Exercise 1

Consider the following function:

$$F(x) = \begin{cases} 0 & x \le -5\\ \frac{\alpha \cdot x + 5}{\beta + |x|} & x > -5 \end{cases}$$

Where α , β are *positive* constants.

1) Determine under what conditions F(x) is a CDF.

Assume from now on that we are in the above conditions.

- 2) Compute the PDF of RV X, whose CDF is F(x).
- 3) Determine under what further conditions E[X] is finite.
- 4) Assuming $\beta = 5$, compute E[X].

Exercise 2

A service is hosted on a system having K identical servers. A job dispatcher routes each arriving job to an idle server, if there exists one, and *rejects it* otherwise. Assume that the service time and interarrival time are exponentially distributed RVs, whose rates are μ , λ respectively.

- 1) Draw the CTMC (or transition rate diagram).
- 2) Compute the stability condition and an expression for the steady-state probabilities.
- 3) Find under what conditions the PMF of the steady-state probability is:
 - a. strictly increasing , i.e., $p_{j+1} > p_j$, $0 \le j < K$.
 - b. strictly decreasing.
 - c. Neither of the above.

Explain your findings.

- 4) Compute the loss probability
- 5) Compute the mean number of busy servers and, from that, the mean response time. Discuss both expressions when $K \rightarrow \infty$.
- 6) Compute the steady-state probabilities and the performance indexes for K = 1. Discuss the result.

Exercise 1 – Solution

1) In order to be a CDF, the following should happen:

- a) F(x) must be monotonic
- b) $\lim_{x \to -\infty} F(x) = 0$
- c) $\lim_{x\to+\infty} F(x) = 1$

b) always holds. c) holds if and only if $\alpha = 1$.

As far as monotonicity is concerned, we observe that F(x) is identically null for $x \le -5$, and that its derivative is:

$$F'(x) = \begin{cases} \frac{\beta + 5}{(\beta - x)^2} & -5 < x < 0\\ \frac{\beta - 5}{(\beta + x)^2} & x > 0 \end{cases}$$

when x > -5. Therefore, monotonicity is guaranteed if $\beta \ge 5$. The conditions requested by 1) are $\alpha = 1, \beta \ge 5$.

2) As per the computations above, it is:

$$f(x) = \begin{cases} 0 & x < -5\\ \frac{\beta + 5}{(\beta - x)^2} & -5 < x < 0\\ \frac{\beta - 5}{(\beta + x)^2} & x > 0 \end{cases}$$

3) It is:

$$E[X] = \int_{-\infty}^{+\infty} x \cdot f(x) dx = \int_{-5}^{0} x \cdot \frac{\beta + 5}{(\beta - x)^2} dx + \int_{0}^{+\infty} x \cdot \frac{\beta - 5}{(\beta + x)^2} dx = (\beta + 5) \cdot \int_{-5}^{0} \frac{x}{(\beta - x)^2} dx + (\beta - 5) \cdot \int_{0}^{+\infty} \frac{x}{(\beta + x)^2} dx$$

Now, the first integral is always finite (since its limits are), whereas the second may not be. After few algebraic passages, we obtain:

$$\int_{0}^{+\infty} \frac{x}{(\beta+x)^2} dx = \int_{0}^{+\infty} \frac{(\beta+x)-\beta}{(\beta+x)^2} dx = \left[\frac{\beta}{\beta+x} + \log|\beta+x|\right]_{0}^{+\infty} = +\infty$$

Therefore, the expectation exists only if $\beta = 5$.

4) Assuming $\beta = 5$, the expectation is equal to:

$$E[X] = (\beta + 5) \cdot \int_{-5}^{0} \frac{x}{(\beta - x)^2} dx + (\beta - 5) \cdot \int_{0}^{+\infty} \frac{x}{(\beta + x)^2} dx$$
$$= 10 \cdot \int_{-5}^{0} \frac{x}{(5 - x)^2} dx = 10 \cdot \left[\frac{-5}{x - 5} + \log|x - 5|\right]_{-5}^{0}$$
$$= 5 - 10 \log(2)$$

Exercise 2 – Solution

1) The diagram is shown below



2) The system is finite, hence always stable. The steady-state probabilities can be computed by writing the local balance equations: $\lambda \cdot p_j = (j+1) \cdot \mu \cdot p_{(j+1)}, \ 0 \le j \le K-1$. From the above, we quickly obtain:

$$p_j = \frac{1}{j!} \cdot \left(\frac{\lambda}{\mu}\right)^j \cdot p_0, \ 0 \le j \le K$$

Call $u = \lambda/\mu$. By imposing the normalization condition, we get:

$$p_0 = \frac{1}{\sum_{i=0}^{K} \frac{u^i}{i!}}, \ p_j = \frac{\frac{u^j}{j!}}{\sum_{i=0}^{K} \frac{u^i}{i!}}$$

3) $p_{j+1} > p_j \Leftrightarrow \frac{\frac{u^{j+1}}{(j+1)!}}{\sum_{i=0}^{K} \frac{u^i}{i!}} > \frac{\frac{u^j}{(j)!}}{\sum_{i=0}^{K} \frac{u^i}{i!}} \Leftrightarrow u > j+1$. Since the above must hold for every j up to K-1 included,

the required condition is u > K. This can be explained by observing that, under this condition, the arrival rate is such that the serving capacity of the system (which is $K\mu$ under a heavy load) is smaller than the arrival rate.

Using the same reasoning, we find that the PMF is decreasing when u < 1. This means that a single server has enough capacity to serve the arrivals, hence it is increasingly more unlikely that more servers will be busy. When $1 \le u \le K$, the PMF is neither (strictly) increasing nor (strictly) decreasing.

4) The loss probability is
$$p_{K} = \frac{\frac{u^{K}}{K!}}{\sum_{i=0}^{K} \frac{u^{i}}{i!}}$$
.

5) The mean number of busy servers is:

$$E[N] = \sum_{j=1}^{K} j \cdot p_{j} = \frac{\sum_{j=1}^{K} j \frac{u^{j}}{j!}}{\sum_{i=0}^{K} \frac{u^{i}}{i!}} = \frac{u \cdot \sum_{j=1}^{K} \frac{u^{j-1}}{(j-1)!}}{\sum_{i=0}^{K} \frac{u^{i}}{i!}} = u \cdot \frac{\left\lfloor \sum_{j=0}^{K} \frac{u^{j}}{j!} - \frac{u^{K}}{K!} \right\rfloor}{\sum_{i=0}^{K} \frac{u^{i}}{i!}} = u \cdot (1 - p_{K})$$

The mean response time can be obtained through Little's theorem, using the *mean* arrival rate $\overline{\lambda} = \lambda \cdot (1 - p_K)$. Thus, the average response time is $E[R] = E[N]/\overline{\lambda} = \frac{u \cdot (1 - p_K)}{\lambda \cdot (1 - p_K)} = \frac{1}{\mu}$. In fact, there is never

any queueing in this system.

When $K \to \infty$ the system becomes an $M/M/\infty$. For this system, the mean number of busy servers is in fact $u = \lambda/\mu$. This can only happen if $\lim_{K\to\infty} p_k = 0$. Note that this condition is necessary of *any* PMF with an infinite support, hence must hold for this PMF as well. We can check it explicitly:

$$\lim_{K \to \infty} p_K = \lim_{K \to \infty} \frac{\frac{u^K}{K!}}{\sum_{i=0}^K \frac{u^i}{i!}} = e^{-u} \cdot \lim_{K \to \infty} \frac{u^K}{K!} = 0$$

6) When K = 1 we have $p_0 = \frac{1}{\sum_{i=0}^{K} \frac{u^i}{i!}} = \frac{1}{1+u}$, $p_1 = \frac{\frac{u^i}{1!}}{1+u} = \frac{u}{1+u}$. It is $p_K = p_1 = \frac{u}{1+u}$, $E[N] = p_1 = \frac{u}{1+u}$, and

also $E[N] = u \cdot (1 - p_K) = u \cdot \left(1 - \frac{u}{1 + u}\right) = u \cdot \left(\frac{1}{1 + u}\right) = \frac{u}{1 + u}$, $E[R] = \frac{1}{\mu}$. In this case, the system is an M / M / 1 / 1

(or a 1-buffer), and the reader can easily check that the above expressions match those of such a system.