

# CE 513

July-Nov 2018

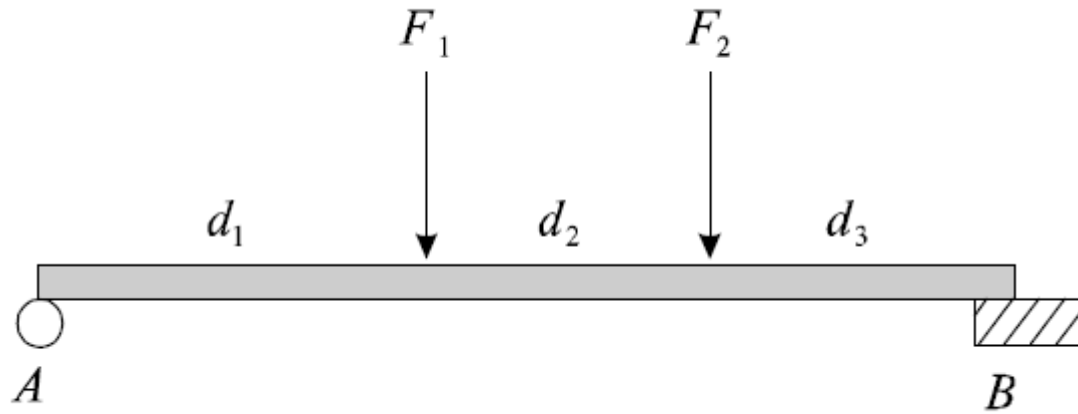
**Dr. Budhadya Hazra**

Room: N-307

Department of Civil Engineering



# Example-1



Given the following loading statistics:

$$\mu_1 = E\{F_1\}$$

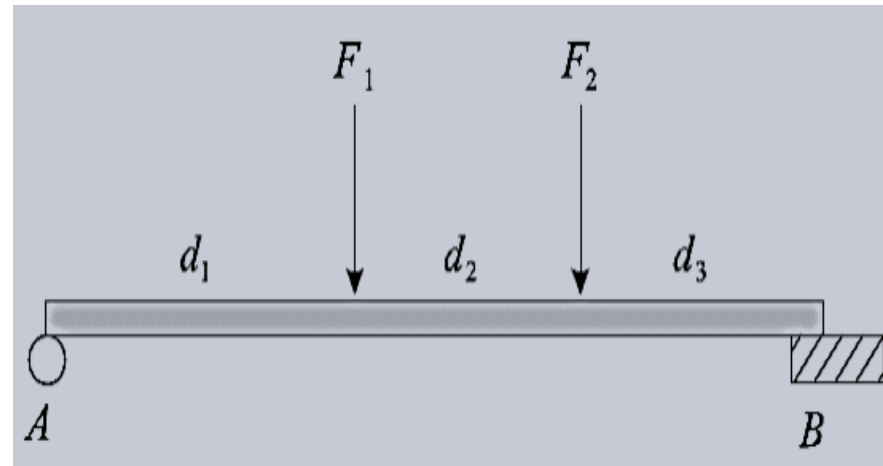
$$\sigma_1 = \sqrt{E\{F_1^2\} - \mu_1^2}$$

$$\mu_2 = E\{F_2\}$$

$$\sigma_2 = \sqrt{E\{F_2^2\} - \mu_2^2}$$

Derive the statistics of the reactions at A and B,  $R_A$  and  $R_B$ , assuming that the forces are statistically independent.

# Example-1



$$R_A = \frac{(F_1 + F_2)d_3 + F_1 d_2}{L} = \frac{F_1(d_3 + d_2) + F_2 d_3}{L}$$

$$R_B = \frac{(F_1 + F_2)d_1 + F_2 d_2}{L} = \frac{F_2(d_1 + d_2) + F_1 d_1}{L}$$

$$\mu_A = \frac{(\mu_1 + \mu_2)d_3 + \mu_1 d_2}{L}$$

$$\mu_B = \frac{(\mu_1 + \mu_2)d_1 + \mu_2 d_2}{L}$$

$$\sigma_A^2 = \frac{1}{L^2} [(d_3 + d_2)^2 \sigma_1^2 + d_3^2 \sigma_2^2]$$

$$\sigma_B^2 = \frac{1}{L^2} [d_1^2 \sigma_1^2 + (d_1 + d_2)^2 \sigma_2^2]$$

$$\begin{aligned}
E\{R_A R_B\} &= E\left\{\frac{[F_1(d_3 + d_2) + F_2 d_3][F_2(d_1 + d_2) + F_1 d_1]}{L^2}\right\} \\
&= \frac{1}{L^2} E\left\{(2d_1 d_3 + d_2 d_3 + d_1 d_2 + d_2^2)F_1 F_2\right. \\
&\quad \left.+ d_1(d_3 + d_2)F_1^2 + d_3(d_1 + d_2)F_2^2\right\} \\
&= \frac{1}{L^2} \left[ (2d_1 d_3 + d_2 d_3 + d_1 d_2 + d_2^2)E\{F_1 F_2\} \right. \\
&\quad \left. + d_1(d_3 + d_2)E\{F_1^2\} + d_3(d_1 + d_2)E\{F_2^2\} \right]
\end{aligned}$$

Since  $F_1$  and  $F_2$  are statistically independent  $E\{F_1 F_2\} = E\{F_1\} E\{F_2\}$

$$\begin{aligned}
E\{F_1^2\} &= \sigma_1^2 + \mu_1^2 \\
E\{F_2^2\} &= \sigma_2^2 + \mu_2^2
\end{aligned}$$

Then,  $E\{R_A R_B\} = a_1 \mu_1 \mu_2 + a_2 (\sigma_1^2 + \mu_1^2) + a_3 (\sigma_2^2 + \mu_2^2)$

The covariance is given by,  $Cov\{R_A, R_B\} = E\{R_A R_B\} - \mu_A \mu_B$

and  $\rho_{R_A R_B} = \frac{Cov\{R_A, R_B\}}{\sigma_A \sigma_B}$



Which can be used for numerical evaluation. Consider the case where  $\sigma_1 = \sigma_2 = \sigma$ ,

$$\sigma_A^2 = \frac{1}{L^2} [(d_3 + d_2)^2 \sigma^2 + d_3^2 \sigma^2] = \frac{\sigma^2 (2d_3^2 + 2d_3 d_2 + d_2^2)}{L^2}$$

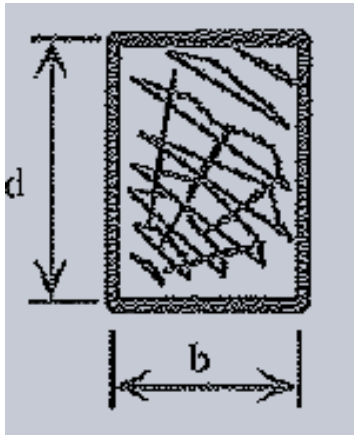
$$\sigma_B^2 = \frac{1}{L^2} [d_1^2 \sigma^2 + (d_1 + d_2)^2 \sigma^2] = \frac{\sigma^2 (2d_1^2 + 2d_1 d_2 + d_2^2)}{L^2}$$

and

$$\rho_{R_A} \rho_{R_B} = \frac{L^2 [a_1 \mu_1 \mu_2 + a_2 (\sigma^2 + \mu_1^2) + a_3 (\sigma^2 + \mu_2^2) - \mu_A \mu_B]}{\sigma^2 \sqrt{(2d_3^2 + 2d_3 d_2 + d_2^2)(2d_1^2 + 2d_1 d_2 + d_2^2)}}$$



# Example-2



Consider the design of a beam *c/s* using bending-stress as a performance criterion. That is:

$$Y = F_b - \frac{6M}{bd^2}$$

$$Y = F(F_b, M, b, d)$$

Goal: Calculate the mean and the variance of *Y*

$$\mu_M = 100,000 \text{ lb} - \text{in}$$

$$\mu_{F_b} = 100,000 \text{ lb} - \text{in}$$

$$\mu_B = 5.6 \text{ in}$$

$$\mu_D = 11.4 \text{ in}$$

$$V_M = 0.12 \Rightarrow \sigma_M = V_M \mu_M = 12,000 \text{ ln} - \text{in}$$

$$V_{F_b} = 0.32 \Rightarrow \sigma_{F_b} = V_{F_b} \mu_{F_b} = 512 \text{ psi}$$

$$V_B = 0.04 \Rightarrow \sigma_B = V_B \mu_B = 0.224 \text{ in}$$

$$V_D = 0.03 \Rightarrow \sigma_D = V_D \mu_D = 0.342 \text{ in}$$

The goal is to calculate the mean and variance of Y

**Solution.** Since Y is a nonlinear function, we must linearize the function about the design point values. We will use the mean values. The linearized form of Y will look like,

$$Y \approx \left[ \mu_{F_b} - \frac{6\mu_M}{\mu_B(\mu_D)^2} \right] + (F_b - \mu_{F_b}) \left. \frac{\partial f}{\partial F_b} \right|_* + (M - \mu_M) \left. \frac{\partial f}{\partial M} \right|_* \\ + (b - \mu_B) \left. \frac{\partial f}{\partial B} \right|_* + (d - \mu_D) \left. \frac{\partial f}{\partial D} \right|_*$$



Where the partial derivatives are evaluated at the design point values (mean values) of the random variables. The partial derivatives are,

$$\begin{aligned}\frac{\partial f}{\partial F_b} = 1 &\Rightarrow \left. \frac{\partial f}{\partial F_b} \right|_* = 1 \\ \frac{\partial f}{\partial M} = -\frac{6}{bd^2} &\Rightarrow \left. \frac{\partial f}{\partial M} \right|_* = -\frac{6}{\mu_B \mu_D^2} \\ \frac{\partial f}{\partial B} = \frac{6M}{b^2 d^2} &\Rightarrow \left. \frac{\partial f}{\partial B} \right|_* = \frac{6\mu_M}{\mu_B^2 \mu_D^2} \\ \frac{\partial f}{\partial D} = \frac{12M}{bd^3} &\Rightarrow \left. \frac{\partial f}{\partial D} \right|_* = \frac{12\mu_M}{\mu_B \mu_D^3}\end{aligned}$$

Substituting these derivatives into the linearized equation, plugging in the mean values of the variables, and rearranging gives the following linearized form of Y:

$$Y = F_b - 0.008244M + 147.2b + 144.6d - 2473$$

$$\sigma_Y^2 = (1)^2 \sigma_{F_b}^2 - (0.008244)^2 \sigma_M^2 + (147.2)^2 \sigma_b^2 + (144.6)^2 \sigma_D^2$$

