

**457.212 Statistics for Civil & Environmental Engineers**  
**In-Class Material: Class 15**  
**Extreme Value Distributions (A&T 4.2.3)**

Given: The distribution model of a random quantity  $X$ , i.e. PDF or CDF  
 Question: From a sample of size  $n$ , the distribution of the minimum or maximum?  
 → Deriving an ( ) distribution  
 e.g. maximum flood (or drought) in the next 100 years, maximum traffic load on bridge in the next 50 years  
 (More generally, the distribution of the  $k$ -th largest or smallest from a sample → “\_\_\_\_\_ Statistics”)

**1. Deriving “Exact Distributions”**

- Maximum:  $Y_n = \max(X_1, X_2, \dots, X_n)$

$$F_{Y_n}(y) = P(X_1 \leq y \quad X_2 \leq y \quad \dots \quad X_n \leq y)$$

Under the assumption that  $X_1, X_2, \dots, X_n$  are statistically ( ) and ( ) distributed,

$$F_{Y_n}(y) = [ \quad ]^n$$

Its corresponding PDF is,

$$f_{Y_n}(y) = \frac{dF_{Y_n}(y)}{dy} = n [ \quad ]^{n-1} f_X(y)$$

- Minimum:  $Y_1 = \min(X_1, X_2, \dots, X_n)$

$$1 - F_{Y_1}(y) = P(X_1 > y \quad X_2 > y \quad \dots \quad X_n > y) = [ \quad ]^n$$

Therefore, the CDF of  $Y_1$  is

$$F_{Y_1}(y) = 1 - [ \quad ]^n$$

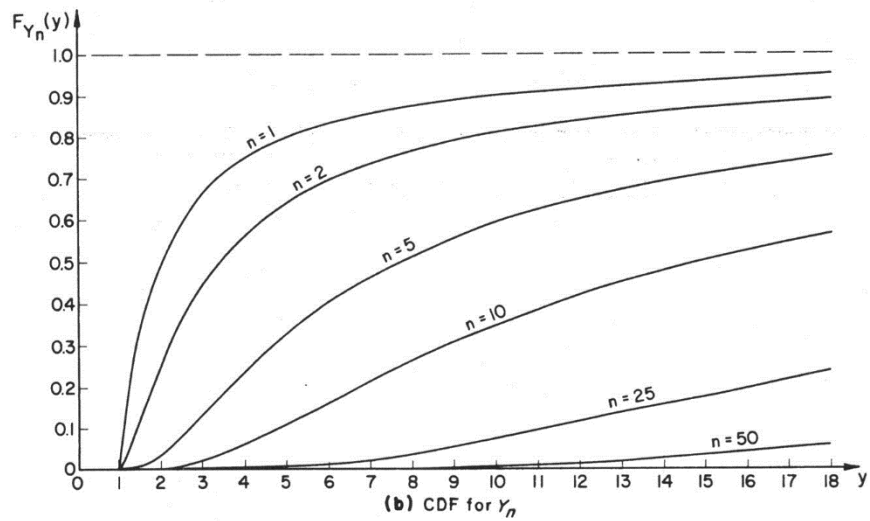
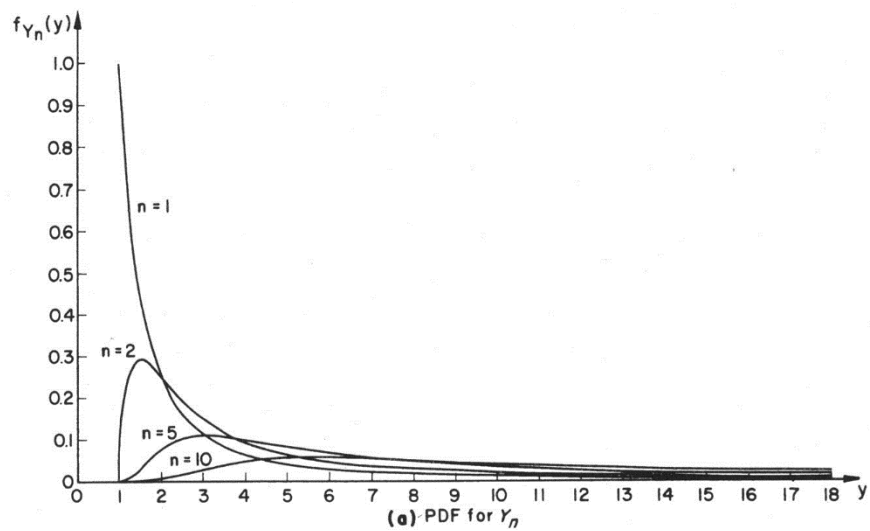
The corresponding pdf is

$$f_{Y_1}(y) = \frac{dF_{Y_1}(y)}{dy} = n [ \quad ]^{n-1} f_X(y)$$

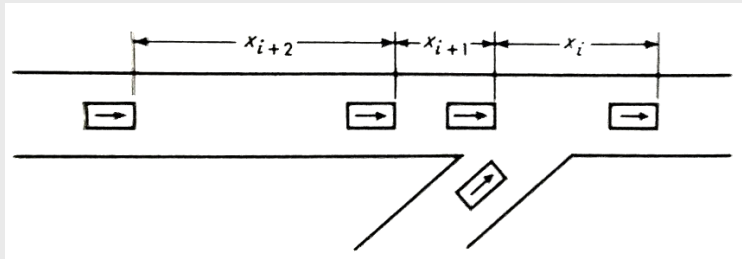
**Example 1:** Suppose the PDF of a random variable  $X$  is given as below.

$$f_X(x) = \frac{1}{x^2}, \quad x \geq 1$$

When someone constructs a sample of size  $n$ , derive the CDF and PDF of the largest in the sample, i.e.  $Y_n = \max(X_1, X_2, \dots, X_n)$ .

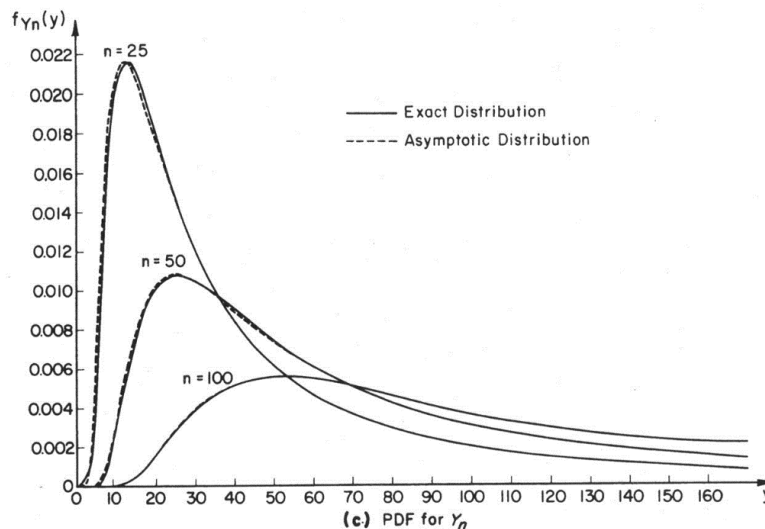


**Example 2:** Suppose a certain class of drivers will merge into freely flowing traffic only if the time between passing cars,  $X_i$  is at least  $y$  sec. If the arrivals of cars are following a Poisson arrival process with the mean occurrence rate  $\nu$ , what is the probability that after  $n$  cars have passed, a driver will not have been able to merge?



## 2. Asymptotic Distributions

An asymptotic distribution can be derived for large samples, i.e.  $n \rightarrow \infty$ , using Cramer's method (1946). For the extreme value distribution in **Example 1**, the exact (i.e. derived) and asymptotic distributions by Cramer's method are compared as follows.



The asymptotic distributions of the extremes tend to converge on certain limiting forms (Gumbel 1958):

- Type I:** The ( ) exponential form,  $\exp[-e^{-A(n)y}]$

Gumbel distribution (largest)

$$F_{Y_n}(y) = \exp[-e^{-\alpha_n(y-u_n)}]$$

$$f_{Y_n}(y) = \alpha_n e^{-\alpha_n(y-u_n)} \exp[-e^{-\alpha_n(y-u_n)}]$$

Note:  $u_n \sim$  location parameter,  $\beta_n = 1/\alpha_n \sim$  scale parameter
- Type II:** The exponential form,  $\exp[-A(n)/y^k]$

Fréchet distribution (largest)

$$F_{Y_n}(y) = \exp\left[-\left(\frac{v_n}{y}\right)^k\right]$$

$$f_{Y_n}(y) = \frac{k}{v_n} \left(\frac{v_n}{y}\right)^{k+1} \exp\left[-\left(\frac{v_n}{y}\right)^k\right]$$
- Type III:** The exponential form with upper/lower bound,  $\exp[-A(n)/(\omega - y)^k]$

Weibull distribution (smallest)

$$F_{Y_1}(y) = 1 - \exp\left[-\left(\frac{y - \varepsilon}{w_1 - \varepsilon}\right)^k\right]$$

The type is determined by the ( ) behavior of the original probability density function.

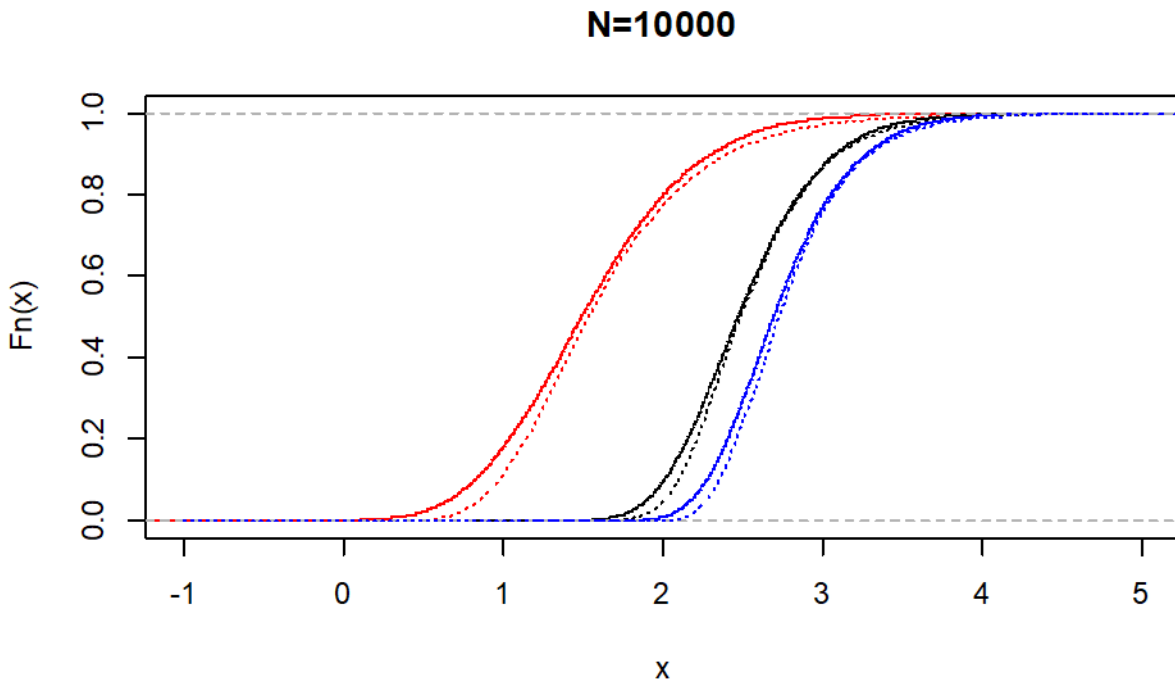
- Exponentially decaying tail (e.g. Normal) → **Type I**
- Polynomial tail (e.g. Example 3) → **Type II**
- Polynomial tail with the limited extreme value → **Type III**

Let us check whether the distribution of the maximum value among  $n = 10, 100$  and  $200$  standard normal random variables converge to the Gumbel distribution:

```
install.packages('VGAM')
library('VGAM')

NormalMaxGen = function(n,N){
  # n: number of samples in each set
  # N: number of sets
  Sample = rnorm(n*N,0,1)
  Sample_matrix = matrix(Sample,n,N)
  SampleMax = apply(Sample_matrix,2,max) # Get maximum of each column
  return(SampleMax)
}
```

```
u = function(n){  
  # location parameter u_n of Gumbel (Example 4.18 in A&T)  
  sqrt(2*log(n))-(log(log(n))+log(4*pi))/2/sqrt(2*log(n))  
}  
  
b = function(n){  
  # scale parameter b_n = 1/alpha_n (See Example 4.18 in A&T)  
  1/sqrt(2*log(n))  
}  
  
x10 = NormalMaxGen(10,10000) # Generate 10 normal random var. 10,000 times  
x100 = NormalMaxGen(100,10000)  
x200 = NormalMaxGen(200,10000)  
  
xv = seq(-1,5,length=100)  
  
plot(ecdf(x10),main="N=10,000",xlim=c(-1,5),col='red')  
lines(xv,pgumbel(xv,u(10),b(10)),col='red',lty='dotted')  
  
lines(ecdf(x100),col='black')  
lines(xv,pgumbel(xv,u(100),b(100)),col='black',lty='dotted')  
  
lines(ecdf(x200),col='blue')  
lines(xv,pgumbel(xv,u(200),b(200)),col='blue',lty='dotted')
```



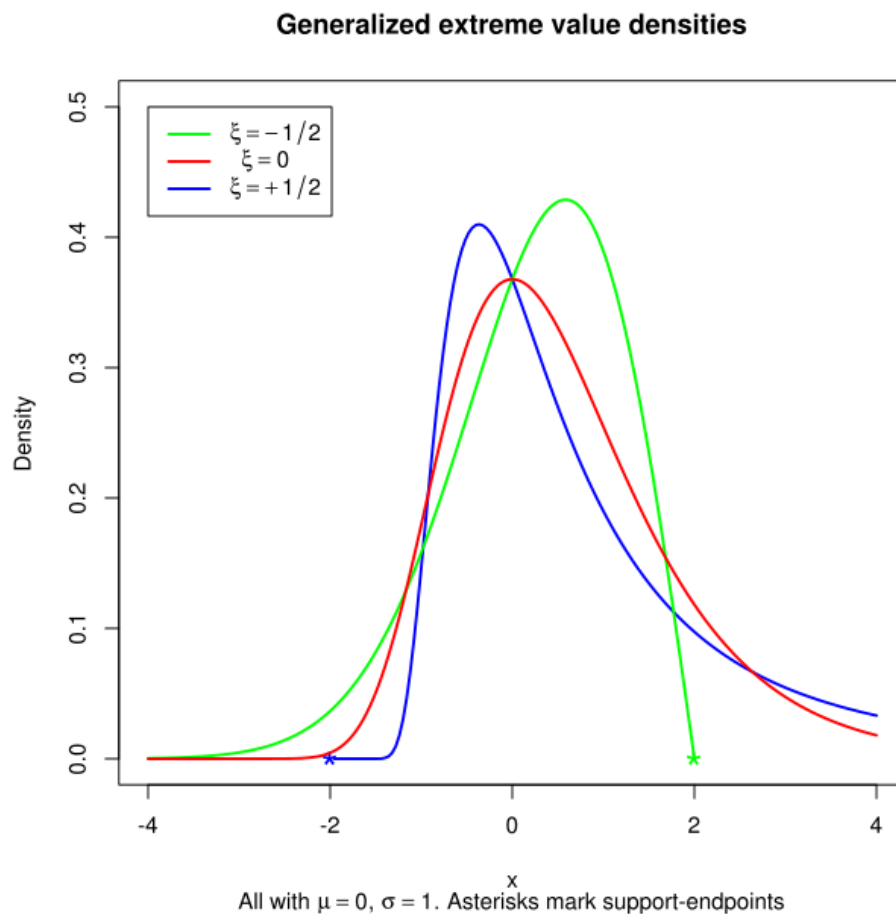
### 3. Generalized Extreme Value (GEV) Distribution

A general extreme value distribution model that can represent Type I (Gumbel), Type II (Fréchet) and Type III (Weibull) distributions.

The CDF of GEV distribution is given as

$$F_Y(y) = \exp \left[ - \left( 1 + \xi \left( \frac{y - u}{\beta} \right) \right)^{-1/\xi} \right]$$

where  $u$ ,  $\beta(> 0)$ , and  $\xi$  are the location, scale and shape parameters. As  $\xi \rightarrow 0$ , GEV becomes Gumbel. For  $\xi > 0$ , GEV is Fréchet distribution. When  $\xi < 0$ , GEV is Weibull distribution. The “VGAM” package of R provides `dgev()`, `pgev()`, `qgev()` and `rgev()` functions. One can estimate the parameters from a data set using the package as well.



\* Image source and more details:

[https://en.wikipedia.org/wiki/Generalized\\_extreme\\_value\\_distribution](https://en.wikipedia.org/wiki/Generalized_extreme_value_distribution)