# Method of Gambling Teams and Waiting Times for Patterns

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We flip a fair coin five times. What pattern is "more difficult" to get: HHHHH or HTHTH? If we give this question to a street-wise guy, the most likely answer is: "the first one". Well, we know that answer is not correct. Both patterns have the same probability to occur – 1/32. However, there is a sense in which the street-wise guy is, in fact, correct. If we flip the coin without stopping, then the average waiting time for the first occurrence of the pattern HHHHH is 62, while for the pattern HTHTH it is 42. So, the pattern HHHHH is indeed "more difficult" to get.

Suppose that Melanie flips a coin until she observes either HHHTH or HTHTH while Kyle flips another coin until he observes either HHHHT or HHHTH. Since Kyle got the two patterns with the shortest waiting times, 32 and 34 versus 34 and 42, one would expect him to have a shorter average waiting time when in fact they are exactly the same – 22 for both Melanie and Kyle.

Consider only two patterns: HHHHT and HHHTH. What is the probability that in a stochastic sequence of heads and tails pattern HHHHT will appear earlier than HHHTH? Since the average waiting times (32 and 34) are close to each other one can think that the probability is reasonably close to 1/2. However, the exact answer is 2/3! As we will see this probability is determined by the relationship between patterns rather than by their individual average waiting times.

#### **Problem Statement**

Let Z be an arbitrary discrete random variable with the set of possible values  $\Sigma$ , and let  $\{Z, Z_k\}_{k \geq 1}$  be a sequence of independent, identically distributed random variables.

Consider a collection of finite patterns over  $\Sigma$ :  $\{A_j\}_{1 \leq j \leq K}$ . Assume that no pattern contains another as a subpattern. We will denote by  $\tau_{A_j}$  the waiting time until  $A_j$  occurs as a run in the sequence  $Z_1, Z_2, \ldots$ 

The objective is to find the expected time of

$$\tau = \min\{\tau_{A_1}, ..., \tau_{A_K}\},\tag{1}$$

and probabilities  $\pi_j = \mathbf{P}(\tau = \tau_{A_j})$ .

# Single Pattern

We flip a fair coin and wait for the pattern A = HTH.

What is  $\mathbf{E} au_A$ ?

# **Key Martingale**

The standard martingale technique is as follows (Li (1981)). Assume that a new gambler arrives just before each time n=1,2,... He bets \$1 that

$$Z_n = H$$
.

If he loses, he leaves the game. If he wins, he gets 2 dollars. Then he bets the whole amount, \$2, on the event that

$$Z_{n+1} = T$$
.

Again if he loses, he leaves. If he wins his total capital is now \$4 dollars, and he bets his whole fortune on the next event

$$Z_{n+2} = H$$
.

If the gambler is lucky and finishes the pattern, he leaves the game with his winnings.

Let  $X_n$  be the net amount of money collected by the casino from all the gamblers up until and including time n. Since the amount of the bets at round n depends only on history up to time n-1, and the odds are fair for each gambler,  $X_n$  is a martingale.

# What is the value of $X_{\tau_A}$ ?

We flip a fair coin until the first time  $\tau_A$  when the pattern A = HTH will occur.

By this moment exactly  $\tau_A$  gamblers entered the game, each of them paid a dollar, and almost all of them lost their money.

Only two gamblers won: the one that entered the game at time  $\tau_A-3$ , and another one who started his betting at time  $\tau_A-1$ . At time  $\tau_A$ , the first gambler has got \$8 and the second \$2.

Thus, we get that  $X_{\tau_A} = \tau_A - 8 - 2$ .

# **Heavy Artillery**

By the Optional Stopping Theorem (Williams, 1991, p. 100) we get that

$$0 = \mathbf{E}(X_0) = \mathbf{E}(X_{\tau_A}) = \mathbf{E}(\tau_A) - 10,$$

and, hence,

$$\mathbf{E}(\tau_A)=10.$$

# **Multiple Patterns**

We flip a fair coin again. But now we wait for one of two patterns:  $A_1 = HTH$  and  $A_2 = HH$ .

Let  $\tau = \min\{\tau_{A_1}, \tau_{A_2}\}$ . What is  $\mathbf{E}\tau_A$ ?

# Methods:

- Martingale approach: Li (1980) and Gerber and Li (1981)
- Markov Chain embedding method: Fu (1996), Fu and Chang (2002), Antzoulacos
  (2001) and other
- Recurrent event theory, combinatorics etc: Feller (1968), Guibas and Odlyzko (1981)
  and other

# First Attempt

Assume now that we have 2 teams of betters, and the first team bets on the pattern  $A_1$ , and the second team — on  $A_2$ .

Let  $X_n$  again be the net gain of the casino at time n. It is a martingale. What is  $X_\tau$  now?

$$X_{\tau} = \begin{cases} 2 \times \tau - & (10 + 2), & \text{if } \tau = \tau_{A_1} \\ 2 \times \tau - & (2 + 6), & \text{if } \tau = \tau_{A_2} \end{cases}$$

After taking the expectation we get that

$$0 = E(X_{\tau}) = 2E(\tau) - 12P(\tau = \tau_{A_1}) - 8P(\tau = \tau_{A_2}).$$

Not good.

#### **Free Parameters**

Let  $y_j$  be the initial amount of money with which each of the gamblers from the j-th team start their betting.

Then

$$X_{\tau} = \begin{cases} (y_1 + y_2) \times \tau - & (10y_1 + 2y_2), & \text{if } \tau = \tau_{A_1} \\ (y_1 + y_2) \times \tau - & (2y_1 + 6y_2), & \text{if } \tau = \tau_{A_2} \end{cases}$$

Let us choose  $y_1$  and  $y_2$  in such way that

that is  $y_1 = 1/14$  and  $y_2 = 1/7$ . As consequence, we get

$$0 = E(X_{\tau}) = (y_1 + y_2)E(\tau) - 1,$$

and

$$E(\tau) = \frac{1}{y_1 + y_2} = 4\frac{2}{3}.$$

## **Wise Gamblers**

Let us consider two other choices of the initial bets  $(y_1, y_2)$ : (0,1) and (1,0). The first choice leads to the equation:

$$0 = \mathbf{E}(\tau) - 2\mathbf{P}(\tau = \tau_{A_1}) - 6\mathbf{P}(\tau = \tau_{A_2}).$$

The other one gives

$$0 = E(\tau) - 10P(\tau = \tau_{A_1}) - 2P(\tau = \tau_{A_2}).$$

That allows us to find that

$$P(\tau = \tau_{A_1}) = \frac{1}{3}, \quad P(\tau = \tau_{A_2}) = \frac{2}{3}.$$

## An Example

Suppose that we have three independent sequences of iid random variables:  $\{Z^{(i)},Z_k^{(i)}\}_{k\geq 1}$  with

$$P(Z^{(i)} = A) = p_i, \quad P(Z^{(i)} = B) = q_i, \quad p_i + q_i = 1, \quad i = 1, 2, 3.$$

Let  $\tau$  be the waiting time for the 2-by-2 block:

For instance, if the realization of  $\{Z^{(i)},Z_k^{(i)}\}_{k\geq 1}, i=1,2,3$  produced the following three sequences:

then  $\tau = 4$ .

What is  $E(\tau)$ ?

#### What can be done?

# • IID Sequence

- Generating function initial bets are  $\alpha^n$
- Moments initial bets are  $n^k$  to get moment of order k+1
- Expected number and generating function of occurrence of subpattern P till observing pattern PB (it works in Markov chain case as well)

## Markov Chain

- Two-state chains of first (or higher) order
- General markov chain?
- Non-homogeneous trials?
- "Conditional" situation?
- Multi-dimensional Patterns?

#### Two-state Markov Chain

Now we take  $\{Z_n, n \geq 1\}$  to be a Markov chain with two states S and F, which may model "success" and "failure." We suppose the chain has the initial distribution  $\mathbf{P}(Z_1 = S) = p_S$ ,  $\mathbf{P}(Z_1 = F) = p_F$  and the transition matrix

$$egin{pmatrix} p_{SS} & p_{FS} \ p_{SF} & p_{FF} \end{pmatrix},$$

where  $p_{SF}$  is shorthand for  $P(Z_{n+1} = F | Z_n = S)$ .

What is  $\mathbf{E}[\tau_{FSF}]$ ?

# **Key Martingale – Watch Then Bet**

Now, when gambler number n+1 arrives he observes first the result of the n-th trial,  $Z_n$ .

So, he knows how to bet on the next letter in the fair way.

# **Too Many Ending Scenarios?**

The problem is that now for one pattern FSF this time we need to consider three different ending scenarios:

- 1. FSF occurs at the beginning of the sequence  $\{Z_n, n \geq 1\}$ , or
- 2. the pattern SFSF occurs, or
- 3. the pattern FFSF occurs.

#### Two Teams for One Pattern

- 1. A gambler from the first team who arrives before round n watches the result of the n-th trial, and then bets  $y_1$  dollars on the first letter in the sequence FSF. If he wins he then bets all of his capital on the next letter in the sequence FSF, and he continues in this way until he either loses his capital or he observes all of the letters of FSF. Such players are called  $straightforward\ gamblers$ .
- 2. The gamblers of the second team make use of the information that they observe. If gambler n+1 observes  $Z_n=S$  just before he begins his play, then he bets just like a straightforward gambler except that he begins by wagering  $y_2$  dollars on the first letter of pattern A. On the other hand, if he observes  $Z_n=F$  when he first arrives, then wagers  $y_2$  dollars on the first letter of the pattern SF. He then continues to wager on the successive letters of SF either until he loses or until he observes SF. Such players are called *smart gamblers*.

# **Stopped Martingale**

If we let  $W_{ij}y_j$  denote the amount of money that team  $j \in \{1,2\}$  wins in scenario  $i \in \{1,2,3\}$ , then the values  $W_{ij}$  are easy to compute, and in terms of these values of stopped martingale  $X_{\tau}$  which represents the casino's net gain is given by

$$X_{\tau} = \begin{cases} (y_1 + y_2)(\tau - 1) - y_1 W_{11} - y_2 W_{12}, & \text{1-st scenario,} \\ (y_1 + y_2)(\tau - 1) - y_1 W_{21} - y_2 W_{22}, & \text{2-nd scenario,} \\ (y_1 + y_2)(\tau - 1) - y_1 W_{31} - y_2 W_{32}, & \text{3-rd scenario.} \end{cases}$$

# **Choosing Initial Bets**

Now, if we take  $(y_1^{\ast},y_2^{\ast})$  to be a solution of the system

$$y_1^*W_{21} + y_2^*W_{22} = 1$$
,  $y_1^*W_{31} + y_2^*W_{32} = 1$ ,

we see that with these bet sizes we have a very simple formula for  $X_{\tau}$ :

$$X_{\tau} = \begin{cases} (y_1^* + y_2^*)(\tau - 1) - y_1^* W_{11} - y_2^* W_{12}, & \text{1-st scenario,} \\ (y_1^* + y_2^*)(\tau - 1) - 1, & \text{2-nd scenario,} \\ (y_1^* + y_2^*)(\tau - 1) - 1, & \text{3-rd scenario.} \end{cases}$$

# **Optional Stopping Theorem Routine**

The optional stopping theorem then gives us

$$0 = (y_1^* + y_2^*)(\mathbf{E}[\tau] - 1) - p_1(y_1^*W_{11} + y_2^*W_{12}) - (1 - p_1),$$

where  $p_1$  is the probability of scenario one. We therefore find

$$\mathbf{E}[\tau] = 1 + \frac{p_1(y_1^* W_{11} + y_2^* W_{12}) + (1 - p_1)}{y_1^* + y_2^*}.$$
 (2)

# Done!

$$\mathbf{E}[\tau_{FSF}] = 1 + \frac{p_S}{p_{SF}} + \frac{1}{p_{SF}^2} + \frac{1}{p_{FS}p_{SF}},$$

## From scan to compound pattern

Scan. Assume that we observe a sequence of Bernoulli trials, and the probability of failure is known and relatively small -5%. We have an alert if we observe too many failures during a short period of time. More specifically, we stop the process if we have at least three failures out of 5 sequential trials.

Compound pattern. We have an alert when the following runs occur first time:

1) 3 out of 3

FFF,

2) 3 out of 4

FFSF, FSFF,

(note that the runs SFFF and FFFS were counted earlier)

3) 3 out of 5

FFSSF, FSFSF, FSSFF.

The expected time is 1608.4 and the standard deviation of the waiting time is 1604.8.

# **Approximations**

exponential

$$P(\tau \le n) \approx 1 - \exp(-(n-l)/\mu),$$

where l is the length of the shortest sequence

gamma

$$\mathbf{P}( au \leq n) pprox rac{1}{\Gamma(a)} \int_0^{(n-l)/b} x^a e^{-x} dx,$$

where l is again the length of the shortest sequence,  $b=\sigma^2/\mu$ , and  $a=\mu/b$ .

• shifted exponential

$$P(\tau \le n) \approx 1 - \exp(-(n + 0.5 + \sigma - \mu))/\sigma),$$

where the 0.5 term is a continuity correction.

# **Numerical Results**

		shifted		upper	lower
n	exponential	exponential	gamma	bound	bound
500	0.01600	0.01589	0.01597	0.01588	0.01589
1000	0.03183	0.03173	0.03179	0.03171	0.03174
1500	0.04741	0.04731	0.04736	0.04729	0.04733
2000	0.06274	0.06265	0.06267	0.06262	0.06267
2500	0.07782	0.07773	0.07775	0.07770	0.07776
3000	0.09266	0.09258	0.09258	0.09254	0.09261
4000	0.12162	0.12155	0.12154	0.12150	0.12169
5000	0.14966	0.14960	0.14957	0.14954	0.14965

Table 1. Fixed window scans: at least 3 out of 10, P(F)=.01,  $\mu=30822$ ,  $\sigma=30815$ 

		shifted		upper	lower
n	exponential	exponential	gamma	bound	bound
50	0.09110	0.07827	0.08268	0.07713	0.07940
60	0.10977	0.09770	0.10059	0.09543	0.09989
70	0.12807	0.11672	0.11828	0.11337	0.11991
80	0.14599	0.13534	0.13573	0.13095	0.13949
90	0.16354	0.15357	0.15292	0.14819	0.15864
100	0.18073	0.17141	0.16985	0.16508	0.17736

Table 2. Fixed window scans: at least 4 out of 20, P(F)=.05,  $\mu=481.59$ ,  $\sigma=469.35$ 

# **THANK YOU**