

# Risk and Safety in Engineering

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# Contents of Today's Lecture

## Methods of structural reliability theory

- Linear Normal distributed safety margins
- Non-linear Normal distributed safety margins
- General case
- FORM improvements
- Monte-Carlo simulation
- Partial safety factors

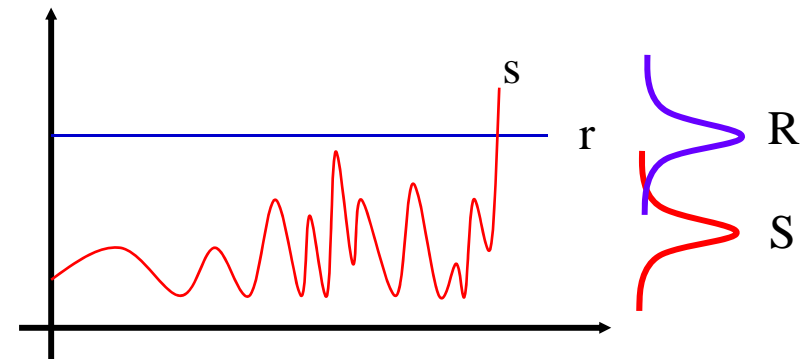
# Structural Reliability Analysis

Reliability of structures cannot be assessed through failure rates because

- Structures are unique in nature
- Structural failures normally take place due to extreme loads exceeding the residual strength

Therefore in structural reliability, models are established for resistances  $R$  and loads  $S$  individually and the structural reliability is assessed through the probability of failure:

$$P_f = P(R - S \leq 0)$$



# Structural Reliability Analysis

If only the resistance is uncertain the failure probability may be assessed by

$$P_f = P(R \leq s) = F_R(s) = P(R/s \leq 1)$$

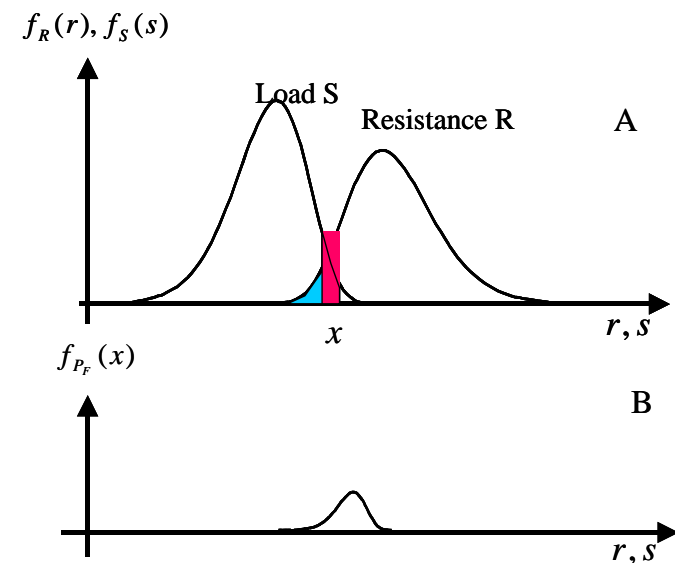
If also the load is uncertain we have

$$P_f = P(R \leq S) = P(R - S \leq 0) = \int_{-\infty}^{\infty} F_R(x) f_S(x) dx$$

where it is assumed that the load and the resistance are independent

This is called the

“Fundamental Case”



# Structural Reliability Analysis

In the case where  $R$  and  $S$  are Normal distributed the safety margin  $M$  is also Normal distributed

$$M = R - S$$

Then the failure probability is

$$P_F = P(R - S \leq 0) = P(M \leq 0)$$

with the mean value of  $M$

$$\mu_M = \mu_R - \mu_S$$

and standard deviation of  $M$

$$\sigma_M = \sqrt{\sigma_R^2 + \sigma_S^2}$$

The failure probability is then

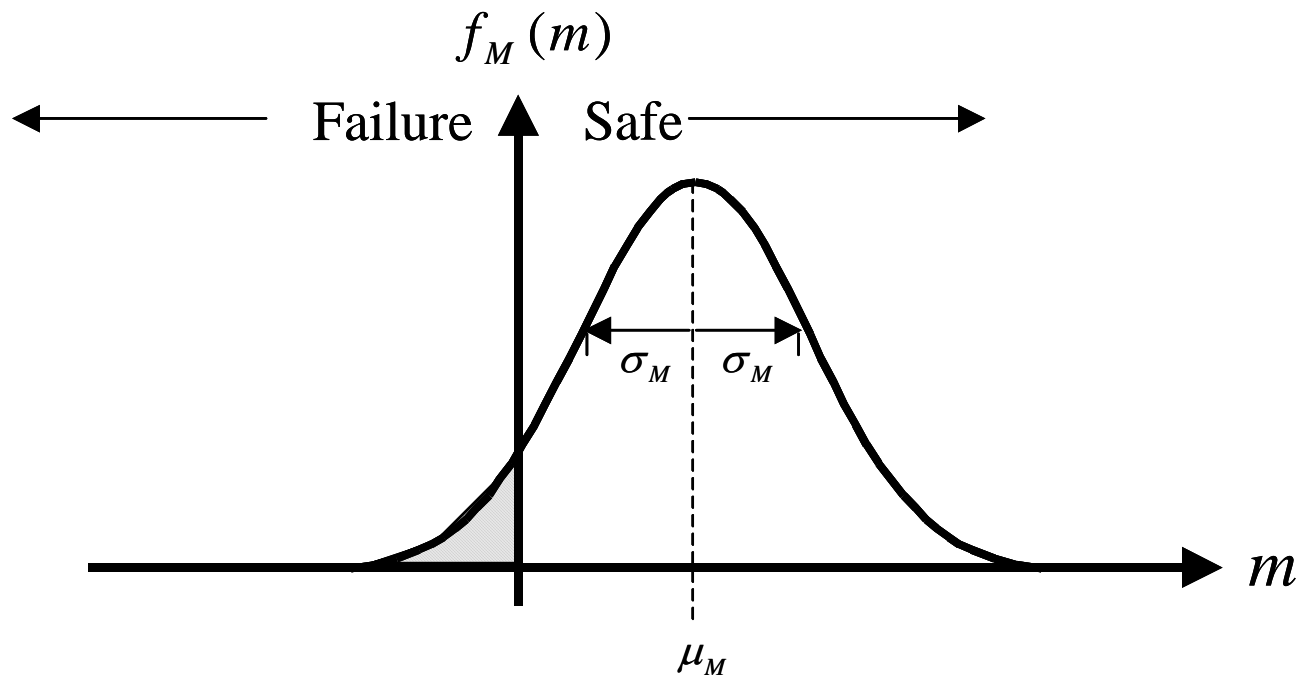
$$P_F = \Phi\left(\frac{0 - \mu_M}{\sigma_M}\right) = \Phi(-\beta)$$

where the reliability index is

$$\beta = \mu_M / \sigma_M$$

# Structural Reliability Analysis

The Normal distributed safety margin  $M$



# Structural Reliability Analysis

In the general case the resistance and the load may be defined in terms of functions

where  $X$  are basic random variables

and the safety margin as

where  $g(\mathbf{x})$  is called the

limit state function

Failure occurs when

$$R = f_1(\mathbf{X})$$

$$S = f_2(\mathbf{X})$$

$$M = R - S = f_1(\mathbf{X}) - f_2(\mathbf{X}) = g(\mathbf{X})$$

$$g(\mathbf{x}) \leq 0$$

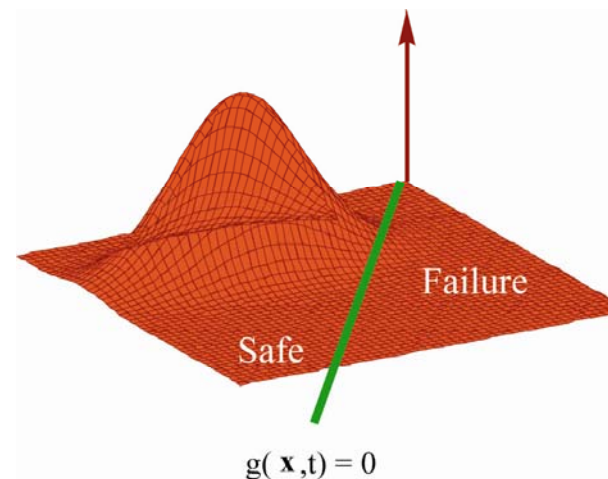
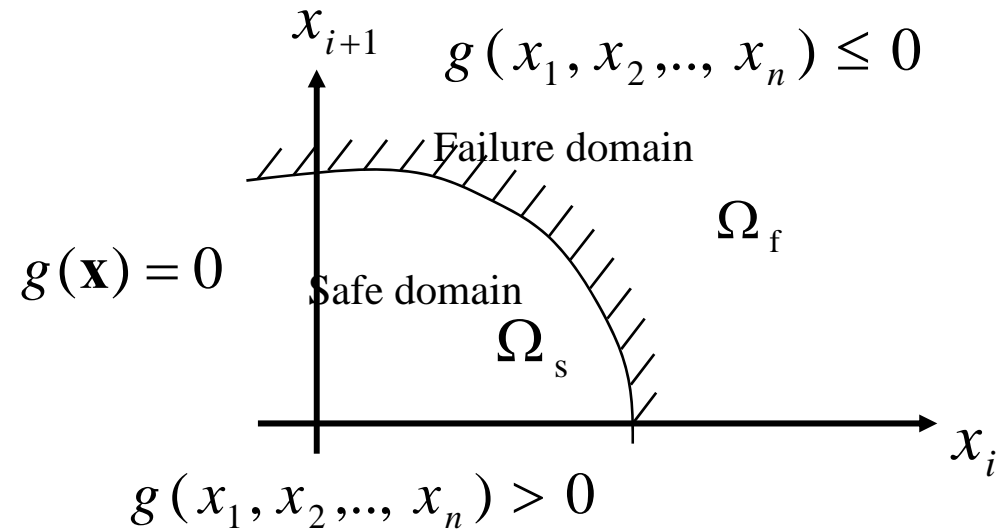
# Structural Reliability Analysis

Setting  $g(\mathbf{x}) = 0$  defines a (n-1) dimensional surface in the space spanned by the  $n$  basic variables  $X$

This is the failure surface separating the sample space of  $X$  into a safe domain and a failure domain

The failure probability may in general terms be written as

$$P_f = \int_{g(\mathbf{x}) \leq 0} f_{\mathbf{X}}(\mathbf{x}) d\mathbf{x}$$



**Failure event**  
 $\mathbf{F} = \{g(\mathbf{x}) \leq 0\}$



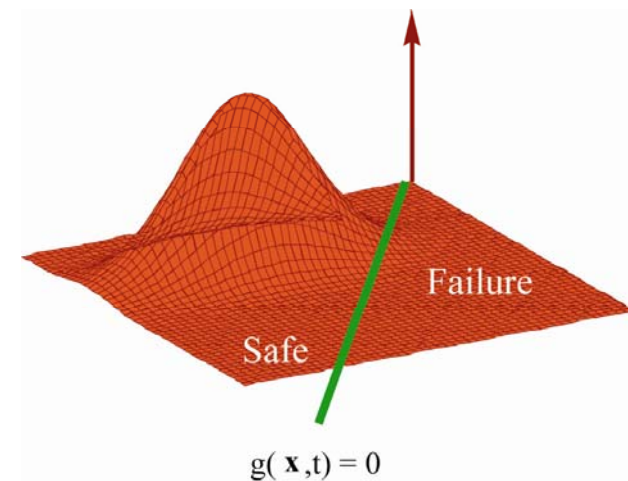
# Basics of Structural Reliability Methods

The probability of failure can be assessed by

where  $f_{\mathbf{x}}(\mathbf{x})$  is the joint probability density function for the basic random variables  $\mathbf{X}$

For the 2-dimensional case the failure probability simply corresponds to the integral under the joint probability density function in the area of failure

$$P_f = \int_{\Omega_f = \{g(\mathbf{x}) \leq 0\}} f_{\mathbf{x}}(\mathbf{x}) d\mathbf{x}$$



# Basics of Structural Reliability Methods

The probability of failure can be calculated using

- numerical integration  
(Simpson, Gauss, Tchebyshev,  
etc.)

but for problems involving dimensions  
higher than say 6 the numerical  
integration becomes cumbersome

$$P_f = \int_{\Omega_f = \{g(\mathbf{x}) \leq 0\}} f_{\mathbf{x}}(\mathbf{x}) d\mathbf{x}$$

**Other methods are necessary !**

# Basics of Structural Reliability Methods

When the limit state function is linear

$$g(\mathbf{x}) = a_0 + \sum_{i=1}^n a_i \cdot x_i$$

the safety margin  $M$  is defined through

$$M = a_0 + \sum_{i=1}^n a_i \cdot X_i$$

with

mean value

$$\mu_M = a_0 + \sum_{i=1}^n a_i \mu_{X_i}$$

and

variance

$$\sigma_M^2 = \sum_{i=1}^n a_i^2 \sigma_{X_i}^2 + \sum_{i=1}^n \sum_{j=1, j \neq i}^n \rho_{ij} a_i a_j \sigma_i \sigma_j$$

# Basics of Structural Reliability Methods

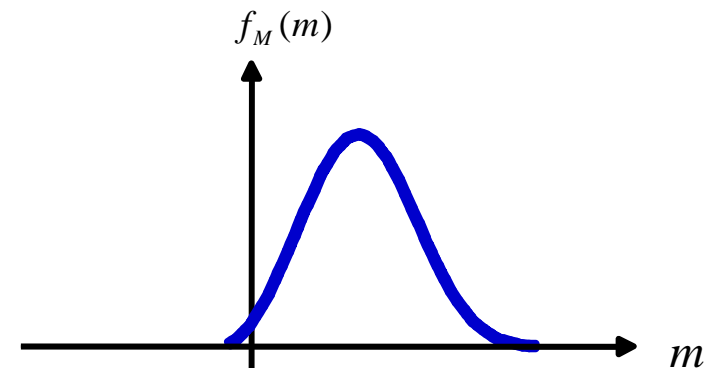
The failure probability can then be written as

$$P_F = P(g(\mathbf{X}) \leq 0) = P(M \leq 0)$$

The reliability index is defined as

$$\beta = \frac{\mu_M}{\sigma_M} \quad (\text{Basler and Cornell})$$

Provided that the safety margin is Normal distributed  
the failure probability is determined as



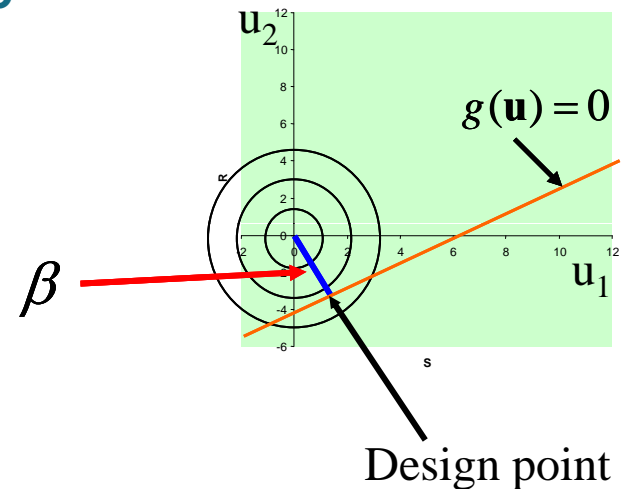
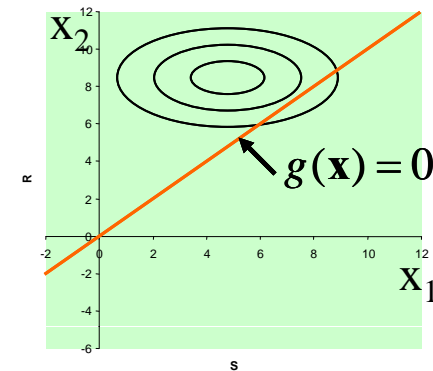
$$P_F = \Phi(-\beta)$$

# Basics of Structural Reliability Methods

The reliability index  $\beta$  has the geometrical interpretation of being the shortest distance between the failure surface and the origin in standard Normal distributed space  $U$

$$U_i = \frac{X_i - \mu_{X_i}}{\sigma_{X_i}}$$

in which case the components of  $U$  have zero means and variances equal to 1



# Basics of Structural Reliability Methods

## Example:

Consider a steel rod with resistance  $r$  subjected to a tension force  $s$

$$g(\mathbf{X}) = R - S$$

$r$  and  $s$  are modeled by the random variables  $R$  and  $S$

$$\begin{aligned}\mu_R &= 350, \sigma_R = 35 \\ \mu_S &= 200, \sigma_S = 40\end{aligned}$$

The probability of failure is required

$$P(R - S \leq 0)$$

# Basics of Structural Reliability Methods

## Example:

Consider a steel rod with resistance  $r$  subjected to a tension force  $s$

$r$  and  $s$  are modeled by the random variables  $R$  and  $S$

The probability of failure is wanted

The safety margin is

The reliability index is then

and the probability of failure

$$g(\mathbf{X}) = R - S$$

$$\begin{aligned}\mu_R &= 350, \sigma_R = 35 \\ \mu_S &= 200, \sigma_S = 40\end{aligned}$$

$$P(R - S \leq 0)$$

$$M = R - S \begin{cases} \mu_M = 350 - 200 = 150 \\ \sigma_M = \sqrt{35^2 + 40^2} = 53.15 \end{cases}$$

$$\beta = \frac{150}{53.15} = 2.84$$

$$P_F = \Phi(-2.84) = 2.4 \cdot 10^{-3}$$

# Basics of Structural Reliability Methods

Usually the limit state function is non-linear

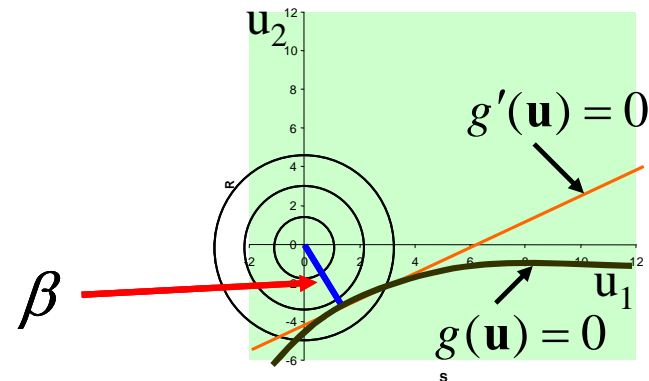
- this small phenomenon caused the so-called invariance problem

Hasofer & Lind suggested to linearize the limit state function in the design point

- this solved the invariance problem

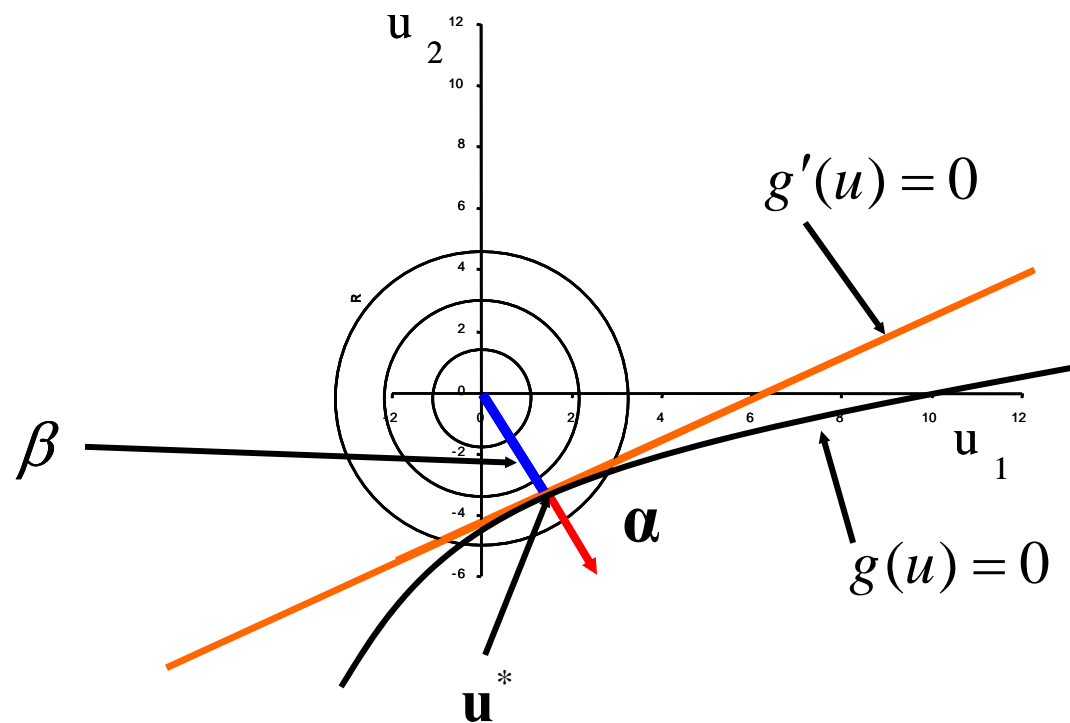
The reliability index may then be determined by the following optimization problem

Can however easily be linearized !



$$\beta = \min_{\mathbf{u} \in \{g(\mathbf{u})=0\}} \sqrt{\sum_{i=1}^n u_i^2}$$





# Basics of Structural Reliability Methods

The optimization problem can be formulated as an iteration problem

1) the design point is determined as

$$\mathbf{u}^* = \beta \cdot \boldsymbol{\alpha}$$

2) the normal vector to the limit state function is determined as

$$\alpha_i = \frac{-\frac{\partial g}{\partial u_i}(\beta \cdot \boldsymbol{\alpha})}{\left[ \sum_{j=1}^n \frac{\partial g}{\partial u_j}(\beta \cdot \boldsymbol{\alpha})^2 \right]^{1/2}}, \quad i = 1, 2, \dots, n$$

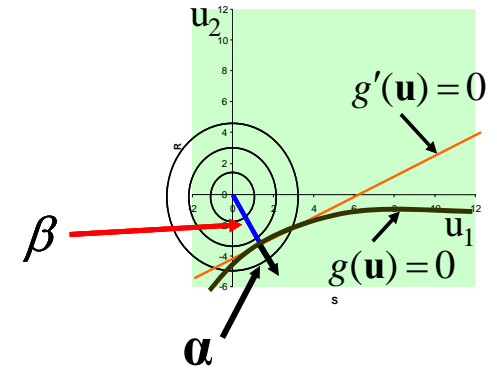
3) the safety index is determined as

$$g(\beta \cdot \alpha_1, \beta \cdot \alpha_2, \dots, \beta \cdot \alpha_n) = 0$$

4) a new design point is determined as

$$\mathbf{u}^* = (\beta \cdot \alpha_1, \beta \cdot \alpha_2, \dots, \beta \cdot \alpha_n)^T$$

5) the above steps are continued until convergence in  $\beta$  is attained



# Basics of Structural Reliability Methods

Example :

Consider the steel rod with cross-sectional area  $a$  and yield stress  $r$

$$h = r \cdot a$$

The rod is loaded with the tension force  $s$

The limit state function can then be written as

$$g(\mathbf{x}) = r \cdot a - s$$

$r$ ,  $a$  and  $s$  are uncertain and modeled by normal distributed random variables

$$\mu_R = 350, \sigma_R = 35 \quad \mu_S = 1500, \sigma_S = 300$$
$$\mu_A = 10, \sigma_A = 1$$

we would like to calculate the probability of failure

# Basics of Structural Reliability Methods

The first step is to transform the basic random variables into standardized Normal distributed space

$$U_R = \frac{R - \mu_R}{\sigma_R}$$

$$U_A = \frac{A - \mu_A}{\sigma_A}$$

$$U_S = \frac{S - \mu_S}{\sigma_S}$$

Then we write the limit state function in terms of the realizations of the standardized Normal distributed random variables

$$\begin{aligned} g(u) &= (u_R \sigma_R + \mu_R)(u_A \sigma_A + \mu_A) - (u_S \sigma_S + \mu_S) \\ &= (35u_R + 350)(u_A + 10) - (300u_S + 1500) \\ &= 350u_R + 350u_A - 300u_S + 35u_R u_A + 2000 \end{aligned}$$

## Basics of Structural Reliability Methods

The reliability index is calculated as

$$\beta = \frac{-2000}{350\alpha_R + 350\alpha_A - 300\alpha_S + 35\beta\alpha_R\alpha_A}$$

the components of the  $\alpha$ -vector are then calculated as

$$\left\{ \begin{array}{l} \alpha_R = -\frac{1}{k}(350 + 35\beta\alpha_A) \\ \alpha_A = -\frac{1}{k}(350 + 35\beta\alpha_R) \\ \alpha_S = \frac{300}{k} \end{array} \right.$$

where

$$k = \sqrt{\alpha_R^2 + \alpha_A^2 + \alpha_S^2}$$

# Basics of Structural Reliability Methods

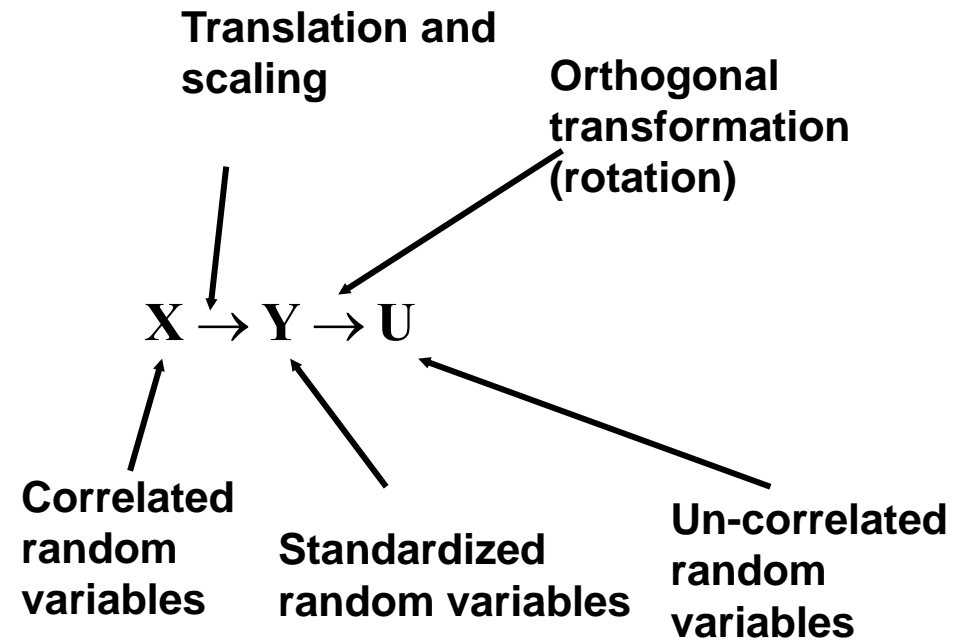
following the iteration scheme  
we get the following iteration  
history

Iteration	Start	1	2	3	4	5
$\beta$	3.0000	3.6719	3.7399	3.7444	3.7448	3.7448
$\alpha_R$	-0.5800	-0.5701	-0.5612	-0.5611	-0.5610	-0.5610
$\alpha_A$	-0.5800	-0.5701	-0.5612	-0.5611	-0.5610	-0.5610
$\alpha_S$	0.5800	0.5916	0.6084	0.6086	0.6087	0.6087

# Basics of Structural Reliability Methods

The procedure can be extended to consider

Correlated random variables



# Basics of Structural Reliability Methods

## Correlated random variables

The covariance matrix for the random variables is given as

$$\mathbf{C}_X = \begin{bmatrix} \text{Var}[X_1] & \text{Cov}[X_1, X_2] \dots & \text{Cov}[X_1, X_n] \\ \vdots & \vdots & \vdots \\ \text{Cov}[X_n, X_1] & \dots & \text{Var}[X_n] \end{bmatrix}$$

and the correlation coefficient matrix is

$$\boldsymbol{\rho}_X = \begin{bmatrix} 1 & \dots & \rho_{1n} \\ \vdots & 1 & \vdots \\ \rho_{n1} & \dots & 1 \end{bmatrix}$$

The first step is the standardization

$$Y_i = \frac{X_i - \mu_{X_i}}{\sigma_{X_i}}, i = 1, 2, \dots, n$$



# Basics of Structural Reliability Methods

## Correlated random variables

The transformation of the correlated random variables into non-correlated random variables can be written as

$$\mathbf{Y} = \mathbf{T}\mathbf{U}$$

where  $\mathbf{T}$  is a lower triangular matrix

then we can write

$$\mathbf{C}_Y = E[\mathbf{Y} \cdot \mathbf{Y}^T] = E[\mathbf{T} \cdot \mathbf{U} \cdot \mathbf{U}^T \cdot \mathbf{T}^T] = \mathbf{T} \cdot E[\mathbf{U} \cdot \mathbf{U}^T] \cdot \mathbf{T}^T = \mathbf{T} \times \mathbf{T}^T = \boldsymbol{\rho}_X$$

with  $T$  standing for transpose matrix

# Basics of Structural Reliability Methods

## Correlated random variables

In the case of 3 random variables we have

$$\mathbf{T} \cdot \mathbf{T}^T = \boldsymbol{\rho}_X = \begin{bmatrix} \rho_{11} & \rho_{12} & \rho_{13} \\ & \rho_{22} & \rho_{23} \\ \text{sym.} & & \rho_{33} \end{bmatrix}$$

As  $\mathbf{T}$  is a lower triangular matrix we have

$$\mathbf{T} \cdot \mathbf{T}^T = \begin{bmatrix} T_{11} & 0 & 0 \\ T_{21} & T_{22} & 0 \\ T_{31} & T_{32} & T_{33} \end{bmatrix} \begin{bmatrix} T_{11} & T_{12} & T_{13} \\ 0 & T_{22} & T_{23} \\ 0 & 0 & T_{33} \end{bmatrix} = \begin{bmatrix} \rho_{11} & \rho_{12} & \rho_{13} \\ & \rho_{22} & \rho_{23} \\ \text{sym.} & & \rho_{33} \end{bmatrix}$$

$$T_{11} = \sqrt{1}$$

$$T_{21} = \rho_{12}$$

$$T_{31} = \rho_{13}$$

$$T_{22} = \sqrt{1 - T_{21}^2}$$

$$T_{32} = \frac{\rho_{23} - T_{31} \cdot T_{21}}{T_{22}}$$

$$T_{33} = \sqrt{1 - T_{31}^2 - T_{32}^2}$$

⋮

# Basics of Structural Reliability Methods

## The normal-tail approximation

$$F_{X_i}(x_i^*) = \Phi\left(\frac{x_i^* - \mu'_{X_i}}{\sigma'_{X_i}}\right)$$

$$f_{X_i}(x_i^*) = \frac{1}{\sigma_{X_i}} \varphi\left(\frac{x_i^* - \mu'_{X_i}}{\sigma'_{X_i}}\right)$$

$$\sigma'_{X_i} = \frac{\varphi(\Phi^{-1}(F_{X_i}(x_i^*)))}{f_{X_i}(x_i^*)}$$

$$\mu'_{X_i} = x_i^* - \Phi^{-1}(F_{X_i}(x_i^*))\sigma'_{X_i}$$

# Basics of Structural Reliability Methods

## Non-normal distributed random variables

$$F_X(x) = F_{X_n}(x_n | x_1, x_2, \dots, x_{n-1}) \cdot F_{X_{n-1}}(x_{n-1} | x_1, x_2, \dots, x_{n-2}) \dots F_{X_1}(x_1)$$

## Rosenblatt Transformation

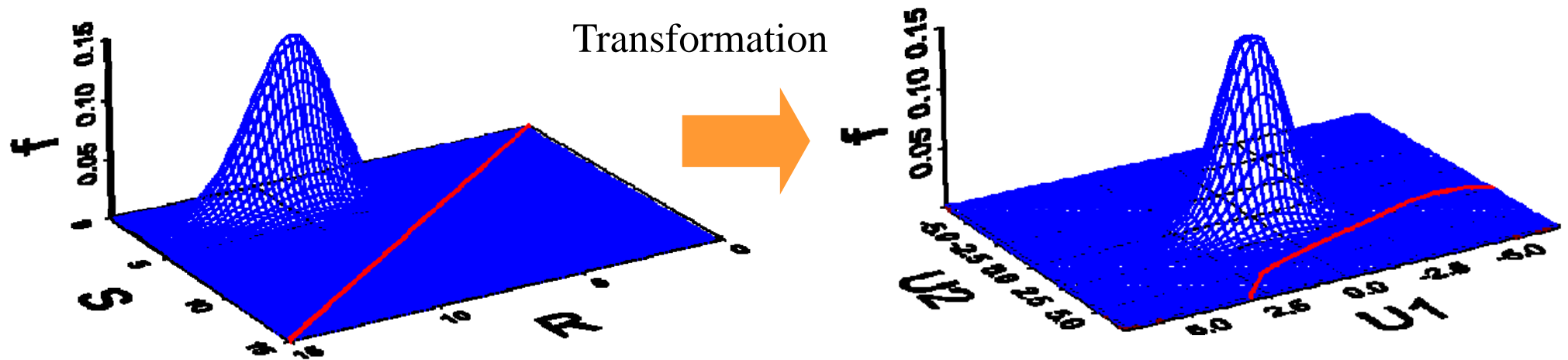
$$\Phi(u_1) = F_{X_1}(x_1)$$

$$\Phi(u_2) = F_{X_2}(x_2 | x_1)$$

⋮

$$\Phi(u_n) = F_{X_n}(x_n | x_1, x_2, \dots, x_{n-1})$$

# Basics of Structural Reliability Methods



$g(Z)$ : linear

$$\mu_{Z1}, \mu_{Z2} \in \mathbb{R}$$

$$\sigma_{Z1}, \sigma_{Z2} \in \mathbb{R}$$

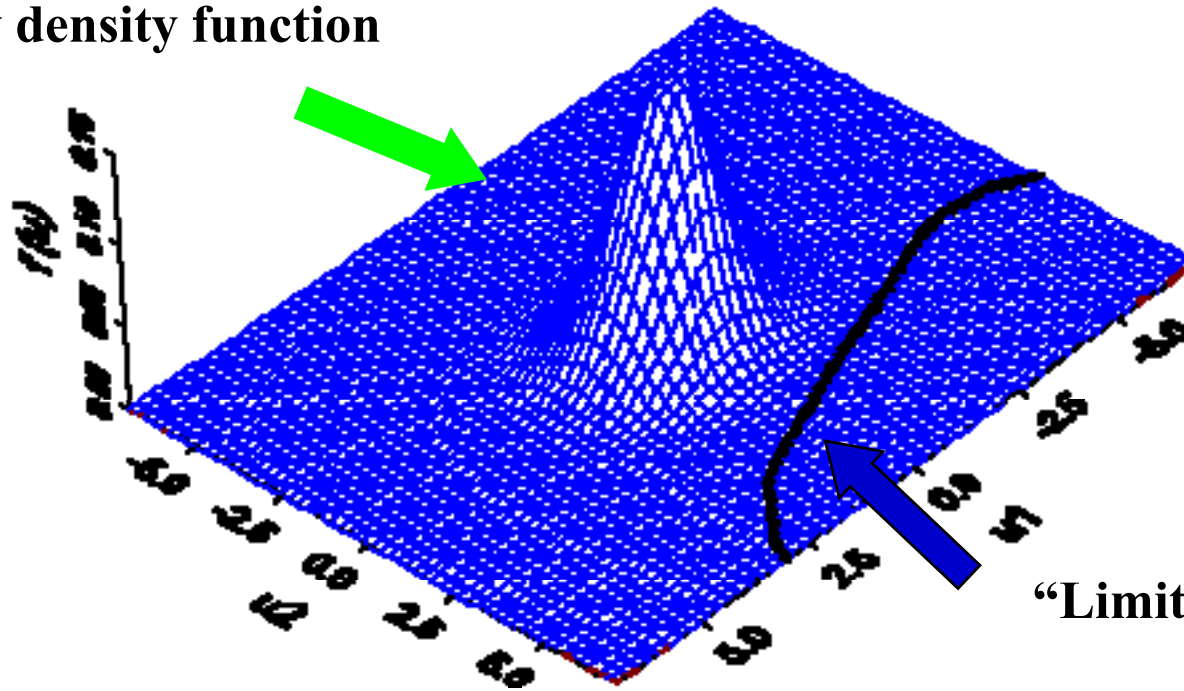
$g(U)$ : non linear

$$\mu_{U1} = \mu_{U2} = 0$$

$$\sigma_{U1} = \sigma_{U2} = 1$$

# Basics of Structural Reliability Methods

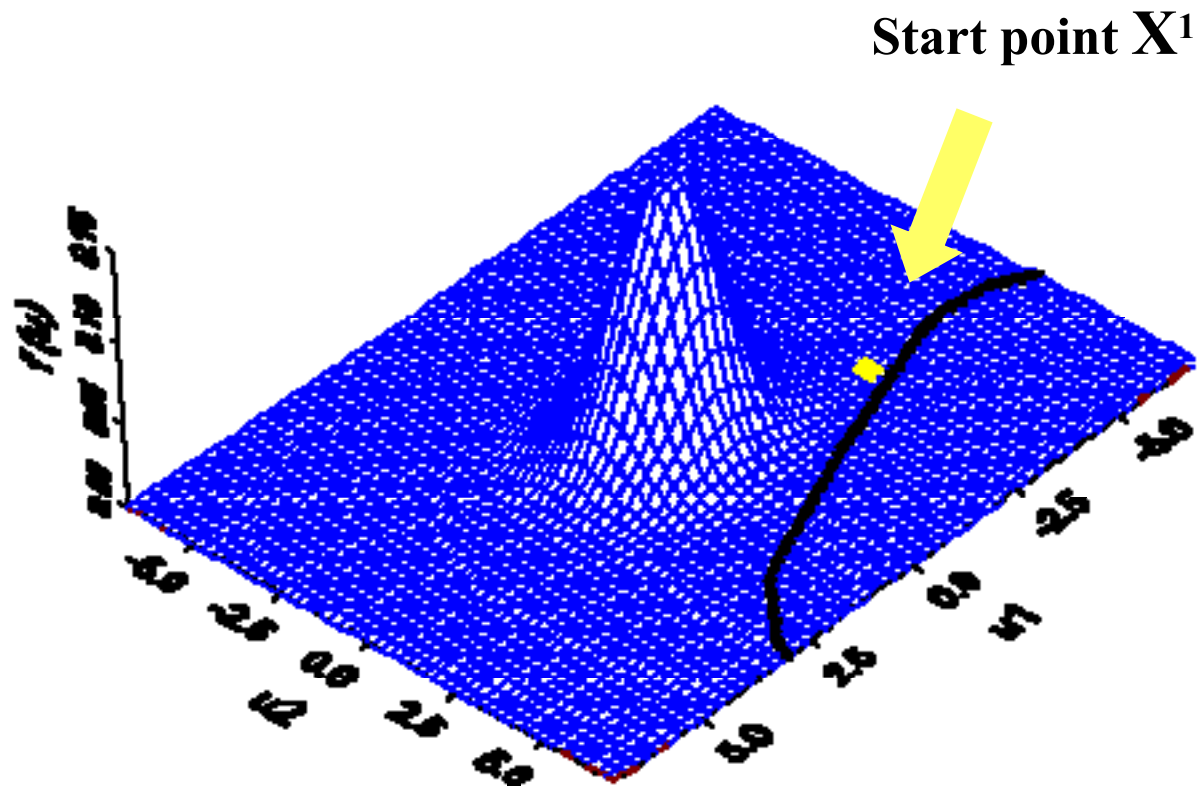
joint probability density function



“Limit state function”

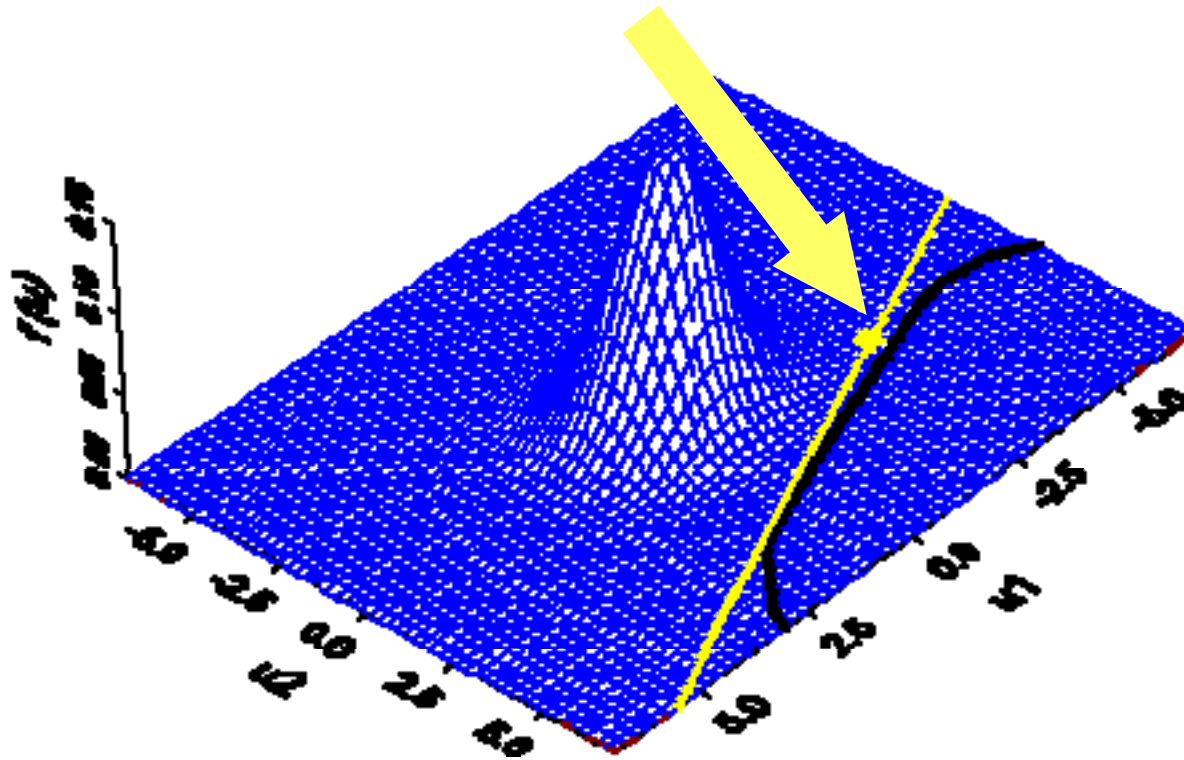
$$g(\mathbf{U}) = R - S$$

# Basics of Structural Reliability Methods



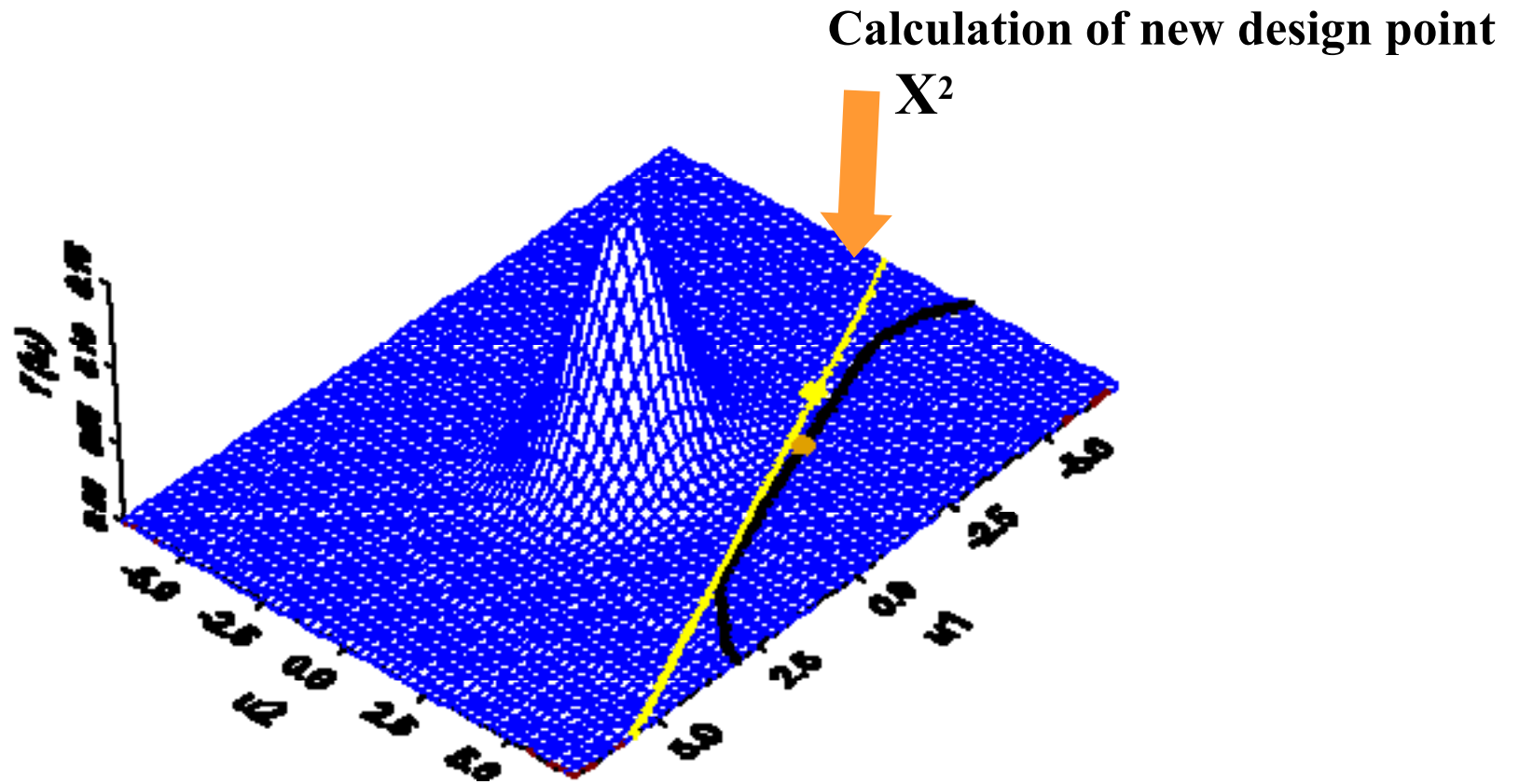
# Basics of Structural Reliability Methods

Linearization of Limit state function in starting point

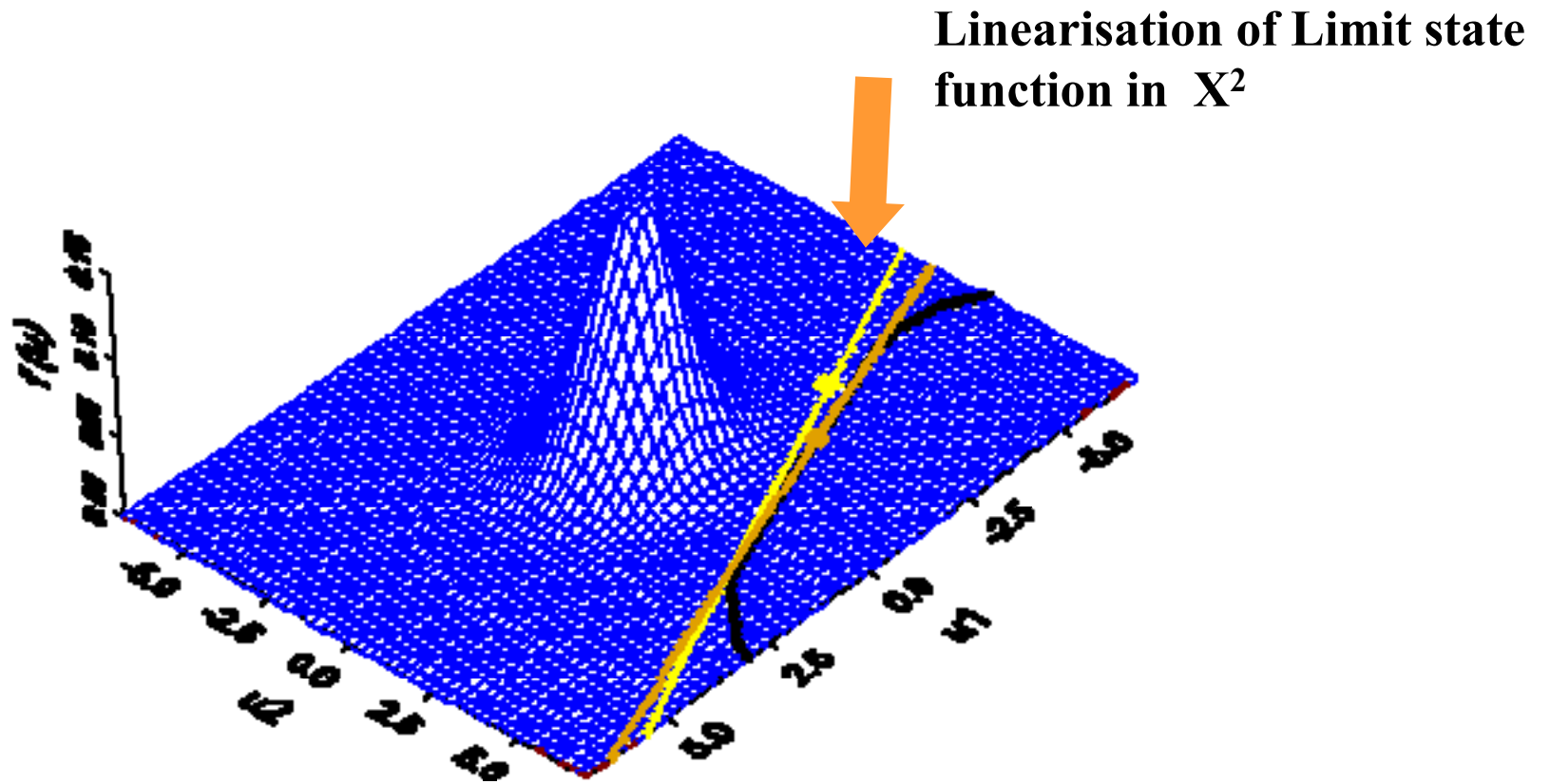




# Basics of Structural Reliability Methods

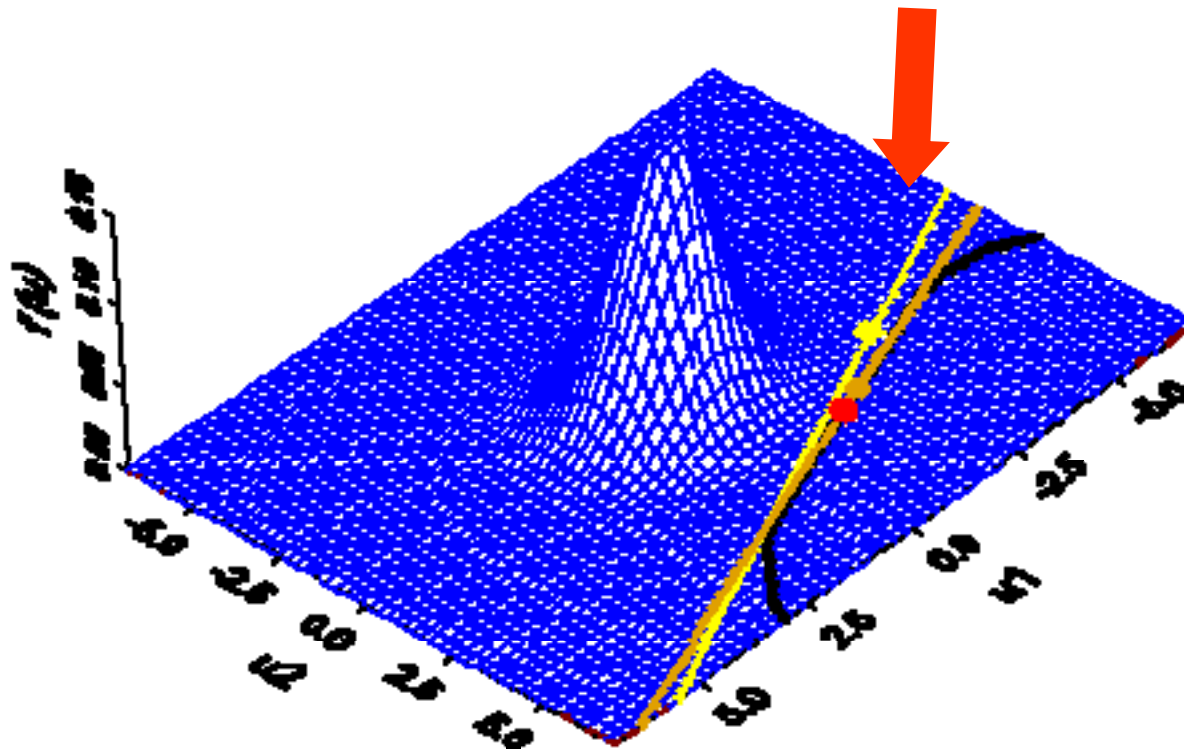


# Basics of Structural Reliability Methods

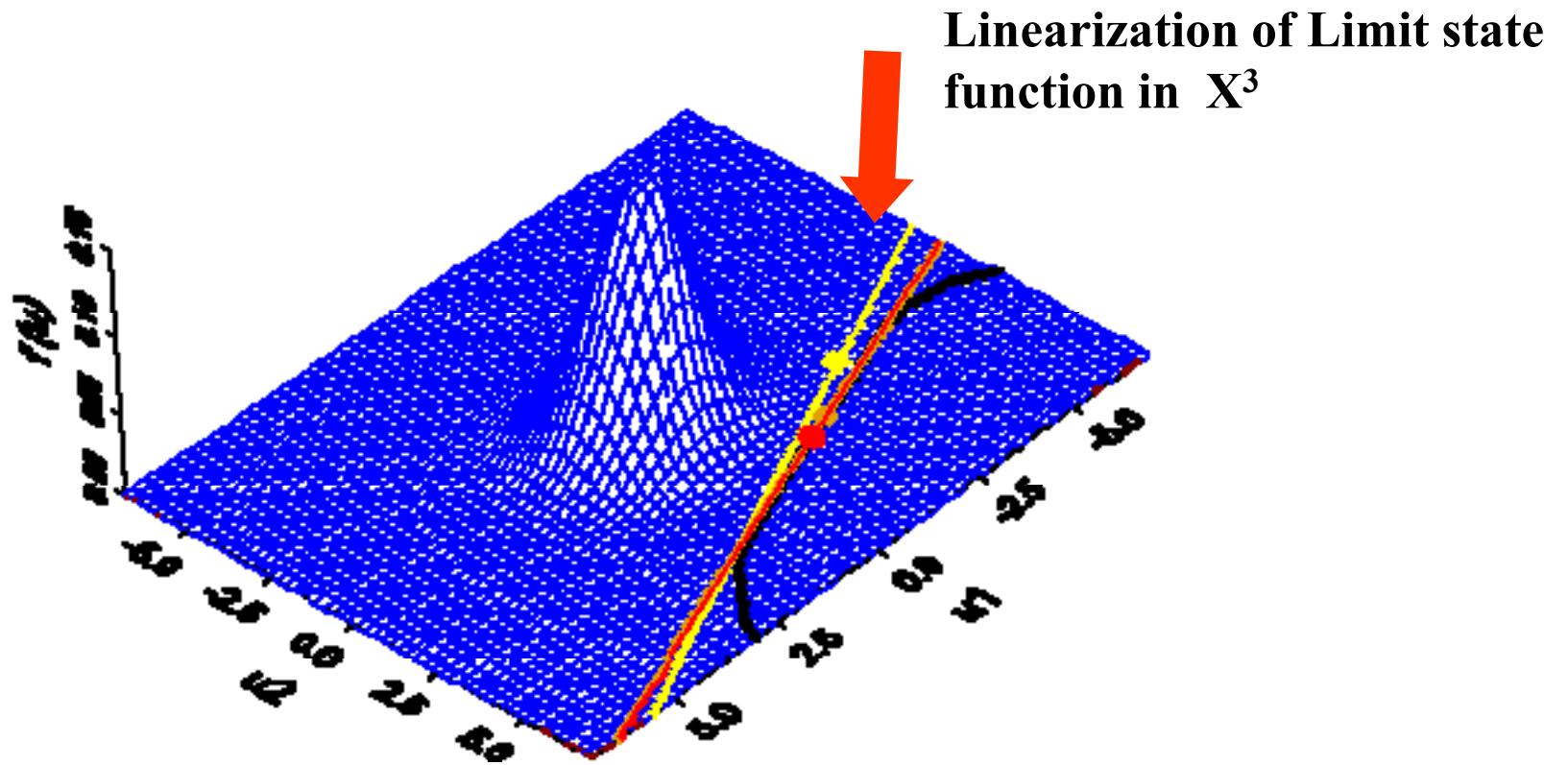


# Basics of Structural Reliability Methods

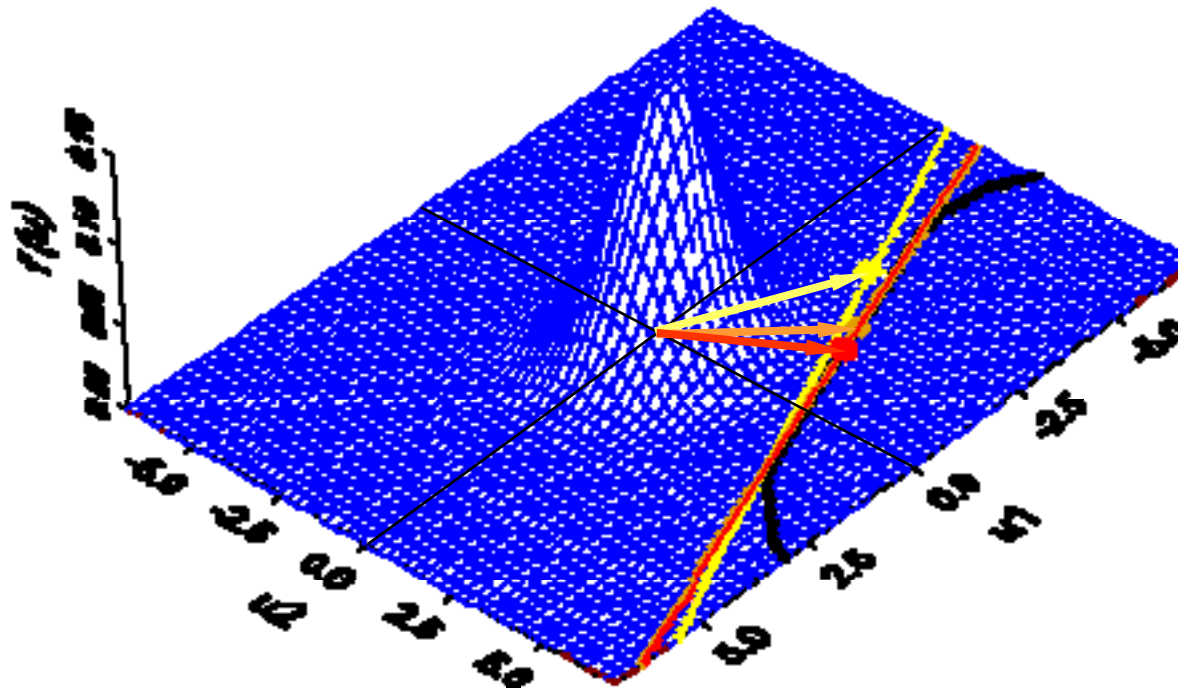
Calculation of new design point  $X^3$



# Basics of Structural Reliability Methods



# Basics of Structural Reliability Methods



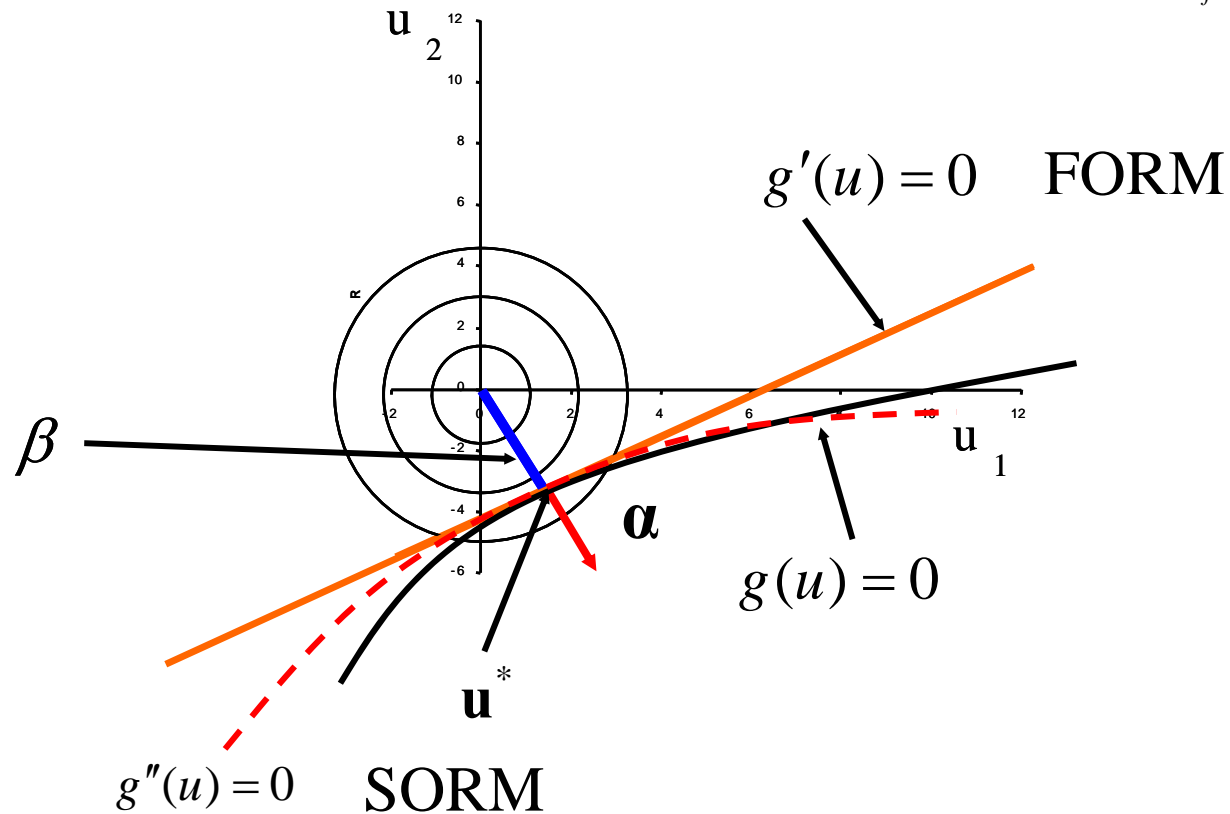
$\beta^1=3.556$   
 $\beta^2=3.607$   
 $\beta^3=3.608$   
 $\beta^4=3.608$

Convergency Criteria: 
$$\Delta\beta = \left| \beta^{n+1} - \beta^n \right| \leq \varepsilon$$

# Basics of Structural Reliability Methods

## SORM Improvements

$$P_f = \int_{\Omega_f = \{g(\mathbf{x}) \leq 0\}} f_{\mathbf{x}}(\mathbf{x}) d\mathbf{x}$$

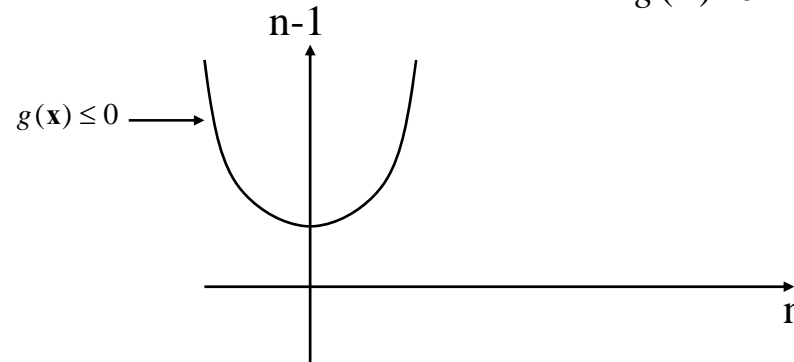


# Basics of Structural Reliability Methods

## SORM Improvements

## Asymptotic Laplace integral solutions

$$I = \int_{g(\mathbf{x}) \leq 0} e^{\lambda h(\mathbf{x})} d\mathbf{x}$$



$$I = \int_{g(\mathbf{x}) \leq 0} e^{\lambda h(\mathbf{x})} d\mathbf{x} \approx \frac{(2\pi)^{(n-1)/2} \lambda^{-(n+1)}}{\lambda^{(n+1)} \sqrt{\prod_{i=1}^{n-1} (1 - \kappa_i)}}$$

$$P_f = \int_{g(\mathbf{x}) \leq 0} \frac{e^{-\frac{1}{2} \sum_{i=1}^n x_i^2}}{(\sqrt{2\pi})^n} d\mathbf{x} \approx \frac{e^{-\beta^2/2}}{\sqrt{2\pi} \beta \sqrt{\prod_{i=1}^{n-1} (1 - \kappa_i)}} = \frac{\varphi(-\beta)}{\beta \sqrt{\prod_{i=1}^{n-1} (1 - \kappa_i)}} \cong \frac{\Phi(-\beta)}{\sqrt{\prod_{i=1}^{n-1} (1 - \kappa_i)}}$$

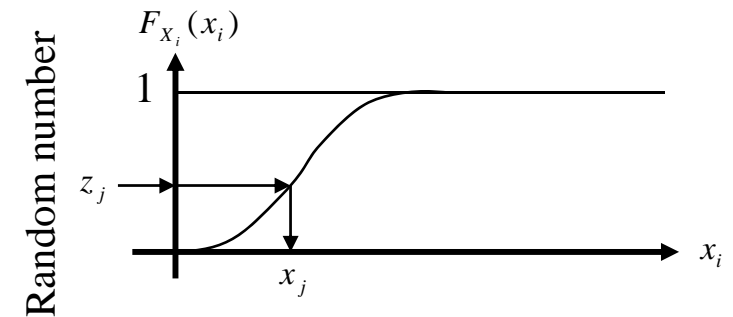
Main curvatures

# Basics of Structural Reliability Methods

Simulation methods may also be used to solve the integration problem

- 1)  $m$  realizations of the vector  $\mathbf{X}$  are generated
- 2) for each realization the value of the limit state function is evaluated
- 3) the realizations where the limit state function is zero or negative are counted
- 4) The failure probability is estimated as

$$P_f = \int_{\Omega_f = \{g(\mathbf{x}) \leq 0\}} f_{\mathbf{X}}(\mathbf{x}) d\mathbf{x}$$



$$n_f$$

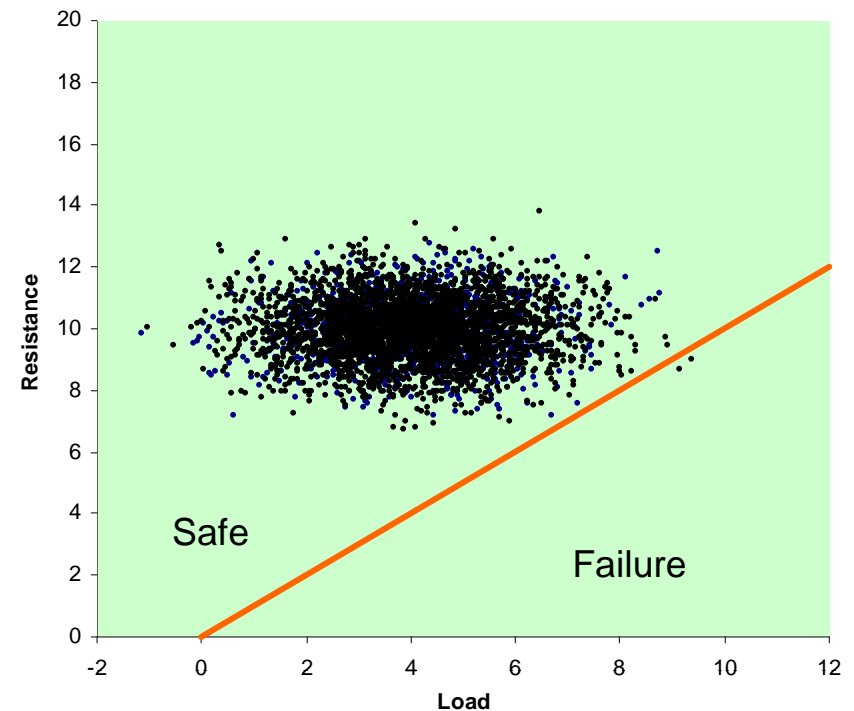
$$p_f = \frac{n_f}{m}$$



# Basics of Structural Reliability Methods

- Estimation of failure probabilities using Monte Carlo Simulation
  - $m$  random outcomes of  $R$  and  $S$  are generated and the number of outcomes  $n_f$  in the failure domain are recorded and summed
  - The failure probability  $p_f$  is then

$$p_f = \frac{n_f}{m}$$



# Basics of Structural Reliability Methods

## Partial safety factors

Design codes prescribe design equations where the design variables (e.g. cross-sections) are to be determined as a function of

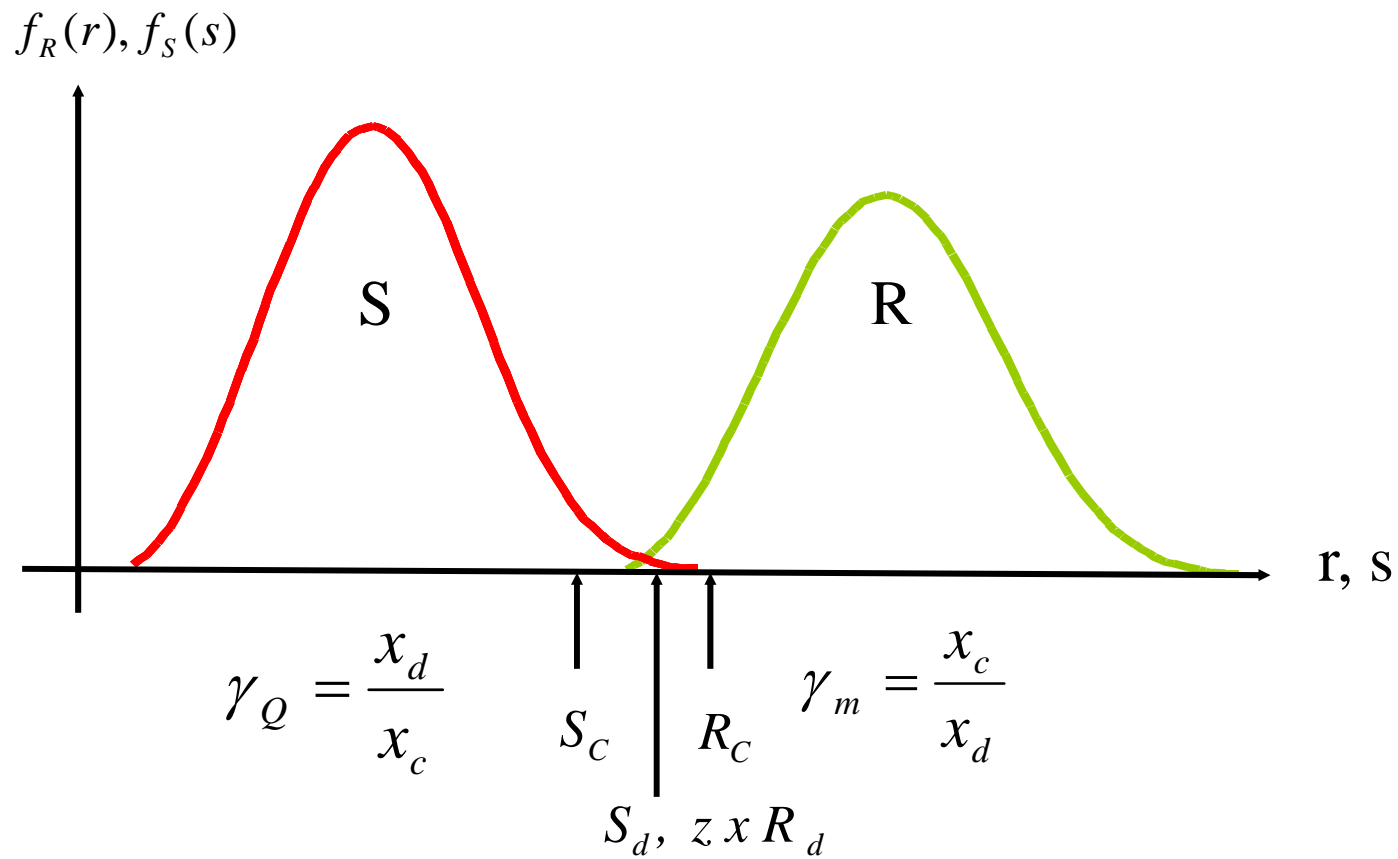
$$zR_c / \gamma_m - (\gamma_{G_a} G_c + \gamma_Q Q_c) = 0$$

- Characteristic values
- Partial safety factors

$$\begin{array}{ccc} R_c & G_c & Q_c \\ \gamma_m & \gamma_G & \gamma_Q \end{array}$$

The design variables are selected such that the design equation is close to zero

# Basics of Structural Reliability Methods



# Basics of Structural Reliability Methods

## Example

Iteration	Start	1	2	3	4	5
$\beta$	3.0000	3.6719	3.7399	3.7444	3.7448	3.7448
$\alpha_R$	-0.5800	-0.5701	-0.5612	-0.5611	-0.5610	-0.5610
$\alpha_A$	-0.5800	-0.5701	-0.5612	-0.5611	-0.5610	-0.5610
$\alpha_S$	0.5800	0.5916	0.6084	0.6086	0.6087	0.6087

$$\mu_R = 350, \sigma_R = 35$$

$$\mu_A = 10, \sigma_A = 1$$

$$\mu_S = 1500, \sigma_R = 300$$

Design value for  $r$

$$r_d = u_R^* \cdot \sigma_R + \mu_R = -0.561 \cdot 3.7448 \cdot 35 + 350.0 = 276.56$$

Characteristic value for  $r$

$$r_c = -1.64 \cdot \sigma_R + \mu_R = -1.64 \cdot 35 + 350 = 292.60$$

Partial safety factor

$$\gamma_R = \frac{292.60}{276.56} = 1.06$$