Long-term correlations in the surface behavior of dolphins

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Abstract. – Here we study the sequence of surface behavioral patterns of dolphins (\textit{Tursiops} sp.) and find long-term correlations. We show that the long-term correlations are not of a trivial nature, i.e. they cannot be explained by the repetition of the same surface behavior many times in a row. Our findings suggest that dolphins have a long collective memory extending back to the 7-th past behavior. As far as we know, this is the first evidence of long-term correlations in the behavior of a non-human species.

Increasing levels of approaches can be followed when studying the relationship between consecutive elements within a sequence. The 1st order approach studies elements neglecting preceding elements. The 2nd order approach studies the relationship between an element and the previous one. The \(m\)-th order approach consists of studying the dependence of one element on the \(m-1\) preceding elements. Sequences of behavioral patterns have been studied in many species. The 2nd order approach has been used for birds [1–4], primates [5–7] and cetaceans [5,6,8–12], to cite some examples. As far as we know, the highest orders that have been achieved are the 3rd [5,6] and the 4th [7] in exceptional studies. Although combinations of signals have been the subject of different studies because of their similarity with human phrases or sentences [13,14], a systematic comparison across species using information theory is found only in the pioneering work by McCowan and collaborators [5,6].

Long-range correlations are found in sequences of many different systems, such as the atmosphere [15,16], DNA [17–22], the human brain [23] and the fossil record [24]. Besides DNA, long-range correlations are found in other symbolic sequences. Of special interest here are literary texts [25–29] and other sequences of human behavior [30], since they constitute, as far as we know, the only evidence of long-distance correlations in behavioral sequences of a species.

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The fact that two behavioral patterns co-occur does not imply, in general, that they are significantly correlated. Statistical physics offers a wide range of measures for detecting long-range correlations. Some examples are Pearson correlation (e.g. [20,31]), information transfer (e.g. [20,31]), Fourier transform (e.g. [20]), wavelet transform (e.g. [32]) or other sophisticated techniques such as detrended fluctuation analysis [33,34]. Here, we chose Shannon’s information transfer for three reasons: simplicity, continuity with previous information theory work in animal behavior [5–7] and known advantages over other measures for detecting correlations within symbolic sequences [31,35] (e.g. the ability to detect non-linear correlations). In particular we will focus on two-point correlations between not necessarily consecutive elements of a sequence (e.g. [20,31]). Correlations between blocks of points have been considered by various studies (e.g. [6, 36]). The goal of the present article is three-fold: (a) going beyond 4th order approaches and entering the domain of long-term correlations that are found in the vast range of different systems summarized above, (b) providing the first report of long-range correlations in the sequence of behavioral patterns of a non-human species and (c) overcoming the statistical limitations of the methods used in previous animal behavior studies and establishing some guidelines for future studies.

Here we study long-term correlations within sequences of surface behavior of dolphin (Tursiops sp.). Surface behavioral patterns are series of body movements that can be unambiguously identified as a unit. The standard classification of surface behavioral patterns we use here has been used to define the ethogram of a population in many previous studies [37–40]. The ethogram was developed by Schneider after watching the population we studied for more than 1,000 hours [41]. For instance, the pattern “tail-stock dive” (TSD) is composed of the following movements: a dolphin surfaces, arches its body above water and increases its angle of re-entrance, only the tail peduncle is lifted out of the water (the tail is not visible above the surface) and the dolphin dives vertically. These behavioral patterns do not represent the entire behavioral repertoire of the population, but are all patterns that always occur at the surface and therefore can always be observed or recorded when performed (the ethogram was censored to ensure that only events that could always been observed were kept for this analysis; the observation point on the boat used for the study was such that other dolphins could not really mask the view of other dolphins).

Little is known about the function of surface dolphin behavioral patterns although it is suspected that some patterns may convey information [8]. On the one hand, aerial behavioral patterns (i.e. jumps) have, for example, been linked to agonistic displays [37,40] and social bound maintenance [42]. On the other hand, percussive patterns (dolphins producing a sound by slapping the water surface with a part of their body) have been shown to convey information about individual’s intentions [8,38,40]. Many surface behavioral patterns can be observed at the same time step since many dolphins can be performing a surface behavioral pattern at the same time step (non-necessarily the same pattern for each individual). Thus, a sequence of surface behavioral pattern consists of a list of bins where each bin contains all the patterns produced by the observed population during a certain time step (Table I). Our data set consists of a collection of 212 sequences recorded between 2000 and 2002 while following bottlenose dolphin (Tursiops sp.) focal schools in Doubtful Sound, Fiordland, New Zealand [8,43,44]. The total number of behavioral patterns produced within these sequences is 30,441. The mean number of patterns per bin is 1.05 ± 0.54. The mean elapsed time between one bin and the next is 15.74 ± 29.08 seconds. The length of a sequence is theoretically unlimited but in practice it is constrained by the observation period, which cannot exceed one hour to avoid observer fatigue. Thus, we cannot expect to find significant correlations beyond the mean sequence length. The mean sequence length in bins is 136.59 ± 105.30.

We define \( S = \{s_1, ..., s_j, ..., s_n\} \) as the set of surface behavioral patterns, where \( n = 33 \) in
Table I – Sample from a sequence of dolphin surface behavioral patterns starting at time $t$. The fragment has 22 occurrences of patterns in 19 time steps. The actual time elapsed between consecutive time steps is not necessarily the same. Each bin contains all the behavioral patterns produced during a certain time step. There are seven different patterns: TSD (tail-stock dive), TO (tail out), AS (active surfacing), FB (fart blow), LT (lob tail), VJ (vertical jump) and FOB (forced blow). At every time step, only one dolphin in the focal school is performing the behavioral pattern, except at time $t + 10$, where there are multiple occurrences of LT at the same time step. That is, all bins, except one, have a single behavioral pattern. In general, when there are multiple patterns at a certain time step, it is possible that one or more individuals produced more than one behavioral pattern. The identity of the producer is not available in our data.

<table>
<thead>
<tr>
<th>Time step</th>
<th>Patterns</th>
</tr>
</thead>
<tbody>
<tr>
<td>$t$</td>
<td>TSD</td>
</tr>
<tr>
<td>$t + 1$</td>
<td>TO</td>
</tr>
<tr>
<td>$t + 2$</td>
<td>AS</td>
</tr>
<tr>
<td>$t + 3$</td>
<td>FB</td>
</tr>
<tr>
<td>$t + 4$</td>
<td>TSD</td>
</tr>
<tr>
<td>$t + 5$</td>
<td>TSD</td>
</tr>
<tr>
<td>$t + 6$</td>
<td>AS</td>
</tr>
<tr>
<td>$t + 7$</td>
<td>TSD</td>
</tr>
<tr>
<td>$t + 8$</td>
<td>AS</td>
</tr>
<tr>
<td>$t + 9$</td>
<td>LT</td>
</tr>
<tr>
<td>$t + 10$</td>
<td>LT</td>
</tr>
<tr>
<td>$t + 11$</td>
<td>LT</td>
</tr>
<tr>
<td>$t + 12$</td>
<td>AS</td>
</tr>
<tr>
<td>$t + 13$</td>
<td>VJ</td>
</tr>
<tr>
<td>$t + 14$</td>
<td>TSD</td>
</tr>
<tr>
<td>$t + 15$</td>
<td>TSD</td>
</tr>
<tr>
<td>$t + 16$</td>
<td>TSD</td>
</tr>
<tr>
<td>$t + 17$</td>
<td>FOB</td>
</tr>
<tr>
<td>$t + 18$</td>
<td>FOB</td>
</tr>
</tbody>
</table>

Our dolphins. Imagine a sequence $X$ of $T$ bins of patterns performed during the same time step defined as $X = x_1, ..., x_i,..., x_T$, where $x_i \subseteq S$. We define the temporal distance (in time steps) between patterns in the bin $x_i$ and patterns in the bin $x_j$ as $|i - j|$, where $|...|$ is the absolute value operator. We define $N_{ij}(d)$ and $p_{ij}(d)$ as the number of times and the proportion of times, respectively, that the pattern $s_i$ has at appeared at temporal distance $d$ before the pattern $s_j$ within our collection of sequences. We have

$$p_{ij}(d) = \frac{N_{ij}(d)}{\sum_{s_i, s_j \in S} N_{ij}(d)}.$$  \hspace{1cm} (1)

We define $p^-_i(d)$ and $p^+_i(d)$, respectively, as the proportion of times that $i$ has appeared at temporal distance $d$ before and after any element of $S$ in the collection of sequences. In other words, we have

$$p^-_i(d) = \sum_{s_j \in S} p_{ij}(d)$$  \hspace{1cm} (2)
and

\[ p^+_{ij}(d) = \sum_{s_j \in S} p_{ij}(d). \]

We define the information transfer between patterns at temporal distance \( d \) as

\[ I(d) = \sum_{s_i, s_j \in S} p_{ij}(d) \log \frac{p_{ij}(d)}{p^+_{ij}(d)}. \]

Previous studies in animal behavior have not taken into account the fact that finite size effects overestimate the value of the actual correlations [31, 35, 45, 46]. Sometimes, the solution adopted in other fields has been to compare the correlations obtained with those obtained on scrambled sequences [20, 47]. This will be the solution we will follow here, where we use \( U \) control collections of sequences. Each control collection is obtained by scrambling all the patterns within each of the sequences in the original collection. Here we use simple shuffling (i.e. just a random permutation of the sequence, e.g. [20]). Other kinds of shuffling preserving some statistical properties of the sequence have been considered (e.g. [36]).

Previous studies of animal behavior [5–7] (and also in other fields, e.g. [20, 21]) have not considered the fact that some long-range correlations are of a trivial nature. The repetition of the same element many times in a row gives long-range correlations but in that case the relevance of any statistically significant correlation is questionable [19]. In the case of dolphins, it could merely indicate synchronization within the school; for instance, all dolphins diving together. To that aim, we will use two analysis: a standard analysis of correlations and another one filtering out the co-occurrence of an element with itself. The latter is obtained using a slightly modified definition of \( I(d) \) where \( N_{ii}(d) = 0 \) is forced when \( s_i \) has appeared more than once \( d \) events before itself. Again, \( p_{ij}(d) \) is defined as in Eq. 1. We denote by \( I_0(d) \) the information transfer when \( N_{ii}(d) = 0 \) is imposed.

Previous animal behavior work has studied the correlation of an element with the \( m - 1 \) preceding elements using an information theory approach [5–7]. Here we will study the correlation of an element with elements that appeared \( d \) events earlier (i.e. that appeared at distance \( d \) before). If \( S \) has \( n \) elements one has to keep track of the co-occurrence of every pattern with at most \( n^{m-1} \) possible combinations of patterns in the first case while in the second case one has to keep track only of the occurrence of every pattern with at most \( n \) different patterns. The first way of measuring correlations is more susceptible to finite size effects than the second.

Fig. 1 A-B and C-D show, respectively, that \( I(d) \) and \( I_0(d) \) in real sequences of surface behavioral patterns are significantly higher than the values obtained for scrambled sequences in a long range. We define \( d^* \) as the maximum value of \( d \) where statistically significant correlations are found. We obtained \( d^* \) approximately as the largest \( d \) where the maximum \( I(d) \) in \( U \) control collections of sequences does not reach the real \( I(d) \). We define \( d^*_0 \) as the value of \( d^* \) obtained on \( I_0(d) \). The probability that \( d^* \) is wrong due to deficient sampling (i.e. due to not having generated enough control collections of sequences) is smaller than \( p = 1/U \). In our case, the null hypothesis is that the real value of \( I(d) \) (or \( I_0(d) \)) and that of control sequences come from the same distribution. \( p \) gives an upper bound for the probability that the null hypothesis is true. In our case, \( U = 10^2 \) gives \( p = 10^{-2} \). We find \( d^* \approx 9 \) and \( d^*_0 \approx 7 \), providing support for the existence of long-term correlations within the sequences of dolphin surface behavioral patterns with \( p \)-value \( < 10^{-2} \) (if we are more conservative, i.e. considering a very small \( p \)-value we get \( d^* \approx 6 \) and \( d^*_0 \approx 7 \) with \( U = 10^3 \) and \( p \)-value \( < 10^{-4} \)). 'Long' here does not mean various orders of magnitude, as in DNA studies (e.g. [20]) but a range that has no precedent in animal behavior studies.
Fig. 1 – A. A detail for short distances of $I(d)$, the information transfer between dolphin surface behavioral patterns at temporal distance $d$. The corresponding mean value over $U = 10^2$ control collections of sequences is also shown (dotted line). Bars indicate the standard error of the estimated mean. The dashed line indicates the maximum value that $I(d)$ has achieved in the $U$ control collections of sequences. B. The same as A up to approximate point where the real series and the error bars of the control series cross. C. A detail for short distance of $I_0(d)$, the information transfer excluding one pattern with itself between dolphin surface behavioral patterns at temporal distance $d$. All series are equivalent to those in A. D. The same as C up to the approximate point where the real and the error bars of the control series cross. Logarithmic scale in both axes was used in B and D. Natural logarithms were used for the calculation of $I(d)$ and $I_0(d)$.

Excluding loops, i.e. co-occurrences of one pattern with itself, may have opposite effects. On the one hand, it eliminates a source of correlations, which alone would result in $I_0(d) < I(d)$. On the other hand, excluding loops constrains the set of possible combinations of two patterns at a certain distance, which alone would give $I_0(d) > I(d)$. Our results (Fig. 1 C-D) suggest that the span of correlations is not very influenced by excluding loops or not. As far as we know, the only species where long-range correlations within sequences of behavioral patterns have been found are humans [25–30] and the dolphins examined here.

We focused on schools of dolphins and not individuals. In spite of being rougher, group sampling is still useful for understanding the communicative value of behavioral patterns and has been used in humans [48–51] and other primates [52, 53]. Patterns could be correlated within and/or between individuals, since we do not have the identity of the producer of each behavioral pattern. The observed correlation could have been the result of a trivial synchronization between members of the school. However, the removal of repetitious patterns did not preclude the appearance of long-range correlation. Our findings suggest that dolphins have a long collective memory extending back to the 7-th past behavioral pattern (we are conservative and choose $d^*$ instead of $d^*$). The possibility that each individual has a long
memory of past behavioral patterns cannot be denied and is a challenge for future studies. One should be careful about the interpretation of our results. The fact that dolphin surface patterns at a certain temporal distance are correlated does not mean that patterns are caused by other patterns. Correlation does not imply causation. Correlations between two surface behavioral patterns could be an epiphenomenon of the fact that the two patterns are associated to the same behavioral state or context. We do not intend to explain the nature of the correlations. Besides, further work should be performed to determine if those long-term correlations are a signature of higher order species or a general phenomenon. Our work suggests some guidelines for future studies of animal sequential behavior.

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