

Appendix C

Evaluation of the Loss Function

The loss function $L(Q)$ is the expected amount a random variable exceeds a fixed value. For example, if the random variable is demand, then $L(Q)$ is the expected amount demand is greater than Q . See Appendix A, Statistics Tutorial, for a more extensive description of the loss function.

This appendix describes how the loss function of a discrete distribution function can be efficiently evaluated. (Appendix A gives one solution method, but it is inefficient.) If you need to evaluate the loss function of a continuous distribution, then convert the continuous distribution into a discrete distribution by “chopping it up” into many pieces. For example, the standard normal table is the discrete (i.e., “chopped up”) version of the continuous standard normal distribution function.

Let N be the number of quantities in the distribution function and let $Q_1, Q_2, Q_3, \dots, Q_N$ be those quantities. For example, take the empirical distribution function in Chapter 11, repeated here for convenience:

| Q | F(Q) | Q | F(Q) | Q | F(Q) |
|-------|--------|-------|--------|-------|--------|
| 800 | 0.0303 | 2,592 | 0.3636 | 3,936 | 0.6970 |
| 1,184 | 0.0606 | 2,624 | 0.3939 | 4,000 | 0.7273 |
| 1,792 | 0.0909 | 2,752 | 0.4242 | 4,064 | 0.7576 |
| 1,792 | 0.1212 | 3,040 | 0.4545 | 4,160 | 0.7879 |
| 1,824 | 0.1515 | 3,104 | 0.4848 | 4,352 | 0.8182 |
| 1,888 | 0.1818 | 3,136 | 0.5152 | 4,544 | 0.8485 |
| 2,048 | 0.2121 | 3,264 | 0.5455 | 4,672 | 0.8788 |
| 2,144 | 0.2424 | 3,456 | 0.5758 | 4,800 | 0.9091 |
| 2,208 | 0.2727 | 3,680 | 0.6061 | 4,928 | 0.9394 |
| 2,304 | 0.3030 | 3,744 | 0.6364 | 4,992 | 0.9697 |
| 2,560 | 0.3333 | 3,808 | 0.6667 | 5,120 | 1.0000 |

$F(Q)$ = Probability demand is less than or equal to the quantity Q

With this distribution function, there are 33 quantities, so $N = 33$ and $Q_1 = 800, Q_2 = 1,184, \dots$, and $Q_{33} = 5,120$. Furthermore, recall that we use μ to represent expected demand, which in this case is $\mu = 3,192$.

We can recursively evaluate the loss function, which means we start with $L(Q_1)$ and then use $L(Q_1)$ to evaluate $L(Q_2)$, and then use $L(Q_2)$ to evaluate $L(Q_3)$, and so forth.

The expected lost sales if we order Q_1 (which in this case is 800 units) are

$$L(Q_1) = \mu - Q_1 = 3,192 - 800 = 2,392$$

Expected lost sales if we order Q_2 are

$$\begin{aligned} L(Q_2) &= L(Q_1) - (Q_2 - Q_1) \times (1 - F(Q_1)) \\ &= 2,392 - (1,184 - 800) \times (1 - 0.0303) \\ &= 2,020 \end{aligned}$$

Expected lost sales if we order Q_3 are

$$\begin{aligned} L(Q_3) &= L(Q_2) - (Q_3 - Q_2) \times (1 - F(Q_2)) \\ &= 2,020 - (1,792 - 1,184) \times (1 - 0.0606) \\ &= 1,448 \end{aligned}$$

In general, the i th expected lost sales are

$$L(Q_i) = L(Q_{i-1}) - (Q_i - Q_{i-1}) \times (1 - F(Q_{i-1}))$$

So you start with $L(Q_1) = \mu - Q_1$ and then you evaluate $L(Q_2)$, and then $L(Q_3)$, up to $L(Q_N)$. The resulting table is

| Q | F(Q) | L(Q) | Q | F(Q) | L(Q) | Q | F(Q) | L(Q) |
|-------|--------|-------|-------|--------|------|-------|--------|------|
| 800 | 0.0303 | 2,392 | 2,592 | 0.3636 | 841 | 3,936 | 0.6970 | 191 |
| 1,184 | 0.0606 | 2,020 | 2,624 | 0.3939 | 821 | 4,000 | 0.7273 | 171 |
| 1,792 | 0.0909 | 1,448 | 2,752 | 0.4242 | 744 | 4,064 | 0.7576 | 154 |
| 1,792 | 0.1212 | 1,448 | 3,040 | 0.4545 | 578 | 4,160 | 0.7879 | 131 |
| 1,824 | 0.1515 | 1,420 | 3,104 | 0.4848 | 543 | 4,352 | 0.8182 | 90 |
| 1,888 | 0.1818 | 1,366 | 3,136 | 0.5152 | 526 | 4,544 | 0.8485 | 55 |
| 2,048 | 0.2121 | 1,235 | 3,264 | 0.5455 | 464 | 4,672 | 0.8788 | 36 |
| 2,144 | 0.2424 | 1,160 | 3,456 | 0.5758 | 377 | 4,800 | 0.9091 | 20 |
| 2,208 | 0.2727 | 1,111 | 3,680 | 0.6061 | 282 | 4,928 | 0.9394 | 8 |
| 2,304 | 0.3030 | 1,041 | 3,744 | 0.6364 | 257 | 4,992 | 0.9697 | 5 |
| 2,560 | 0.3333 | 863 | 3,808 | 0.6667 | 233 | 5,120 | 1.0000 | 1 |

Q = Order quantity

$F(Q)$ = Probability demand is less than or equal to the order quantity

$L(Q)$ = Loss function (the expected amount demand exceeds Q)

With this empirical distribution example, the quantities differ by more than one unit, for example, $Q_2 - Q_1 = 384$. Now suppose the demand forecast is the Poisson distribution with mean 1.25. The distribution function is given in Table A.1 but is repeated here for convenience:

| Q | $f(Q)$ | $F(Q)$ |
|-----|---------|---------|
| 0 | 0.28650 | 0.28650 |
| 1 | 0.35813 | 0.64464 |
| 2 | 0.22383 | 0.86847 |
| 3 | 0.09326 | 0.96173 |
| 4 | 0.02914 | 0.99088 |
| 5 | 0.00729 | 0.99816 |
| 6 | 0.00152 | 0.99968 |
| 7 | 0.00027 | 0.99995 |
| 8 | 0.00004 | 0.99999 |
| 9 | 0.00001 | 1.00000 |

Now we have $Q_1 = 0$, $Q_2 = 1$, and so forth. We find the expected lost sales with the same process: $L(Q_1) = 1.25 - 0 = 1.25$ and

$$\begin{aligned}
 L(Q_2) &= L(Q_1) - (Q_2 - Q_1) \times (1 - F(Q_1)) \\
 &= 0.53650 - (2 - 1) \times (1 - 0.64469) \\
 &= 0.18114
 \end{aligned}$$

Completing the table yields

| Q | $f(Q)$ | $F(Q)$ | $L(Q)$ |
|-----|---------|---------|---------|
| 0 | 0.28650 | 0.28650 | 1.25000 |
| 1 | 0.35813 | 0.64464 | 0.53650 |
| 2 | 0.22383 | 0.86847 | 0.18114 |
| 3 | 0.09326 | 0.96173 | 0.04961 |
| 4 | 0.02914 | 0.99088 | 0.01134 |
| 5 | 0.00729 | 0.99816 | 0.00221 |
| 6 | 0.00152 | 0.99968 | 0.00038 |
| 7 | 0.00027 | 0.99995 | 0.00006 |
| 8 | 0.00004 | 0.99999 | 0.00001 |
| 9 | 0.00001 | 1.00000 | 0.00000 |