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Review A fresh look at the "hot hand" paradox

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ARTICLE INFO

Article history: Received 4 April 2022 Received in revised form 22 September 2022 Accepted 27 September 2022 Available online 5 October 2022 Communicated by Dmitry Pelinovsky

Keywords: First-passage Hot hand Backward Kolmogorov equation Waiting time

ABSTRACT

We use the backward Kolmogorov equation approach to understand the apparently paradoxical feature that the mean waiting time to encounter distinct fixed-length sequences of heads and tails upon repeated fair coin flips can be different. For sequences of length 2, the mean time until the sequence HH (heads-heads) appears equals 6, while the waiting time for the sequence HT (heads-tails) equals 4. We give complete results for the waiting times of sequences of lengths 3, 4, and 5; the extension to longer sequences is straightforward (albeit more tedious). We also derive moment generating functions, from which any moment of the mean waiting time for specific sequences can be found. Finally, we compute the mean waiting times T_{2nH} for 2n heads in a row, as well as the moment generating function for this sequence, and $T_{n(HT)}$ for *n* alternating heads and tails. For large *n*, $T_{2nH} \sim 3T_{n(HT)}$. Thus distinct sequences of coin flips of the same length can have very different mean waiting times.

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1. Introduction

This article is dedicated to Charlie Doering, who left this world much too soon. While I did not have the opportunity to collaborate with him, I did have the pleasure of many fruitful and engaging scientific and social interactions with him over nearly four decades. It was always enjoyable to discuss science, and

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https://doi.org/10.1016/j.physd.2022.133551 0167-2789/© 2022 Elsevier B.V. All rights reserved.

indeed almost any topic, with Charlie because of his enthusiasm, his insights, and his ability to make you feel good about what you were presenting to him. He was a true mensch; the world would be a much better place if he were still with us.

The account of the "hot hand" phenomenon that is presented here is a "golden oldie" that has been extensively investigated in both the serious and the popular literature. Even though this topic was not close to Charlie's recent interests, I am pretty sure



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that he would have liked this story and that he would have had insightful suggestions that would have improved this article.

In repeated flips of a fair coin, the outcomes H (heads) or T (tails) occur equiprobably. Thus in a long string of N coin flips, the number of heads and tails, N_H and N_T , will be nearly equal, with $|N_H - N_T|$ of the order of \sqrt{N} . Given that H and T appear equiprobably, the average frequencies of specific fixed-length sequences of H's and T's must be the same; for example, the sequence HHTH occurs with the same frequency as HTTH. One might then expect that the waiting time before encountering either of these sequences should be the same. Surprisingly, this expectation is false! This phenomenon is now known as the "hot hand" paradox or the "hot hand" fallacy.

The counterintuitive nature of this "hot hand" paradox appears to have first been studied in a systematic way by Gilovich et al. [1]. These authors sought to understand if scoring streaks of individual players in basketball games was merely a manifestation of random fluctuations or whether a player's scoring could be characterized by well defined "hot" and "cold" streaks. Their conclusion was that a "hot hand" did not exist in basketball scoring statistics.

However, it was later found that there does exist a hot hand paradox when the problem is formulated in an appropriate way. Namely, in a long string of fair coin flips, the waiting times for the next appearance of distinct sequences of equal-length strings of H's and T's (for example, HHTH and HTTH), can be different. This form of the hot hand paradox is perhaps best stated in the following stark way (as mentioned at the outset of the article by Miller and Sanjurjo [2]): suppose one flips a fair coin *N* times. Whenever an H occurs (there should be roughly $\frac{1}{2}N$ such H's), one then records the outcome of the next coin flip. Naively, one expects that that number of recorded H's should be one-half of the total number of H's, i.e., $\frac{1}{4}N$ such events. In fact, this number is less than $\frac{1}{4}N$!

The seemingly paradoxical nature of the "hot hand" phenomenon has spawned considerable discussion and literature that has ultimately resolved the paradox (see, e.g., Refs. [2–10]). However, the approaches given in some of these references are complicated and the simplicity of the mechanism that underlies the paradox can be lost in calculational details; a notable exception, however is Ref. [11], which provides an extraordinarily simple and appealing way to determine the mean waiting time for an arbitrary sequence of heads and tails of arbitrary length.

Here we give an alternative route to understand the hot hand paradox that is based on the backward Kolmogorov equation [12–14]. This formulation has proved to be extremely useful in a variety of first-passage phenomena. By recasting the hot hand paradox as a first-passage problem, we can use the backward Kolmogorov equation to compute the waiting time for specific sequences of H's and T's of length up to 5. This method can be straightforwardly extended to longer sequences if so desired. We also give an intuitive reason why different sequences of the same length do not occur with the same frequency.

We then extend the backward Kolmogorov approach to derive the moment generating function for all sequences of length up to 3; again, this approach could be readily extended to longer sequences if so desired. From the generating function, arbitraryorder moments of the waiting time can be easily extracted. These results seem to have not been derived previously. Finally, we also compute the waiting time for particularly simple sequences of arbitrary length, namely, the sequence of 2*n* consecutive H's and the sequence of *n* consecutive (HT)'s. We find that $T_{2nH} \sim 3T_{n(HT)}$, so that 2*n* heads in a row is three times less frequent than *n* (HT)'s in a row.

To complete this introduction, we now present the simple idea that underlies the backward Kolmogorov equation. Consider a Markov process that is currently in a particular state *S*. We want to compute the average time $T_{S \rightarrow F}$ until the process reaches a specified final state *F*. Suppose that there are two possible outcomes at each stage of the process that occur with equal probability. That is, from state *S*, the process transitions either to state *S'* or to *S''*, each with probability $\frac{1}{2}$. Suppose further that the time required for each transition equals 1. Since the Markov process has no memory, when either of the states *S'* or *S''* are reached, the process starts anew. Consequently, the hitting time when starting from *S* is just the average of the hitting times starting from either *S'* or *S''* plus the time spent in the transition itself. That is

$$T_{S \to F} = \frac{1}{2} (T_{S' \to F} + 1) + \frac{1}{2} (T_{S'' \to F} + 1).$$
(1)

We will use this basic equation to compute the waiting time for specific sequences of H's and T's of a given length as a result of repeated flips of a fair coin.

A powerful aspect of the backward Kolmogorov approach is that it applies to *any functional* of the first-passage time and not just the average first-passage time. Thus one can write equations similar to (1) for the mean-square time, the mean-cube time, etc. Even more striking, we can write an equation of the form of (1) for the moment generating function, $\langle e^{-st} \rangle$, where the angle brackets denote taking the average over all possible sequences of heads and tails, from which arbitrary-order moments can be obtained merely by a Taylor-series expansion.

2. Average waiting times

2.1. Doublets

Let us start with the simplest example of length-2 sequences. The possible sequences are HH, HT, TH, and HH. Because the coin is fair, we obtain the same statistics by the substitution $H \leftrightarrow T$, so that the waiting time for the sequences TT and TH is the same as that for HH and HT. Consequently, we only consider the first two sequences. How long does one have to wait before encountering each of these sequences in a long string of fair coin flips?

Starting with Eq. (1), we first compute the mean waiting time $T_{\rm HH}$ to encounter an HH sequence. For this purpose, we introduce the auxiliary restricted times:

- *A*, the average waiting time for the sequence HH starting with an H.
- *B*, the average waiting time for the sequence HH starting with a T.

These two times obey the backward equations

$$A = \frac{1}{2} \times 2 + \frac{1}{2}(1+B)$$

$$B = \frac{1}{2}(1+B) + \frac{1}{2}(1+A),$$
(2a)

which express the waiting times *A* and *B* as the average time to reach the desired final state after a coin flip, plus the time for the coin flip itself. Thus in the equation for *A*, the first term accounts for the next coin flip being H (which occurs with probability $\frac{1}{2}$) after which the sequence HH has been generated. The factor 2 counts the two coin flips that are need to generate the sequence HH from scratch. The second term accounts for the next coin flip being T. Again, the probability for this event is $\frac{1}{2}$. Once a *T* appears, the waiting time to generate an HH sequence is *B* by definition. Consequently, the factor (1 + B) accounts for the time spent in making a single coin flip plus the waiting time when the sequence string starts with T. Solving these two equations gives A = 5 and B = 7. Since H and T appear equiprobably in a long series of fair coin flips, on average, the average waiting time for the sequence HH is $T_{HH} = \frac{1}{2}(A + B) = 6$.

For the waiting time T_{HT} , we now define

- *A*, the average waiting time for the sequence HT starting with H.
- *B*, the average waiting time for the sequence HT starting with T.

Using the same reasoning as above, these two times obey the backward equations

$$A = \frac{1}{2} \times 2 + \frac{1}{2}(1+A)$$

$$B = \frac{1}{2}(1+A) + \frac{1}{2}(1+B),$$
(2b)

from which (A, B) = (3, 5). Again, because H and T appear equiprobably in a long series of coin flips, $T_{HT} = \frac{1}{2}(A + B) = 4$.

Why are these two times different? The key lies in the second term on the right in the first lines of Eqs. (2a) and (2b), which account for a "mistake". For example, in Eq. (2a), if the next coin flip is T, one has to "start over" to generate HH. The soonest that the next HH can happen immediately after a T is after two more coin flips. In contrast, in Eq. (2b), if the next coin flip is H (again a mistake), the process "starts over". Now, however, the next HT sequence can appear in after only one more coin flip.

2.2. Triplets

We now generalize to triplet sequences. The $2^3 = 8$ distinct triplets are HHH, HHT, HTH, and THH and their counterparts obtained by H \leftrightarrow T. By left/right symmetry, the triplets HHT and THH have identical statistics, so the only distinct sequences are HHH, HHT, and HTH. Let T_{HHH} be the average waiting time to encounter the sequence with three consecutive H's. To compute this time, we define the auxiliary restricted times:

- *A*, the average waiting time for HHH when the current state is H;
- *B*, the average waiting time for HHH when the current state is HH;
- *C*, the average waiting time for HHH when the current state is T.

Following the same reasoning that led to Eqs. (2a), the above times satisfy

$$A = \frac{1}{2}(1+B) + \frac{1}{2}(1+C)$$

$$B = \frac{1}{2} \times 2 + \frac{1}{2}(1+C)$$

$$C = \frac{1}{2}(1+A) + \frac{1}{2}(1+C).$$
(3)

The first term in the equation for *B* merits explanation. From the state HH, the desired sequence HHH is obtained with probability $\frac{1}{2}$, and the time for this event is 2 because the time is measured starting *before* the second H has been added to the sequence. The solution to (3) is (*A*, *B*, *C*) = (13, 9, 15). Since the probability to find an H or a T are equal, the average waiting time to encounter the sequence HHH is just the average of the times to find HHH when starting with an H or starting with a T. Thus $T_{\text{HHH}} = \frac{1}{2}(A + C) = 14$.

Similarly, let T_{HHT} be the average waiting time to encounter the sequence HHT. Now we introduce the auxiliary times:

- *A*, the average waiting time for HHT when the current state is H;
- *B*, the average waiting time for HHT when the current state is HH;
- *C*, the average waiting time for HHT when the current state is T.

These times satisfy

$$A = \frac{1}{2}(1+B) + \frac{1}{2}(1+C)$$

$$B = \frac{1}{2} \times 2 + \frac{1}{2}(1+B)$$

$$C = \frac{1}{2}(1+A) + \frac{1}{2}(1+C),$$
(4)

Again, there is a subtlety in the second equation: if the initial state is HH, then after adding an H, the current state is still HH, so that the second term involves *B*. This feature that the initial state consists of a subsequence of length greater than one plays an increasing role for longer sequences (see Appendices A and B). The solution to (4) is (*A*, *B*, *C*) = (7, 3, 9). Thus the mean waiting time to encounter the sequence HHT is the average of the times to find HHT after an H or after a T, which gives $T_{\text{HHT}} = \frac{1}{2}(A+C) = 8$.

Finally, let T_{HTH} be the average waiting time to encounter the sequence HTH. We introduce the auxiliary times:

- *A*, the average waiting time for HTH when the current state is H;
- *B*, the average waiting time for HTH when the current state is HT;
- *C*, the average waiting time for HHT when the current state is T.

These times satisfy

$$A = \frac{1}{2}(1+A) + \frac{1}{2}(1+B)$$

$$B = \frac{1}{2} \times 2 + \frac{1}{2}(1+C)$$

$$C = \frac{1}{2}(1+A) + \frac{1}{2}(1+C),$$
(5)

with solutions (A, B, C) = (9, 7, 11). The mean waiting time to encounter the sequence HTH is the average of the times to find HTH after an H or after a T, which gives $T_{\text{HTH}} = \frac{1}{2}(A+C) = 10$. To summarize, $T_{\text{HHH}} = 14$, $T_{\text{HHT}} = 8$, and $T_{\text{HTH}} = 10$, in agreement with known results. The corresponding results for quartet and quintet sequences are given in Appendices A and B.

3. Mean-square waiting times

We now extend the backward Kolmogorov approach to compute the mean-square waiting time for specific sequences. As we will show in the next section, deriving the moment generating function requires less computation than the second moment, and it becomes increasingly laborious to directly derive moments beyond the second. Thus we present the calculation of the second moment for doublet sequences as an illustration only, and then proceed to the moment generating functions for doublets and longer sequences.

Let us start with the sequence HH. In analogy with the discussion of the mean waiting time, we now introduce the auxiliary restricted mean-square times:

- *A*₂, the mean-square waiting time for HH starting with an H.
- B_2 , the mean-square waiting time for HH starting with a T.

These two quantities obey the backward equations

$$A_{2} = \frac{1}{2} \times 4 + \frac{1}{2} \langle 1 + \mathcal{B} \rangle^{2}$$

$$B_{2} = \frac{1}{2} \langle 1 + \mathcal{B} \rangle^{2} + \frac{1}{2} \langle 1 + \mathcal{A} \rangle^{2}.$$
(6a)

Here the notations A and \mathcal{B} denote the time to reach the sequence HH from a specific sequence realization when starting from an H or a *T* respectively, and the angle brackets denote an average over all sequences. According to this notation, $A = \langle A \rangle$ and $A_2 = \langle A^2 \rangle$.

We expand the quadratic inside the angle brackets to give

$$A_{2} = 2 + \frac{1}{2}(1 + 2B + B_{2})$$

$$B_{2} = \frac{1}{2}(1 + 2B + B_{2}) + \frac{1}{2}(1 + 2A + A_{2}),$$
(6b)

and then substitute in the solutions A = 5 and B = 7 from Eq. (2a) to obtain $A_2 = 45$ and $B_2 = 71$. Since H and T occur with equal probability, the mean-square waiting time to find the sequence HH is the average of A_2 and B_2 : $T_{HH}^2 = 58$.

For the sequence HT, we introduce

• *A*₂, the mean-square waiting time for HT starting with H.

• B_2 , the mean-square waiting time for HT starting with T.

These two times obey the backward equations

$$A_{2} = \frac{1}{2} \times 4 + \frac{1}{2} \langle 1 + A \rangle^{2} = 2 + \frac{1}{2} (1 + 2A + A_{2})$$

$$B_{2} = \frac{1}{2} \langle 1 + A \rangle^{2} + \frac{1}{2} \langle 1 + B \rangle^{2}$$

$$= \frac{1}{2} (1 + 2A + A_{2}) + \frac{1}{2} (1 + 2B + B_{2}).$$
(7)

Using A = 3 and B = 5 from Eq. (2b), we now find $A_2 = 11$ and $B_2 = 29$. Then the mean-square waiting time for the sequence HT is $T_{\text{HT}}^2 = \frac{1}{2}(A_2 + B_2) = 20$.

Clearly, this same approach can be extended to higher moments, but the calculations become progressively more tedious because the equations for the *n*th moment involves all lowerorder moments. As shown below, however, we can compute the moment generating function with less computational labor than that for the second moment, and from this generating function all moments are obtained by a simple Taylor series expansion.

4. The moment generating function

4.1. Doublet sequence HH

To begin, we define the moment generating functions

$$G_A = \langle e^{-s\mathcal{A}} \rangle \qquad \qquad G_B = \langle e^{-s\mathcal{B}} \rangle,$$

where again A and B denote the time to reach the sequence HH from a specific sequence realization when starting from an H or a T respectively. These generating functions are especially useful because they contain all moments of the waiting time by a Taylor series expansion. For example, the moments of the time A to encounter the sequence HH when starting with an H are

$$G_A = 1 - sA + \frac{1}{2}s^2A_2 - \frac{1}{3!}s^2A_3 + \cdots$$

For the sequence HH, the backward Kolmogorov equation for the moment generating functions are obtained by taking each of the terms in Eqs. (2a) and putting it inside the exponential like so: $e^{-s \times each \text{ term}}$. This immediately gives

$$G_{A} = \frac{1}{2} \langle e^{-2s} \rangle + \frac{1}{2} \langle e^{-s(1+\mathcal{B})} \rangle = \frac{1}{2} (e^{-2s} + G_{B} e^{-s})$$

$$G_{B} = \frac{1}{2} \langle e^{-s(1+\mathcal{B})} \rangle + \frac{1}{2} \langle e^{-s(1+\mathcal{A})} \rangle = \frac{1}{2} (G_{B} e^{-s} + G_{A} e^{-s}).$$
(8)

The structure of these equations mirror those of Eqs. (2a) for the mean waiting time. Because of this close correspondence, solving the backward equations for the moment generating functions has the same degree of difficulty as solving for the mean waiting times. The solution to (8) is

$$G_A = \frac{(2 e^s - 1)e^{-s}}{4 e^{2s} - 2 e^s - 1} \qquad G_B = \frac{e^{-s}}{4 e^{2s} - 2 e^s - 1}.$$
 (9)

Taylor expanding these two generating functions gives

$$G_A = 1 - 5s + \frac{45}{2}s^2 - \frac{629}{3!}s^3 + \cdots$$

$$G_B = 1 - 7s + \frac{71}{2}s^2 - \frac{1015}{3!}s^3 + \cdots,$$
(10)

from which $T_{HH} = \frac{1}{2}(A + B) = 6$, $T_{HH}^2 = 2 \times \frac{1}{2}(A_2 + B_2) = 58$, and $T_{HH}^3 = 3! \times \frac{1}{2}(A_3 + B_3) = 822$, etc.

Finally, we note that the moment generating functions in (9) both have simple poles at $s^* \approx -0.21194$. This implies that the

long-time tail of the distribution of waiting times, P(t), has an exponential decay of the form $e^{-|s^*|t} = e^{-t/\tau}$, with $\tau \approx 4.718$. Even though the times T_{HH} and T_{HT} are numerically different, both times are governed by a single characteristic scale.

4.2. Doublet sequence HT

Building on the above perspective, we write the backward equations for the moment generating functions for the sequence HT by merely reading off from Eqs. (2b):

$$G_A = \frac{1}{2}(e^{-2s} + G_A e^{-s})$$

$$G_B = \frac{1}{2}(G_B e^{-s} + G_A e^{-s}),$$
(11)

with solutions

$$G_A = \frac{e^{-s}}{2e^s - 1}$$
 $G_B = \frac{e^{-s}}{(2e^s - 1)^2}$. (12)

Taylor expanding these two generating functions gives

$$G_A = 1 - 3s + \frac{11}{2}s^2 - \frac{51}{3!}s^3 + \cdots$$

$$G_B = 1 - 5s + \frac{29}{2}s^2 - \frac{197}{3!}s^3 + \cdots,$$
(13)

from which $T_{HT} = \frac{1}{2}(A + B) = 4$, $T_{HT}^2 = 2 \times \frac{1}{2}(A_2 + B_2) = 20$, and $T_{HT}^3 = 6 \times \frac{1}{2}(A_3 + B_3) = 124$, etc.

Corresponding results for triplet sequences are given in Appendix C.

5. Simple arbitrary length sequences

5.1. Mean waiting time for n consecutive H's

While the calculational details for longer sequences are straightforward, they become progressively more tedious as the sequence length is increased. However, for the sequence of *n* consecutive H's, the equations for the restricted times are sufficiently systematic in character that they can be readily solved. To this end, we first define the following set of restricted times:

- A_k, the average waiting time for nH starting from k consecutive H's;
- *B*, the average waiting time for *n*H starting from T.

These times satisfy

$$A_{1} = \frac{1}{2}(1 + A_{2}) + \frac{1}{2}(1 + B)$$

$$A_{2} = \frac{1}{2}(1 + A_{3}) + \frac{1}{2}(1 + B)$$

$$\vdots$$

$$A_{n-2} = \frac{1}{2}(1 + A_{n-1}) + \frac{1}{2}(1 + B)$$

$$A_{n-1} = \frac{1}{2} \times 2 + \frac{1}{2}(1 + B)$$

$$B = \frac{1}{2}(1 + A_{1}) + \frac{1}{2}(1 + B).$$
(14)

Because of the linear and recursive structure of Eqs. (14) they can be solved one by one, and the final result, for the waiting time to *n*H when starting from an H or a *T* respectively, is

$$A_1 = 2^{n+2} - 3$$
 $B = 2^{n+2} - 1.$ (15)

Finally, T_{nH} is the average of these two waiting times:

$$T_{\rm nH} = \frac{1}{2}(A_1 + B) = 2^{n+2} - 2.$$
 (16)

A

5.2. n Consecutive (HT)'s

We can carry out a similar calculation for the sequence of n consecutive (HT)'s. Here, we first define the following set of restricted times:

- A_{2k-1} , the average waiting time for n(HT) starting from (k 1)1)(HT)H;
- A_{2k} , the average waiting time for n(HT) starting from k(HT);
- *B*, the average waiting time for *n*(HT) starting from T.

These times satisfy

$$A_{1} = \frac{1}{2}(1 + A_{2}) + \frac{1}{2}(1 + A_{1})$$

$$A_{2} = \frac{1}{2}(1 + A_{3}) + \frac{1}{2}(1 + B)$$

$$A_{3} = \frac{1}{2}(1 + A_{4}) + \frac{1}{2}(1 + A_{1})$$

$$A_{4} = \frac{1}{2}(1 + A_{5}) + \frac{1}{2}(1 + B)$$

$$\vdots$$

$$A_{2n-3} = \frac{1}{2}(1 + A_{2n-2}) + \frac{1}{2}(1 + A_{1})$$

$$A_{2n-2} = \frac{1}{2}(1 + A_{2n-1}) + \frac{1}{2}(1 + B)$$

$$A_{2n-1} = \frac{1}{2} \times 2 + \frac{1}{2}(1 + A_{1})$$

$$B = \frac{1}{2}(1 + A_{1}) + \frac{1}{2}(1 + B)$$
Solving these equations recursively, we find

Solving these equations recursively, we find

$$A_1 = \frac{4}{3}(2^{2n} - 1) - 1 \qquad B = 2 + A_1.$$
(18)

The average waiting time $T_{n(HT)} = \frac{1}{2}(A_1 + B) = A_1 + 1$ now is

$$T_{n(\rm HT)} = \frac{4}{2}(2^{2n} - 1). \tag{19}$$

It is instructive to compare the times T_{nH} and $T_{n(HT)}$. The fair comparison is between T_{2nH} and $T_{n(HT)}$; i.e., between strings of the same length. Asymptotically, Eq. (16) gives $T_{2nH} \sim 4 \cdot 2^{2n}$, while (19) gives $T_{n(\text{HT})} \sim \frac{4}{3} \cdot 2^{2n}$. One has to wait three times as long, on average, to encounter a sequence of 2n H's in a row compared to a sequence of n (HT)'s in a row.

5.3. Moment generating function for n consecutive H's

We now use the approach outlined in Section 4 to compute the moment generating function for the occurrence of *n* consecutive H's. Following the notation of Section 5.1, we define G_k as the moment generating function for the time to reach the state of nH's when the sequence starts with k consecutive H's, while G_B is the moment generating function to reach the state nH when the sequence starts with a T. In close analogy with Eqs. (14), these moment generating functions satisfy

$$G_{1} = x G_{2} + x G_{B}$$

$$G_{2} = x G_{3} + x G_{B}$$

$$\vdots$$

$$G_{n-2} = x G_{n-1} + x G_{B}$$

$$G_{n-1} = y + x G_{B}$$

$$G_{B} = x G_{1} + x G_{B},$$
(20)

where for notational simplicity we introduce $x \equiv \frac{1}{2}e^{-s}$ and $y \equiv$ $\frac{1}{2}e^{-2s}$. The last equation gives $G_B = x G_1/(1-x)$, while from the penultimate equation we can replace the factor $x G_B$ everywhere with $G_{n-1} - y$.

Following similar steps as those used to solve (14), the moment generating functions G_1 and G_B are

$$G_1 = \frac{x^{n-1}(1-x)^2 y}{(1-x)^2 - x^2(1-x^n)} \qquad G_B = \frac{x^n(1-x)y}{(1-x)^2 - x^2(1-x^n)}.$$
 (21)

The moment generating function for the time T_{nH} to encounter the sequence *n*H is the average of G_1 and G_R ; that is

$$\frac{1}{2}(G_1 + G_B) = \frac{1}{2} \frac{x^{n-1}(1-x)y}{(1-x)^2 - x^2(1-x^n)}$$
(22)

Finally, we may expand this generating function in a power series to obtain the moments of the time to reach the sequence *n*H. The first few moments are:

6. Concluding comments

While many of the results given here are already quite well known, the backward Kolmogorov approach provides a fresh perspective to calculate average waiting times for specific sequences of H's and T's in a long string of repeated flips of a fair coin. Once one understands the underlying idea of the Kolmogorov approach, computing waiting times for specific sequences is straightforward and direct.

Another important aspect of this approach is that it also allows one to compute any *functional* of the waiting time, such as higher moments, and even the characteristic function, $\langle \exp(-sT) \rangle$. We showed how to compute the moment generating function for short specific sequences, as well as for special arbitrary-length sequences of heads and tails, from which arbitrary moments of the waiting time for these sequences can easily be derived. This approach can be readily extended to longer sequences. These results about higher moments and the moment generating function appear to have not been treated previously. An open challenge is whether there exists a simple approach of the spirit given in [11], that allows one to compute the moment generating function for any sequence of arbitrary length.

The surprising outcome of repeated fair coin flips is that the average waiting times for specific sequences of H's and T's of the same length are different even though the average frequency of these two sequences are the same. The effect is especially pronounced for a long string of 2n H's compared to the string of n (HT)'s. For large *n*, one has to wait three times longer to encounter the former sequence compared to the latter.

As a final note, although our approach unambiguously demonstrates the existence of distinct waiting times for distinct fixedlength sequences, this seemingly paradoxical phenomenon requires careful thought to appreciate intuitively.

Declaration of competing interest

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests: Sidney Redner reports financial support was provided by Santa Fe Institute. Sidney Redner reports a relationship with Santa Fe Institute that includes: employment.

Data availability

N/A

Acknowledgments

I thank David Atkinson and Porter Johnson for helpful suggestions while this manuscript was being written, Paul Krapivsky and Michael Mauboussin for their encouragement, and Ivan Corwin for helpful advice. I also gratefully acknowledge financial support from NSF, United States Grants DMR-1608211 and DMR-1910736.

Appendix A. Calculational details for guartet sequences

The six distinct quartets are HHHH, HHHT, HHTH, HHTT, HTHT, and HTTH. The calculation $T_{\rm HHHH}$ was given in Section 5.1 and here we continue with T_{HHHT} . For T_{HHHT} , we define A, B, C, and *D* as the average waiting time for HHHT when the current state is H, HH, HHH, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+B) + \frac{1}{2}(1+D)$$

$$B = \frac{1}{2}(1+C) + \frac{1}{2}(1+D)$$

$$C = \frac{1}{2} \times 2 + \frac{1}{2}(1+C)$$

$$D = \frac{1}{2}(1+A) + \frac{1}{2}(1+D),$$

(A.1)

with solution (A, B, C, D) = (15, 11, 3, 17), from which $T_{\text{HHHT}} =$ $\frac{1}{2}(A+D) = 16.$

For $T_{\rm HHTH}$, we define A, B, C, and D as the average waiting time for HHTH when the current state is H, HH, HHT, and T, respectively. These times satisfy

 $A = \frac{1}{2}(1+B) + \frac{1}{2}(1+D)$ $B = \frac{1}{2}(1+B) + \frac{1}{2}(1+C)$ (A.2) $C = \frac{1}{2} \times 2 + \frac{1}{2}(1+D)$ $D = \frac{1}{2}(1+A) + \frac{1}{2}(1+D),$

with solution (*A*, *B*, *C*, *D*) = (17, 13, 11, 19), from which $T_{\rm HHTH} =$ $\frac{1}{2}(A+D) = 18.$

For T_{HHTT} , we define A, B, C, D as the average waiting time for HHTT when the current state is H, HH, HHT, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+B) + \frac{1}{2}(1+D)$$

$$B = \frac{1}{2}(1+B) + \frac{1}{2}(1+C)$$

$$C = \frac{1}{2} \times 2 + \frac{1}{2}(1+A)$$

$$D = \frac{1}{2}(1+A) + \frac{1}{2}(1+D),$$

(A.3)

with solution (A, B, C, D) = (15, 11, 9, 17), from which $T_{\text{HHTT}} =$ $\frac{1}{2}(A+D) = 16.$

For T_{HTHT} , we define A, B, C, D as the average waiting time for HTHT when the current state is H, HT, HTH, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+A) + \frac{1}{2}(1+B)$$

$$B = \frac{1}{2}(1+C) + \frac{1}{2}(1+D)$$

$$C = \frac{1}{2} \times 2 + \frac{1}{2}(1+A)$$

$$D = \frac{1}{2}(1+A) + \frac{1}{2}(1+D),$$

(A.4)

with solution (*A*, *B*, *C*, *D*) = (19, 17, 11, 21), from which T_{HTHT} = $\frac{1}{2}(A+D) = 20.$

Finally, for $T_{\rm HTTH}$, we define A, B, C, D as the average waiting time for HTTH when the current state is H, HT, HTT, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+A) + \frac{1}{2}(1+B)$$

$$B = \frac{1}{2}(1+A) + \frac{1}{2}(1+C)$$

$$C = \frac{1}{2} \times 2 + \frac{1}{2}(1+D)$$

$$D = \frac{1}{2}(1+A) + \frac{1}{2}(1+D),$$

(A.5)

with solution (A, B, C, D) = (17, 15, 11, 19), from which $T_{\rm HTTH} =$ $\frac{1}{2}(A+D) = 18.$

In summary, the quartet average waiting times in reverse time order are $T_{4H} = 30$, $T_{HTHT} = 20$, $T_{HHTH} = T_{HTTH} = 18$, $T_{HHHT} =$ $T_{\rm HHTT} = 16.$

Appendix B. Quintet sequences

The nine distinct quintets are: HHHHH, HHHHT, HHHTH, HHTHH, HHHTT, HHTHT, HTHHT, HTHHT, and HTTHH. There are additional non-independent sequences that are obtained by either the interchange $H \leftrightarrow T$ or by reading the above sequences in reverse order. Again, the calculation $T_{\rm HHHHH}$ was given in Section 5.1 and we continue with T_{HHHHT} . For T_{HHHHT} , we define A, B, C, D, E as the average waiting time for HHHHT when the current state is H, HH, HHH, HHHH, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+B) + \frac{1}{2}(1+E)$$

$$B = \frac{1}{2}(1+C) + \frac{1}{2}(1+E)$$

$$C = \frac{1}{2}(1+D) + \frac{1}{2}(1+E)$$

$$D = \frac{1}{2} \times 2 + \frac{1}{2}(1+D)$$

$$E = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$

(B.1)

with solution (A, B, C, D, E) = (31, 27, 19, 3, 33), and we obtain $T_{\rm HHHHT} = \frac{1}{2}(A+E) = 32.$

For T_{HHHTH} , we define A, B, C, D, E as the average waiting time for HHHTH when the current state is H, HH, HHH, HHHT, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+B) + \frac{1}{2}(1+E)$$

$$B = \frac{1}{2}(1+C) + \frac{1}{2}(1+E)$$

$$C = \frac{1}{2}(1+C) + \frac{1}{2}(1+D)$$

$$D = \frac{1}{2} \times 2 + \frac{1}{2}(1+E)$$

$$E = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$

(B.2)

with solution (*A*, *B*, *C*, *D*, *E*) = (33, 29, 21, 19, 35), and we obtain $T_{\rm HHHTH} = \frac{1}{2}(A+E) = 34.$

For *T*_{HHTHH}, we define *A*, *B*, *C*, *D*, *E* as the average waiting time for HHTHH when the current state is H, HH, HHT, HHTH, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+B) + \frac{1}{2}(1+E)$$

$$B = \frac{1}{2}(1+B) + \frac{1}{2}(1+C)$$

$$C = \frac{1}{2}(1+D) + \frac{1}{2}(1+E)$$

$$D = \frac{1}{2} \times 2 + \frac{1}{2}(1+E)$$

$$E = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$

(B.3)

with solution (*A*, *B*, *C*, *D*, *E*) = (37, 33, 31, 21, 39), and we obtain $T_{\rm HHTHH} = \frac{1}{2}(A+E) = 38.$

For *T*_{HHHTT}, we define *A*, *B*, *C*, *D*, *E* as the average waiting time for HHHTT when the current state is H, HH, HHH, HHHT, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$

$$B = \frac{1}{2}(1+C) + \frac{1}{2}(1+E)$$

$$C = \frac{1}{2}(1+C) + \frac{1}{2}(1+D)$$

$$D = \frac{1}{2} \times 2 + \frac{1}{2}(1+A)$$

$$E = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$

(B.4)

with solution (*A*, *B*, *C*, *D*, *E*) = (31, 27, 19, 17, 33), and we obtain $T_{\rm HHHTT} = \frac{1}{2}(A+E) = 32.$

For *T*_{HHTHT}, we define *A*, *B*, *C*, *D*, *E* as the average waiting time for HHTHT when the current state is H, HH, HHT, HHTH, and T,

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respectively. These times satisfy

$$A = \frac{1}{2}(1+B) + \frac{1}{2}(1+E)$$

$$B = \frac{1}{2}(1+B) + \frac{1}{2}(1+C)$$

$$C = \frac{1}{2}(1+D) + \frac{1}{2}(1+E)$$

$$D = \frac{1}{2} \times 2 + \frac{1}{2}(1+B)$$

$$E = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$
(B.5)

with solutions (*A*, *B*, *C*, *D*, *E*) = (31, 27, 25, 15, 33), and we obtain $T_{\text{HHTHT}} = \frac{1}{2}(A + E) = 32.$

For T_{HTHHT} , we define *A*, *B*, *C*, *D*, *E* as the average waiting time for HTHHT when the current state is H, HT, HTH, HTHH, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+A) + \frac{1}{2}(1+B)$$

$$B = \frac{1}{2}(1+C) + \frac{1}{2}(1+E)$$

$$C = \frac{1}{2}(1+D) + \frac{1}{2}(1+B)$$

$$D = \frac{1}{2} \times 2 + \frac{1}{2}(1+A)$$

$$E = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$
(B.6)

with solution (*A*, *B*, *C*, *D*, *E*) = (35, 33, 27, 19, 37), and we obtain $T_{\text{HTHHT}} = \frac{1}{2}(A + E) = 36$.

For $T_{\rm HTHTH}$, we define A, B, C, D, E as the average waiting time for HTHTH when the current state is H, HT, HTH, HTHT, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+A) + \frac{1}{2}(1+B)$$

$$B = \frac{1}{2}(1+C) + \frac{1}{2}(1+E)$$

$$C = \frac{1}{2}(1+A) + \frac{1}{2}(1+D)$$

$$D = \frac{1}{2} \times 2 + \frac{1}{2}(1+E)$$

$$E = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$
(B.7)

with solution (*A*, *B*, *C*, *D*, *E*) = (41, 39, 33, 23, 43), and we obtain $T_{\text{HTHTH}} = \frac{1}{2}(A + E) = 42$.

For $T_{\rm HTTHH}$, we define *A*, *B*, *C*, *D*, *E* as the average waiting time for HTTHH when the current state is H, HT, HTT, HTTH, and T, respectively. These times satisfy

$$A = \frac{1}{2}(1+A) + \frac{1}{2}(1+B)$$

$$B = \frac{1}{2}(1+A) + \frac{1}{2}(1+C)$$

$$C = \frac{1}{2}(1+D) + \frac{1}{2}(1+E)$$

$$D = \frac{1}{2} \times 2 + \frac{1}{2}(1+B)$$

$$E = \frac{1}{2}(1+A) + \frac{1}{2}(1+E)$$
(B.8)

with solution (*A*, *B*, *C*, *D*, *E*) = (33, 31, 27, 17, 35), and we obtain $T_{\text{HTTHH}} = \frac{1}{2}(A + E) = 34$.

In summary, the quintet average waiting times in reverse time order are: $T_{5H} = 62$, $T_{HTHTH} = 42$, $T_{HHTHH} = 38$, $T_{HTHHT} = 36$, $T_{HHHTH} = T_{HTTHH} = 34$, $T_{HHHTH} = T_{HHTHT} = 32$. All the results for quartets and quintets agree with those given in [7].

Appendix C. Moment generating functions for triplet sequences

For the sequence HHH, we define the moment generating functions

$$G_A = \langle s^{-s\mathcal{A}} \rangle$$
 $G_B = \langle e^{-s\mathcal{B}} \rangle$ $G_C = \langle e^{-s\mathcal{C}} \rangle$

where A, B, and C denote the time to reach the sequence HHH from a specific sequence realization when starting from H, HH, or *T* respectively. Reading off from Eqs. (3), the backward equations for these moment generating functions are:

$$G_{A} = \frac{1}{2}(G_{B} e^{-s} + G_{C} e^{-s})$$

$$G_{B} = \frac{1}{2}(e^{-2s} + G_{C} e^{-s})$$

$$G_{C} = \frac{1}{2}(G_{A} e^{-s} + G_{C} e^{-s}),$$
(C.1)

with solutions

$$G_{A} = \frac{e^{-s}(2e^{s} - 1)}{8e^{3s} - 4e^{2s} - 2e^{s} - 1}$$

$$G_{B} = \frac{e^{-s}(4e^{2s} - 2e^{2} - 1)}{8e^{3s} - 4e^{2s} - 2e^{s} - 1}$$

$$G_{C} = \frac{e^{-s}}{8e^{3s} - 4e^{2s} - 2e^{s} - 1}.$$
(C.2)

Taylor expanding these generating functions gives

$$G_A = 1 - 13s + \frac{309}{2}s^2 - \frac{11053}{3!}s^3 + \cdots$$

$$G_B = 1 - 9s + \frac{201}{2}s^2 - \frac{7161}{3!}s^3 + \cdots$$

$$G_C = 1 - 15s + \frac{367}{2}s^2 - \frac{13167}{3!}s^3 + \cdots$$

from which $T_{HHH} = \frac{1}{2}(A + C) = 14$, $T_{HHH}^2 = 2 \times \frac{1}{2}(A_2 + C_2) = 338$, and $T_{HHH}^3 = 6 \times \frac{1}{2}(A_3 + C_3) = 12110$, etc.

For the sequence HHT, we read off from Eqs. (4) to give the backward equations for the moment generating functions:

$$G_{A} = \frac{1}{2}(G_{B} e^{-s} + G_{C} e^{-s})$$

$$G_{B} = \frac{1}{2}(e^{-2s} + G_{B} e^{-s})$$

$$G_{C} = \frac{1}{2}(G_{A} e^{-s} + G_{C} e^{-s}),$$
(C.3)

with solutions

$$G_{A} = \frac{e^{-s}}{4 e^{2s} - 2 e^{s} - 1} = 1 - 7s + \frac{71}{2} s^{2} - \frac{1015}{3!} s^{3} + \cdots$$

$$G_{B} = \frac{e^{-s}}{2 e^{s} - 1} = 1 - 3s + \frac{11}{2} s^{2} - \frac{51}{3!} s^{3} + \cdots$$

$$G_{C} = \frac{e^{-s}}{(4 e^{2s} - 2 e^{s} - 1)(2 e^{2} - 1)}$$

$$= 1 - 9s + \frac{105}{2} s^{2} - \frac{1593}{3!} s^{3} + \cdots,$$
(C.4)

from which $T_{HHT} = \frac{1}{2}(A + C) = 8$, $T_{HHT}^2 = 2 \times \frac{1}{2}(A_2 + C_2) = 88$, and $T_{HHT}^3 = 6 \times \frac{1}{2}(A_3 + C_3) = 1304$, etc.

For the sequence HTH, we read off from Eqs. (5) to give the backward equations for the moment generating functions:

$$G_{A} = \frac{1}{2}(G_{A} e^{-s} + G_{B} e^{-s})$$

$$G_{B} = \frac{1}{2}(e^{-2s} + G_{C} e^{-s})$$

$$G_{C} = \frac{1}{2}(G_{A} e^{-s} + G_{C} e^{-s}),$$
(C.5)

with solutions

$$G_{A} = \frac{e^{-s}(2e^{s} - 1)}{8e^{3s} - 4e^{2s} - 2e^{s} - 1}$$

= 1 - 9s + $\frac{137}{2}s^{2} - \frac{3129}{3!}s^{3} + \cdots$
$$G_{B} = \frac{e^{-s}(2e^{s} - 1)^{2}}{8e^{3s} - 4e^{2s} - 2e^{s} - 1}$$

= 1 - 7s + $\frac{103}{2}s^{2} - \frac{2359}{3!}s^{3} + \cdots$
$$G_{C} = \frac{e^{-s}}{8e^{3s} - 4e^{2s} - 2e^{s} - 1}$$

= 1 - 11s + $\frac{179}{2}s^{2} - \frac{4139}{3!}s^{3} + \cdots$, (C.6)

from which $T_{HTH} = \frac{1}{2}(A + C) = 10$, $T_{HTH}^2 = 2 \times \frac{1}{2}(A_2 + C_2) = 158$, and $T_{HTH}^3 = 6 \times \frac{1}{2}(A_3 + C_3) = 3634$, etc.

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