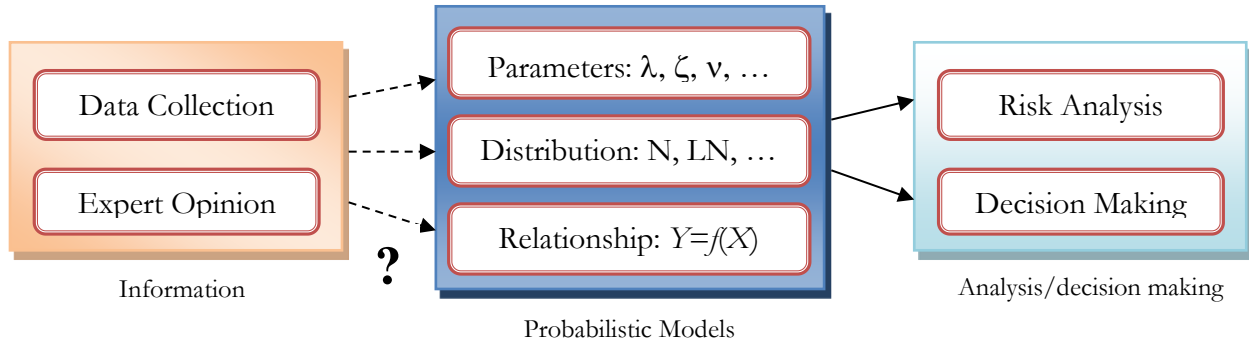


**457.212 Statistics for Civil & Environmental Engineers**  
**In-Class Material: Class 19**  
**Point Parameter Estimation (1) (A&T: 6.1-6.2)**

**1. Statistical Inference**



**2. Point Estimation of Parameters**

(a) Sample statistics  $\hat{\theta}$  : Estimator of true parameter  $\theta$



Example: Sample mean  $\bar{X} = \frac{1}{N} \sum_{i=1}^N x_i$  is an estimator of true mean  $\mu$ , i.e.  $\hat{\mu} = \bar{X}$

Sample standard deviation  $\hat{\sigma} = s$

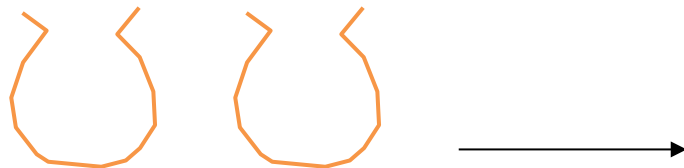
(b) Desirable properties of a point estimator

(i) Unbiased:  $E[\hat{\theta}] = \theta$  ~ Average of point estimates is the same as the true parameter.



\*Recall: “biased” (1/N) and “unbiased” (1/(N-1)) sample standard deviation

(ii) Consistent:  $\lim_{N \rightarrow \infty} \text{Var}[\hat{\theta}] = 0$  ~ As the size of the sample increases, it converges.



(iii) Efficient:  $\text{Var}[\hat{\theta}]$  as small as possible (for the same  $N$ ).

**Example 1:** Is the sample mean  $\bar{X}$  an “unbiased” and “consistent” estimator?

### 3. Point Estimation by “Method of Moments”

**Step 1:** Find the relationship between the true parameter and moments

$$\theta = g(E[X], E[X^2], \dots)$$

**Step 2:** Estimate the moments by

$$\hat{E}[X^m] = \frac{1}{N} \sum_{i=1}^N x_i^m$$

**Step 3:** Substitute the estimated moments into the relationship in Step 1

$$\hat{\theta} = g(\hat{E}[X], \hat{E}[X^2], \dots)$$

(a) Point estimate of mean ( $\mu$ ) by Method of Moments

$$\mu =$$

$$\hat{E}[X] =$$

$$\hat{\mu} =$$

(b) Point estimate of variance by M.M.

$$\sigma^2 =$$

$$\hat{E}[X] =$$

$$\hat{E}[X^2] =$$

$$\hat{\sigma}^2 = \hat{E}[X^2] - \hat{E}[X]^2 = \frac{1}{N} \left( \sum_{i=1}^N x_i^2 - N\bar{X}^2 \right) = s_{\text{biased}}^2$$

\*Note: use “unbiased” one:

(c) Point estimate on other distribution parameters by M.M.

~ Use the relationship between parameters and moments (Table 6.1 A&T; Supp. #6)

(i) Lognormal distribution parameters  $\lambda$  and  $\zeta$

$$\hat{\lambda} = \ln \hat{\mu} - 0.5 \ln \left( 1 + \frac{\hat{\sigma}^2}{\hat{\mu}^2} \right), \quad \hat{\zeta} = \sqrt{\ln \left( 1 + \frac{\hat{\sigma}^2}{\hat{\mu}^2} \right)}$$

(ii) Exponential distribution parameter  $\nu$ :  $\mu = \frac{1}{\nu}$

$$\text{Therefore, } \hat{\nu} = \frac{1}{\hat{\mu}}$$

**Example 2:** Table E6.1 A&T: Concrete crushing strength of 25 test data  $\{x_1, \dots, x_{25}\}$

$$\sum_{i=1}^{25} x_i = 140, \quad \sum_{i=1}^{25} x_i^2 = 794.52$$

(a) Point estimate of mean by M.M.:  $\hat{\mu}$

(b) Point estimate of standard deviation by M.M.:  $\hat{\sigma}$

(c) Point estimates of lognormal distribution parameters by M.M.:  $\hat{\lambda}$  and  $\hat{\xi}$

(d) Point estimates of Gamma distribution parameters by M.M.:  $\hat{\nu}$  and  $\hat{k}$

Hint:  $\mu = k / \nu$  and  $\sigma^2 = k / \nu^2$  for a Gamma distribution.

**Example 3 (Modified A&T Example 6.2):** Data for the fatigue life of 75 S-T Aluminum available. Its sample mean and variances are

$$\bar{X} = 26.75 \text{ million cycles}$$

$$s^2 = 360.0 \text{ million cycles}$$

If the fatigue life follows a Gumbel distribution (See Table 6.1 A&T), what are the point estimates on the distribution parameters  $u$  and  $\alpha$  by M.M.?

$$\text{Hint: } \mu = u + \frac{0.5772}{\alpha}, \quad \sigma^2 = \frac{\pi^2}{6\alpha^2}$$