tracked using passive integrated transponder tags in a wandering reach of the San Juan River, British Columbia, Canada, to assess particle movement around three major bars in the river. In-channel topographic changes were monitored through repeat LiDAR surveys during this period and used in concert with the tracer data set to assess the relationship between particle displacements and changes in channel morphology, specifically, the development and re-working of bars. This has direct implications for virtual velocity and morphologic based estimates of bedload flux, which rely on accurate estimates of the variability and magnitude of particle path lengths over time. Tracers were deployed in the river at three separate locations in the Fall of 2015, 2016, 2017, and 2018, with recovery surveys conducted during the summer low-flow season the year after tracer deployment and multiple mobilizing events. Tracers exhibited path length distributions reflective of both morphologic controls and year to year differences related to the annual flow regime. Annual tracer transport was restricted primarily to less than one riffle-pool-bar unit, even during years with a greater number of peak floods and duration of competent flow. Tracer deposition and burial was focused along bar margins, particularly at or downstream of the bar apex, reflecting the downstream migration and lateral bar accretion observed on Digital Elevation Models of difference. This highlights the fundamental importance of bar development and re-working underpinning bedload transport processes in bar-dominated channels.

1. Introduction

In gravel-bed rivers there is a natural feedback between channel morphology and bedload transport. The morphology of the channel is developed through the movement and deposition of individual bed particles and in turn the spatial patterns of bed material transport are controlled at least in part by the morphology of the channel (Church, 2006; Church & Ferguson, 2015). Therefore, attempts at calculating bed material transport rates, or more generally in studying bedload processes, need to consider morphologic controls on bed particle dynamics. This is particularly relevant when employing the virtual velocity approach to estimating bedload flux because it relies upon an accurate measure of the distribution and variability of particle path lengths (travel distances), which may differ greatly in channels of different morphology (Ashmore and Church, 1998; Vázquez-Tarrío et al., 2018). One idea suggested by Neill (1987), is that bed particle path lengths may be related to, and inferred directly from, the channel morphology. Depositional features such as bars are self-formed through individual particle displacements, so it follows that over the long-term the predominant particle path lengths must be related to the scale and spacing of the bars. This idea is appealing because with sufficient data tying particle path length and burial with the morphological development of bars, it may eventually allow path length to be estimated from morphology without the need for time-consuming and resource intensive direct particle tracking. However, evidence from field-based studies to support the link between bar morphology and particle path length is currently weak (Hassan & Bradley, 2017).

Throughout the bedload tracking literature, the idea that hydraulic forcing is the primary control on particle transport is prevalent, and functional relationships between average particle travel distances and the combination of flow strength and/or grain size have been developed. Hassan and Church (1992) demonstrated that mean tracer path length and excess stream power are positively correlated for single discharge events. More recently, Phillips and Jerolmack (2014) used an impulse framework to describe the effects of flow strength on particle transport. Church and Hassan (1992) showed that there is a nonlinear relationship between mean particle path length and the size of the particle scaled by the median size of local subsurface...
bed material, whereby travel distances of particles smaller than the median are relatively insensitive to increasing grain size, but that there is a rapid decline in path length with increasing grain size for particles larger than the median grain size of local subsurface bed material. These findings have since been re-affirmed with data from studies across a range of channel types (Hassan & Bradley, 2017; Vázquez-Tarrío et al., 2018), though data from larger, bar-dominated channels is lacking. Milan (2013) demonstrated that this effect is, at least in part, caused by differences in the duration of competent flow for different grain sizes. Many tracer studies, however, have noted differences between path length distributions and theoretical models because of tracers accumulating at distinct regions related to the river morphology (e.g., Bradley & Tucker, 2012). One example of this is the tendency of tracers to be preferentially transported to and stored in gravel bars in channels with riffle-pool-bar morphologies, especially over longer time-scales (Ferguson et al, 2002; Haschenburger, 2013).

In a literature review and re-analysis of published tracer data, Pyrce and Ashmore (2003a) found that the positively skewed path length distributions consistently reported in the literature occurred during moderate discharge events or in smaller channels lacking well-developed sedimentary structures or bar morphology. However, they found that in bar-dominated channels, high magnitude flows (i.e., those capable of altering or forming bars) lead to bi- or multimodal distributions related to the location of bars. Pyrce and Ashmore’s (2003b) flume experiments of an alternate bar channel aligned with these findings, as the authors demonstrated that during bar-forming flows most tracers were deposited on the first bar downstream from the upstream pool in which particles were seeded. Only during lower flows at the critical discharge for gravel entrainment, were positively skewed distributions, with path lengths shorter than bar spacing, observed. Using the same tracer data set, Pyrce and Ashmore (2005) showed that during bar formation and development bed particle path lengths are commensurate with the spacing of erosion and depositional sites, and that deposition locations tied to bar development processes. In another flume experiment, Kasprak et al. (2015) demonstrated that tracer path lengths were closely related to erosional and depositional processes associated with bar development in a braided channel, with average path lengths on the scale of confluence-diffuence spacing. Similar results have been yielded from more recent modeling of braided channels (Middleton et al., 2019; Peirce, 2017). The question remains however, as to whether these observations are seen in full-scale rivers where conditions are less controlled.

In a synthesis and re-analysis of previously published field-based tracer data, Vázquez-Tarrío et al. (2018) explored the influence of both hydraulic and morphologic controls on particle transport for a range of channel types. They noted that there was a weak positive correlation between stream power and average travel distance for the data set. However, when travel distance was scaled by a morphological length scale for each channel type (i.e., the spacing between macroscale bedforms), the scatter in the relationship was reduced, indicating that tracer transport has some dependence on channel morphology. Furthermore, analyses of empirical predictors of path length have pointed toward channel width as the most significant control on travel distance (Beechie, 2001; Vázquez-Tarrío & Batalla, 2019). For bar-dominated channels, this may imply that bar spacing exerts a control on path length because the longitudinal spacing of bars is proportion- al to channel width. These analyses are the starting point to investigating the relationship between path length, bar development and channel scale, but currently this lacks tracer-based data collected in larger bar-dominated channels where morphologic control is expected to be most significant. Therefore, there remains uncertainty as to whether the principles of bed particle dynamics and statistics of displacements, derived from smaller rivers, such as step-pool, plane-bed, and low amplitude pool-riffle channels (Montgomery & Buffington, 1997), are applicable to bar-dominated channels with more complex morphology and higher rates of morphological change, and further, if spatial patterns of tracer deposition and burial are tied to bar development.

The paucity of tracer data collected in larger rivers may be explained in part by logistical challenges in searching such large areas of channel, and the potentially deep burial of tracers resulting in low recovery rates. One solution that is increasingly being used to track bed particle movement in larger channels, is the use of passive integrated transponder (PIT) tags (Hassan & Bradley, 2017). PIT tags are small, glass, cylindrical capsules that operate using radio frequency identification (RFID) technology. Several factors make PIT tags an effective technology for bedload tracking including their long lifespan, resistance to abrasion and breakage (Cassel, Piégay, & Lavé, 2017), and the ability to distinguish individual particles from one another.
using unique codes. Furthermore, as smaller PIT tags are being developed, an increasingly wide range of sediment sizes can be tracked (Hassan & Roy, 2016). Technological improvements in PIT tag technology, such as the increased read range of antennas (Arnaud et al., 2015), innovative surveying strategies (Arnaud et al., 2017) and the development of “wobblestones” (Cain & MacVicar, 2020; Papangelakis et al., 2019), have made it more possible to track bed particle movement in larger rivers. Active ultra-high frequency (a-UHF) RFID tags have also been used to explore active layer depths (Brousse et al., 2018) and particle paths (Misset et al., 2020) in wandering/braided channels, providing the benefit of larger detection ranges than PIT tags. Due to their large size however, aUHF tags can only be fit into natural particles with a b-axis of at least 70 mm or molded into synthetic pebbles (Cassel, Dépret, & Piégay, 2017; Cassel et al., 2020).

The primary objective of this study was to explore the relationship between channel morphology and particle path lengths in a large, wandering gravel bed river—the San Juan River, British Columbia, Canada. Wandering channels are irregularly sinuous and can display aspects of both meandering and braided channels. These channels are characterized by complex bar development and some degree of lateral instability (Beechie et al. 2006; O’Connor et al., 2003). Typically, the most common bar morphology is a lateral bar and the dominant mode of deposition is lateral bar accretion (Desloges & Church, 1987; Rice et al., 2009). Church and Rice (2009) describe a pattern of bar evolution whereby vertical growth is limited by the height at which sediment can be elevated, and the longitudinal growth of bars is limited by the length-scale of the channel, resulting in bars primarily growing laterally. This pattern of rapid lateral accretion may persist for decades (Rice et al., 2009) and is accompanied by the erosion of the opposite bank, producing a laterally unstable channel with less systematic migration than true meandering channels (Fuller et al., 2003; McLean et al., 1999). We expect this pattern of bar and channel evolution to be reflected in spatial patterns of tracer displacements for the San Juan River, as bars are by definition an expression of the displacement, transport and deposition of individual bed particles. If path lengths are tied to morphology in bar-dominated channels, then particle displacements and burial should be tied to patterns of morphological change over a defined period. To address this objective, PIT tags were used to track bed particle movements and repeat LiDAR surveys were conducted to measure topographic change and bar development during the tracer monitoring period. Combining tracers with topographic change captured at high resolution is uncommon in the literature and allows a more comprehensive interpretation of the process-form coupling of bedload transport and channel morphology than can be achieved via either method separately (Vericat et al., 2017).

2. Materials and Methods

2.1. Study Site

The San Juan River, also known by its First Nation name, the Pacheedaht, is located on southern Vancouver Island, British Columbia and drains an area of about 730 km² (Figure 1a). The main channel is over 50 km long with a total relief of 690 m. The San Juan River valley follows a major east-west fault with distinct topography and bedrock geology on the north and south sides. Bedrock north of the river consists of a series of volcanic and intrusive units, whereas the south side of the valley is underlain almost exclusively by metamorphic rocks of the Leech River Complex (BCGS, 2018). The river outlets to the Strait of Juan de Fuca, near the town of Port Renfrew (Figure 1a).

Forest harvesting in the San Juan River Watershed dates back to the early 1900s and has been linked to changes in physical habitat and channel morphology in the mainstem and tributaries (NHC Ltd., 1994). This study was guided by watershed management objectives, to provide detailed information on the current sediment dynamics and morphologic changes in the San Juan River which will help inform future restoration decision making.

The study focused on the lower alluvial reach of the San Juan River beginning near Red Creek, downstream of a canyon reach (Figure 1b). The alluvial channel exhibits a wandering morphology, as defined by Mollard (1973) and Neill (1973), with an active width varying between 50 and 150 m and a reach-averaged slope of 0.0011 m m⁻¹. During low-flows the river has a single identifiable main channel though it displays a multichannel pattern during higher-flows when secondary channels are active. Riffle-pool-bar sequences are the primary macroscale bedforms in the alluvial reach, with bars typically on the order of several hundred meters long and up to 100 m wide. Bars are composed primarily of gravel, cobble, and sand...
and there is a general trend of downstream fining of surface sediment caliber both within and between bars. For referencing purposes, mainstem bars were numbered, with Bar 1 being the farthest upstream and subsequent bars numbered in ascending order downstream. Particle tracking focused on Bars 6, 7, and 15, the most accessible sites along the river (Figure 1b). These bars are representative of the typical length and width, overall appearance, and grain sizes found in the alluvial reach. The morphology of the study bars is summarized in Table 1.

The closest gauging station to the study sites is the Water Survey of Canada hydrometric station 08HA010, which is installed on the lower San Juan River, approximately 2.5 km downstream from Bar 15 and 7.5 km from Bar 6 (Figure 1b). The 2-, 10-, and 100-year floods are approximately 800, 1,050, and 1,200 m³ s⁻¹.
respectively, though the upper end of the rating curve is uncertain due to the difficulty of obtaining discharge measurements during peak floods (Figure 2). Mean monthly discharge varies from a high of 97 m³ s⁻¹ in January to a low of 4.5 m³ s⁻¹ in August, with a mean annual discharge of 49 m³/s (WSC, 2019). The discharge regime closely follows the seasonal trend in rainfall because 99% of annual precipitation at low elevations falls as rain (ECCC, 2019), with only transient snow accumulations (no seasonal snowpack) at higher elevations within the watershed.

2.2. Tracer Stone Deployment and Tracking

Half duplex low-frequency (134.2 kHz) PIT tags were inserted into individual gravel bed particles to track annual particle movements around three major bars in the San Juan River. Low-frequency tags are ideal for tracking in coastal and fluvial environments because their signal can pass through water and can penetrate most nonmetallic objects (sediment, wood, etc.) better than high and ultra-high frequency tags (Chapuis et al., 2014; Schneider et al., 2010). In 2015, 100 tracers were installed along a cross-section at the head of Bar 6, while between 2016 and 2018 a further 1,199 tracers were installed across the three study sites (Bars 6, 7, and 15) with between 125 and 142 tracers per site annually. A combination of 12, 23, and 32 mm long PIT tags were used in the original 2015 deployment. However, by 2017 only the 32 mm tags were used because of their larger read range (Chapuis et al., 2014) and higher recovery rates during the first recovery survey.

Wolman particle size counts were conducted at each site to characterize the size distribution of surficial bed material (Table 2) (Wolman, 1954). Native stones were then collected from the San Juan River and brought back to the lab for preparation, which included drilling a cavity in each particle, inserting and epoxying an RFID tag in place, and painting the stone (similar to methods to described in Eaton et al., 2008). Particles were selectively chosen to reflect the size distribution of bed surface material in the channel as best as possible (Figure 3). However, PIT tags did not fit into particles smaller than 22 mm, thus the fine end of the bed material distribution was not well represented. Differences between the size distribution of bed material and tracers was greatest for Bar 15, which had the finest material of the three study bars (Table 2). However, the Bar 15 tracer distribution does reflect well the size distribution of local surface bed material greater than 22 mm (Figure 3c). The use of 32 mm tags in 22–32 mm tracers biased this size class toward particles with an a-axis longer than 35 mm.

Tracers were deployed annually in the fall, prior to winter flooding, along launch lines perpendicular to the direction of flow. Each launch line spanned the bar head, riffle, and tail of the opposite (upstream) bar, providing the opportunity to observe tracer dispersion around major bars, and to observe differences in particle mobility and path lengths across different morphologic units. While particle path lengths are unlikely to be influenced by seeding position across the channel cross-section in smaller, plane-bed type streams, it has been shown to play a significant role in particle movement in riffle-pool channels with well-developed bar morphology (Liébault et al., 2012). Clusters of tracers were deployed at one to two m intervals along the launch line by replacing particles on the surface of the riverbed with tracer stones to mimic natural positions and local bed texture. Tracers starting on the surface of the riverbed are more exposed to flow than buried or locked particles, and therefore, observed travel distances in this study are likely to be overestimates of annual transport distances for the bed as a whole.

Tracer recovery surveys were conducted annually during the low-flow seasons at the end of July through August for 2016 to 2019. All recovered tracers were removed from the channel and redeployed at their original launch line the following fall so that each year the deployment strategy was a repeat of the previous one. The winter storm season with several mobilizing flows is treated as an annual event producing annual particle

| Table 1 |
| Dimensions of Study Bars |

| Location | Length (m) | Average width (m) | Relief (m) | Slope (m m⁻¹) |
|----------|------------|-------------------|------------|---------------|
| Bar 6    | 550        | 100               | 4.2        | 0.0038        |
| Bar 7    | 540        | 55                | 3.9        | 0.0031        |
| Bar 15   | 585        | 45                | 3.2        | 0.0009        |

Abbreviation: DEM, Digital Elevation Model.
Notes. Bar length was measured relative to the channel thalweg. All measurements are based on 2015 Lidar DEMs.
displacements in relation to morphology. This allowed us to assess the morphologic influence on tracer displacements through repeat tests. This decision was also influenced by the practicalities of tracking sediment in a large river because tracking after every event would be onerous and leaving tracers in the channel would likely necessitate increasing the downstream extent surveyed every year, which would quickly become untenable in a river of this size. Furthermore, excavating and removing tracers from the channel allowed us to directly measure tracer burial depths which provides valuable information on active layer dimensions and gives context to particle deposition in relation to morphological development. The maximum downstream extent of survey differed for each location, though generally the first 2 bars downstream of each seeding site were searched. The deepest portions of pools were omitted from the searching process because they were not wadable, even at low-flow.

Two antennas were used to search for tracer stones. A small handheld wand antenna, the “BP Plus Portable,” and a larger “Cord Antenna System,” both were purchased from BioMark® (Figure 4). Based on testing in the lab, the wand antenna had a maximum read range around 40–50 cm for the 32 mm PIT tags, though the tag signals are anisotropic, and the read range was as low as 10 cm for certain orientations. The wand antenna was used as the sole antenna for tracer recovery in 2016, resulting in a low recovery rate (33%), and was used only as a supplementary tool to the larger antenna for subsequent years. Since PIT tag signals interfere with one another when in close proximity (Chapuis et al., 2014; Lamarre et al., 2005), the wand antenna was still a useful tool for distinguishing between PIT tags in areas where tracers were densely concentrated—typically the launch line, as a fraction of the tracer population remained immobile. The wand antenna was also effective for refining the position of tracers after detection with the cord antenna.

The cord antenna system consisted of the cord antenna cable, secured to a 15’ × 5’ (5 m × 1.5 m) rectangular PVC pipe frame, mounted to a backpack frame using a series of ropes, pulleys and cams (Figure 4b). The frame held the cable in a (semi-) rigid structure, stabilizing the antenna and allowing it to keep a high current. The operator stood in the center of the rectangle wearing the backpack and could manipulate the height of each corner of the antenna to help navigate obstacles and changing topography in the field (Figure 4b). The cord antenna covered a much larger surface area than the wand antenna, making it an ideal tool for searching large areas efficiently. It also had a much larger range of detection than the wand antenna, with a maximum read range around 1.75 m, and thus could detect tracer stones buried at greater depths.

Once detected, tracer positions were recorded using one of two methods, and dug up (where possible) to determine a burial depth. For tracers that moved only a short distance (less than 20 m or so), a measuring tape was used to directly measure transport distances from the launch line. For tracers moving larger distances, a handheld Garmin GPS unit was used to record tracer locations. GPS waypoint errors were on the order of...
of two to three m, which was considered acceptable for the purpose of determining typical particle path lengths, since average path lengths were generally around 100 m, resulting in less than 5% error.

2.3. Geomorphic Change Detection

2.3.1. Topographic Surveying

In addition to particle tracking, repeat aerial LiDAR surveys, flown by Terra Remote Sensing Inc. in 2015, 2018, and 2019, were contracted for the study and other projects related to management of the river. Each LiDAR survey was conducted using the Reigl LMS-1780 sensor from an airborne platform. Ground accuracy tests, performed by Terra Remote Sensing Inc., involved both internal and external horizontal and vertical checks (TRS Inc., 2018a, 2018b, 2019). Internal checks were conducted via comparison of intra and inter-flight areas of overlap. External checks consisted of two components: comparison of the LiDAR ground surface with control stations not used in the calibration process, and with a series of check points collected on open surfaces using Post Processed Kinematic (PPK) GPS. A root mean square error for vertical precision (RMSEv) of less than 10 cm was reported for all surveys (Table 3) (TRS Inc., 2018a, 2018b, 2019). Survey point densities were spatially variable across the study sites. Average point densities ranged from 12 to 38 points per square meter with differences based largely on ground cover and topography (Table 3). LiDAR point clouds were filtered by ground and nonground classes by Terra Remote Sensing Inc. (TRS Inc., 2018a, 2018b, 2019). LiDAR point clouds were used to generate a series of raster-based digital elevation models (DEMs) of the study sites. LiDAR DEMs were produced for each survey at a 10 cm spatial resolution by converting point clouds to a Triangulated Irregular Network (TIN), from which concurrent raster DEMs were generated using linear interpolation. Topographic changes between survey dates were then calculated by processing the LiDAR DEMs using the Geomorphic Change Detection (GCD) software (Wheaton et al., 2010) to produce DEMs of difference (DoDs). In-channel areas that were inundated in both the old and new DEMs, where point cloud returns were either nonexistent or affected by refraction were not used in building DEMs, restricting change detection analysis to above-water areas. To capture bank erosion, a minimum level of detection of 1 m was used in change detection analysis for areas that were wet in the new DEM but were vegetated.

### Table 3

Lidar Survey Metadata (Sourced From TRS Inc., 2018a, 2018b, 2019)

| Year of survey | RMSEv (m) | σv (m) | Average point density (points/m²) |
|---------------|-----------|--------|----------------------------------|
| 2015          | 0.026     | 0.036  | 12.8 ± 7.2                       |
| 2018          | 0.032     | 0.037  | 28 ± 16.0                        |
| 2019          | 0.069     | 0.062  | 38 ± 11.0                        |

Abbreviations: RMSEv, vertical root mean square error; σv, standard deviation of vertical error.
banks in the old DEM. This threshold allowed us to observe changes in bank position, as the riverbanks are 2 m or taller throughout the alluvial reach.

The LiDAR-derived DoDs (Difference of DEMs between successive surveys) were used to interpret patterns of tracer displacement and burial depths, and to provide information on morphological development of the bars during the study period. They were not used to calculate complete reach-scale sediment budgets due to the lack of in-channel topographic data and stage differences during each LiDAR survey affecting the relative portion of the river bed that was exposed. Currently, collecting reach-scale bathymetric data in large channels is challenging and relies on either boat-based multibeam echo sounding (MBES) systems or green wavelength LiDAR sensors (Tomsett & Leyland, 2019).

2.3.2. DEM Analysis

To account for uncertainty in the DEMs, a spatially variable uncertainty analysis was conducted using the GCD ArcMap extension. This involves three main steps: (1) an estimate of uncertainty for each individual DEM; (2) propagation of these errors through the DoD; and (3) an assessment of the statistical significance of these uncertainties in distinguishing real geomorphic change from noise (Wheaton et al., 2010). A major appeal of this method is that it requires little to no additional survey error information other than the survey data itself. Further, accounting for spatially variable error allows for recovery of information in areas with low elevation uncertainties that would otherwise be lost.

For each DEM, two surfaces were generated for uncertainty analysis using the built-in tools in the GCD software: a point density raster and a slope raster. The rationale behind using these surfaces was that steep areas with low point density have high elevation uncertainty, whereas flat areas with high survey point density will generally have lower elevation uncertainty (Wheaton et al., 2010). These surfaces were then combined on a cell-by-cell basis, using a fuzzy inference system (FIS), to produce an elevation uncertainty surface. Uncertainty surfaces associated with individual DEMs were then combined using simple error propagation (Brasington et al., 2003) to produce a single propagated error surface for the DoD. The GCD software then uses probabilistic thresholding to determine the statistical significance of these uncertainties. The probability that the elevation change associated with each individual cell of the DoD is then assessed at a user-defined confidence interval, in this case 80%. Originally, 95% was chosen, however upon examination of output thresholded DoDs, this limit appeared too restrictive, with real change being removed from areas of eroding banks and in obvious areas of deposition.

2.4. Hydrological Analysis

Discharge data from the WSC hydrometric station were used to characterize differences in the hydrologic conditions between study years. A bankfull discharge (Qbf) of 500 m³ s⁻¹ was visually estimated from time-lapse imagery, roughly the 1 year return interval flood (Figure 5), with Qbf being defined as the discharge at which the entire active width of the channel was inundated. This was used as a reference discharge to approximate the number of mobilizing flow events per year. While previous research indicates that flows
less than bankfull may mobilize coarse bed sediment (Pfeiffer et al., 2017; Phillips & Jerolmack, 2019; Ryan et al., 2002), a threshold discharge for gravel entrainment could not be accurately established for this study because tracers were exposed to multiple potentially mobilizing events each year, rendering it impossible to determine which specific events caused tracer movement. Further, the complexity of channel morphology and variability in tracer grain size makes a single threshold discharge difficult to define for this study. Despite this, $Q_{bf}$ still provides a relative basis for the number of potential mobilizing flows, and the hydrograph for the period of the tracer study illustrates that even if the reference discharge were shifted substantially, say 100 $m^3 s^{-1}$, the number of mobilizing events would change very little (Figure 6).

Previous tracer studies have used the total excess flow energy ($\Omega_T$) as a metric to capture the intensity and duration of competent flow for multiple flood events (Haschenburger, 2013; Papangelakis & Hassan, 2016). However, this requires knowledge of the critical discharge, which was unknown for this study. Instead, a modified $\Omega_T$ was used in analysis for this study, whereby the total flow above estimated bankfull was integrated over the period between tracer deployment ($t_d$) and recovery ($t_r$):

$$\Omega_T = \rho g S \int_{t_d}^{t_r} (Q - Q_{bf}) \, dt$$

where $\rho$ is the density of water (1,000 $kg \, m^{-3}$), $g$ is the acceleration due to gravity (9.81 $m \, s^{-2}$), and $S$ is the reach-average slope. Discharge data was integrated at 5 min intervals (as collected at the WSC hydrometric station). The modified $\Omega_T$, along with annual peak instantaneous discharge ($Q_{max}$), and number of potentially mobilizing events ($Q > Q_{bf}$) were recorded for each study year to give context to differences in tracer mobility, path lengths, and burial data that might be the result of different hydrologic conditions between years. The primary purpose of this analysis is to identify any differences in annual path lengths that can be attributed to differences in the annual flow regime, and to then compare any observed differences with morphologic constraints on path lengths that may occur over the longer term related to, for example, deposition on bars.

3. Results

Tracer recovery rates were generally high for a river of the size and scale of the San Juan River (see Table 1 in Chapuis et al., 2015), with annual recovery rates exceeding 65% for all but one of the deployments (Table 4). The low recovery rate (33%) of tracers from the original 2015 deployment was a result of the surveying approach used in searching for tracers, as only the smaller handheld wand antenna was used in the initial recovery survey in 2016. The low recovery rate from this deployment limited interpretation and
analysis of the data from this year, and as such was removed from analyses presented in this section unless specifically noted.

### 3.1. Discharge Effects on Particle Dynamics

A total of 21 events with peaks greater than 500 m$^3$ s$^{-1}$ occurred from 2015 to 2019, ranging from four to six events per study year (Table 4). The annual instantaneous $Q_{\text{max}}$ during the tracer study ranged from 749 m$^3$ s$^{-1}$ (2016–2017) to 1,022 m$^3$ s$^{-1}$ (2015–2016) corresponding to roughly 2 and 6-years return interval floods (Table 4; Figure 2). Using the modified $\Omega_T$ as a metric, the “wettest” year (i.e., greatest amount of flow exceeding $Q_{bf}$) was 2017–2018, the “driest” year was 2016–2017, while 2015–2016 and 2018–2019 had $\Omega_T$ values in between (Table 4). The sensitivity of $\Omega_T$ to the defined $Q_{bf}$ was tested, and shifting $Q_{bf}$ to 400 m$^3$ s$^{-1}$ or 600 m$^3$ s$^{-1}$ made no difference to the relative ranking of $\Omega_T$ between study years, and made little difference to proportional differences between years.

To assess annual differences in tracer movement related to discharge, $\Omega_T$ was plotted against the mobility rate, median path length, and mean burial depth for each site and study year (Figure 7). The relative mobility of tracers ($r_m$) appeared insensitive to differences in $\Omega_T$. For each bar, more than 80% of tracers were mobilized each year, with the exception of 2016–2017 Bar 15 tracers (48% mobile) (Figure 7a). The low mobility of these tracers relative to other deployments may be a result of both low $\Omega_T$ and also local morphodynamics (see Section 3.3).

The median path length of tracers ($L_{50}$), scaled by the local bar length, showed a general positive relationship with $\Omega_T$, though this was not statistically significant when pooling the data between sites ($R^2 < 0.05$) (Figure 7b). Tracers seeded at each bar tended to show different responses to increasing $\Omega_T$. The influence of $\Omega_T$ on Bar 6 and Bar 7 tracers was unclear, as the wettest year of the study, 2017–2018, did not produce the highest $L_{50}$ for either site (both had greater $L_{50}$'s in 2018–2019). However, the path length of Bar 15 tracers sharply increased with $\Omega_T$ for the 3 years of study (Figure 7b). Despite any increases in annual path length associated with $\Omega_T$, most tracers were limited to transport distances less than one bar-length. This implies that any longer-term morphologic constraints on path length, such as deposition within bars, is unlikely to

| Year        | Number of events | $Q_{\text{max}}$ (m$^3$ s$^{-1}$) | $\Omega_T$ (MJ m$^{-1}$) | Recovery rate, R (%) | Mobility rate, $r_m$ | Median path length, $L_{50}$ (m) | Median scaled path length, $L_{10}$ (m) | Maximum scaled path length, $L_{max}$ (m) | Average burial depth, $B_{avg}$ (cm) |
|-------------|------------------|-------------------------------|------------------------|----------------------|---------------------|----------------------------------|-------------------------------------|--------------------------------------|-------------------------------|
| 2015–2016   | 6                | 1,022                         | 571                    | 33                   | 0.82                | 38                               | 0.07                                | 0.46                                 | 9.8                           |
| 2016–2017   | 4                | 749                           | 331                    | 66                   | 0.89                | 69                               | 0.12                                | 0.78                                 | 5.7                           |
| 2017–2018   | 6                | 1,003                         | 890                    | 71                   | 0.91                | 130                              | 0.23                                | 2.02                                 | 16.4                          |
| 2018–2019   | 5                | 942                           | 479                    | 75                   | 0.88                | 155                              | 0.28                                | 0.79                                 | 19.6                          |
| 2016–2017   | 4                | 749                           | 331                    | 76                   | 0.80                | 97                               | 0.18                                | 0.92                                 | 4.4                           |
| 2017–2018   | 6                | 1,003                         | 890                    | 69                   | 0.87                | 187                              | 0.35                                | 0.94                                 | 14.9                          |
| 2018–2019   | 5                | 942                           | 479                    | 79                   | 0.93                | 219                              | 0.41                                | 1.41                                 | 13.0                          |
| 2016–2017   | 4                | 749                           | 180                    | 70                   | 0.48                | 155                              | 0.27                                | 0.96                                 | 5.1                           |
| 2017–2018   | 6                | 1,003                         | 483                    | 65                   | 0.92                | 308                              | 0.54                                | 1.27                                 | 8.4                           |
| 2018–2019   | 5                | 942                           | 260                    | 75                   | 0.94                | 197                              | 0.32                                | 1.34                                 | 8.5                           |

Table 4: Summary of Annual Floods and Tracer Recovery Results

Notes. Mobility rate is calculated as the fraction of tracers that were recovered more than 10 m downstream of their initial position. Tracer path lengths are scaled by the length of the bar at which they were seeded. Average burial depths include surface tracers.
be substantially affected by differences in the typical annual flow regime. However, this may not hold true for extreme events capable of major morphologic change in the bars and river morphology.

Overall, there was an increase in burial depth with increasing $\Omega_T$ when pooling the data across all sites ($R^2 = 0.49$) (Figure 7c). The largest deviation from the general linear trend was the 2018–2019 deployment of tracers at Bar 6 ($B_{ave} = 20$ cm). Bar 15 exhibited the lowest average burial depths, while Bar 6 and 7 tracers were typically buried deeper (Figure 7c). Tracer burial is explored in more detail in Section 3.3 with respect to topographic changes observed on DoDs.

Path length frequency distributions for tracers deployed at Bars 6, 7, and 15 are presented in Figure 8 and maps illustrating the final position of recovered tracers are presented in Figures 9–11 respectively. The path length distributions for 2016–2017, the “driest” year, were positively skewed for all three sites (Figure 8) with the lowest median path lengths ($L_{50}$) of any of the years of study (Table 4). The following year, 2017–2018, was the “wettest” of the study period. The Bar 6 path length distribution was also positively skewed this year, though the $L_{50}$ increased from 69 to 130 m downstream reflective of the increased hydrologic conditions (Table 4). The 2017–2018 Bar 7 and Bar 15 path length distributions followed roughly symmetrical distributions, with the primary mode of tracer deposition occurring at the bend apex next to Bar 7 (Figure 10b), and just downstream of the apex at Bar 15 (Figure 11b). In 2018–2019, the year with moderate $\Omega_T$, Bar 6 tracers exhibited a bi-modal distribution with the primary mode of deposition occurring just upstream of the bend apex and a secondary mode reflecting short-transport distances (Figure 9b). Bar 7 and Bar 15 path length distributions for 2018–2019 were positively skewed, though there was a minor peak in the Bar 7 distribution, reflecting tracers accumulating on the bar tail (Figures 8b and 10c). Overall, the shape of path length distributions generally reflected discharge conditions, whereby positively skewed distributions were observed for the driest year, bi-modal distributions were observed for two of the three sites for intermediate hydrologic conditions, and roughly symmetrical distributions, centered around the bend apex at each bar were observed for two of the three sites during the wettest year.

### 3.2. Morphologic Effects on Particle Displacements

Tracer path lengths were scaled by the length of the bar at which they were seeded to better compare particle displacements between the three bars (Figure 12). The distribution of scaled path lengths were significantly different (Kruskall-Wallis test, $p < 0.05$) among all three sites whereby Bar 6, Bar 7, and Bar 15 had median scaled path lengths ($L_{50}$) of 0.20, 0.34, and 0.41 “bar-lengths” respectively (Table 5). Bar 6 tracers, while having the highest mobility rate (89% mobilized), had the lowest $L_{50}$, with 90% of tracers depositing upstream of the bar apex (which is at approximately $L = 0.50$) (Figures 9 and 12). For both 2016 and 2018 Bar 6 deployments, there was a clustering of tracers just upstream of the bend apex that appears to be related to the growth of a coarse gravel sheet with a slip face roughly 1 m high, which migrated downstream from the bar head between 2015 and 2019 (Figure 13). While transport past the apex was more common for Bars 7 and 15, these tracers still tended to be deposited within the initial bar in which they were seeded, with...
deposition focused along bar margins (Figures 10 and 11). This is also highlighted by the tracer escape rates, that is, the fraction of mobile tracers recovered downstream of the initial bar in which they were seeded. Less than 1% of tracers escaped Bar 6 annually, 5% escaped Bar 7, and 4% escaped Bar 15 (Table 5) indicating that annual particle displacements on the San Juan River tend to be within one riffle-pool-bar unit.

Tracers recovered in the wetted channel, closer to the thalweg, may be more likely to be remobilized by future events, and as such represent potential future frontrunners, while those stored on gravel bars are less likely to be remobilized, and may become trapped in the bars with future transport limited by bar development and re-working. For Bar 6, 66% of tracers were deposited in or on bars, while 34% were recovered in the wetted channel. Similarly, 71% of Bar 7 tracers were recovered on bars versus 29% in the wetted channel. Bar 15 appears to trap less sediment than the other sites, as only 46% of tracers were recovered on gravel bars versus 54% in the wetted channel, as Bar 15 is less-developed laterally than the other study bars. For each of the sites however, the proportion of tracers recovered on gravel bars reflects more than just those transported to and deposited on bars, it also reflects tracers that were originally seeded on bar tails and remained there. The role of initial seeding morphology was explored by partitioning the data by the morphologic unit in which tracers were originally deployed (Table 6). The Bar 6 2016 launch was treated separately from the rest of the Bar 6 data as the launch line was located 120 m downstream of the other years, with tracers seeded across the bar edge and adjacent pool. In general, tracers seeded on bar tails tended to have lower relative mobility than those seeded on the bar heads or in the wetted channel (Table 6). The exception to this was that 80% of bar tail tracers were mobilized for Bar 15 versus 73% mobility in the wetted channel (Table 6). When breaking this data down further however, 92% and 93% of wetted channel tracers were mobilized at Bar 15 during 2017 and 2018 deployments respectively, and only the 2016 tracers exhibited low mobility in the wetted channel (40% mobile). Tracers deployed on bar heads were almost always mobilized (100% mobile for Bar 6, and 98% for Bar 15), suggesting that these areas are active erosional sites. These differences

Figure 8. Path length frequency distribution for (a) Bar 6, (b) Bar 7, and (c) Bar 15 mobile tracers (i.e., transported more than 10 m downstream).
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highlight the importance that deployment strategy and channel morphology can have on tracer dispersion, as recently shown by McDowell and Hassan (2020).

Box plots of tracer scaled path lengths for each seeding morphologic unit for each study site is presented in Figure 14. For Bar 6 (2015, 2017, and 2018 deployments), tracers seeded on the bar tail had a lower median scaled path length than those seeded on the bar head or in the wetted channel (Table 6). However, differences in the distributions between seeding morphologies was not statistically significant (Kruskall Wallis test, \( p > 0.05 \)). Bar 6 tracers tended to be deposited upstream of the bar apex regardless of where they were initially seeded, with clustering coincident with the development of the migrating gravel sheet terminating at the bar apex (Figure 13). For the 2016 Bar 6 deployment, there was a statistically significant difference (Mann Whitney U test, \( p < 0.001 \)) between the path length distribution of tracers seeded in the wetted channel (specifically a pool in this case) and those seeded on the bar edge. Bar edge tracers were restricted to path lengths less than 0.3 bar lengths downstream, while those seeded in the pool more frequently traveled farther downstream, with a maximum observed path length of 0.78 bar lengths (Table 6). Again, bar edge tracers appear to become incorporated into the migrating gravel sheet at this location. The scaled path length distributions were significantly different (Mann Whitney U test, \( p < 0.001 \)) for Bar 7 tracers seeded in the wetted channel relative to those seeded on the adjacent bar tail, with median path lengths of 0.39

Figure 9. Tracer recovery locations for Bar 6 deployed in (a) 2015, (b) 2016, (c) 2017, and (d) 2018. Note that these maps include tracers recovered in surveys two or more years after deployment.
and 0.21 bar lengths for wetted channel and bar tail tracers respectively. Similarly, Bar 15 tracers seeded on the bar tail had a significantly different path length distribution than either those seeded on the bar head \((p < 0.001)\) or bar tail \((p < 0.05)\) based on a Kruskall Wallis test and Dunn’s post hoc comparison. Overall, tracers seeded on bar tails tended to be both less mobile and were displaced shorter distances on average than those seeded either in the wetted channel or on bar heads, underlining the spatially variable nature of bed-material transport in bar-dominated channels.

In summary, the results indicate that particle dynamics are directly tied to the local channel morphology. Median tracer path lengths were consistently focused upstream of the bar apex of the first downstream bar, and less than 5% of tracers escaped the first downstream bar after 1 year of transport. Furthermore, Bar 6 and Bar 7 tracers were preferentially deposited on the bars relative to the wetted channel, with clustering along bar margins and near the bar apex region. Bar 15 tracers were more evenly split between the wetted channel and bar surfaces. In addition to this, differences were observed between the dynamics of tracers installed in different morphologic units. Those installed on bar tails were less likely to be entrained than those beginning in the wetted channel or on bar heads, and had shorter median path lengths.

### 3.3. Morphologic Change and Particle Burial

Path length data demonstrated a relation with channel morphology which can be further explored in relation to the geomorphic change detection analysis and by incorporating tracer burial depth data. DoDs are presented for 2015–2018 and 2018–2019 for the Bars 6–7 reach in Figure 15 with tracer burial depths and locations overlaid on top. Bar 15 DoDs with buried tracers for the same periods are presented in Figure 16. Tracers that were located using the cord antenna system but were too deep to be detected by the wand antenna (or physically recovered), were estimated to be deeper than 30 cm (a conservative estimate of the maximum detection range of the wand antenna). Tracers recovered in pools were not physically recovered so burial depth was unknown and they were not included in burial depth analyses.
For the Bar 6–7 reach, a general pattern of downstream bar and bend migration was observed over the period of study. More than 80% of bar surface area was net depositional for both the 2015–2018 and 2018–2019 DoDs for Bar 6 and Bar 7, with average values of 0.4–0.6 m of net vertical deposition (Table 7). On Bar 6 the greatest magnitude of net deposition (1.7–2.1 m) occurred at the downstream end of the advancing gravel sheet near the bar apex (Figure 13; Table 7). The greatest net erosion on Bar 6 occurred in two locations: erosion of the inner bank near the bar head (2015–2018), and pools scoured at the bar tail just upstream of a bedrock outcrop (2018–2019) (Table 7). For Bar 7 a similar pattern occurred with maximum net deposition observed near the bar apex (∼1.1 m of vertical deposition) and maximum net erosion focused at the inner bank near the bar head (Table 7). In addition to the changes occurring on bar surfaces, the banks opposite of Bar 6 and Bar 7 retreated over the study period, though the magnitude of bank erosion was greatest opposite of Bar 6. In summary, from 2015 to 2019 there was a pattern of lateral bar accretion and downstream bar migration coupled with bank erosion.

The Bar 15 DoDs exhibited a similar pattern of topographic change to the Bar 6 and Bar 7 reach. Overall, the surface area of Bar 15 was net depositional for both the 2015–2018 and 2018–2019 DoDs with average net vertical deposition of 0.3–0.4 m (Table 7). The greatest magnitude of net deposition recorded on the surface of Bar 15 (∼1.6 m) occurred mid-way down the bar and appears associated with a large fallen tree residing in the area (Table 7). The maximum net erosion on Bar 15 focused on the inner bank near the bar head (Table 7), and bank retreat was observed.

Figure 11. Tracer recovery locations for Bar 15 deployed in (a) 2016, (b) 2017, and (c) 2018. Note that these maps include tracers recovered in surveys two or more years after deployment.

Figure 12. Box plots of scaled tracer path lengths for each study site. Note: \( n_m \) is the number of mobile tracers; the box represents the 25th and 75th percentiles; the line inside the box is the median; vertical lines represent the 10th and 90th percentiles; black dots represent the maximum observed path length. Path length data are scaled by the length of the bar at which they were seeded.
opposite the downstream portion of Bars 15 and 16 (Figure 16), leading to a pattern of downstream bar and bend migration.

The distribution of tracer burial depths for Bar 6, 7, and 15 are presented in Figure 17. The maximum burial depth for Bar 6 tracers was 52 cm, although 34% of tracers were not physically recovered, and we suspect that they are buried deeper than 30 cm (because they were not detected with the wand antenna) (Figure 17a). Assuming that these tracers were buried beyond this depth, the median burial depth for Bar 6 tracers was 25 cm. A maximum burial depth of 47 cm was recorded for Bar 7 tracers, with 30% of tracers not physically recovered, likely due to deep burial, resulting in an overall median burial depth of 18 cm (Figure 17b). For both locations, more than 40% of tracers were recovered at depths exceeding the commonly quoted assumed active layer depth of $\sim 2D_{90}$ of the surface (Hassan & Bradley, 2017; Wilcock & McArdell, 1997). Tracer burial data from Bars 6 and 7 suggests that maximum active layer thickness may be 50 cm or greater locally. This indicates that in this type of channel the particle exchange during bed material transport may operate at depths beyond a surface layer a few grains thick for at least a portion of the bed and active layer depth is governed by the distribution of bar-scale erosion and deposition. The median burial depth of 10 cm for Bar 15 tracers was the lowest of the three study sites (Figure 17c). The lower tracer burial on Bar 15 relative to the other sites aligns with the lower magnitudes of topographic change observed on DoDs relative to the Bar 6 and Bar 7 reach. Because morphologic change drives burial depth, and local bar dynamics and morphology differ between sites, this produces differences in tracer dispersion and burial between sites.

In general, the spatial pattern of tracer deposition was well-reflected in the DoDs. For Bar 6, 49% of tracers were recovered in areas corresponding to known geomorphic change on the DoDs (Table 8). Of these tracers 88% resided in depositional cells and 12% in erosional cells. The deepest buried tracers (>30 cm) tended to be deposited near the apex of Bar 6, which aligns with the downstream extent of the migrating gravel sheet in this area (Figure 13). For Bar 7, 65% of recovered tracers were located in areas of known geomorphic change, with 91% in depositional cells and 9% in erosional cells (Table 8). The deepest buried Bar 7 tracers were recovered either near the bar apex or on the bar at the launch line (Figure 15). Due to the lack of topographic data captured in the wetted channel, only 23% of tracers from Bar 15 were recovered in areas of known geomorphic change, in large part due the high proportion of Bar 15 tracers that remained in the initial riffle in which they were seeded. For those located in areas of known change, 70% were located in depositional cells, and 30% in erosional cells (Table 8). Across all three study sites, tracers tended to be recovered in areas of deposition on DoDs when there was known geomorphic change, which supports the idea that particle transport and deposition is tied to overall channel morphodynamics. For the portion of tracers recovered in areas of indeterminate change on DoDs the link between particle deposition and topographic change cannot be established. However, these areas were most commonly bar margins and riffles (Figures 15 and 16), areas that are likely to be depositional environments.

There was no correlation observed between tracer burial depth and the corresponding elevation change from DoDs for any of the three sites (Figure 18). This is perhaps not surprising since the timing between tracer deployments and recoveries does not match the time between topographic surveys used to produce the DoDs and because tracers could only be recovered within approximately 50 cm of the bed surface whereas scour and fill occurred at depths beyond this (Table 7). Overall, the mean burial depth of tracers recovered in depositional cells was greater than those in recovered in erosional cells for each site, though the low number of tracers recovered in erosional cells prevents the results from being statistically significant (Table 8).

### Table 5

|               | Bar 6 | Bar 7 | Bar 15 |
|---------------|-------|-------|--------|
| $n_m$         | 259   | 250   | 218    |
| $r_m$         | 0.89  | 0.87  | 0.78   |
| $L_{50}$      | 0.22  | 0.34  | 0.40   |
| $n_{L>1.0}$   | 1     | 12    | 8      |
| $r_{L>1.0}$   | 0.00  | 0.05  | 0.04   |
| $L_{\text{max}}$ | 2.02  | 1.41  | 1.34   |

Notes: $n_m$, number of mobile tracers; $r_m$, mobility rate; $L_{50}$, median scaled path length; $n_{L>1.0}$, number of tracers transported more than 1 bar length downstream; $r_{L>1.0}$, tracer escape rate; $L_{\text{max}}$, maximum observed scaled path length.
In summary, the results indicate that individual particle deposition and burial is related to the spatial pattern of bar development and re-working. Bars were net depositional over the study period, with lateral bar accretion and downstream bar migration observed for each site. Tracers were preferentially deposited in net depositional areas on DoDs relative to erosional areas, at average depths close to or exceeding twice the local $D_{90}$. This evidence suggests that particle exchange in this type of river operates at depths beyond a thin bed surface layer.

### Table 6
**Tracer Mobility and Path Length Breakdown by Initial Morphologic Unit**

| Initial morphologic unit | Bar head | Wetted channel | Bar tail |
|--------------------------|----------|----------------|----------|
| $r_m$                    | 1.00     | 0.96           | 0.77     |
| $L_{50}$                 | 0.27     | 0.21           | 0.21     |
| $L_{\text{max}}$         | 0.86     | 0.92           | 2.02     |

| Initial morphologic unit | Bar 6 (2016) | Bar edge | Wetted channel |
|--------------------------|--------------|----------|----------------|
| $r_m$                    | 0.75         | 0.96     |
| $L_{50}$                 | 0.11         | 0.17     |
| $L_{\text{max}}$         | 0.28         | 0.78     |

| Initial morphologic unit | Bar head | Wetted channel | Bar tail |
|--------------------------|----------|----------------|----------|
| $r_m$                    | 0.94     | 0.82           |          |
| $L_{50}$                 | 0.39     | 0.21           |          |
| $L_{\text{max}}$         | 1.24     | 1.41           |          |

| Initial morphologic unit | Bar 6 (2016) | Bar edge | Wetted channel |
|--------------------------|--------------|----------|----------------|
| $r_m$                    | 0.98         | 0.73     | 0.80           |
| $L_{50}$                 | 0.50         | 0.42     | 0.27           |
| $L_{\text{max}}$         | 0.93         | 1.27     | 1.34           |

**Figure 14.** Box plots of scaled tracer path lengths as a function of initial morphologic unit. Note: the box represents the 25th and 75th percentiles; the line inside the box is the median; vertical lines represent the 10th and 90th percentiles. Path length data are scaled by the length of the bar at which they were seeded.
3.4. Grain Size Effects on Tracer Mobility and Path Length

The path length of San Juan River tracers exhibited a similar relationship with grain size as described previously in the literature (Church & Hassan, 1992; Hassan & Bradley, 2017), with particles larger than the local D_{50} exhibiting a steep decline in mean path length with increasing grain size, and particles smaller than...
Table 7
Summary of Net Elevation Changes on Bar Surfaces for 2015–2018 and 2018–2019 DoDs

| Location | 2015–2018 | 2018–2019 |
|----------|-----------|-----------|
|          | % Erosion | % Deposition | % Erosion | % Deposition |
| Bar 6    | 20        | 80        | 7         | 93          |
| Bar 7    | 11        | 89        | 4         | 96          |
| Bar 15   | 23        | 77        | 14        | 86          |

| Location | Net elevation change (m) | 2015–2018 | 2018–2019 |
|----------|--------------------------|-----------|-----------|
|          | Eros<sub>avg</sub> | Eros<sub>max</sub> | Dep<sub>avg</sub> | Dep<sub>max</sub> | Eros<sub>avg</sub> | Eros<sub>max</sub> | Dep<sub>avg</sub> | Dep<sub>max</sub> |
| Bar 6    | −0.47 | −2.02 | 0.59 | 2.14 | −0.70 | −3.45 | 0.41 | 1.71 |
| Bar 7    | −0.45 | −1.41 | 0.42 | 1.12 | −0.48 | −2.24 | 0.41 | 1.16 |
| Bar 15   | −0.45 | −2.79 | 0.34 | 1.61 | −0.72 | −3.85 | 0.43 | 1.64 |

Abbreviation: DoD, DEMs of difference.

Notes. The percentage of erosion and deposition reflects the relative proportion of cells that were net depositional and erosional on bar surfaces. Average erosion and deposition values are likely to be inflated relative to the true changes occurring on bar surfaces, as the change detection algorithm better captures high-magnitude changes compared to low-magnitude changes which may be indistinguishable from noise.

Figure 17. Tracer burial depths for (a) Bar 6, (b) Bar 7, and (c) Bar 15.
the local D$_{50}$ showing less variation in mean path length with increasing grain size (Figure 19). However, while the mean travel distance of each tracer size fraction follow this nonlinear trend, there is a large amount of scatter in the data suggesting that other factors, such as channel morphology and hydrologic conditions, play an important role on particle transport. The relative mobility of each tracer size fraction is presented in Figure 20, where the tracer size class is scaled by the local surface D$_{50}$. In general, relative mobility decreased with increasing relative grain size. This trend was moderately strong for the 2016–2017 tracer displacements ($R^2 = 0.69$) and weak for the 2017–2018 and 2018–2019 displacements ($R^2 = 0.29$ and 0.36 respectively). The 2016–2017 season was the “driest” year during the study period, in terms of both Q$_{max}$ and the total duration of competent flow, suggesting that hydrologic conditions limited mobility of larger sizes compared to the “wetter” years when most of the bed was mobilized regardless of grain size.

4. Discussion

The results from particle tracking in the San Juan River reveal insights into bed particle dynamics in a wandering style gravel-bed river, which has seldom been a focus in previous studies. The recovery rate of tracers in this study combined with topographic change data is unique for a channel of this size and as such provides important information on the nature of bedload transport in these systems. This data can also help to develop tracer displacement statistics and models for this type of river with respect to morphologic change and bar development.

The intrabedform, or intrabar, transport that was observed on the San Juan River, aligns with results from the few previous bedload tracking results on larger, bar-dominated channels. On a meandering section of the Ain River, France, Rollet et al. (2008) conducted a PIT tag tracer study, recovering 36% of tracers after 1 year, with an average travel distance of 50 m, less than 1 bar length downstream. While interpretation of particle transport was limited by the low recovery rate, the authors noted that over the tracer monitoring period, a thick sedimentary layer had accreted on the edge of the gravel bar immediately downstream of the tracer injection site and posited that lost tracers were likely buried at this location, beyond their antenna’s maximum range of detection. This description of bar growth is similar to changes observed on Bar 6 of the San Juan River, where tracer clustering and lateral accretion was focused at the bar apex. In another PIT tag study, conducted on a wandering riffle-pool reach of the Durance River, France, Chapuis et al. (2015) recovered 40% of tracers after four months, with an average particle path length of 83 m, again indicative of transport within one riffle-pool unit. As observed on the San Juan River, the authors noted that particles deployed on bar tails were either immobile or only displaced short distances, while those seeded closer to the thalweg were transported farther downstream. Similar variations in transport conditions between morphologic units were reported from PIT tag and painted tracer data collect from a wandering reach of the Parma River, Italy (Brenna et al., 2019). The spatial variability in bedload transport, intrinsic to the riffle-pool morphology, means that deployment strategy has a strong influence on tracer mobility and resultant path length distributions. In the cases of the Ain, Durance, and San Juan Rivers, tracers tended to remain within one morphological unit, with minimal transport downstream. This provides direct evidence that in addition to hydraulic

| Location | Fraction of recovered tracers | Fraction of recovered tracers | $B_{avg}$ (cm) | Fraction of recovered tracers | $B_{avg}$ (cm) |
|----------|-------------------------------|-------------------------------|----------------|-------------------------------|----------------|
| Bar 6    | 0.51                          | 0.06 (0.12)                   | 12.1           | 0.43 (0.88)                   | 18.5           |
| Bar 7    | 0.35                          | 0.06 (0.09)                   | 5.1            | 0.59 (0.91)                   | 10.3           |
| Bar 15   | 0.73                          | 0.08 (0.30)                   | 5.7            | 0.19 (0.70)                   | 10.6           |

Abbreviation: DoD, DEMs of difference.

Note. The fraction of recovered tracers in brackets reflect recovery in areas of known topographic change (i.e., tracers deposited in areas of indeterminate change removed).

Table 8

Summary of Tracer Recovery in Areas of Topographic Change

Figure 18. Tracer burial depth plotted against net elevation change in the DoD for the cell in which the tracer was buried.
controls, channel morphology influences particle dynamics in these systems. Particle trapping and burial in association with bar development appears to be an important consideration in modeling sediment behavior in this type of river, and therefore in the bedload transport process more generally. This emphasizes the morphologic control on bed particle transport, a point that was also raised in a re-analysis of bedload tracking data by both Vázquez-Tarrío et al. (2018) and McDowell and Hassan (2020).

The pattern of particle displacement and deposition on the San Juan River reflected a combination of morphologic controls and seasonal variations in the flow regime. During the 2017–2018 winter, the wettest study year, bedload deposition focused at the apex of the first bar downstream for Bars 7 and 15, and either slightly positive or symmetrical distributions around the bar apex were observed. However, during years with more moderate floods and lower competent flow volume, path length distributions tended to be positively skewed, with lower tracer mobility. It should be noted here that these results specifically pertain to particles seeded on the bed surface near the bar head. At least this is the case for the first movement of tracers, while subsequent events in the same year could be acting on both surficial and buried tracers. Differences in the spatial distribution of tracers between sites also appears related to channel shape and bar morphology. At Bar 6, the high-amplitude bend, and migration of a coarse bedload sheet, appear to restrict transport to the bar tail. Whereas tracer deposition on the bar tail was more common for Bars 7 and 15, bars that are less well-developed laterally. Overall, the results from this study provide some validation to the findings from Pyrce and Ashmore's flume experiments (2003a; 2005) whereby they observed bi- and multimodal path length distributions during bar-forming discharges, with modes coincident with bar apexes, and spatial patterns of tracer deposition related to bar-scale patterns of erosion and deposition.

To further compare tracer transport on the San Juan River with results from the literature, Figure 4 from Vázquez-Tarrío et al. (2018), has been re-created in Figure 21 with data from Bars 6, 7, and 15 on the San Juan River. The original figure contains travel distances of tracers starting from both unconstrained (i.e., initial movement after seeding, as done in this study) and constrained (movement after tracers have been incorporated into the bed) positions. In this graph, dimensionless stream power ($\omega^*$) was calculated as in Eaton and Church (2011), to compare flow strength and particle transport across rivers of different scales. Mean scaled travel distances were scaled using a morphologic length scale (i.e., the spacing of macroscale bedforms), which for the San Juan River meant the riffle-riffle spacing. Data from the 2015–2016 deployment on Bar 6 were not plotted on this graph due to the low recovery rate (33%) and uncertainty in the statistics of particle displacement for this year. The San Juan River data appear in line with results from riffle-pool channels, showing a positive relationship between dimensionless stream power and mean scaled travel distance, though travel distances on the San Juan River were relatively low (Figure 21). What is more interesting though, and as noted by Vázquez-Tarrío et al. (2018), is that riffle-pool channels share a common trait in that they rarely exhibit average travel distances beyond 1–2 length-scale units (regardless of constrained or unconstrained starting position), generally at the lower end of the range for other channel types. It appears that riffle-pool channels have limited transport distances compared with other channels, presumably a result of bars and riffles constraining particle movement. Over the long-term, path lengths are therefore likely to be limited by the rate of bar development and re-working, which in turn plays an important role in overall channel evolution and stability (Reid et al., 2019; Rice et al., 2009). The path length distributions presented in this study reflect annual particle transport over a 4 year period. The annual displacements occur from multiple events in the well-defined high-flow season from October to March. This is consistent each year although exact magnitudes vary. Particles are “slaved” to the bar development and

![Figure 19. Tracer path lengths (L) as a function of scaled grain size (D) for each size fraction of tracers. Path length is scaled by the mean path length of the size fraction containing the local $D_{50}$. Grain size is scaled by the $D_{50}$ of local bed surface material. Colored dots represent data summarized for a specific size fraction of tracers, hollow gray dots represent the raw data. Note that previous studies have scaled particle size by the local subsurface $D_{50}$ whereas the San Juan data was scaled by the local surface $D_{50}$ as no bulk grain size sampling was carried out.](image-url)
few move through to the next bar any given year. The preferential deposition and incorporation of tracers in bars has been previously demonstrated to hold true over longer time-scales in gravel-bed rivers in general (e.g., Ferguson et al., 2002; Haschenburger, 2013). Therefore, we expect that these annual path lengths are representative of the normal bar development except in the case of a high magnitude flood sufficient to disrupt the existing bar and channel configuration, in which case displacements would be expected to reflect that larger scale morphological change. For morphological flux calculation this would produce an annual flux based on net morphological change and tracer distances scaled to the bars.

The tracer burial data provide context to the path length analysis and revealed insights into patterns of sediment transfer and storage. Burial depths up to, and likely exceeding, 50 cm were recorded at each of the three sites, although the absolute magnitude of maximum particle burial depth was not obtained in this study due to methodological constraints. However, the development of “wobblestone” technology looks to provide a promising solution to estimating particle burial without the need to physically recover the particles and disturb the bed (Papangelakis et al., 2019). Overall, tracer burial aligned with the patterns of bar-scale deposition observed on DoDs, which raises two important points. First, this suggests that as individual particles are transported to and buried in local areas of deposition, they become incorporated into the channel morphology, and future movement will be limited by the rate of bedform (or bar) migration. This relates back to Neill’s (1987) original speculation that over the long term, average particle path lengths may be inferred from the channel morphology. Second, the fact that more than 30% of buried tracers were recovered at depths beyond twice the local D90 for each site suggests that the concept of a shallow active layer less than two particles deep does not reflect the nature of bedload processes occurring in this type of channel. It should be noted however, that the winter storm season was treated as a single event for this study, and that there were multiple events each year that may have produced the reported burial depths. As described by Ashmore et al. (2018), it appears that bed particle deposition and burial in larger, more dynamic rivers is controlled by bar-scale patterns of deposition, whereby the active layer is spatially nonuniform and maximum burial depths occur on the scale of vertical changes in bed level associated with those dominant morphological processes. This is important because the dimensions of the active layer, when combined with the virtual velocity of bed particles, gives the morphological bedload transport rate (Haschenburger & Church, 1998; Liébault & Laronne, 2008; Mao et al. 2017; Vázquez-Tarrío and Menéndez-Duarte, 2014; Vericat et al., 2017).

5. Conclusion

Annual tracking of bed particles over 4 years in a large bar-dominated wandering gravel-bed river, combined with repeat LiDAR surveys of topographic change, show that particle displacements are closely tied to the bar scale of the river and to patterns of bar development and accretion. This has implications for annual and event-based path length when applied to virtual velocity and morphological estimates of bedload, and points to possible differences in path length characteristics and controls between rivers of different morphology and size.

Annual tracer path lengths were tied directly to the local channel morphology with tracer deposition and burial reflected by areas of deposition on DoDs over the time period of tracer tracking. Year to year differences in tracer path lengths related to the annual flow regime were also observed, though appear secondary to morphologic constraints because...
tracers were rarely transported past the bar immediately downstream of their injection site. Tracer deposition focused along bar margins and to a lesser extent, the surface of the downstream portion of bars, reflecting the downstream bar migration and lateral bar accretion observed on DoDs. Burial information indicates that active layer depths are tied to topographic change and exceed the grain exchange depth that is typically assumed in bedload analysis and modeling in gravel-bed rivers. This highlights the fundamental importance of bar development and re-working underpinning bedload transport processes in bar-dominated channels and supports recent analyses of morphological effects in tracer dispersion as indicated by Vázquez-Tarrio et al. (2018). Ultimately, short-term particle displacements in bar-dominated channels are linked with the morphological style of channel evolution and therefore differs from bedload processes in small, plane-bed channels, for which much previous analysis and theory have been developed.

Data Availability Statement

The data used in this study can be found in the Scholars Portal Dataverse (https://doi.org/10.5683/SP2/UQGZCG).

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