

## TIMID PLAY WHEN LARGE BETS ARE PROFITABLE

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The total variation of a simple, symmetric random walk with absorbing barrier at zero, is stochastically larger than the total variation of any other nonnegative, integer-valued supermartingale with the same initial position. This strengthens a result of David Freedman on the optimality of *timid play* for maximizing the time to bankruptcy in certain gambling situations.

Let  $S = (S_0, S_1, \dots)$  be a sequence of random variables and let  $X_i = S_i - S_{i-1}$  for  $i \geq 1$ . Define the total variation of  $S$  by

$$V(S) = \sum_{i=1}^{\infty} |X_i|.$$

To state the main result of this note, associate with every positive integer  $k$  the collection  $M(k)$  of all nonnegative, integer-valued supermartingales  $S$  which start at  $S_0 = k$ . Let  $T(k)$  be that member of  $M(k)$  which corresponds to a simple, symmetric random walk with absorbing barrier at 0. Recall that a random variable  $X$  is stochastically larger than a random variable  $Y$  if  $P[X \geq t] \geq P[Y \geq t]$  for every real number  $t$ .

**THEOREM 1.** *When  $S$  varies over  $M(k)$ , the total variation  $V(S)$  is stochastically maximized at  $S = T(k)$ .*

To relate Theorem 1 to a result of Freedman [2], let  $S$  be in  $M(k)$  and denote by  $\tau(S)$  the first time, if ever, at which  $S$  reaches 0. The supermartingale  $S$  is said to have *no pauses* if all its increments are almost surely nonzero prior to the absorption time  $\tau(S)$ . For such an  $S$ ,  $\tau(S)$  is known to be almost surely finite.

**COROLLARY (Freedman [2]).** *Let  $S$  be in  $M(k)$ . If  $S$  has no pauses, then  $\tau(T(k))$  is stochastically larger than  $\tau(S)$ .*

**PROOF.** Observe that  $\tau(S) \leq V(S)$  and conclude from Theorem 1 that  $V(T(k))$  is stochastically larger than  $\tau(S)$ . However,  $V(T(k)) = \tau(T(k))$ .  $\square$

In gambling terms, Freedman's result says that for a subfair gambling situation on the nonnegative integers, timid play stochastically maximizes playing time. Theorem 1 strengthens this result by asserting that even though large

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bets yield a bigger immediate reward in terms of variation, timid play is still optimal.

PROOF OF THEOREM 1. It is convenient to introduce, for each  $v \geq 0$ , the following gambling problem.

On the fortune space  $F = \{(k, c) : k = 0, 1, \dots; c \geq 0\}$ , define the utility function  $u$  by

$$u(k, c) = \begin{cases} 0 & c < v \\ 1 & c \geq v, \end{cases}$$

and the set  $\Gamma(k, c)$  of available gambles at  $(k, c)$  in  $F$ , as the set of all distributions of  $(k + X, c + |X|)$ , where  $X$  is any integer-valued random variable with nonpositive mean and support in  $[-k, \infty)$ . Heuristically, a gambler in state  $k$  with cash  $c$ , when choosing the lottery  $X$ , moves to the new state  $k + X$  and his cash becomes  $c + |X|$ . The gambler's goal, as reflected by his utility function  $u$ , is to obtain cash of at least  $v$ . How should he play so as to maximize the probability of reaching his goal? One available strategy is *timid play*, which selects at each stage before absorption at 0, the lottery corresponding to the random variable  $X$  which assumes the values  $\pm 1$  with equal probabilities. The process  $T(k)$  in  $M(k)$  can thus be viewed as the sequence of states of a gambler with initial state  $k$  who plays timidly, while  $V(T(k))$  corresponds to the additional cash accumulated under the timid strategy. Furthermore, every  $S$  in  $M(k)$  can be generated, in a similar way, by a strategy available in this gambling problem, and  $V(S)$  will then correspond to the additional cash resulting from this strategy. Thus Theorem 1 is equivalent to the optimality of timid play for the gambling problem. To prove this optimality, let

$$(1) \quad Q(k, c) = P[c + V(T(k)) \geq v]$$

be the probability of reaching the goal by playing timidly from the initial fortune  $(k, c)$  in  $F$ .

Another strategy available to the gambler at the fortune  $(k, c)$  consists of selecting some  $X$  at the first stage, and then, from the new position  $k + X$ , with the new cash  $c + |X|$ , playing timidly. The probability of attaining the goal (cash  $\geq v$ ) with such a strategy is  $EQ(k + X, c + |X|)$ . If timid play is optimal, then, clearly,

$$(2) \quad EQ(k + X, c + |X|) \leq Q(k, c).$$

Conversely, if (2) holds then, since obviously  $Q \geq u$ , Theorem 2.12.1 of Dubins and Savage [1] applies to show that timid play is indeed optimal. To prove Theorem 1, it thus suffices to establish (2) for all  $(k, c)$  in  $F$  and all integer-valued random variables  $X \geq -k$  with  $EX \leq 0$ .

The proof of (2) proceeds in two stages. First (2) is reduced to two-valued  $X$  and then it is proved for such  $X$ . For the reduction stage, recall that every distribution with finite mean is an average of two-point distributions with the same mean. In fact, it follows easily from Freedman [3, Lemma (108), page 68],

that given a random variable  $X$  with finite mean and support in  $K = \{-k, -k + 1, -k + 2, \dots\}$ , there is a sequence,  $\{X_t; t = 1, 2, \dots\}$ , of two-valued random variables with support in  $K$  and the same mean as  $X$ , and a probability measure  $\mu$  on  $\{1, 2, \dots\}$  such that,

$$(3) \quad Eg(X) = \int Eg(X_t) d\mu(t)$$

for all bounded, real-valued functions  $g$  defined on  $K$ . In particular, (3) holds for all the functions  $g = g_{k,c}$ ,  $(k, c) \in F$ , defined by  $g(x) = Q(k + x, c + |x|)$ ,  $x \in K$ . Thus it suffices to prove (2) for two-valued  $X$ .

Let, therefore,  $X$  be a two-valued random variable with support in  $K$  and  $EX \leq 0$ . Suppose  $X = -a$  with probability  $\alpha$  and  $X = b$  with the complementary probability  $\beta = 1 - \alpha$ , where  $a, b$  are integers,  $0 \leq a \leq k$  and  $0 \leq b$ . Since  $EX \leq 0$ ,  $\alpha = b/(a + b) + \delta$  and  $\beta = a/(a + b) - \delta$  for some  $\delta \geq 0$ . Consequently,

$$(4) \quad \begin{aligned} EQ(k + X, c + |X|) &= \alpha Q(k - a, c + a) + \beta Q(k + b, c + b) \\ &= \frac{b}{a + b} Q(k - a, c + a) + \frac{a}{a + b} Q(k + b, c + b) \\ &\quad - \delta(Q(k + b, c + b) - Q(k - a, c + a)). \end{aligned}$$

For  $k, l \geq 0$ , almost every path of  $T(k + l)$  visits  $k$ . Also, the time to reach  $k$  from  $k + l$  is at least  $l$ . It follows that  $\tau(T(k + l)) \equiv V(T(k + l))$  stochastically majorizes  $\tau(T(k)) + l \equiv V(T(k)) + l$ . In particular,  $V(T(k + b))$  is stochastically larger than  $V(T(k - a)) + (a + b)$ . Hence,  $Q(k + b, c + b) \geq Q(k - a, c + a)$  and so, by (4),

$$(5) \quad \begin{aligned} \frac{b}{a + b} Q(k - a, c + a) + \frac{a}{a + b} Q(k + b, c + b) \\ \geq EQ(k + X, c + |X|). \end{aligned}$$

Let  $\tau$  be the first time for which the process  $T(k)$  reaches either  $k - a$  or  $k + b$  and denote by  $s_\tau$  the state of the process at time  $\tau$ . Let  $E_a$  be the event that  $s_\tau = k - a$  and let  $E_b$  be the event that  $s_\tau = k + b$ . Now calculate

$$(6) \quad \begin{aligned} Q(k, c) &= P[V(T(k)) \geq v - c] \\ &= P[V(T(s_\tau)) \geq v - c - \tau] \\ &= \frac{b}{a + b} P[V(T(s_\tau)) \geq v - c - \tau | E_a] \\ &\quad + \frac{a}{a + b} P[V(T(s_\tau)) \geq v - c - \tau | E_b] \\ &\geq \frac{b}{a + b} P[V(T(k - a)) \geq v - c - a] \\ &\quad + \frac{a}{a + b} P[V(T(k + b)) \geq v - c - b] \\ &= \frac{b}{a + b} Q(k - a, c + a) + \frac{a}{a + b} Q(k + b, c + b). \end{aligned}$$

The first equality is by (1); the second holds because  $V(T(k))$  has the same distribution as  $\tau + V(T(s_\tau))$ ; the third is true because  $s_\tau$  has the same distribution as  $k + Y$ , where  $Y$  is the two-valued random variable with mean zero and values  $-a$  and  $b$ ; the inequality uses the strong Markov property of the process  $T(k)$  together with the facts that  $\tau \geq a$  on  $E_a$  and  $\tau \geq b$  on  $E_b$ ; the final equality is again by (1). Inequality (2) now follows from (6) and (5). This completes the proof of Theorem 1.

REMARK. For nonnegative integers  $k \leq n$ , let  $M_n(k)$  be the collection of all *martingales* which start at  $k$  and have their values in  $\{0, 1, \dots, n\}$ . Theorem 1 and its corollary remain valid if in their statements  $M(k)$  is replaced by  $M_n(k)$  while at the same time  $T(k)$  is modified so as to be absorbed at  $n$  and at 0. In particular, for  $S$  in  $M_n(k)$ ,  $EV(S) \leq EV(T(k))$ . This assertion would be false if supermartingales on the grid  $\{0, 1, \dots, n\}$  were included in  $M_n(k)$ .

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