# MAT370 Applied Probability Sample Student Project

## The Extreme Value of a Sample from an Exponentially Distributed Population

#### Overview

An extreme value from a sample (a maximum or a minimum) is often of interest. For example, when designing a sewer system or a dam, an engineer would like to make predictions about the magnitudes of future precipitation events based upon available data. A dam, for instance, should be able to withstand future water levels without being absurdly immense.

This project considers the relatively easy case of exponentially distributed random variables. A theoretical model for the distribution of the maximum value from a sample of size n is developed, and checked against data from samples of size 2.

#### Theory

Consider a random variable, X, which is exponentially distributed. That is, one with the following probability density function

$$f(x) = \begin{cases} \alpha e^{-\alpha x} & \text{if } 0 \le x < \infty \\ 0 & \text{otherwise} \end{cases}$$

In this case X has the cumulative distribution function

$$F(x) = \begin{cases} 1 - e^{-\alpha x} & \text{if } 0 \le x < \infty \\ 0 & \text{otherwise} \end{cases}$$

where  $F(x) = Prob[X \le x]$ .

Suppose now that we have a family of *iid* random variables  $X_i$ . That is, the random variables are *independent*, *identically distributed* with common cumulative distribution function F(x). We calculate the cumulative distribution for  $max(X_1, X_2, \ldots, X_n)$ .

Consider first Z = max(X, Y) for independent random variables X and Y with common cumulative distribution function F(x). The cumulative distribution function for Z is defined to be the function

$$F_Z(x) = Prob[Z \le x].$$

But  $Z \leq x$  if and only if  $X \leq x$  and  $Y \leq x$ . That is,

$$Prob[Z \le x] = Prob[X \text{ and } Y \le x] = Prob[X \le x]Prob[Y \le x]$$

since X and Y are independent. Thus, since X and Y have common cumulative distribution function F(x)

 $Prob[Z \le x] = F_Z(x) = (F(x))^2.$ 

And finally, since the probability density function

$$f_Z(x) = F'_Z(x),$$

we have

$$f_Z(x) = 2(F(x))^1 F'(x).$$

For the general case, where  $Z = max(X_1, X_2, \ldots, X_n)$ , we have

$$f_Z(x) = n(F(x))^{(n-1)}F'(x).$$

For example, suppose X and Y have exponential distribution with  $\alpha = 1$ . Then their common density is given by

$$f(x) = \begin{cases} e^{-x} & \text{if } 0 \le x < \infty \\ 0 & \text{otherwise} \end{cases}$$

and their cumulative distribution function

$$F(x) = \begin{cases} 1 - e^{-x} & \text{if } 0 \le x < \infty \\ 0 & \text{otherwise} \end{cases}$$

Then, according to the previous discussion, Z = max(X, Y) will have cumulative distribution

$$Prob[Z \le x] = F_Z(x) = (F(x))^2 = \begin{cases} (1 - e^{-x})^2 & \text{if } 0 \le x < \infty \\ 0 & \text{otherwise} \end{cases}$$

and probability density function  $f_Z(x) = F'_Z(x)$ 

$$f_Z(x) = 2(F(x))^1 F'(x) f(x) = \begin{cases} 2(1 - e^{-x})(e^{-x}) & \text{if } 0 \le x < \infty \\ 0 & \text{otherwise} \end{cases}$$

### Computation

The following MATLAB code produced the accompanying graphs.

```
alpha = 1;
dataX = exprnd(alpha, 100000, 1);
dataY = exprnd(alpha, 100000, 1);
dataZ = max(dataX, dataY);
intervalWidth = 0.2;
subplot(2,2,1)
h = hist(dataZ, 0.1:intervalWidth: 7.1);
bar(0.1:intervalWidth:7.1,h,'y')
title('Histogram of Maximum Random Data')
subplot(2,2,2)
h = h/sum(h);
bar(0.1:intervalWidth:7.1,h,'y')
title('RFH of Maximum Random Data')
subplot(2,2,3)
h = h/intervalWidth;
bar(0.1:intervalWidth:7.1,h,'y')
title('MRFH of Maximum Random Data')
axis([-1\ 7\ -0.1\ 0.8])
subplot(2,2,4)
x = 0:0.001:12;
theoryZ = 2^{*}(1 - \exp(-x)).^{*}(\exp(-x));
bar(0.1:intervalWidth:7.1,h,'y')
title('MRFH: Theory Superimposed')
hold on
plot(x, theoryZ)
```

 $axis([-1\ 7\ -0.1\ 0.8])$ 



