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## Biologging to identify nesting and non-nesting emergences for four species of imperiled sea turtles

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Quantifying sea turtle nesting behavior is essential for recovery planning and evaluating management actions. Traditional monitoring approaches, based on nest counts from beach surveys, can misclassify non-nesting emergences, obscure true fecundity, and underestimate clutch frequency, metrics that directly influence population models and regulatory decisions. Here, we demonstrate that high-resolution acceleration data loggers (ADLs) can reliably discriminate nesting from non-nesting emergences across four imperiled species of sea turtles at sites in the Gulf of America, southeast USA, and Caribbean. From 60 recovered ADL deployments on green (Chelonia mydas; N = 10), hawksbill (Eretmochelys imbricata; N = 7), Kemp's ridley (Lepidochelys kempii; N = 21), and loggerhead sea turtles (Caretta caretta; N = 22) lasting on average 17.5  $\pm$  8.7 days (range 2-43 days), we identified 54 nesting events and 76 non-nesting emergences, with >97% accuracy when compared to direct observations. These data provide the first observer-validated, species-specific behavioral signatures of nesting phases and reveal correlations between egglaying duration and clutch size. All non-nesting emergences occurred within 72 hours of subsequent nesting, allowing managers to anticipate nest deposition windows. By refining inter-nesting intervals and fecundity estimates, ADLs offer a practical path to reduce error in clutch frequency estimates. The integration of ADL-derived algorithms with satellite-transmitting tags would enable the remote, real-time monitoring of nesting activity, creating a system for the remote monitoring of inter-nesting intervals and nest fecundity that are crucial to quantify the impacts of climate change and other threats to sea turtle nesting habitat.

KEYWORDS

behavior, accelerometer, remote sensing, marine turtle, conservation

### 1 Introduction

Imperiled sea turtles have complex life cycles that include the use of disparate habitats in both marine and terrestrial environments. Their conservation requires diverse strategies that protect vulnerable life stages, which are outlined in species-specific U.S. federal recovery plans (NMFS and USFWS, 2008; NMFS and USFWS, 2011; NMFS and USFWS, 2015). The most-well-studied component of sea turtle populations is nesting females, as they are accessible to researchers on land, where they return on average every 2-5 years to nest. Although many populations have stabilized and certain populations have started to rebound (Balazs and Chaloupka, 2004; Troëng and Rankin, 2005; Chaloupka et al., 2008; Weber et al., 2014), others, such as Northwest Atlantic loggerhead turtles (Caretta caretta), have not shown trends in recovery (Ceriani et al., 2019; Blumenthal et al., 2021), and Kemp's ridley turtles (Lepidochelys kempii) remain critically endangered (NMFS et al., 2011; NMFS and USFWS, 2015; Wibbels and Bevan, 2019).

Trends in sea turtle nesting numbers are used by managers to assess population status (refer to Esteban et al., 2017; Witherington et al., 2009; Brost et al., 2015). Together, nesting frequency and egg counts allow for the estimation of the overall fecundity per individual, a key vital rate in sea turtle life tables (Crouse et al., 1987; Richards et al., 2011) that affects trajectories of population projections. In Florida, USA, counts of individual females as well as measures of their reproductive output (i.e., number of eggs) are collected on a suite of "index" nesting beaches, and interpreted annually to assess trends across years. However, accurate measures of remigration interval and clutch frequency are lacking for many sea turtle populations (Richards et al., 2011; Lamont et al., 2014; Ceriani et al., 2019), as tagging of individual nesting females requires more resources than the widely used morning counts of nests. Still, these surveys for tracks left by nesting females are timeconsuming, costly, and often include demarcation of erroneous "nests" unless the presence of eggs is verified (Casale and Ceriani, 2019). Relying on the presence of tracks to determine the occurrence of nesting events can also be problematic due to the fact that many environmental conditions (e.g., high tides, vegetation, and high winds) can obscure tracks, sometimes completely masking all evidence that a nest was laid.

Tools such as satellite and GPS tags have been increasingly used in recent years by researchers worldwide to document the frequency of nesting in individual turtles (Hart et al., 2010; Tucker, 2010; Esteban et al., 2017), a key metric in sea turtle population estimation (Weber et al., 2013; Esteban et al., 2017). When turtles emerge on the beach, the satellite tag typically sends high-quality Argos locations (e.g., location classes 3, 2, and 1) and/or GPS fixes. However, location data alone cannot reveal how long a turtle is on the beach or behavioral information such as whether the turtle successfully nested or instead exhibited a non-nesting emergence (NNE; also known as an "emerge no lay" or "false crawl"). In a recent analysis of high-resolution GPS data to assess clutch frequency for tagged green turtles (Chelonia mydas), Esteban et al. (2017) assessed GPS locations and haul-out periods to

deduce the numbers of actual female turtles laying nests in an Indian Ocean study site. Their findings showed that the true number of nests laid per female was poorly estimated with direct female observations alone, and they suggested that robust estimates could be derived using biologging tools. However, because some inferences derived from the interpretation of satellite tag data may not accurately represent whether a turtle landfall event was a nesting activity or an NNE, behavioral classifications of female landfall events are still needed.

Nesting behavior in sea turtles can involve as many as 50 distinct action patterns (Lindborg et al., 2019), but the main phases of these behaviors consistently include approaching and ascending the beach, digging a nest, egg laying, nest covering, and descending the beach and returning to the ocean (Hailman and Elowson, 1992; Miller and Dinkelacker, 2008). Egg laying and hole filling do not occur during NNE events, and so fine-scale records of body movements are sufficient to differentiate nesting from NNE events. Such records have typically been documented directly by human observers, but animal-borne biologging tools can now provide such records in the absence of direct observation.

In recent years, biologging tags that include acceleration data loggers (ADLs) have been increasingly used to study the fine-scale behavior and energetics of many marine animals (Cooke et al., 2004) and have been repeatedly used to quantify animal body pitch as well as the force and frequency of body movements associated with locomotion during diving for marine turtles (Yasuda and Arai, 2009; Fossette et al., 2010, 2012; Hart et al., 2016; Okuyama et al., 2009, 2012, 2013). Anti-predator behavior (Hounslow et al., 2020), as well as nesting behaviors in sea turtles, has also been studied using ADLs. Acceleration and depth data from green sea turtles allowed for the identification of five components of nesting behavior with 86% accuracy (Nishizawa et al., 2013). More recently, Jeantet et al. (2022) showed that automatic classification of acceleration data could detect six behaviors associated with nesting events in green sea turtles with 95% accuracy, and the number of eggs laid could be calculated with 7% error. Thus, ADLs and other biosensors have the potential to provide unbiased estimates of nesting activity, NNE frequency, and fecundity of turtles in the wild without on-site human monitoring effort.

We applied animal-borne ADLs to four sea turtle species in order to 1) observe and describe the fine-scale nesting behaviors of these species using acceleration-derived metrics, 2) differentiate nesting events and NNEs based on logger data, 3) compare rates of ADL-derived NNEs to field observations, 4) calculate precise inter-nesting intervals for each individual using this tool, and 5) when possible, estimate fecundity. This effort represents a move toward more remote collection of key data using state-of-the-art technologies.

### 2 Methods

### 2.1 Sea turtle capture and tagging

We tagged adult female sea turtles nesting along beaches in Texas, Alabama, North Carolina, Florida, and the U.S. Virgin

Islands during 2012-2017. Sites included Padre Island National Seashore, Texas; Bon Secour National Wildlife Refuge, Alabama; Bald Head Island, North Carolina; Dry Tortugas National Park, Florida; T.H. Stone Memorial St. Joseph Peninsula State Park, Florida; and Buck Island Reef National Monument, U.S. Virgin Islands (Figure 1). Turtles included four species: loggerheads, green turtles, Kemp's ridleys, and hawksbills (Eretmochelys imbricata). After an individual completed any nesting or NNE activity and was returning to the water, we captured and restrained each one in a beach corral or transported the turtle (i.e., Kemp's ridleys) to a laboratory nearby. We then scanned individuals for passive integrated transponder (PIT) tags and looked for flipper tags to determine if they were previously marked and tagged them if we detected no tags according to accepted protocols (NMFS SEFSC, 2008). We took standard morphometric measurements [curved carapace length (CCL), curved carapace width (CCW), straight carapace length (SCL), and straight carapace width (SCW)], and blood and tissue samples for other analyses. In some cases, nests were relocated according to established protocols (refer to Shaver and Caillouet, 2015; Shaver et al., 2016), and egg counts were obtained from the relocation data and used for fecundity analyses.

# 2.2 Tag programming, attachment, and data processing

We programmed ADLs (Wildlife Computers, Redmond, WA, USA; TDRX or DD-X) to record acceleration at either 16 or 32 Hz and depth (meters) at 1 Hz, while we set temperature (Celsius), wet–dry status (arbitrary tag units), and light levels (arbitrary tag

units) to record every 7 seconds. We additionally measured swim speed and 3D magnetic field strength for turtles that received a daily diary tag (DD-X). ADLs recorded acceleration data in m/s2, with all acceleration-derived metrics being recorded using these same units. We affixed ADLs to the anterior carapace, usually on the third or fourth vertebral scute (Figure 2), posterior to the satellite tag, following standard protocols for attachment (NMFS SEFSC, 2008; Hart et al., 2021). We outfitted turtles with either SPOT and SPLASH model satellite tags from Wildlife Computers or Kiwisat KS-101 model tags from Lotek (Havelock North, New Zealand). We attached ADLs and satellite tags using Superbond cool-setting epoxy (refer to Hart et al., 2021); we released turtles after the epoxy dried (approximately 1 hour later) and allowed them to return to the ocean. Observers recorded detailed, time-stamped notes (all watches, computers, and ADLs synced to time.gov prior to deployment) of turtle behavior after ADLs were affixed and subsequently whenever an ADL-tagged animal was sighted. We retrieved ADLs from turtles by recapturing individuals on subsequent nesting events. Returning turtles carrying tags were typically allowed to nest before we recaptured them for identification and tag removal.

We downloaded ADL data and processed them in Igor Pro (Wavemetrics) and R (R Development Core Team 2019). We calculated static acceleration (indicative of body position) using a 4-second running mean of the raw acceleration, while we calculated dynamic acceleration (indicative of body movement) by subtracting the static acceleration from the raw acceleration (e.g., Shepard et al., 2008a, b). Flipper beats were assessed as the dynamic acceleration of the X axis, which corresponds to a pitching and surging motion. We summed the absolute value of the dynamic acceleration along all

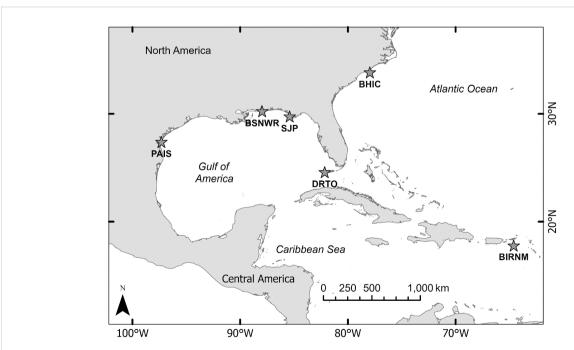


FIGURE 1
Study site map where accelerometer data loggers were deployed on nesting females. Abbreviations are as follows: PAIS, Padre Island National Seashore, TX; BSNWR, Bon Secour National Wildlife Refuge, AL; SJP, T.H. Stone Memorial St. Joseph Peninsula State Park, FL; DRTO, Dry Tortugas National Park, FL; BHIC, Bald Head Island, NC; BIRNM, Buck Island Reef National Monument, St. Croix, U.S. Virgin Islands.

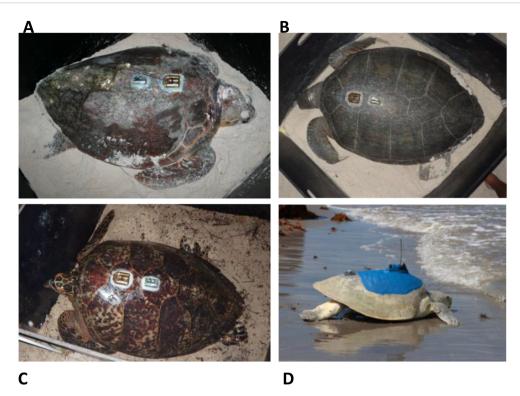


FIGURE 2
Satellite tag (anterior, with antenna) and accelerometer (ADL), which is posterior to satellite tag, shown epoxied to the carapace of a loggerhead (A), green (B), hawksbill (C), and Kemp's ridley (D) sea turtles for this study. Photo credits: U.S. Geological Survey (A–C), and National Park Service (D). ADL. acceleration data logger.

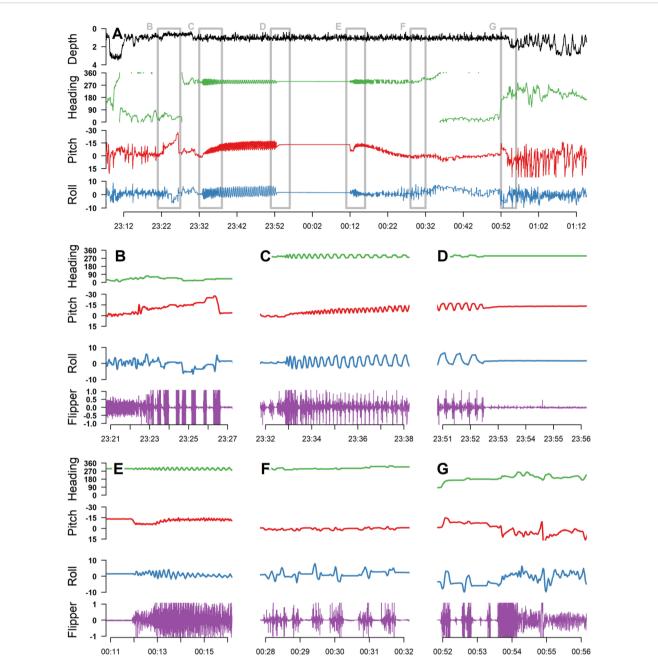
three axes to derive overall dynamic body acceleration (ODBA; Wilson et al., 2006). We then aggregated ODBA over 1 second to match the temporal frequency of depth data. Additionally, we interpolated temperature and wet–dry status to be at the same frequency as the depth data. For daily diary ADLs, we corrected magnetometer data for the hard iron offset, and we calculated the two-dimensional heading of the animal after correcting for body pitch and roll, determined using an accelerometer.

## 2.3 Validating ADL data signatures through direct observation

We used observer notes of general turtle behaviors to describe and ground-truth the behaviors logged in the ADL data. To distinguish patterns in behavior that would differentiate nesting events  $\nu s$ . NNE events, we visually examined each turtle dataset, noted the start and end times of observed events, and compared these data-based classifications to notes from direct observations on the beach. Tag data from observer-confirmed events were then used to generate threshold criteria to automatically identify nesting and NNE events from the remaining tag data. Manual event classification from ADL data included validating nesting and NNE signatures with direct observation, after which we examined each individual's entire ADL dataset for signs of these events. Visual scanning of the ADL data showed distinct patterns when animals were on land compared to in the water. We characterized time on

land by a shallow depth near 0 m, the wet-dry sensor reading dry, and often a change in temperature as the animal emerged from the water. Although similar sensor values could occur when an animal was in the water floating at the surface, floating animals showed constant motion due to water movement, whereas turtles on land always displayed at least a short period where acceleration data indicated a cessation of movement.

Nesting behavior was comprised of similar phases across all four species (refer to Figure 3), and each phase produced characteristic data signatures across multiple channels (Table 1). Although the behavioral phases of nesting have been characterized in previous studies (e.g., Bustard and Greenham 1969; Hailman and Elowson, 1992; Lindborg et al., 2019; Jeantet et al., 2022), we sought to describe phases that were consistent across species, were inclusive of a turtle's entire time on land, and could be reliably identified based on ADL data alone. The typical turtle displayed relatively constant flipper beats as it actively swam toward the nesting beach. Once on land, the turtle displayed high-amplitude acceleration spikes indicative of crawling with intermittent periods of rest (Figure 3b). During this initial crawling, an increase in body pitch was typically observed as the animal crawled up the beach and over the berm. After crawling, the turtle began to dig, characterized by repetitive oscillations in body pitch and alternating side-to-side rolling as the animal excavated the nest chamber with one rear flipper before pitching forward and rolling to the other side to excavate with the other rear flipper (Figure 3c). During this process, the turtle's mean body pitch would gradually increase (i.e.,



A complete nesting event from loggerhead sea turtle Sally (A), which was outfitted with a daily diary tag. Panels B–G are zoomed-in sections of panel A, corresponding to the location of the gray boxes, labelled in panel (A) The event began with the turtle swimming in a northerly direction toward the beach, displaying consistent flipper beats until 03:23 when it landed on the beach (B). Once on land, the turtle displayed high-amplitude acceleration indicative of crawling, followed by periods of rest; the steadily increasing and then quickly decreasing body pitch indicates the animal crawling up the beach and over a berm (B). At 03:33, the turtle began to dig, indicated by pitching up and rolling to its side to excavate the nest chamber and then pitching forward and rolling to its other side to excavate with its other rear flipper (C). During this process, the turtle's body pitch gradually increased, indicating that the hole was getting deeper and its body was oriented increasingly in a head-up posture. At 03:52, the turtle stopped excavating the nest chamber and remained still, indicating that it started to lay eggs (D). At 04:14, the turtle began to move again, signifying the end of egg deposition, and then began to fill in the nest with sand using high-intensity flipper beats and less rolling and pitching than observed during excavation (E). As the turtle filled in the nest, its body pitch returned to level, and the turtle began to show more intermittent movement, using its flippers to swipe and cover the nest with sand (F). Finally, the turtle showed intermittent movement as it crawled back to the water and began swimming away at 04:54 (G).

increasingly head up) as the hole deepened. In addition to rolling, magnetometer data showed that turtles would shift their heading right to left depending on which rear flipper they were using to excavate. Eventually, the turtle would stop excavating the hole and

remain still, indicating that it had started to lay eggs (Figure 3d). After a lengthy, stationary, egg-laying period, the turtle began filling the hole (Figure 3e), producing a locomotory pattern that typically had more constrained oscillations in pitch and roll movements

TABLE 1 Typical data signatures associated with nesting behaviors in four species of sea turtle.

Behavior	Duration	Depth (pressure)	Body position (accel)	Flipper beats* (accel)	Heading (magnetometer)	Temperature
Swimming approach to beach	minutes	Depth at bottom of dives steadily decreases as turtle approaches shore	Pitch positive during ascents; negative during descents	Constant flipper beats indicative of swimming	Constant, oriented toward shore	Constant
Crawling (landing on beach, crawling through sand over berm)	5–50 min	Depth = 0 m	Pitch positive as turtle crawls up the beach	Intermittent, high- amplitude FB indicative of crawling; followed by periods of rest	Constant	Initial temp. drop (from air temperatures and evaporation) as turtle moves from water to air
Digging (digging and scooping with rear flippers, alternating between right and left)	10–50 min	Depth = 0 m	Pitch and roll oscillate as turtle alternates rear flippers in digging motion; pitch gradually increases as posterior of body descends into hole	High-amplitude FB with relatively constant periodicity indicative of digging	Heading shifts back and forth depending on which rear flipper (right or left) is scooping sand	Constant for most; slow increase for Kemp's (daytime nesters)
Egg laying (not present in NNE events)	~6–37 min	Depth = 0 m	Constant	None	Constant	Constant
Filling hole (not present in NNE events)	NA; combined with crawling down beach	Depth = 0 m	Pitch gradually decreases back to horizontal as posterior of body rises	High-amplitude FB with relatively constant periodicity indicative of digging	Heading changes with each FB as head goes from side to side while digging	Constant
Crawling down beach and returning to water	<5 min	Depth = 0 m until water is reached; then again shows variable values associated with diving	Pitch negative as turtle crawls down the beach	Intermittent, high- amplitude FB indicative of crawling	Constant, oriented offshore	Abrupt temp. change in some cases, depending on environment

<sup>\*</sup>Dynamic acceleration or quick movements in the dorso-ventral acceleration axis that are caused by flipper movement but, on land, are very different than typical "flipper beats" during swimming. Accel, accelerometer-derived; NNE, non-nesting emergence; FB, flipper beat.

compared to the excavation phase. As the turtle filled the hole with its back flippers, the mean body pitch would gradually return to approximately horizontal. After the hole was filled, the turtle would display intermittent locomotion (Figure 3f), during which time observers noted the turtle using its front flippers to swipe and cover the nest with sand. The turtle would then turn back to the ocean and intermittently crawl back into the sea, where it would resume normal swimming behavior (Figure 3g; Table 1).

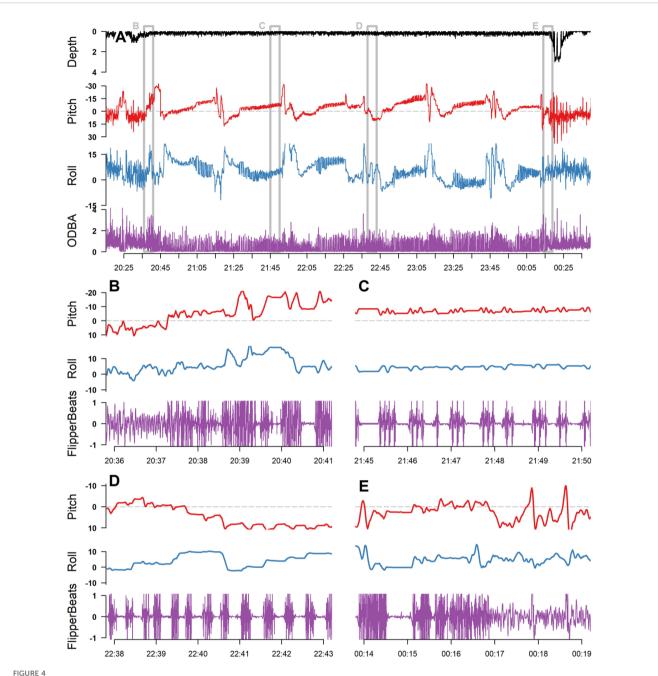
NNE events showed the same emergence behaviors and data signatures as nesting events, but typically involved turtles returning to the sea with no signs of digging or other nesting behaviors such as covering the nest (Table 1; Figure 4). Some NNE events showed movements indicative of excavation, but the behavior would eventually be abandoned with no sign of a stationary egg-laying period. Some nesting and NNE events were utilized as an opportunity for ADL removal and recovery, and thus, data signatures for complete events, including the "return to ocean" phase, were often not obtained when observers were present. However, six complete nesting events and one NNE were observed while animals carried tags (Table 2), and behaviors observed after tags were removed (crawling in the sand) were

frequently validated through observations earlier in the NNE or nesting events, as well as from the original tagging events.

## 2.4 Automatic classification and comparison with satellite telemetry

We created a custom algorithm to automatically classify nesting events and NNEs *a posteriori* from ADL sensor data using criteria from our direct observations. Specifically, we classified time on land as any period with a wet–dry sensor reading >80 (proprietary tag manufacturer units) that lasted at least 120 seconds in duration. To further distinguish this from long surface intervals at sea, we also required the dry period to include at least 8 consecutive seconds with an ODBA <  $0.2 \text{ m/s}^2$ , and it must have exhibited a mean depth <4 m over the previous 10 minutes. Nesting events were separated from NNE events as containing a stationary period (ODBA was never above  $0.2 \text{ m/s}^2$  for more than 3 consecutive seconds) that lasted for at least 300 consecutive seconds. This stationary period was indicative of egg laying.

In order to compare our ADL-based technique of identifying nesting and NNE events with a more traditional technique based on



A non-nesting emergence (NNE) from Ali, a female green turtle in U.S. Virgin Islands (A). Panels (B–E) are zoomed-in sections of panel A, corresponding to the location of the gray boxes, labelled in panel (A) The event began with the turtle reaching the beach at 00:37, with high-amplitude flipper beats and increasing body pitch angle indicating that it was crawling up the beach (B). It showed several bouts of digging, indicated by high periods of intense flipper beats punctuated with resting periods (e.g., (C, D)), with large but gradual changes in body pitch indicative of the animal's rear flippers lowering into an excavated hole (A). No long periods of rest that would indicate egg laying were observed during any of the excavation bouts, and the event ended with the animal returning to the water at 04:17, nearly 4 hours after it emerged. ODBA, overall dynamic body acceleration; depth is in m.

satellite tag data, we analyzed spatial data from satellite tags and recorded high-quality, land-based locations (as in Tucker, 2010; Hart et al., 2010; see Hart et al., 2025). These locations and times were deemed landfall events for the purposes of further analyses and characterized as either nesting events or NNEs using methods described by Hart et al. (2010). We excluded Kemp's ridleys from spatial analyses.

R was used to conduct all statistical analyses. The duration of emergences (nesting and NNE), specific behaviors (crawling, digging, and egg laying), and inter-nesting intervals were compared among species using a series of analysis of variance (ANOVA) tests. Each turtle's number of false crawls was standardized by their deployment duration. This false crawl rate was compared among species using ANOVA. For all among-species

TABLE 2 The number of nesting, egg-laying, and non-nesting emergence (NNE) events for each species based on visual examination of data logged in accelerometers (ADLs).

Species	Total	Complete	Duration (min)	Min	Max	
Nesting overall						
Green turtle	7 (4)	4 (1)	171.5 ± 35.2	139.2	219.6	
Hawksbill	5 (3)	2 (0)	138	129.8	146.1	
Kemp's ridley	17 (16)	1 (0)	43.4	43.4	43.4	
Loggerhead	25 (13)	17 (5)	100.5 ± 38.5	56.2	234.5	
Crawling up beach <sup>1</sup>						
Green turtle	7 (4)	7 (4)	44.9 ± 24.2	27.3	97	
Hawksbill	5 (3)	5 (0)	31.2 ± 16.4	14.5	51.8	
Kemp's ridley	17 (8)	17 (8)	10.8 ± 3.3	5.5	15.8	
Loggerhead	25 (8)	25 (8)	17.6 ± 10	8.2	46.9	
Digging hole <sup>2</sup>						
Green turtle	7 (4)	7 (4)	29 ± 13	13.8	50.1	
Hawksbill	5 (3)	5 (0)	26.1 ± 8	16.8	35.8	
Kemp's ridley	17 (8)	17 (8)	17.4 ± 3.9	10.8	26.5	
Loggerhead	25 (8)	25 (8)	15.9 ± 3.9	8.1	25.3	
Laying eggs						
Green turtle	7 (4)	7 (4)	18.9 ± 6.8	6.7	25.4	
Hawksbill	5 (3)	2 (0)	24.5	22	27	
Kemp's ridley	17 (8)	17 (8)	11.9 ± 2.4	7.6	18.6	
Loggerhead	25 (8)	25 (8)	16.5 ± 5.9	8.9	36.7	
Filling and crawling o	down beach <sup>3</sup>					
Green turtle	4 (1)	4 (1)	56.8 ± 17.6	40.6	78.8	
Hawksbill	2 (0)	2 (0)	33.2	31.5	34.9	
Kemp's ridley	1 (0)	1 (0)	6.6			
Loggerhead	17 (5)	17 (5)	46.8 ± 27.2	18.3	141.5	
NNE overall						
Green turtle	13 (3)	10 (0)	28.1 ± 67.6	2	220.4	
Hawksbill	20 (2)	18 (0)	5.3 ± 3.9	2.3	17.4	
Kemp's ridley	5 (0)	5 (0)	11 ± 7.7	2.2	23.1	
Loggerhead	38 (4)	35 (1)	10.4 ± 8.4	2.3	37.3	

Complete denotes that the animal carried out the full behavior, including return to the ocean, while carrying an ADL. Numbers in parentheses represent the number of events that were validated by a human observer. Mean, min, and max duration were only calculated for complete events.

ADL, acceleration data logger.

comparisons, if ANOVAs found significant variation, Tukey's honestly significant difference (HSD) test was used for pairwise comparisons to determine which species differed from each other. The correlation between egg-laying duration and clutch size for Kemp's ridleys in Texas and loggerheads in Alabama, where nest relocation efforts are permitted (see above), was assessed.

### **3 Results**

A total of 96 adult female sea turtles (loggerhead, 41; Kemp's ridley, 32; green turtle, 14; hawksbill, 9) were tagged, and ADLs were recovered from 60 individuals (loggerhead, 22; Kemp's ridley, 21; green turtle, 10; and hawksbill, 7; Table 3) for a 62.5% recovery

Period of crawling until first instance of digging.

<sup>&</sup>lt;sup>2</sup> Only refers to the final bout of digging if there were multiple bouts detected.

<sup>&</sup>lt;sup>3</sup> Distinguishing hole-filling movements from crawling down beach was difficult, so these two behaviors were combined.

rate. ADLs remained on turtles 10–722 days, with 90% of tags removed within 40 days. ADL batteries expired before tag recovery for 21 individuals, producing a mean recording duration of 17.5 ± 8.7 days (range 2–43 days). Two individuals were recaptured on nesting beaches 2 years after being tagged, and ADLs were recovered at that time. ADLs deployed on green turtles had the shortest mean recording duration (10.6 days), whereas tags deployed on hawksbills, Kemp's ridleys, and loggerheads had longer mean recording durations between 17.6 and 20 days (Table 3), reflecting variability in species-specific internesting intervals.

Several nesting and NNE events detected by ADLs were confirmed by direct observation or from satellite tag data that confirmed time on land (Hart et al., 2025). We directly observed a total of 45 emergence events (N = 36 nesting; N = 9 NNE) of ADL-equipped turtles, and at least two instances of each behavior (nesting or NNE) were observed for each species except for Kemp's ridleys, for which no NNE events were witnessed (Table 2; Hart et al., 2025). Satellite telemetry for loggerheads, green turtles, and hawksbills was used to confirm an additional 18 landfall events (N = 15 nesting; N = 4 NNE; Hart et al., 2025) that were detected by ADLs but not witnessed by observers.

## 3.1 Frequency of nesting and NNE behaviors across species

A total of 54 nesting events and 76 NNE events were identified from the ADL data of 60 individual turtles (Table 2), with up to four nesting events recorded in a single individual (Figure 5; Hart et al., 2025). Our manual identification of events (nesting or NNE) from the ADL data was confirmed for all 45 emergence events (N = 36 nesting; N = 9 NNE) that were also directly observed, with no mischaracterized events.

A total of 24 "complete" nesting events, in which turtles returned to the ocean after nesting while still carrying their tags, were recorded by ADL data, and six of these events were documented by observers. Complete nesting events were predominantly from loggerheads (N=17), followed by green turtles (N=4), hawksbills (N=2), and Kemp's ridley (N=1; Table 2). Of the 76 NNE events, 68 were "complete" (where tagged

TABLE 3 The number of individuals tagged, the number of recovered accelerometers (ADL), and the average recording duration for each species.

Species	# Tagged	# Recovered	ADL record (days)
Green turtle	14	10	10.5 ± 3.9
Hawksbill	9	7	19.9 ± 8.1
Kemp's ridley	32	21	19.6 ± 5.8
Loggerhead	41	22	17.5 ± 11.3
Total	96	60	

ADL, acceleration data logger.

turtles returned to the ocean), but only one of these complete events was witnessed by observers (Table 2).

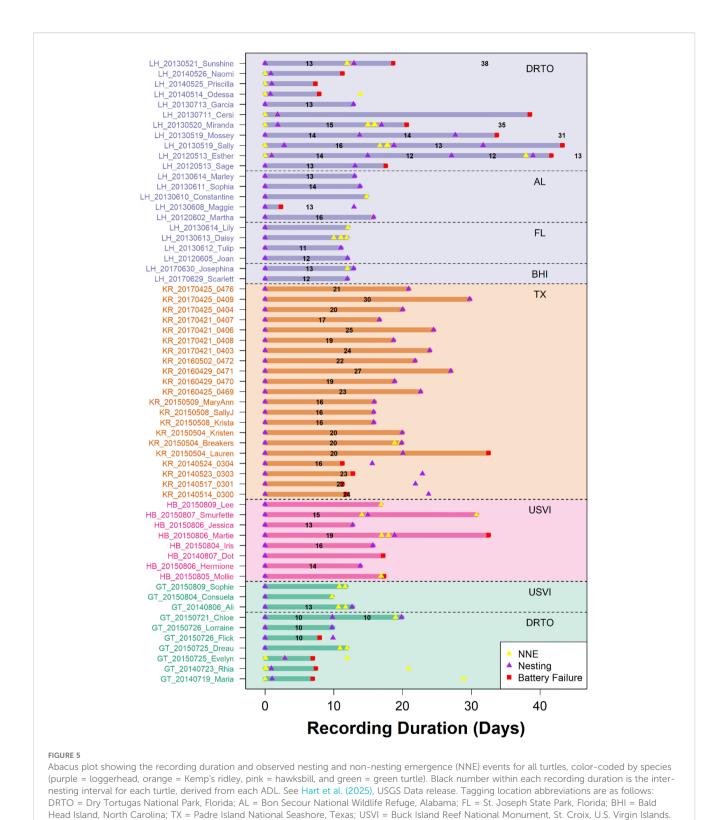
ADL-derived nesting data allowed for the calculation of 44 inter-nesting intervals, with significant differences in inter-nesting period among species (ANOVA F3,40 = 27, p < 0.001). Green turtles had the shortest inter-nesting period (N = 4, 10.75  $\pm$  1.5 days), followed by loggerheads (N = 18, 13.3  $\pm$  1.32 days), hawksbills (N = 5, 15.2  $\pm$  2.3 days), and Kemp's ridleys (N = 17, 20.7  $\pm$  3.9 days; Table 4). In addition to the longest inter-nesting period, Kemp's ridleys displayed the largest variation in internesting period among individuals, from 16 days to a maximum of 30 days (Table 4). For loggerheads, variation was observed within individuals as well, with one turtle displaying inter-nesting periods of both 16 and 13 days, and another individual showing intervals of 14, 12, and 12 days between nesting events (Figure 5).

# 3.2 Nesting event duration and correlation to fecundity

Loggerheads and hawksbills had mean nesting durations greater than 100 minutes, whereas green turtles had longer average nesting durations of 171  $\pm$  35 min (Table 2). While we were only able to identify one complete nesting event for a Kemp's ridley, this was the shortest nesting event observed out of all the species, at just 43 minutes. Species-specific differences in nesting durations were significant (ANOVA: F3,20 = 5.41, p = 0.006), with Kemp's ridleys (post-hoc Tukey's, p = 0.027) and loggerheads (Tukey's HSD, p = 0.012) having shorter nesting durations than green turtles. Loggerheads additionally had the largest variation in their nesting duration, with their shortest nesting event lasting 56 minutes and the longest lasting 234 minutes, the longest of any turtle in this study.

Species differences were also apparent across different phases of the nesting process (Figure 6), with significant differences among species in the duration of crawling (F3,50 = 15.2, p < 0.001), digging (F3,50 = 11, p < 0.001), and egg laying (F3,47 = 6.4, p < 0.001). Kemp's ridley and loggerhead turtles had the shortest duration of crawling up the beach before initiating digging, and both had significantly shorter durations than green sea turtles (Tukey's HSD, loggerhead-green sea turtle: p < 0.001; Kemp's-green sea turtle: p < 0.001). Similarly, both Kemp's ridley and loggerhead turtles had the shortest digging durations and were significantly faster than green turtles (Tukey's HSD, loggerhead-green sea turtle: p < 0.001; Kemp's-green sea turtle: p < 0.001) and hawksbill turtles (Tukey's HSD, loggerhead-hawksbill sea turtle: p < 0.01; Kemp'shawksbill sea turtle: p = 0.03). Kemp's ridleys had shorter egg-laying durations than other species (Tukey's HSD, loggerhead: p = 0.02; green turtle: p = 0.01; hawksbills: p = 0.008). We did not find any significant differences across species in the duration of post-egglaying behavior.

We analyzed the correlation of nesting phase duration with fecundity for 20 nests (16 Kemp's ridley and four loggerheads) for which both ADL and clutch size data were available. Total clutch size was positively correlated with the egg deposition period ( $F_{1,18}$  =



7.9, p = 0.01), which accounted for 30% of the variation observed in clutch size. Clutch size was not significantly correlated with duration of any other phase of nesting, nor with overall nesting event duration.

ADL, acceleration data logger,

### 3.3 NNE duration and timing

Non-nesting emergences were relatively short in duration, with 75% of complete NNEs (N = 68) lasting less than 10 minutes and

TABLE 4 Inter-nesting intervals (days) calculated from accelerometer (ADL) data.

Species	Mean <u>+</u> SD	Min	Max
Green turtle	10.5 ± 1.4	9.7	12.7
Hawksbill	15.2 ± 2.3	12.7	18.8
Kemp's ridley	20.7 ± 3.9	15.8	29.8
Loggerhead	13.2 ± 1.3	11	15.9

ADL, acceleration data logger.

averaging just 11  $\pm$  25 minutes. There was no difference across species in the duration of NNE events (ANOVA: F3,64 = 1.69, p = 0.17). One notable NNE performed by a green sea turtle 48 hours prior to a nesting event lasted for over 220 minutes and appeared to involve digging and then abandoning five different egg chambers. This was not directly observed by researchers, but the authors have witnessed similar long-duration NNE events in this species in the past (KMH, personal observ.), as well as in loggerheads in NW Florida (MML, personal observ.).

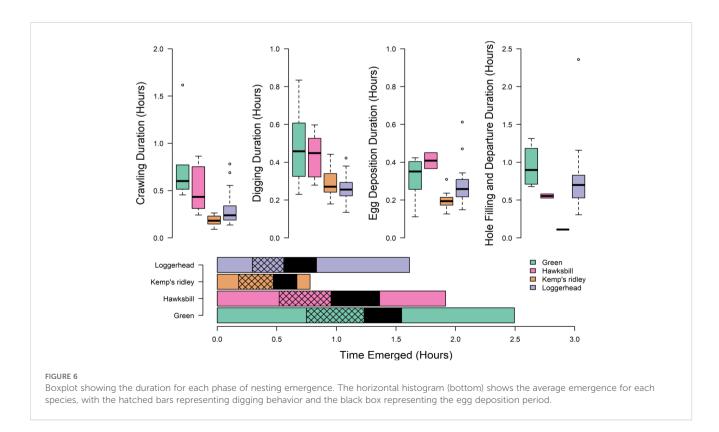
At least five NNEs were detected in each species, and all NNEs occurred within 72 hours prior to a nesting event (Figure 7). NNEs occurred an average of  $18.2 \pm 15.9$  hours before a nesting event, with 36% occurring the same night as a nesting event and 50% occurring within 24 hours before a nesting event. Species had different NNE rates (ANOVA: F3,57 = 5.2, p = 0.003), with Kemp's ridleys conducting fewer NNEs than green turtles or hawksbills (Tukey's HSD, p = 0.003, p = 0.05), but not loggerheads (p = 0.11).

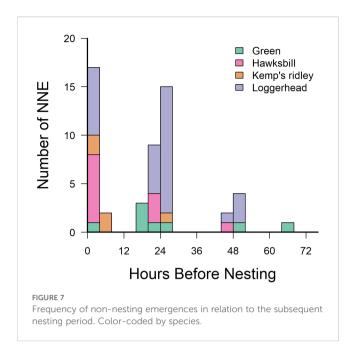
Nesting events and NNEs occurred primarily during nighttime hours (1900-0500 hours) for loggerheads, green turtles, and

hawksbills, whereas all Kemp's ridley events occurred during daylight hours (Figure 8). Nesting and NNE events showed temporal overlap for all species, but nesting events were more clustered in time, whereas NNE events occurred over a larger range of hours. Hawksbills predominantly nested between the hours of 2100 and 0000, but the majority of NNEs occurred later in the morning between 0100 and 0500. For Kemp's ridleys, the majority of false crawls occurred early in the morning (0600–0800 h), before nesting events peaked (1000–1400 h). For green turtles, the latest nesting event occurred at 0630, but 31% (4 of 13) of NNEs occurred later than this, with three occurring during daylight hours of the early morning. There was little difference in the timing of nesting and NNE events for loggerheads (Figure 8).

# 3.4 Automatic event classification and comparison to satellite telemetry

Automatic classification using parameters based on acceleration, depth, and wet-dry sensor data (refer to Methods) identified 118 times that a turtle was on land. Of these, 45 were classified as nesting events, and 73 were classified as NNE. Two individuals had faulty accelerometers that introduced noise in the sensor every 32nd point, which falsely increased their ODBA and thus did not meet classification criteria for being on land. The tags of two other turtles had faulty wet-dry sensors, obscuring the analysis. When these four individuals were removed from the analysis, automatic classification correctly identified 45 out of 46 nesting events (97.8% accuracy rate), and no NNEs were identified as nesting events, producing a 0% false positivity rate. The one





incorrect assignment was from a nesting Kemp's ridley that repositioned itself during egg laying and was thus classified (incorrectly) as an NNE event instead of a nesting event.

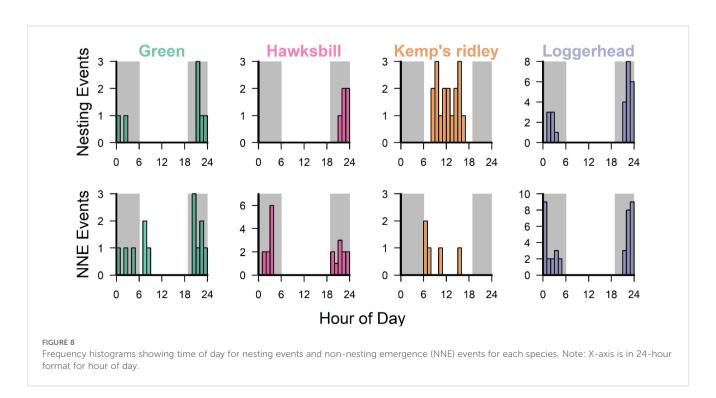
The ability to identify nesting and NNE events from satellite tag data was compared to our manual ADL method using high-quality, land-based satellite locations that were available for 12 turtles across 17 nesting events; no Kemp's ridleys were included in this analysis. Satellite-derived classification of nesting events produced identical results to the manual ADL method for loggerheads (100%, N=12) and green turtles (100%, N=3). However, the satellite-based method incorrectly identified unobserved nesting events (0%, N=12)

2) for hawksbills. In one case, the satellite estimate suggested that the turtle nested 3 days before the ADL-derived nesting event, while in the other instance, there were no satellite-derived estimates that suggested that the turtle nested.

Analysis of satellite telemetry locations showed that locations alone severely under-represented NNE events, both by missing entire nights in which NNEs occurred and by lacking the resolution to detect multiple NNE events for a turtle in a single night. Our ADL data revealed that individuals could have multiple NNE events (up to four or five) in a single night, which were combined into a single NNE event by satellite analysis, if they were detected at all. Overall, only 6.3% of NNE events (four out of 63) were documented by satellite analysis, and only 35% (13 of 37) of nights in which an NNE event occurred were correctly classified by satellite analysis (N = 4) and direct observation (N = 9) combined.

### 4 Discussion

We used ADLs to discriminate the fine-scale phases of the nesting and non-nesting beach emergences for four imperiled species of sea turtles across multiple locations. This approach allowed the precise calculation of inter-nesting periods, a parameter that directly influences fecundity estimates used in demographic models. The distinct nesting stages we observed aligned with previously published green sea turtle behaviors (Nishizawa et al., 2013; Jeantet et al., 2021), and we now present these phases for green turtles in two additional locations and for three additional species. By developing species-specific behavioral signatures (refer to Hart et al., 2025), we provide the basis for future tag deployments with potential onboard event-detection algorithms that can automate nesting counts and reduce observer dependence.



Our labor-intensive methodology—tagging individuals, recording second-by-second behavioral observations, matching with ADL and satellite telemetry data, and recapturing turtles to retrieve tags—served as a critical calibration experiment. These efforts lay the foundation for the remote, unbiased detection of nesting and NNE events using pop-off technologies, especially in areas with low recapture rates (e.g., Northern Gulf; Lamont et al., 2014). By validating acceleration signatures against ground-truthed observations, we show how biologging can move beyond general nest monitoring to generate data directly relevant to population assessment and regulatory decisions.

There are currently only a few indirect methods for improving estimates of sea turtle reproductive output. Satellite telemetry revealed higher clutch frequencies (mean 5.4 nests/female) than direct beach patrol data (2.2 nests/female) in Florida loggerheads (Tucker, 2010). Analysis of other capture–recapture data that go beyond raw nest counts per female has also improved estimates of clutch frequencies (Rivalan et al., 2006). More recent genetic tagging improved clutch frequency estimates (Shamblin et al., 2021). ADLs now add a complementary tool with sub-second resolution, offering direct behavioral identification of nesting events with further refinement of clutch counts when paired with satellite or GPS tags. While previous methods estimate nesting frequency by dividing breeding duration by average inter-nesting interval, our ADL approach provides direct behavioral identification, improving the accuracy of fecundity estimates that support population models and conservation policy.

The use of ADLs can refine inter-nesting interval estimates beyond what previous technologies offered, as shown in our data (Table 4). These metrics can now be tracked over time to assess environmental or anthropogenic influences on nesting behaviors. Deploying ADLs before the nesting season could capture first clutch dates or even early mating behavior. Compared to standard satellite tags, ADLs provide higher-resolution behavioral data and greater insight into specific on-land activities, despite GPS tags' haul-out capabilities that can have higher failure rates (Hays et al., 2021). Our findings complement those of Nishizawa et al. (2013), with similar nesting time means and phase distinctions. However, our larger sample sizes and diverse sites enabled broader species comparisons, reinforcing how deploying ADLs in wild populations can provide additional baseline data. Integrating ADLs in future monitoring could test hypotheses regarding fitness benefits of natural vs. built beaches (Pike, 2008; Wilmers et al., 2015) and detect behavioral changes from nearby human disturbances.

In aquatic settings, ADLs are being applied to assess turtle responses to ocean noise (Tyson et al., 2017) and thermoregulation behaviors linked to reproductive output (Fossette et al., 2012). ADLs could also be used to document fine-scale turtle behavior near highrisk areas, including renourished beaches. Although U.S. Army Corps of Engineers protocols require such monitoring, biologging tools are not yet standard. ADLs combined with satellite or GPS tags could reveal how turtles use renourished beaches and nearshore habitats. For instance, Lamont and Houser (2014) found that loggerhead nest density correlated with nearshore currents and sand movement, which change during renourishment projects. ADLs can track how turtles respond behaviorally to such changes.

Future directions point to greater autonomy and remote data retrieval. Advanced algorithms and multi-sensor biologging (e.g., event-detection tags), paired with satellite transmission capabilities, could enable the identification of nesting and NNE events without recapture, allowing clutch frequency to be estimated from a single deployment. Prototype pop-off ADL packages are already in use (KMH, USGS, modeled after Lear and Whitney, 2016; Whitmore et al., 2016) and are capable of collecting fine-scale data for days to weeks before pop-off and tag recovery.

#### 5 Conclusion

Clutch frequency is critical for modeling sea turtle population trajectories, but variations in reproductive output, nest site fidelity, and nesting intervals complicate estimates of nesting female abundance (Casale and Ceriani, 2020). By pairing ADLs with satellite/GPS tags and genetic mark-recapture methods (Shamblin et al., 2021), researchers can generate robust estimates of reproductive output without repeated recaptures. Recent validation work (e.g., Hounslow et al., 2022; Jeantet et al., 2020) provides essential reference data for interpreting ADL outputs across species.

Future studies could apply existing behavior classification algorithms to assess whether nesting turtles forage during internesting intervals. This could expand our understanding of resource use and energetics during the reproductive season. This collaborative work enhances sample sizes for species comparisons (Hays et al., 2004) and promotes data sharing (Hart et al., 2025). Our findings demonstrate that ADLs improve estimates of nesting and NNE behaviors and may eventually provide estimates of fecundity as well. Integrating this information into population models will likely sharpen assessments of anthropogenic impacts on reproduction and directly inform conservation strategies.

### Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found below: https://doi.org/10.5066/P14PODHE.

#### Ethics statement

The animal study was approved by USGS Institutional Animal Care and Use committee and National Park Service Institutional Animal Care and Use committee. The study was conducted in accordance with the local legislation and institutional requirements.

### **Author contributions**

KH: Methodology, Data curation, Investigation, Validation, Conceptualization, Resources, Writing – review & editing, Supervision, Visualization, Project administration, Formal analysis,

Writing – original draft, Funding acquisition. CW: Visualization, Investigation, Writing – review & editing, Methodology, Writing – original draft, Formal analysis, Validation. DS: Methodology, Conceptualization, Validation, Project administration, Funding acquisition, Supervision, Writing – review & editing, Resources, Data curation. ML: Validation, Investigation, Data curation, Resources, Project administration, Writing – review & editing, Methodology, Supervision. MC: Investigation, Data curation, Writing – review & editing, Validation, Formal analysis, Project administration. AC: Investigation, Data curation, Writing – review & editing, Methodology, Formal analysis. NW: Conceptualization, Visualization, Resources, Validation, Project administration, Formal analysis, Writing – original draft, Investigation, Supervision, Writing – review & editing, Data curation, Methodology.

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### Conflict of interest

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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