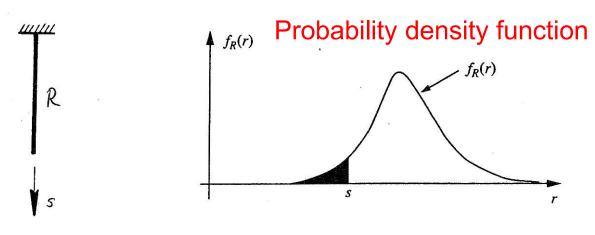
Structural Reliability Analysis – A Simple Case

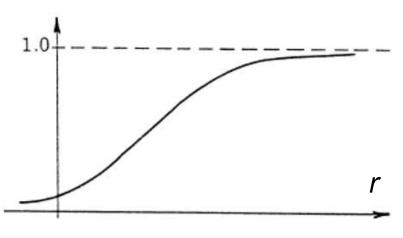
Bar with random strength subjected to a known axial load

Strength Load

- random variable R
- → deterministic value s



F_R(r) Probability
Distribution
Function



$$F_R(r) = P(R \le r)$$

Probability of failure
$$= p_f$$

$$= \text{Prob } (R \le s)$$

$$= \text{Prob } (R - s \le 0)$$

$$= F_R(s)$$

$$= \int_{-\infty}^s f_R(r) \ dr$$

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$$= 1 - p_f$$

Example 1: Involving one random variable – Simple Case

Consider a steel bar which is to be tested in the laboratory in a standard tensile rig. A fixed axial load is applied at one end, whilst the bar is rigidly held at the other end. We wish to estimate the probability that the bar will yield under a deterministic (accurately measured by a load cell) applied load of s kN. In the following, s will take different numerical values in order to illustrate how the probability against yield changes as s changes.

Assume that the yield stress of the material is a normally distributed random variable with a mean value of 420 MPa and a standard deviation of 25 MPa. Further, assume that the cross section of the bar is accurately measured and equal to 144 mm² (i.e. 12mm x 12mm).

Example 1: Involving one random variable – Simple Case

Clearly, $R = \text{Yield Stress x Area} = f_v \times A$ the resistan

the resistance random variable

where $f_y = N (420, 25) \{ in MPa \}$

- a normally distributed random variable

 $A = 144 \text{ mm}^2$

a deterministic quntity, a constant

Thus, $R = N (60.48, 3.60) \{ in kN \}$

i.e. a normally distributed random variable with mean $\mu_R = 60.48$ kN and

standard deviation, $\sigma_R = 3.60 \text{ kN}$.

Reminder: The above is obtained using E[aX] = a E[X] and $Var[aX] = a^2 Var[X]$ where a is constant and X is a random variable (see expectation properties); moreover, a variable which is a linear function of normally distributed variables is also a normal variable.

Example 1: Involving one random variable – Simple Case

In this case the load is deterministic, taking any particular value, say s. The probability of failure can then be written as (see Example 1)

$$p_f = \text{Prob } (R \le s)$$

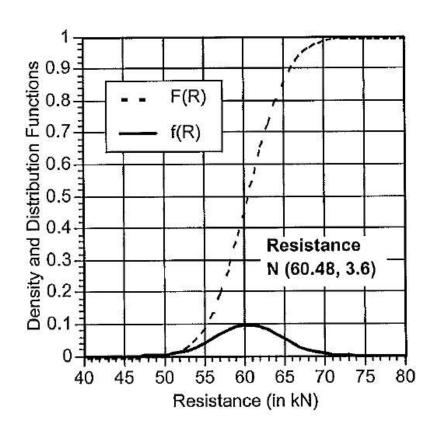
$$= \int_{-\infty}^{s} f_R(r) dr = F_R(s)$$

where $f_R(r)$ is the probability density function of R given by

$$f_R(r) = \frac{1}{\sigma_R \sqrt{2\pi}} \exp\left[-\frac{1}{2} \frac{(r - \mu_R)^2}{\sigma_R^2}\right]$$
 for $-\infty \le r \le \infty$

with $\mu_R = 60.48$ kN and $\sigma_R = 3.60$ kN

and $F_R(s)$ is the distribution function of R evaluated at R = s.



Example 1: Involving one random variable – Simple Case

As in Example 1, for this case, where there is only one random variable, the probability of failure is found by integrating the probability density of R from $-\infty$ to s; or by evaluating the cumulative distribution of R at s.

In this example, R is normal, hence in order to perform this calculation we need to transform R to a standard normal variable, which we will denote with Z_R .

Reminders:

(1) a standard normal variable is N (0,1), i.e. it has zero mean and unit standard deviation;

(2) any normal variable can become a standard normal through a linear transformation;

(3) for estimating probabilities involving a normal variable all we need is the standard normal table

$$R \to Z_R = \frac{R - \mu_R}{\sigma_R}$$

Thus,

Prob
$$[R \le s] = \text{Prob } [Z_R \le \frac{s - \mu_R}{\sigma_R}] = \text{Prob } [Z_R \le \frac{s - 60.48}{3.6}] = \Phi(\frac{s - 60.48}{3.6})$$