

## 8. Reliability Chains

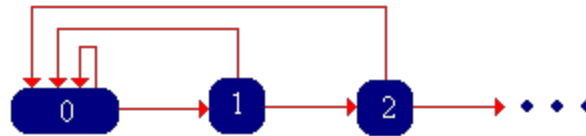
### The Success-Runs Chain

Suppose that we have a sequence of **trials**, each of which results in either **success** or **failure**. Our basic assumption is that if there have been  $x$  consecutive successes, then the probability of success on the next trial is  $p(x)$  where  $p : \mathbb{N} \rightarrow (0, 1)$ . Whenever there is a failure, we start over, independently, with a new sequence of trials. Appropriately enough,  $p$  is called the **success function**. Let  $X_n$  denote the length of the run of successes after  $n$  trials.

1. Argue that  $X = (X_0, X_1, X_2, \dots)$  is a **Markov chain** with state space  $\mathbb{N}$  and transition probability function

$$P(x, x+1) = p(x), \quad P(x, 0) = 1 - p(x), \quad x \in \mathbb{N}$$

This Markov chain is called the **success-runs chain**. The state graph of is given below:



Now let  $T$  denote the trial number of the first failure, starting with a fresh sequence of trials. Note that in the context of the success-runs chain  $X$ ,  $T = T_0$ , the **first return time** to state 0, starting in 0. Note that  $T$  takes values in  $\mathbb{N}_+ \cup \{\infty\}$ , since, presumably, it is possible that no failure occurs. Let  $r(n) = \mathbb{P}(T > n)$  for  $n \in \mathbb{N}$ , the probability of at least  $n$  consecutive successes, starting with a fresh set of trials. Let  $f(n) = \mathbb{P}(T = n)$  for  $n \in \mathbb{N}_+$ , the probability of exactly  $n - 1$ , consecutive successes, starting with a fresh set of trails.

2. Show that

- $p(x) = \frac{r(x+1)}{r(x)}$  for  $x \in \mathbb{N}$
- $r(n) = \prod_{x=0}^{n-1} p(x)$  for  $n \in \mathbb{N}$
- $f(n) = (1 - p(n-1)) \prod_{x=0}^{n-2} p(x)$  for  $n \in \mathbb{N}_+$
- $r(n) = 1 - \sum_{x=1}^n f(x)$  for  $n \in \mathbb{N}$
- $f(n) = r(n-1) - r(n)$  for  $n \in \mathbb{N}_+$

Thus, the functions  $p$ ,  $r$ , and  $f$  give equivalent information. If we know one of the functions, we can construct the other two, and hence any of the functions can be used to define the success-runs chain.

3. The function  $r$  is the **reliability function** associated with  $T$ . Show that it is characterized by the following properties:
- $r$  is positive.
  - $r(0) = 1$
  - $r$  is strictly decreasing.

4. Show that the function  $f$  is characterized by the following properties:
- $f$  is positive.
  - $\sum_{x=1}^{\infty} f(x) \leq 1$

Essentially,  $f$  is the **probability density function** of  $T$ , except that it may be **defective** in the sense that the sum of its values may be less than 1. The leftover probability, of course, is the probability that  $T = \infty$ . This is the critical consideration in the classification of the success-runs chain, which we consider in the next paragraph.

5. Verify that each of the following functions has the appropriate properties, and then find the other two functions:
- $p$  is a constant in  $(0, 1)$ . Thus, the trials are **Bernoulli trials**.
  - $r(n) = \frac{1}{n+1}$  for  $n \in \mathbb{N}$ .
  - $r(n) = \frac{n+1}{2n+1}$  for  $n \in \mathbb{N}$ .
  - $p(x) = \frac{1}{x+2}$  for  $x \in \mathbb{N}$ .



6. The **success-runs applet** is a simulation of the success-runs chain based on Bernoulli trials. Run the simulation 1000 times for various values of  $p$ , and note the limiting behavior of the chain.

### Recurrence and the Remaining Life Chain

7. From the state graph, show that the success-runs chain is **irreducible** and **aperiodic**.

Recall that  $T$  has the same distribution as the first return time to 0 starting at state 0. Thus, the classification of the chain as recurrent or transient depends on  $\alpha = \mathbb{P}(T = \infty)$ . Specifically, the success-runs chain is transient if  $\alpha > 0$  and recurrent if  $\alpha = 0$ . Thus, we see that the chain is recurrent if and only if a failure is sure to occur. We can compute the parameter  $\alpha$  in terms of each of the three functions that define the chain.

8. Show that

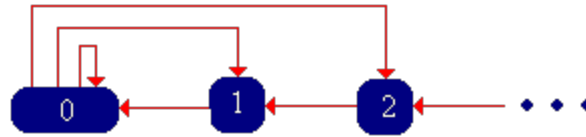
$$\alpha = \prod_{x=0}^{\infty} p(x) = \lim_{n \rightarrow \infty} r(n) = 1 - \sum_{x=1}^{\infty} f(x)$$

When the success-runs chain  $X$  is recurrent, we can define a related random process. Let  $Y_n$  denote the number of trials remaining until the next failure, after  $n$  trials.

9. Argue that  $Y = (Y_0, Y_1, Y_2, \dots)$  is a Markov chain with state space  $\mathbb{N}$  and transition probability function

$$Q(0, x) = f(x+1), \quad Q(x+1, x) = 1, \quad x \in \mathbb{N}$$

10. The Markov chain  $Y$  is called the **remaining life** chain. Verify the state graph below and show that this chain is also irreducible, aperiodic, and recurrent.



11. Compute  $\alpha$  and determine whether the success-runs chain  $X$  is transient or recurrent for each of the cases in [Exercise 5](#).



12. Run the simulation of the **success-runs chain** 1000 times for various values of  $p$ , starting in state 0. Note the return times to state 0.

### Positive Recurrence and Limiting Distributions

Let  $\mu = \mathbb{E}(T)$ , the expected trial number of the first failure, starting with a fresh sequence of trials.

13. Show that the success-runs chain  $X$  is **positive recurrent** if and only if  $\mu < \infty$ , in which case the invariant distribution has probability density function  $g$  given by

$$g(x) = \frac{r(x)}{\mu}, \quad x \in \mathbb{N}$$

14. Show that

- If  $\alpha > 0$  then  $\mu = \infty$
- If  $\alpha = 0$  then  $\mu = \sum_{n=1}^{\infty} n f(n)$
- $\mu = \sum_{n=0}^{\infty} r(n)$

15. Suppose that  $\alpha = 0$ , so that the remaining life chain  $Y$  is well-defined. Show that this chain is also positive recurrent if and only if  $\mu < \infty$ , with the same invariant distribution as  $X$  (with probability density function  $g$  given in the previous exercise).

16. Determine whether the success-runs chain  $X$  is transient, null recurrent, or positive recurrent for each of the cases in [Exercise 5](#). If the chain is positive recurrent, find the invariant probability density function.



17. The success-runs chain corresponding to Bernoulli trials has a geometric distribution as the invariant distribution. Run the simulation of the **success-runs chain** 1000 times for various values of  $p$ . Note the apparent convergence of the empirical distribution to the invariant distribution.

### Time Reversal

Suppose that  $\mu < \infty$ , so that the success-runs chain  $X$  and the remaining-life chain  $Y$  are positive recurrent.

18. Show that  $X$  and  $Y$  are **time reversals** of each other, and use this fact to show again that  $g$  is the invariant probability density function.

19. Run the simulation of the **success-runs chain** 1000 times for various values of  $p$ , starting in state 0. If you imagine watching the simulation backwards in time, then you can see a simulation of the remaining life chain.

---

[Virtual Laboratories](#) > [16. Markov Chains](#) > [1](#) [2](#) [3](#) [4](#) [5](#) [6](#) [7](#) **8** [9](#) [10](#) [11](#) [12](#)

[Contents](#) | [Applets](#) | [Data Sets](#) | [Biographies](#) | [External Resources](#) | [Key words](#) | [Feedback](#) | ©