



# First Insights Into the Fine-Scale Movements of the Sandbar Shark, *Carcharhinus plumbeus*

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The expanding use of biologging tags in studies of shark movement provides an opportunity to elucidate the context and drivers of fine-scale movement patterns of these predators. In May 2017, we deployed high-resolution biologging tags on four mature female sandbar sharks Carcharhinus plumbeus at Ningaloo Reef for durations ranging between 13 and 25.5 h. Pressure and tri-axial motion sensors within these tags enabled the calculation of dive geometry, swimming kinematics and path tortuosity at fine spatial scales (m-km) and concurrent validation of these behaviors from video recordings. Sandbar sharks oscillated through the water column at shallow dive angles, with gliding behavior observed in the descent phase for all sharks. Continual V-shaped oscillatory movements were occasionally interspersed by U-shaped dives that predominately occurred around dusk. The bottom phase of these U-shaped dives likely occurred on the seabed, with dead-reckoning revealing a highly tortuous, circling track. By combining these fine-scale behavioral observations with existing ecological knowledge of sandbar habitat and diet, we argue that these U-shaped dives are likely to be a strategy for bentho-pelagic foraging. Comparing the diving geometry of sandbar sharks with those of other shark species reveals common patterns in oscillatory swimming. Collectively, the fine-scale movement patterns of sandbar sharks reported here are consistent with results of previous biologging studies that emphasize the role of cost-efficient foraging in sharks.

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# INTRODUCTION

An understanding of the movement ecology of marine megafauna requires investigation of patterns at a range of spatial and temporal scales across an individual's lifetime (Nathan et al., 2008). Broad-scale movements determined by acoustic and satellite tracking can reveal migratory pathways, environmental preferences and patterns in habitat use and residency (Braccini et al., 2017; Ferreira et al., 2019). Fine-scale movements recorded by advanced biologging tags allow the quantification of patterns in activity, feeding and energetics (e.g., Gleiss et al., 2013; Brownscombe et al., 2014; Brewster et al., 2018), and can provide behavioral and physiological context to other studies of broad-scale movements (Whitney et al., 2012, 2019).

Recent biologging studies on whale sharks (*Rhincodon typus*) and oceanic whitetip sharks (*Carcharhinus longimanus*) have shown that the foraging, energetics and movement patterns of sharks are likely linked through behaviors that minimize energy outputs, while maximizing

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foraging opportunities (cost-efficient foraging) (Gleiss et al., 2011b; Meekan et al., 2015; Papastamatiou et al., 2018a). To date, the applicability of this concept to other sharks is difficult to determine, as relatively few species have been studied using a biologging approach.

Here, we used biologging tags to investigate the fine-scale movement behaviors of adult sandbar sharks (Carcharhinus plumbeus) for the first time. This species is globally distributed in tropical and warm temperate coastal waters, and is commercially important throughout its range (McAuley et al., 2007; Ebert et al., 2013). K-selected species traits, such as low fecundity and late age at maturity, targeted fishing and predicted population declines have led to the sandbar shark being listed as Vulnerable on the IUCN Red List (Musick et al., 2009). Largely because of its commercial importance, there exists a robust understanding of the broad-scale (10-1,000 km) distribution, movements and diet of the species (Stevens and McLoughlin, 1991; McElroy et al., 2006; Ellis and Musick, 2007; Grubbs et al., 2007; McAuley et al., 2007; Barnes et al., 2016; Braccini and Taylor, 2016; Braccini et al., 2017). Tracking studies have found that sandbar sharks are generally wide-ranging, undergo large migrations (>1,000 km) but also display evidence of site-fidelity (Grubbs et al., 2007; Papastamatiou et al., 2010; Barnes et al., 2016; Braccini et al., 2017). Diet studies in a number of regions have found a preference for feeding on benthic prey, predominately demersal fishes, cephalopods and crustaceans (Cliff et al., 1988; Stevens and McLoughlin, 1991; McElroy et al., 2006; Ellis and Musick, 2007). Our use of biologging tags allowed our study to record fine spatial and temporal scale (m and seconds) behavior that linked foraging to movement. Our aim was to determine the extent to which these movements conformed to patterns of cost-efficient foraging seen in other species monitored with this type of technology. Patterns consistent across species support the contention that this type of foraging is a key determinant of the vertical movements of sharks and other megafauna in marine environments.

# MATERIALS AND METHODS

## **Data Collection**

In Western Australia, mature sandbar sharks are predominately found in north and north-west tropical waters (McAuley et al., 2007; Braccini and Taylor, 2016), and some degree of site fidelity have been revealed around Ningaloo Reef (**Figure 1A**) (Braccini et al., 2017). We captured sandbar sharks at this locality using baited drumlines inside the reef lagoon (22.99°S, 113.8°E) in May 2017 (**Figure 1A**). Drumlines were equipped with a single 20/0 circle hook baited with fish scraps. Three drumlines were deployed ~100 m apart at depths of 8–12 m between 7:00 and 16:00, and were monitored continuously. Once a shark was caught, it was secured alongside a 5.8 m vessel by the leader of the fishing line and a tail rope. Each shark was measured and its sex recorded before a biologging tag was clamped to the base of its dorsal fin.

Either a CATS Diary Tag (Customized Animal Tracking Solutions, Australia) or CATS Cam Tag was deployed on

sandbar sharks (Figures 1B,C). Both are equipped with triaxial accelerometers, magnetometers and gyroscopes, depth, temperature, and light sensors. In addition, the Cam tag housed a HD video camera, and speed sensors were present, but not functional, in the diary tags. Active sensors recorded all parameters at 20 Hz, and video recorded at pre-programmed hours of the day. CATS tags were attached to a stainless steel spring clamp [CATS, Australia (Gleiss et al., 2009; Chapple et al., 2015)] via docking pin and a corrodible galvanic timed release (GTR, Ocean Appliances, Australia) allowing the tags to be rigidly attached to the dorsal fin. The GTRs were designed to dissolve in seawater after 24 h. After this time the tag would release from the clamp and float to the surface. Floating tag packages were tracked down in the days following tagging using a hand-held VHF receiver operated from the vessel (Lear and Whitney, 2016). The magnesium sleeve on the clamp also dissolved after 7 days, detaching the clamp from the shark.

#### Data Processing Depth Record

Zero offset corrections were applied to the depth record based on times when sharks were being tagged at the surface. The depth record was then split into vertical swimming phases ("ascending," "descending," and "level swimming") using vertical velocity (VV). To do this the depth record was firstly smoothed using a 10 s running mean and the average VV was calculated by taking the difference of this smoothed depth between successive points at 1 s intervals. Ascents and descents were defined where VV exceeded an absolute value of 0.05 m/s for more than 10 s, and level where this value was not exceeded (Whitney et al., 2016).

## Tri-axial Sensor Data

Data recorded by the accelerometer (acceleration) and gyroscope (angular velocity) were analyzed using Igor Pro ver. 7.0.4.1 (Wavemetrics, Inc. Lake Oswego, USA) and Ethographer (Sakamoto et al., 2009). The gravitational component of acceleration (static acceleration) was determined using a 3 s box smoothing window on the raw acceleration data (Shepard et al., 2008). Shark body pitch angles (orientation of the shark with regard to the horizontal plane) were derived by calculating the arcsine of the static acceleration in the surging (posterioranterior) axis. To correct for the tag attachment angle on each individual shark, we determined the pitch when the shark was swimming at a constant depth (when VV was equal to zero), and subtracted this value from all pitch estimates (Kawatsu et al., 2009). The dynamic component of acceleration was calculated by subtracting the gravitational component from the raw acceleration for each axis. We then used a continuous wavelet transformation on the dynamic component of the sway (lateral) axis to calculate the acceleration signal amplitude and frequency of tailbeats. Using these same methods, amplitude and frequency were calculated using the angular velocity data, and the resulting signals were compared with those derived from the acceleration data to determine the best measure of tailbeat kinematics.

The angular velocity data produced the clearest tailbeat signal and consequently was used to quantify the incidence of gliding behavior (cessation of tailbeats for more than 1 s) using

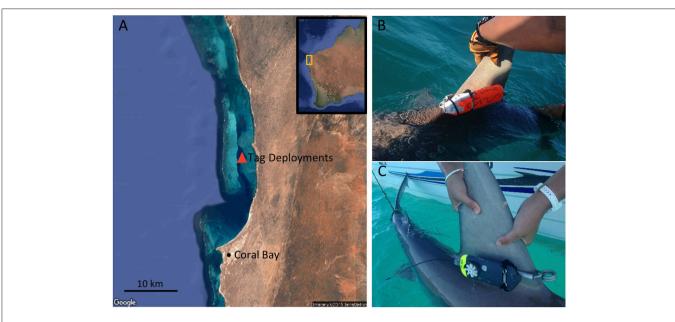


FIGURE 1 | Tagged sandbar sharks at Ningaloo Reef. (A) Location of tag deployments. (B) CATS Carn tag clamped to the dorsal fin. (C) CATS Diary tag clamped to the dorsal fin.

a combination of two methods. Gliding behavior was firstly computed using the k-means cluster function in Ethographer. This algorithm clusters the spectra computed by the wavelet transformation based on similarity of shape. The behavioral spectrum with the lowest peaks in amplitude was assumed to represent gliding behavior (Nakamura et al., 2011). The incidence of the resulting cluster was then inspected against the dynamic sway data. As this cluster did not match with gliding behavior in some individuals (i.e., tailbeats evident in sway data classified to be gliding and vice versa), a mask was created using the characteristics of the tailbeat amplitude and frequency where the cluster was judged to correctly classify gliding behavior (from visual inspection of the dynamic sway and concurrent videos). This mask was then used to extract glides from all sharks, and an additional manual quality control was undergone in the case where the mask obviously misclassified glides.

Tailbeat signals were also examined for evidence of a recovery period following the stress of capture following Whitney et al. (2016).

#### Shark Heading and Pseudo-Track Calculation

Acceleration and magnetometer data were used to calculate head yaw angle (hereafter referred to as "heading") and pseudo-tracks in the software Framework 4 (Walker et al., 2015; Whitney et al., 2019). Heading calculations required input of orientation of the sensors in the tags, and magnetometer and acceleration data. The orientation of the device was corrected using orientation specifications provided by CATS (N. Liebsch pers. comm). Pitch and roll calculated from the accelerometer data were used to correct for tilt (Walker et al., 2015). In addition to the calculations applied by Framework 4, we applied a 3 s box smoothing window to the heading data to filter out the dynamic movements of the sharks caused by tail beating that created significant yaw.

Heading was then used to calculate pseudo-tracks for each shark using dead-reckoning (Mitani et al., 2003; Wilson et al., 2007). These calculations operate under the principle that the heading and speed of an individual at time "t" can be used to calculate the position at time "t + 1" (Walker et al., 2015). As we had no estimate of speed, a fixed value was used to compute the pseudo-tracks. Because a fixed value of speed is unlikely to be representative of individual sharks, the dead-reckoning will accumulate errors over the course of the track. Therefore, tracks were used only for plotting short-term movements of the sharks in three-dimensional space ("pseudo-tracks") and could not be used to estimate geographical positions (Wilson et al., 2007; Walker et al., 2015). Heading data were also used to calculate the turning angles of individuals on a second-by-second basis. Data were resampled to a 1 s frequency and converted to a 0-360° scale. The minimum difference in angle between consecutive observations was used to determine turning angle.

#### Window Size and Statistics

The time window used for analysis was selected by determining the window for which the highest variance in turning angle was observed. This time window was estimated to be 15 min (900 s), and within this period, a number of vertical movement parameters were summarized including mean and maximum depth, ascent pitch, descent pitch, ascent VV, and descent VV. The proportion of time spent moving vertically (ascending and descending), termed the "diving ratio," (**Table 1**) was also calculated for each window as per:

 $Diving Ratio = \frac{Time \ vertically \ moving \ in \ window \ (seconds)}{Total \ time \ in \ sampling \ window}$ 

This parameter allowed us to discriminate between sampling windows where vertical movements consisted of continuous oscillations through the water column or extended periods of level swimming between dives. In addition, heading data were used to calculate measurements of path tortuosity and direction. Firstly, turning angles were summed for each window to obtain an estimate of tortuosity. This was also separately calculated for each vertical phase (level swimming, ascending, and descending), and standardized by the time spent in the relative phase within a sampling window (i.e., the mean). Secondly, the Circular package in R was used to calculate the mean resultant length, or  $\overline{R}$  of the heading, a measure of the concentration of unimodal circular data (Pewsey et al., 2013). When  $\overline{R}$  is close to or equal to one, points are closely clustered around the mean direction and are highly directional, and as  $\overline{R}$  approaches zero, points spread more evenly in a circle and become tortuous.

#### Video Analysis

Video recorded by the tags was analyzed in BORIS (Behavioral Observation Research Interactive Software, Friard and Gamba, 2016). This open-source software allows the user to set an ethogram and record the timing of these events. We set an ethogram to mark burst swimming events, presence of seabed in the field of view, habitat type, and other notable behaviors. Video was also used to validate parameters recorded and calculated from the tri-axial sensor data, such as tailbeat frequency, bursts and glides.

### **Statistical Analysis**

To investigate if changes in vertical movements were related to changes in horizontal movements, we constructed a series of

**TABLE 1** | Terms and definitions used repetitively throughout manuscript.

Terms	Definitions
Diving ratio	The proportion of time spent vertically moving (ascending or descending) within a 15 min sampling window
R	Mean resultant length. A measure of the concentration of the heading data. When $\overline{R}$ is equal to or close to one, points are closely clustered around the mean direction. When $\overline{R}$ approaches zero, points are spread more evenly in a circle.
Pseudo-track	An approximation of a sandbar sharks horizontal track.
Inshore	Inside the reef in depths of <25 m.
Offshore	Outside the reef in depths of $>25$ m.
VV	Vertical velocity (m/s)

#### TABLE 2 | Summary details of tagged sandbar sharks.

Shark ID Tag ID Deployment date **Deployment duration** Pre-caudal length (cm) Fork length (cm) Total length (cm) Girth (cm) Sex Shark 1 CC2 11/5/2017 8:54 25 h 33 min 141 F 156 185 82.5 Shark 2 CD1 11/5/2017 10:27 13 h 5 min 137 152 189 79 F Shark 3 CC1 11/5/2017 14:10 17 h 46 min 149 180 84 F 138 Shark 4 CD1 18/5/2017 12:29 17 h 15 min 125 140 176 75 F

CC: CATS camera tag, CD: CATS diary tag.

generalized linear models (GLMs) for each individual sandbar shark (N = 4). Diving ratio within a sampling window (see Window Size and Statistics) was set as the response variable in all models, and turning (a proxy for tortuosity) as the explanatory variable. The first set of models used summed turning angles to explore the general relationship between tortuosity and diving behavior in tagged sharks. For the remaining models, mean turning angle for each phase of vertical movement (level swimming, ascent and descent, see Window Size and Statistics) was sequentially set as the explanatory variable. Inspection for correlation at lag = 1 found no serial correlation in the data. Statistical analysis was performed in R 3.4.0 (R Core Team, 2017).

## RESULTS

Four mature female sandbar sharks ranging in length from 1.76 to 1.89 m TL were caught and tagged at Ningaloo Reef in May 2017 (two camera tag and two diary tag deployments, **Table 2**). Time from capture to release did not exceed 25 min for any of the tagged sharks. All deployed tags were recovered after a mean attachment duration of 18 h (range 13–25.5 h). All sensors recorded continuously for all four tag deployments, excluding the video camera, which only recorded a total 5.5 h and 30 min in two deployments.

The quantification of the post-release recovery period for sandbar sharks using tailbeat data was confounded by the immediate movement into deeper water by three of the tagged individuals (**Supplementary Figures 1**, **2**). For these sharks, increases in tailbeat frequency were associated with deeper dives. Shark 4 remained inshore and no recovery period was observed i.e., there was no apparent effect of capture on activity (**Supplementary Figure 1**). As a result, the entire recorded track from shark release to tag detachment for each individual was used in the analysis.

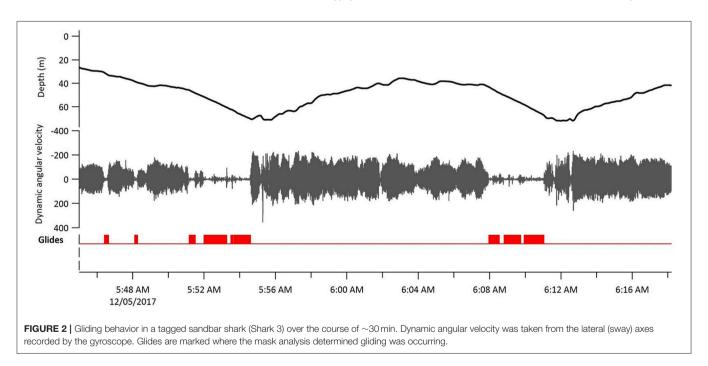
#### **Vertical Movements**

Tagged sandbar sharks swam at a mean depth of  $23.4 \pm 17.5$  m, reaching a maximum depth of 96.6 m (**Table 3**). Sharks moved vertically in an oscillating pattern for a mean  $57 \pm 17.6\%$  of their track at a mean ascent angle of  $11.4 \pm 2.2^{\circ}$  and descent angle of  $-11.5 \pm 3.5^{\circ}$  (**Table 3**). Sharks displayed a mean tailbeat frequency of  $0.65 \pm 0.19$  Hz, averaging  $0.72 \pm 0.12$  Hz on ascent and  $0.57 \pm 0.13$  Hz on descent. Gliding behavior was observed throughout  $13.2 \pm 14.3\%$  of the descent phase (range 0-75%), with a maximum continuous glide time of 210 s (**Figure 2**; **Table 3**). Sparse tailbeats often interrupted glides during descents

TABLE 3 | Summary statistics calculated for the entire dataset, and each tagged individual.

	Overall	Shark 1	Shark 2	Shark 3	Shark 4	
Mean depth (m)	23.4 ± 17.5	21.9 ± 18.5	34.6 ± 19.7	29.7 ± 13.8	10.7 ± 4.4	
Maximum depth (m)	96.6	96.6	73.3	73.7	19.9	
Mean temperature (°C)	$25.9 \pm 0.8$	$26.2 \pm 0.5$	$27 \pm 0.4$	$25.6 \pm 0.3$	$24.9\pm0.5$	
Temperature range	20-27.5	20-27.4	22.4-27.5	24.9-26.3	24–26	
Mean TBF (Hz)	$0.65\pm0.19$	$0.67 \pm 0.13$	$0.74 \pm 0.22$	$0.60\pm0.16$	$0.61\pm0.22$	
Mean ascent TBF (Hz)	$0.72 \pm 0.12$	$0.73\pm0.06$	$0.82 \pm 0.14$	$0.68 \pm 0.12$	$0.69\pm0.13$	
Mean descent TBF	$0.57\pm0.13$	$0.60\pm0.08$	$0.66\pm0.19$	$0.53\pm0.15$	$0.52\pm0.09$	
Mean level TBF	$0.66 \pm 0.13$	$0.68\pm0.07$	$0.77\pm0.17$	$0.59\pm0.13$	$0.62\pm0.12$	
Mean ascent angle ( $^{\circ}$ )	$11.4 \pm 3.5$	$10 \pm 4.7$	$11.7 \pm 2.1$	$13.2 \pm 2.5$	$11.4 \pm 1.8$	
Mean descent angle (°)	$11.5 \pm 2.2$	$11.8 \pm 2.5$	$11.6 \pm 2.3$	$11.1 \pm 2.2$	$11.3 \pm 1.4$	
Mean ascent VV (m/s)	$0.15\pm0.04$	$0.14\pm0.05$	$0.17\pm0.04$	$0.16\pm0.04$	$0.15\pm0.02$	
Mean descent VV (m/s)	$0.15\pm0.04$	$0.16\pm0.04$	$0.17\pm0.05$	$0.14\pm0.04$	$0.13\pm0.03$	
Vertical ratio	$0.57\pm0.18$	$0.58\pm0.15$	$0.62\pm0.17$	$0.55 \pm 0.17$	$0.53\pm0.21$	
% Descent gliding	$13.2\pm14.3$	$13.2 \pm 13.6$	$8.4 \pm 13.9$	$17.5 \pm 16.7$	$12.4 \pm 11.6$	
Max glide time (s)	210	196	210	179	131	
Mean R	$0.51 \pm 0.22$	$0.54 \pm 0.19$	$0.44 \pm 0.18$	$0.71 \pm 0.12$	$0.30\pm0.16$	
R range	0.007-0.96	0.13-0.96	0.08-0.88	0.39–0.95	0.007-0.76	
Mean sum of turning angles	$7,001 \pm 4,715$	$5,024 \pm 3,156$	$12,290 \pm 5,297$	$4,339 \pm 2,115$	$8,643 \pm 4,214$	
Mean turning angle level swim	$8.1 \pm 5.9$	$5.6 \pm 3.9$	$14.2 \pm 7.6$	$5.3 \pm 2.9$	$10.1 \pm 5$	
Mean turning angle ascending	$6.5 \pm 5.7$	$4.4 \pm 3.1$	$12.5 \pm 8.8$	$3.5 \pm 1.5$	$8.3 \pm 4.1$	
Mean turning angle descending	$7.7 \pm 5.4$	$6.0 \pm 3.7$	$14.1 \pm 7.1$	$4.8 \pm 2.4$	$8.6 \pm 3.8$	

Statistics were firstly calculated for 15 min windows, then summarized for the entire tagging duration. TBF, tailbeat frequency; VV, vertical velocity; R, mean circular length.



(Figure 2). For other summary statistics of vertical movements, see Table 3.

The depth record for each shark consisted of continual V-shaped oscillatory vertical movements, interspersed by occasional U-shaped dives with extended bottom periods

(Figure 3). These U-shaped dives occurred predominately around dusk for all individuals. Through a combination of seabed observations in the video, dead-reckoning and the gradual increase in window maximum depth as three sharks moved offshore, we assumed that the oscillations were limited

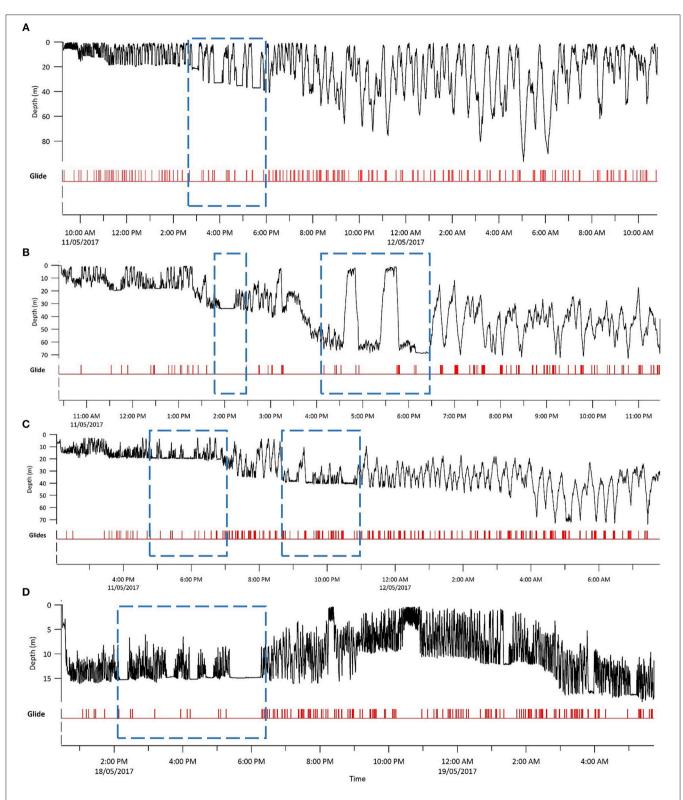


FIGURE 3 | Entire depth record and gliding behavior for each tagged sandbar shark. Glides are marked where the mask analysis determined gliding was occurring. Dashed blue squares denote extended periods of U-shaped diving, and associated looping behavior. (A) Shark 1; (B) Shark 2; (C) Shark 3; and (D) Shark 4. Note that both *x*- and *y*-axes scales differ between sharks.

vertically by the seabed. Comparing these maximum depths with bathymetry of the study site, three sharks moved offshore (maximum depths >70 m), and one shark appeared to remain inside the reef (Shark 4, maximum depth <20 m).

## **Tortuosity and Three-Dimensional Paths**

All sharks displayed periods of both directional and tortuous travel, with an overall mean  $\overline{R}$  of 0.51  $\pm$  0.22 (range among indivduals 0.007-0.96, Table 3). The GLMs revealed a significant negative relationship between tortuosity and diving ratio for three of the four individual sharks, with turning explaining 53-66% of the variation in diving ratio (Table 4; Supplementary Figure 3). Shark 2 showed no significant relationship between tortuosity and diving ratio, with this individual observed to display relatively high levels of tortuosity throughout the track in comparison to the other sharks (Table 3). Further investigation of the windows with the highest tortuosity (high sum of turning angles and low  $\overline{R}$ ), found tortuous windows to be associated with U-shaped dives (Figure 4). Due to extended periods of level swimming, the diving ratio was relatively low in sampling windows containing U-shaped dives. For all tagged sharks, pseudo-tracks displayed repeated circling on the bottom phase of these dives, irrespective of depth (Figure 5). For example, U-shaped dives dives recorded throughout dusk periods had circling phases at depths of 36.5, 67, 67, and 14.6 m for each individual shark, respectively (Figure 5). These bottom phases lasted a maximum of 55 min.

As sampling windows of low or high diving ratio represented U- or V-shaped dives, respectively, models were taken a step further to investigate changes in turning angle for each phase of vertical movement. Significant negative relationships were revealed between diving ratio and turning angles for both level and ascent phases in Sharks 1, 3 and 4 (**Supplementary Table 1**). The relationship was stronger for level swimming phases ( $R^2 = 0.32-0.42$ ) than ascent phases ( $R^2 = 0.12-0.19$ ). More tortuous ascents were evident in U-shaped dives, and directional ascents in V-shaped dives (**Figure 6**). No relationship was found between descent turning angles and diving ratio, and similar patterns of descent were found in both V- and U-shaped dives (e.g., **Figure 6**).

# Video

The 6 h of video validated gliding and turning behavior, and identified seabed habitat at the bottom of dives. No interactions with prey or conspecifics were observed, and the only other animals observed were trevally accompanying the shark, and one school of baitfish briefly swimming through the field of view. For Shark 1, a 24 min video recorded a U-shaped dive at dusk. The video confirmed the bottom phase of the dive occurred at the seabed in a sandy habitat with patchy weed and reef, and validated the circling behavior (**Supplementary Video 1**).

# DISCUSSION

Our deployment of biologging tags on adult female sandbar sharks revealed oscillatory and tortuous movement paths. Tagged sharks underwent continual V- and U-shaped dives through the water column, diving at shallow angles and displaying tortuous circling behaviors throughout the bottom phase of U-shaped dives.

# **Tortuous Circling Behavior**

As traveling in a straight line is the most energetically efficient form of movement, the higher energetic costs of turning should be offset by other benefits, such as enhanced foraging success (Wilson et al., 2013). Because turn costs increase linearly with angle, Wilson et al. (2013) predicted that more acute turns should be strongly motivated, and would have a tendency to be clustered in time and space. The tortuous circling behavior observed in all sandbar sharks supported these predictions, occurring on the seabed throughout the bottom phase of U-shaped dives, and predominately occurring around dusk. Possible explanations for this behavior include intra-specific interactions, the presence of favorable oceanographic conditions, or foraging behavior. Given that no conspecifics were observed in any of the recorded video, and circling behavior was observed in both well-mixed offshore and protected inshore environments, it is unlikely that intraspecific interactions or oceanographic conditions are acting as the primary drivers of the recorded patterns. The most likely explanation for such behavior is that it aids in foraging in this species.

Tortuous movement paths across a range of both aquatic and terrestrial taxa have been used as a proxy to identify foraging and may be associated with prey pursuit, or search for prey either upon encountering a cue or based on past learning experiences (Austin et al., 2006; Wilson et al., 2007; Byrne and Chamberlain, 2012; Papastamatiou et al., 2012; Adachi et al., 2017). Sandbar shark diets are dominated by demersal fauna, such as benthopelagic fishes, cephalopods and crustaceans (McElroy et al., 2006; Ellis and Musick, 2007), and at dusk many nocturnal benthic species emerge from their diurnal refuges (Helfman, 1986). The circling behavior that sandbar sharks were recorded to display on the seabed may therefore represent a strategy to increase the chance of capturing prey at this time. In addition to undertaking circling behavior on the bottom phase of Ushaped dives, relatively higher turning angles were recorded on the ascent phase of U-shaped dives in comparison to V-shaped dives. It is possible that sandbar sharks may also be searching for prey throughout the ascent phase of U-shaped dives while moving between benthic foraging patches. On ascent, sharks may be able to search for prey that are backlit and silhouetted against the surface, and potentially ambush them from below (Martin and Hammerschlag, 2012). No changes in turning angle were observed between dive types during descent, indicating that individuals may be less motivated to search for prey in this phase.

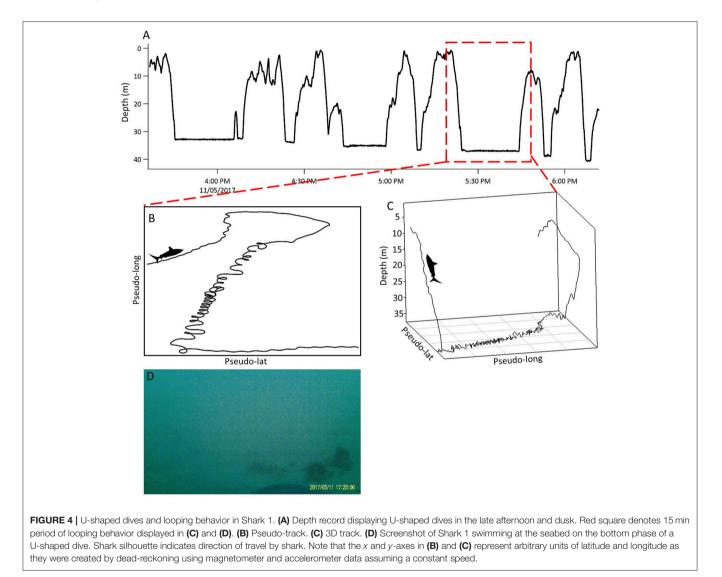
# Vertical Movements

Previous studies reporting vertical movements of sandbar sharks have found these animals to reach maximum depths of over 250 m and display diel patterns in movement (Papastamatiou et al., 2006; Barnes et al., 2016). Barnes et al. (2016) found that sandbar sharks tagged off eastern Australia exhibited a preference for deeper waters throughout daylight hours. Our fine-scale movement data showed that sandbar sharks displayed continual

Shark ID	Model	df	Slope	AIC	∆AIC	R <sup>2</sup>	р
Shark 1	Diving ratio $\sim$ turning Diving ratio $\sim$ 1	101 102	-0.03	-169 -91	-66	0.54	< 0.001
Shark 2	Diving ratio $\sim$ turning Diving ratio $\sim$ 1	51 52	-0.006	-37 -36	-1	0.04	0.15
Shark 3	Diving Ratio $\sim$ turning Diving ratio $\sim$ 1	70 71	-0.05	-97 -44	-53	0.53	< 0.001
Shark 4	Diving Ratio $\sim$ turning Diving ratio $\sim$ 1	68 69	-0.03	-93 -18	-75	0.66	< 0.001

**TABLE 4** | Linear model comparisons using Akaike's Information Criterion (AIC), R<sup>2</sup>, and *p*-values.

AAIC displays deviance in AIC scores from null model. Diving ratio is modeled as a function of turning, where turning is the sum of the turning angles for a sampling window. Model results where turning is standardized by vertical phase of movement are shown in **Supplementary Table 1**.



patterns of oscillatory vertical movements in both inshore and offshore habitats that were likely bathymetrically limited, reaching a maximum depth of 96.6 m. Shallow dive angles occurred throughout the tracks, with relatively symmetrical patterns in ascent and descent. Due to the limited duration of tag attachment, we were not able to explore diel patterns in vertical movement, particularly as patterns during the day were biased to the shallow inshore waters in which the sharks were tagged.

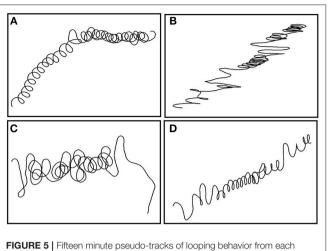
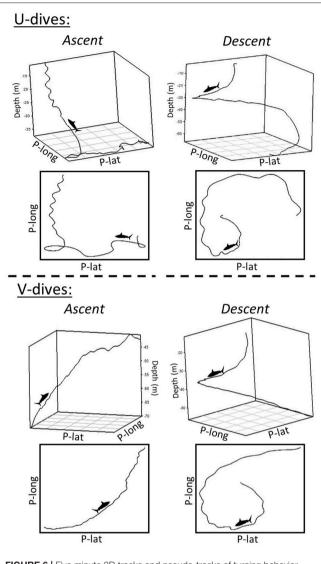


FIGURE 5 | Hitteen minute pseudo-tracks of looping behavior from each sandbar individual around dusk. (A) Shark 1 at 17:30 at depth 36 m; (B) Shark 2 at 18:08 at depth 67 m; (C) Shark 3 at 17:47 at depth 67 m; and (D) Shark 4 at 17:42 at depth 15 m.

U-shaped dives, however, occurred in both inshore and offshore habitats, predominately around dusk. Several other species of elasmobranch have been found to display diel patterns in foraging-related behaviors, with crepuscular periods commonly being associated with enhanced activity rates (Gleiss et al., 2013, 2017a; Papastamatiou et al., 2015, 2018b). Despite a potential preference for foraging at this time, sandbar sharks may still feed opportunistically throughout other diel periods, as occurs in gray reef sharks (*Carcharhinus amblyrhynchos*) (Papastamatiou et al., 2018b). This may account for occasional U-shaped dives observed throughout the night in sandbar shark depth traces (**Figure 3**). However, we again highlight the limited duration of tag attachment in our study and longer tag deployments are required to determine if these patterns occur on a diel cycle.

## **Species Comparisons**

Repetitive oscillatory movements through the water column are a common feature of the tracks of nearly all epipelagic fishes for which fine-scale depth traces have been recorded (Carey and Scharold, 1990; Klimley et al., 2002; Schaefer et al., 2009). Recent studies using biologging tags on tiger sharks (Galeocerdo cuvier) and oceanic whitetip sharks (Carcharhinus longimanus) suggested that these dives likely function as a means of searching for prey (Nakamura et al., 2011; Papastamatiou et al., 2018a). However, if sandbar sharks are predominately searching for prev on the seabed, continual hunting forays through the water column are unlikely to be the primary motivator of these movements. Alternatively, these animals may be indirectly searching for benthic prey by oscillating through vertically stratified layers for olfactory cues in between patches (Carey and Scharold, 1990; Klimley et al., 2002). In addition, data from triaxial sensors suggest that they are doing so in an energetically efficient manner. Shallow dive angles and gliding behavior appear to be common feature of the vertical movements of all large epipelagic fishes for which detailed data records are available



**FIGURE 6 |** Five minute 3D tracks and pseudo-tracks of turning behavior throughout the ascents and descents of U-shaped dives and V-shaped dives. *X* and *Y* axes represent arbitrary units of latitude (P-lat) and longitude (P-long) as they were created by dead-reckoning using magnetometer and accelerometer data assuming a constant speed. The Z axes in the 3D plots represents depth (m).

(Furukawa et al., 2011; Gleiss et al., 2011a,b; Meekan et al., 2015; Papastamatiou et al., 2018a). Given the need to oscillate through the water column to search for cues, shallow dive angles and intermittent gliding behavior have been calculated to minimize the cost of transport, culminating in patterns of cost-efficient foraging (Gleiss et al., 2011b; Papastamatiou et al., 2018a). Sandbar sharks descended at very shallow dive angles (mean 11.4  $\pm$  2.2 and  $-11.5 \pm 3.5^{\circ}$  for ascent and descent, respectively), with uninterrupted glides of up to 210 s in duration. These angles are comparable with mean dive angles for tiger sharks, oceanic whitetip sharks, and whale sharks (*Rhincodon typus*). Mean ascent and descent angles for tiger sharks were recorded at 9.4 and  $-11^{\circ}$ , respectively (Andrzejaczek, unpublished data),

and rarely exceeded absolute dive angles of 11° for oceanic whitetip sharks (Papastamatiou et al., 2018a). For whale sharks mean descent angles of 12–13° were documented, with highly variable ascent angles varying 10–25° depending on dive type (Gleiss et al., 2011b). The shallower dive angles were estimated to minimize the cost of horizontal travel for whale sharks, and may have been near the minimum angle required for gliding to be possible. However, in contrast to sandbar sharks, whale sharks would continuously glide on a majority of descents. A number of factors may be responsible for these inter-specific differences in gliding behavior, including body composition (i.e., ratio lipids to lean tissue) (Gleiss et al., 2017b), morphometrics influencing lift and drag, and the increased drag incurred by smaller sharks carrying relatively large biologging tags.

## **Limitations and Future Considerations**

Despite our limited sample size and tag attachment durations, consistent patterns emerged among tagged individuals. Populations of sandbar sharks at Ningaloo mostly consist of adult females (Braccini and Taylor, 2016; Braccini et al., 2017), therefore tagging in other locations will likely be required to obtain a more representative sample of both sexes and all size classes. Durations of attachment (>24 h) in our study were limited by our need to recover biologging tags to download the data. Offshore swimming behavior of the tagged individuals meant that the distances involved in tag recovery following 24 h of attachment were at the limit of our VHF radio range and offshore boat license in two cases (25-30 km). Longer durations of tag attachment will provide a greater understanding of diel cycles of movement in this species, and an increased capacity to record and store video will aid in revealing more about the drivers of movements, particularly in relation to foraging behaviors. We also acknowledge that the recovery rate of sharks from capture processes can be difficult to define, and that we may not be recording natural behaviors in our current study. Although continuous glides of up to 210s were recorded by sandbar sharks, sparse tailbeats frequently interrupted gliding descents. The possibility that drag due to the tag (Whitney et al., 2019) may be playing a role in inhibiting natural gliding behaviors in this and other studies that use relatively large biologging devices requires future investigation. Lastly, accurate speed measurements and GPS anchor points would considerably enhance dead-reckoning analyses, allowing further insight into fine-scale locational and habitat preferences (Walker et al., 2015).

## Conclusions

Evaluating the results of our study in the context of broad scale ecological knowledge for the species provided new ecological insights into the fine spatial and temporal scale movement behaviors of the sandbar shark. Collectively, patterns recorded

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Adachi, T., Costa, D. P., Robinson, P. W., Peterson, S. H., Yamamichi, M., Naito, Y., et al. (2017). Searching for prey in a three-dimensional environment: hierarchical movements enhance foraging success in by biologging tags on sandbar sharks suggested energetically efficient oscillatory swimming behaviors whilst moving between benthic hunting patches, consistent with biologging studies emphasizing the role of cost-efficient foraging. Larger sample sizes with extended tag recording durations will enable greater ecological insight into these behaviors.

# ETHICS STATEMENT

All methods were performed in accordance with approved guidelines by the University of Western Australia Animal Ethics Committee (RA/3/100/1437), and under permit numbers 2,881 (WA Department of Fisheries) and 08-000322-3 (WA Department of Parks and Wildlife).

# **AUTHOR CONTRIBUTIONS**

SA, MM, and AG conceived the study. SA and AG performed fieldwork/data collection. SA, AG, CP and MM analyzed and interpreted the data. SA led the writing of the manuscript. All authors read and approved the final manuscript.

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## SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fmars. 2018.00483/full#supplementary-material

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**Conflict of Interest Statement:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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