

**Exercise 1**

A box contains a number of coins. Of these, some (a fraction  $\alpha \leq 1$ ) are *fair*, whereas for the remaining ones the probability of *heads* is  $p \neq 1/2$ .

Assume you pick a coin *at random* from the box, and you flip it 100 times.

- 1) What is the probability that the first flip yields *heads*?
- 2) Let  $X$  be a RV that counts the total number of *heads* in 100 flips. What is its PMF?
- 3) Compute  $E[X]$ .
- 4) Compute  $\alpha$  as a function of  $p$  when  $E[X] = 25$ . Discuss your findings.
- 5) Assume you obtain 50 heads in 100 flips. Compute the probability that your coin is a fair one.

**Exercise 2**

The output interface of a network router serves packets at a constant speed. Packets have a transmission time which is exponentially distributed, with a rate  $\mu$ . Packets arrive at the interface with exponential interarrival times. However, the rate of arrivals depends on the number of packets already in the system, with the following law:

$$\lambda_n = \begin{cases} \frac{N - n}{N \cdot (n + 1)} & n < N \\ 0 & n \geq N \end{cases}$$

- 1) Model the system and draw its CTMC
- 2) Compute the stability condition and the steady-state probabilities [*suggestion: spend a little time tweaking the SSPs, it will pay off later on*]
- 3) Compute the SS probabilities seen by an arriving packet
- 4) Compute the throughput
- 5) Compute the mean number of packets at the interface and the mean response time of the latter.

Some formulas you may find useful:

- $k \binom{N}{k} = N \binom{N-1}{k-1}$
- $\binom{N-j}{j+1} \binom{N}{j} = \binom{N}{j+1}$
- $\sum_{k=0}^N \binom{N}{k} \cdot \frac{1}{x^k} = \left(1 + \frac{1}{x}\right)^N$

**Exercise 1 – solution**

1) According to the law of total probability, we have:

$$P\{\text{heads}\} = P\{\text{heads}|\text{fair}\} \cdot P\{\text{fair}\} + P\{\text{heads}|\text{unfair}\} \cdot P\{\text{unfair}\} = \frac{1}{2} \cdot \alpha + p \cdot (1 - \alpha)$$

2) By the same token,

$$\begin{aligned} p_k &= P\{X = k\} = P\{X = k|\text{fair}\} \cdot P\{\text{fair}\} + P\{X = k|\text{unfair}\} \cdot P\{\text{unfair}\} \\ &= \binom{100}{k} \left(\frac{1}{2}\right)^k \left(\frac{1}{2}\right)^{100-k} \cdot \alpha + \binom{100}{k} p^k \cdot (1-p)^{100-k} \cdot (1-\alpha) \\ &= \binom{100}{k} \left[ \left(\frac{1}{2^{100}}\right) \cdot \alpha + p^k \cdot (1-p)^{100-k} \cdot (1-\alpha) \right] \end{aligned}$$

$$3) E[X] = \sum_{k=0}^{100} k \cdot p_k = \sum_{k=0}^{100} k \cdot [P\{X = k|\text{fair}\} \cdot P\{\text{fair}\} + P\{X = k|\text{unfair}\} \cdot P\{\text{unfair}\}]$$

Therefore,

$$E[X] = \alpha \cdot 100 \cdot \frac{1}{2} + (1 - \alpha) \cdot 100 \cdot p = 100 \left[ \alpha \cdot \left(\frac{1}{2} - p\right) + p \right]$$

4) From the above, we have:

$$100 \left[ \alpha \cdot \left(\frac{1}{2} - p\right) + p \right] = 25$$

i.e.,

$$\alpha = \frac{1 - 4p}{2 - 4p}$$

Since  $\alpha, p \in [0; 1]$ , the above has solutions only when  $p \in [0; \frac{1}{4}]$ , which makes sense (you cannot have  $E[X] = 25$  if *all* your coins have  $p > \frac{1}{4}$ ). When  $p \in [0; \frac{1}{4}]$ ,  $\alpha$  is decreasing. When  $p = 0$  it is  $\alpha = \frac{1}{2}$ , which is intuitively correct, and when  $p = \frac{1}{4}$  it is  $\alpha = 0$ , which is also intuitively correct.

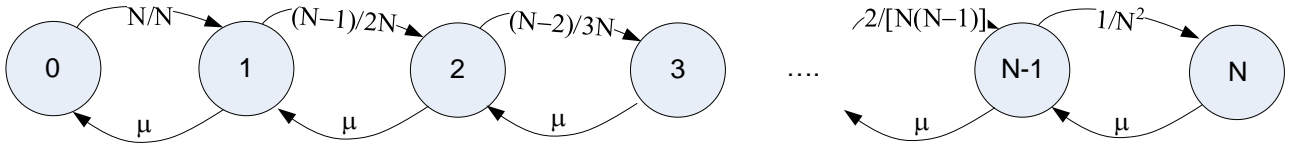
5) By Bayes' theorem:

$$\begin{aligned} P\{\text{fair}|X = 50\} &= \frac{P\{X = 50|\text{fair}\} \cdot P\{\text{fair}\}}{P\{X = 50\}} \\ &= \frac{\binom{100}{50} \left[ \left(\frac{1}{2^{100}}\right) \cdot \alpha \right]}{\binom{100}{50} \left[ \left(\frac{1}{2^{100}}\right) \cdot \alpha + p^{50} \cdot (1-p)^{50} \cdot (1-\alpha) \right]} = \frac{\left(\frac{1}{2^{100}}\right) \cdot \alpha}{\left(\frac{1}{2^{100}}\right) \cdot \alpha + p^{50} \cdot (1-p)^{50} \cdot (1-\alpha)} \\ &= \frac{1}{1 + [4p(1-p)]^{50} \cdot \frac{1-\alpha}{\alpha}} \end{aligned}$$

The above probability is significantly different from 1 only when the second addendum at the denominator is nonnegligible, which happens if  $p \cong \frac{1}{2}$  or  $\alpha \cong 0$ . This matches intuition.

**Exercise 2 – solution**

The CTMC is the following:



The system is stable, since it has a finite number of states.

Since the transitions are nearest-neighbor, we can write the usual equation that connects  $p_k$  to  $p_0$ :

$$p_k = \prod_{j=0}^{k-1} \frac{\lambda_j}{\mu_{j+1}} \cdot p_0 = \frac{1}{\mu^k} \cdot \prod_{j=0}^{k-1} \frac{N-j}{N \cdot (j+1)} \cdot p_0 = \binom{N}{k} \cdot \frac{1}{(N\mu)^k} \cdot p_0$$

Which holds for  $0 \leq k \leq N$ . Therefore, we can write normalization as follows:

$$\sum_{k=0}^N \binom{N}{k} \cdot \frac{1}{(N\mu)^k} \cdot p_0 = 1$$

With the help of one of the suggested formulas, we obtain:

$$p_0 = \frac{1}{\left(1 + \frac{1}{N\mu}\right)^N}$$

Therefore, we have:

$$\begin{aligned} p_k &= \binom{N}{k} \cdot \frac{1}{(N\mu)^k} \cdot \frac{1}{\left(1 + \frac{1}{N\mu}\right)^N} = \binom{N}{k} \cdot \left(\frac{1}{N\mu}\right)^k \cdot \left(\frac{N\mu}{1 + N\mu}\right)^N \\ &= \binom{N}{k} \cdot \left(\frac{1}{N\mu}\right)^k \cdot \left(\frac{N\mu}{1 + N\mu}\right)^{N-k} \cdot \left(\frac{N\mu}{1 + N\mu}\right)^k = \binom{N}{k} \cdot \left(\frac{1}{1 + N\mu}\right)^k \cdot \left(\frac{N\mu}{1 + N\mu}\right)^{N-k} \end{aligned}$$

Calling  $\alpha = \frac{1}{1+N\mu}$ , we have  $p_k = \binom{N}{k} \alpha^k \cdot (1 - \alpha)^{N-k}$ .

This system is non-PASTA. Therefore, we expect  $r_k \neq p_k$ . It is, in fact,  $r_k = \frac{\lambda_k}{\bar{\lambda}} \cdot p_k$ . Computing  $\bar{\lambda}$  is tedious, unless one observes that  $\bar{\lambda} = \gamma = \mu(1 - p_0)$ . Given the above, we have:

$$\begin{aligned} r_k &= \frac{N-k}{N \cdot (k+1)} \cdot \frac{1}{\mu(1-p_0)} \cdot p_k = \frac{N-k}{N \cdot (k+1)} \cdot \frac{1}{\mu(1-(1-\alpha)^N)} \cdot \binom{N}{k} \alpha^k \cdot (1-\alpha)^{N-k} \\ &= \frac{1}{N\mu} \cdot \binom{N}{k+1} \alpha^k \cdot (1-\alpha)^{N-k} \cdot \frac{1}{(1-(1-\alpha)^N)} \end{aligned}$$

Moreover, we have:  $\gamma = \mu(1 - (1 - \alpha)^N)$

The mean number of packets at the interface is the mean of a binomial distribution with a probability of success equal to  $\alpha$ , i.e.

$$E[N] = N \cdot \alpha = \frac{N}{1 + N\mu} = \frac{1 - \alpha}{\mu}$$

By Little's Law, we get that

$$E[R] = \frac{E[N]}{\gamma} = \frac{1 - \alpha}{\mu} \cdot \frac{1}{\mu(1 - (1 - \alpha)^N)} = \frac{1 - \alpha}{\mu^2(1 - (1 - \alpha)^N)}$$