

# St. Petersburg paradox

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**What is the probability of running out of money in a game with infinite expected return?**

## Question formulation

We consider a repeatable game, where the player pays a fixed fee  $m$  to enter. The pot starts at \$1, and doubles each time a head (H) is thrown. The game stops when the first tail (T) appears, and the player receives the current pot value, i.e.

$$r(T) = \$1, \quad r(HT) = \$2, \quad r(HHT) = \$4, \quad r(HHHT) = \$8, \quad \dots$$

The player begins with a balance of  $s$ , and can only play the game if  $s \geq m$  (i.e. if they have enough money to pay the entry fee). We want to find  $p_m(s)$ , which is the probability that the gambler can play the game indefinitely (or that they do not run out of money).

## The paradox

This game is paradoxical because the expected return,  $\langle r \rangle$ , is infinite, so any rational gambler would play, no matter the cost of the entry fee  $m$ :

$$\begin{aligned} r_n &= 2^n && \text{Return after } n \text{ heads,} \\ p_n &= \left(\frac{1}{2}\right)^n \left(1 - \frac{1}{2}\right) && \text{Probability of } n \text{ heads followed by a tail,} \\ \langle r \rangle &= \left(\sum_{n=0}^{\infty} p_n r_n\right) - m = \left(\frac{1}{2} \sum_{n=0}^{\infty} 1\right) - m = \infty && \text{Expected return.} \end{aligned}$$

## Solution: Markov chain

The repeated game can be formulated as a Markov chain. Let  $t$  denote the number of games played, and  $P_i(t)$  be the probability that we have a balance of  $i$  after  $t$  games ( $i \in \{1, 2, \dots\}$ ,  $t \in \{0, 1, 2, \dots\}$ ). Therefore,  $P_i(0) = \delta_{i,s}$ . We now want to express  $P_i(t+1)$  as a function of  $P_i(t)$ . As the states  $i < m$  are absorbing (we can't play any more if our balance is less than the entry fee  $m$ ), we write down an equation for the absorbing states and an equation for the remainder:

$$P_i(t+1) = \sum_{n: i-(2^n-m) \geq m} p_n P_{i-(2^n-m)}(t) + P_i(t) \quad \text{for } 1 \leq i < m, \quad (1a)$$

$$P_i(t+1) = \sum_{n: i-(2^n-m) \geq m} p_n P_{i-(2^n-m)}(t) \quad \text{for } m \leq i < \infty, \quad (1b)$$

where the limit on the sum accounts for no transitions out of the absorbing states.

This Markov chain can be forward-integrated, starting from  $P_i(0)$ , to find the distribution of balances after  $t$  games. The probability that we can continue playing ( $i > m$ ) after this time is then

$$p_m(s) = \lim_{t \rightarrow \infty} \left[ 1 - \sum_{i=1}^{m-1} P_i(t) \right]. \quad (2)$$

To solve the Markov chain computationally we have to impose an artificial upper bound on the size of the state space, as well as on the number of iterations  $t$ .

## Mathematica code

```
(*Construct the transition matrix for cost m and state-space size nStates*)
m = 2;
nStates = 2^10;
weights = Table[(p^n (1 - p)) /. p -> 0.5, {n, 0, Floor[Log2[nStates]]}];
dim = {nStates, nStates};

mat =
Total[
  Map[
    SparseArray[
      Table[{i, m + i - 2^#} -> weights[[# + 1]], {i, 2^#,
        Min[nStates - m + 2^#, nStates]}], dim] &,
    Range[0, Floor[Log2[nStates]] ]
  ]
];
mat += SparseArray[Table[{i, i} -> 1., {i, 1, m - 1}], dim];

(*Build the initial state vector*)
s = 5;
P = Normal[SparseArray[s -> 1, nStates]];
(*Apply the matrix nIter times to the state vector*)
nIter = 1000;
Pfinal = Nest[Dot[mat, #] &, P, nIter];
(*Calculate the probability in the states i>m*)
1 - Sum[Pfinal[[k]], {k, 1, m - 1}]
```

## Solution: Linear system

An alternative approach to this problem is to consider a linear system for the variables  $p_m(s)$  themselves. i.e.

$$p_m(s) = \sum_{n=0}^{\infty} p_n p_m(s + 2^n - m), \quad (3)$$

with  $p_m(s) = 0$  for  $s < m$ . The solution Eq. (3) is directly related to the stationary solution of Eqs. (1).

Computationally, as above, we have to put an upper bound on the state space