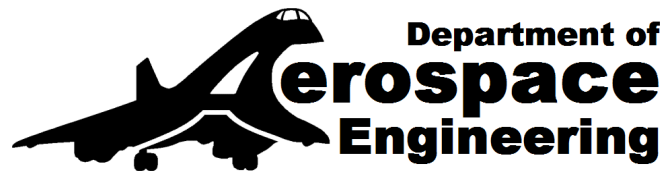


# Reliability Analysis in High Dimensions

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# Outline of the presentation

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- Introduction to structural reliability analysis
- Limitation of current methods in high dimension
- Asymptotic distribution of quadratic forms
- Strict asymptotic formulation
- Weak asymptotic formulation
- Numerical result
- Open problems & discussions

# Reliability analysis: basics

Probability of failure

$$P_f = (2\pi)^{-n/2} \int_{g(\mathbf{x}) \leq 0} e^{-\mathbf{x}^T \mathbf{x} / 2} d\mathbf{x}$$

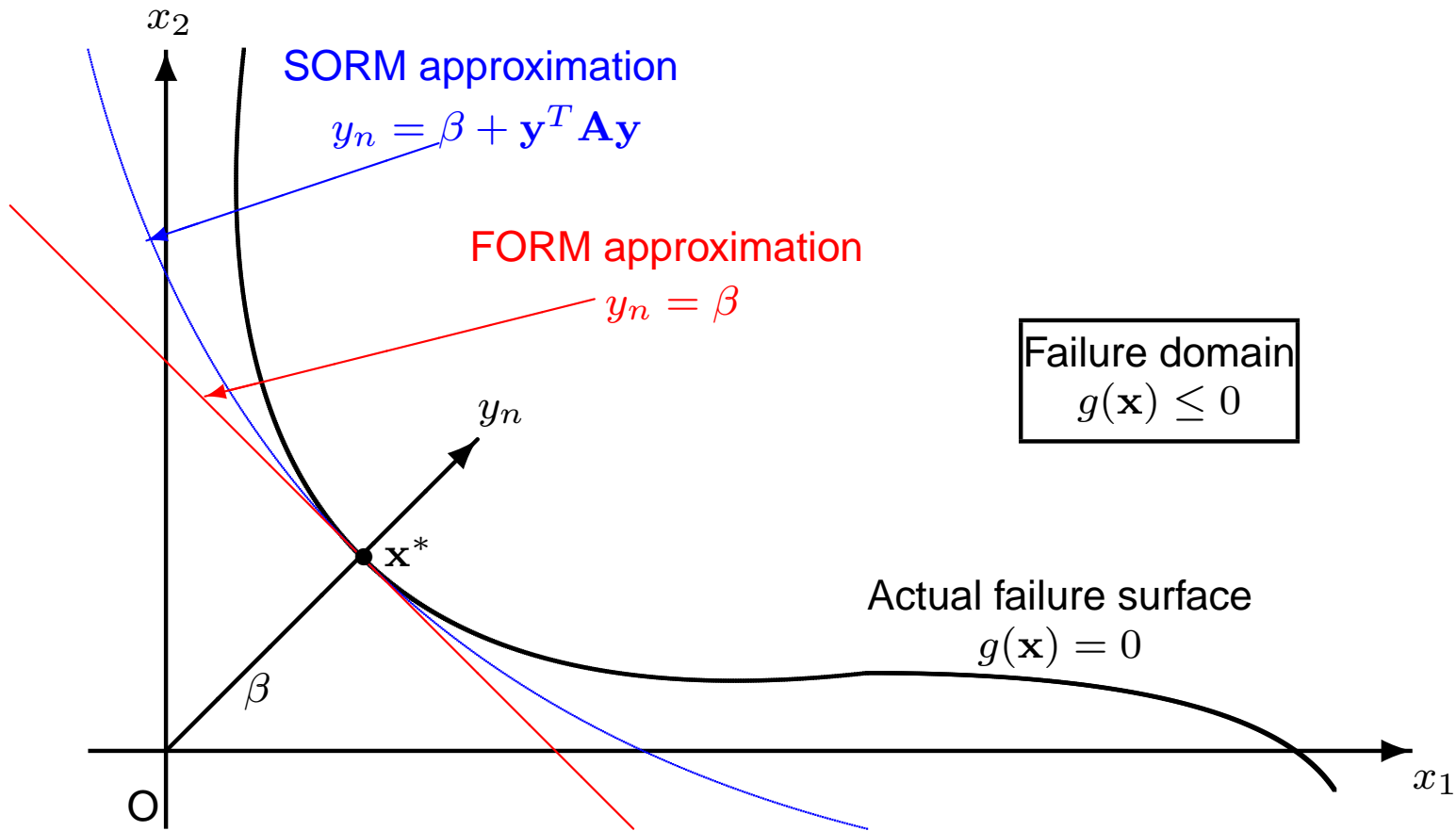
$\mathbf{x} \in \mathbb{R}^n$ : Gaussian parameter vector

$g(\mathbf{x})$ : failure surface

Maximum contribution comes from the neighborhood where  $\mathbf{x}^T \mathbf{x} / 2$  is minimum subject to  $g(\mathbf{x}) \leq 0$ . The design point  $\mathbf{x}^*$ :

$$\mathbf{x}^* : \min\{(\mathbf{x}^T \mathbf{x}) / 2\} \quad \text{subject to} \quad g(\mathbf{x}) = 0.$$

# Graphical explanation



$$\frac{\mathbf{x}^*}{\beta} = -\frac{\nabla g}{|\nabla g|} = \boldsymbol{\alpha}^*$$

# FORM/SORM approximations

$$P_f \approx \text{Prob} [y_n \geq \beta + \mathbf{y}^T \mathbf{A} \mathbf{y}] = \text{Prob} [y_n \geq \beta + U] \quad (1)$$

where

$$U : \mathbb{R}^{n-1} \mapsto \mathbb{R} = \mathbf{y}^T \mathbf{A} \mathbf{y},$$

is a quadratic form in Gaussian random variable. The eigenvalues of  $\mathbf{A}$ , say  $a_j$ , can be related to the principal curvatures of the surface  $\kappa_j$  as  $a_j = \kappa_j/2$ . Considering  $\mathbf{A} = \mathbf{O}$  in Eq. (1), we have the FORM:

$$P_f \approx \Phi(-\beta)$$

# SORM approximations

Breitung's asymptotic formula (1984):

$$P_f \rightarrow \Phi(-\beta) \|\mathbf{I}_{n-1} + 2\beta\mathbf{A}\|^{-1/2} \quad \text{when } \beta \rightarrow \infty$$

Hohenbichler and Rackwitz's improved formula (1988):

$$P_f \approx \Phi(-\beta) \left\| \mathbf{I}_{n-1} + 2 \frac{\varphi(\beta)}{\Phi(-\beta)} \mathbf{A} \right\|^{-1/2}$$

# The curse of dimensionality

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- If  $n$ , i.e. the dimension is large, the computation time to obtain  $P_f$  using any tools will be high (no magic is possible!)

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# The curse of dimensionality

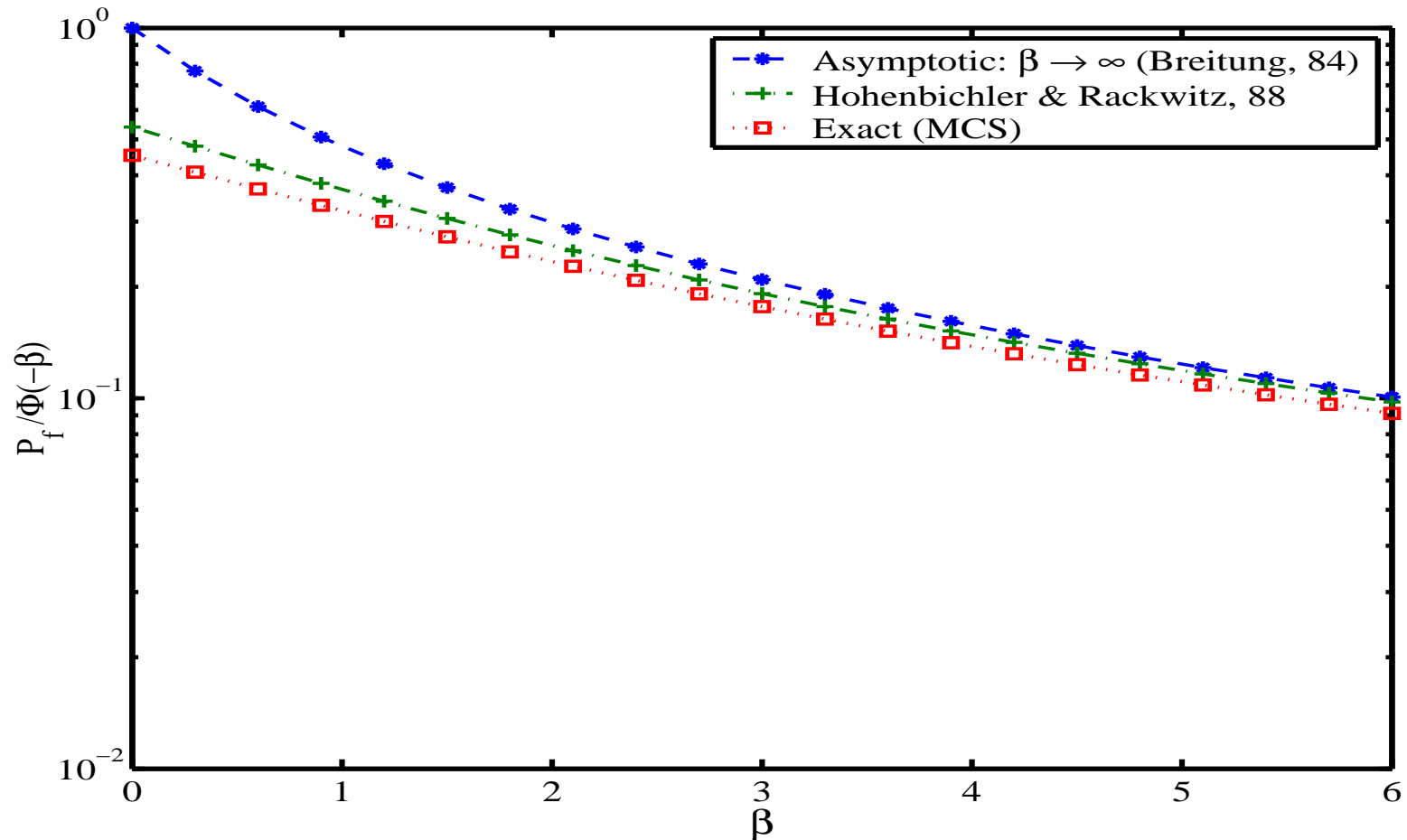
- If  $n$ , i.e. the dimension is large, the computation time to obtain  $P_f$  using any tools will be high (no magic is possible!)
- **Question 1:** What is a ‘high dimension’?
- **Question 2:** Suppose we have followed the ‘normal route’ and did all the calculations (i.e.,  $x^*$ ,  $\beta$  and  $A$ ). Can we still trust the results from classical FORM/SORM in high dimension?

# Numerical example

Consider a problem for which the failure surface is **exactly** parabolic:  $g = -y_n + \beta + \mathbf{y}^T \mathbf{A} \mathbf{y}$

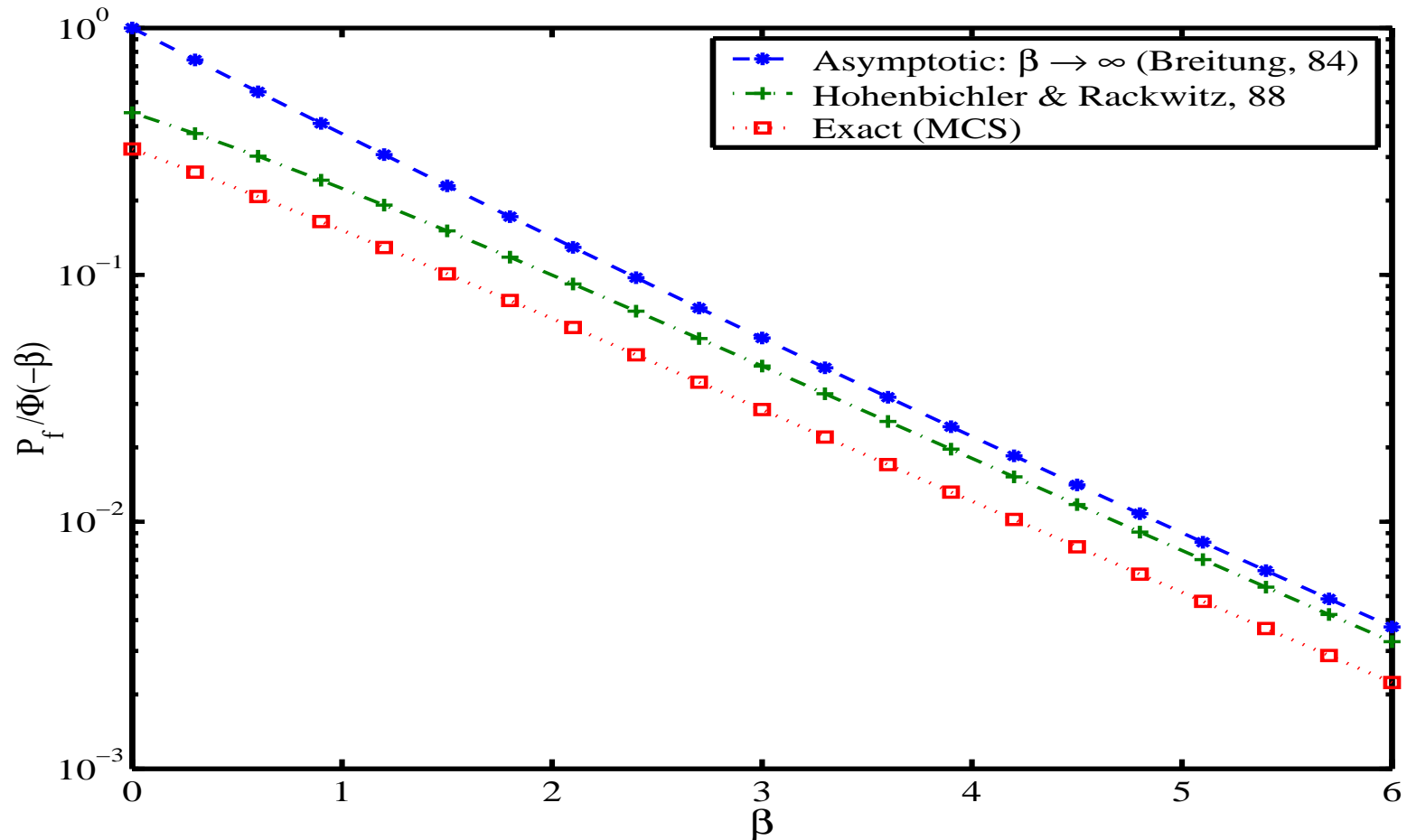
- We choose  $n$  and the value of  $\text{Trace}(\mathbf{A})$
- When  $\text{Trace}(\mathbf{A}) = 0$  the failure surface is effectively linear. Therefore, the more the value of  $\text{Trace}(\mathbf{A})$ , the more non-linear the failure surface becomes.
- It is assumed that the eigenvalues of  $\mathbf{A}$  are uniform random numbers.

# $P_f$ for small $n$



Failure probability for  $n - 1 = 3$ ,  $\text{Trace}(\mathbf{A}) = 1$

# $P_f$ for large $n$



Failure probability for  $n - 1 = 100$ ,  $\text{Trace}(\mathbf{A}) = 1$

# Asymptotic distribution of quadratic forms

Moment generating function:

$$M_U(\mathbf{s}) = \|\mathbf{I}_{n-1} - 2\mathbf{s}\mathbf{A}\|^{-1/2} = \prod_{k=1}^{n-1} (1 - 2sa_k)^{-1/2}$$

Now construct a **sequence** of new random variables  $q = U/\sqrt{n}$ . The moment generating function of  $q$ :

$$M_q(\mathbf{s}) = M_U(\mathbf{s}/\sqrt{n}) = \prod_{k=1}^{n-1} (1 - 2sa_k/\sqrt{n})^{-1/2}$$

# Asymptotic distribution

Truncating the Taylor series expansion:

$$\ln (M_q(s)) \approx \text{Trace} (\mathbf{A}) s / \sqrt{n} + (2 \text{Trace} (\mathbf{A}^2)) s^2 / 2n$$

We assume  $n$  is large such that the following conditions hold

$$\frac{2}{n} \text{Trace} (\mathbf{A}^2) < \infty$$

and  $\frac{2^r}{n^{r/2} r} \text{Trace} (\mathbf{A}^r) \rightarrow 0, \forall r \geq 3$

# Asymptotic distribution

Therefore, the moment generating function of  $U = q\sqrt{n}$  can be approximated by:

$$M_U(s) \approx e^{\text{Trace}(\mathbf{A})s + \left(2 \text{Trace}(\mathbf{A}^2)\right) s^2/2}$$

From the uniqueness of the Laplace Transform pair it follows that  $U$  asymptotically approaches a Gaussian random variable with mean  $\text{Trace}(\mathbf{A})$  and variance  $2\text{Trace}(\mathbf{A}^2)$ , that is

$$U \simeq \mathbb{N}_1 \left( \text{Trace}(\mathbf{A}), 2 \text{Trace}(\mathbf{A}^2) \right) \quad \text{when } n \rightarrow \infty$$

# Minimum number of random variables

The error in neglecting higher order terms:

$$\frac{1}{r} \left( \frac{2s}{\sqrt{n}} \right)^r \text{Trace} (\mathbf{A}^r), \text{ for } r \geq 3.$$

Using  $s = \beta$  and assuming there exist a small real number  $\epsilon$  (the error) we have

$$\frac{1}{r} \frac{(2\beta)^r}{n^{r/2}} \text{Trace} (\mathbf{A}^r) < \epsilon \text{ or } n > \frac{4\beta^2}{\sqrt[r]{r^2\epsilon^2}} \left( \sqrt[r]{\text{Trace} (\mathbf{A}^r)} \right)^2$$



# Strict asymptotic formulation

We rewrite (1):

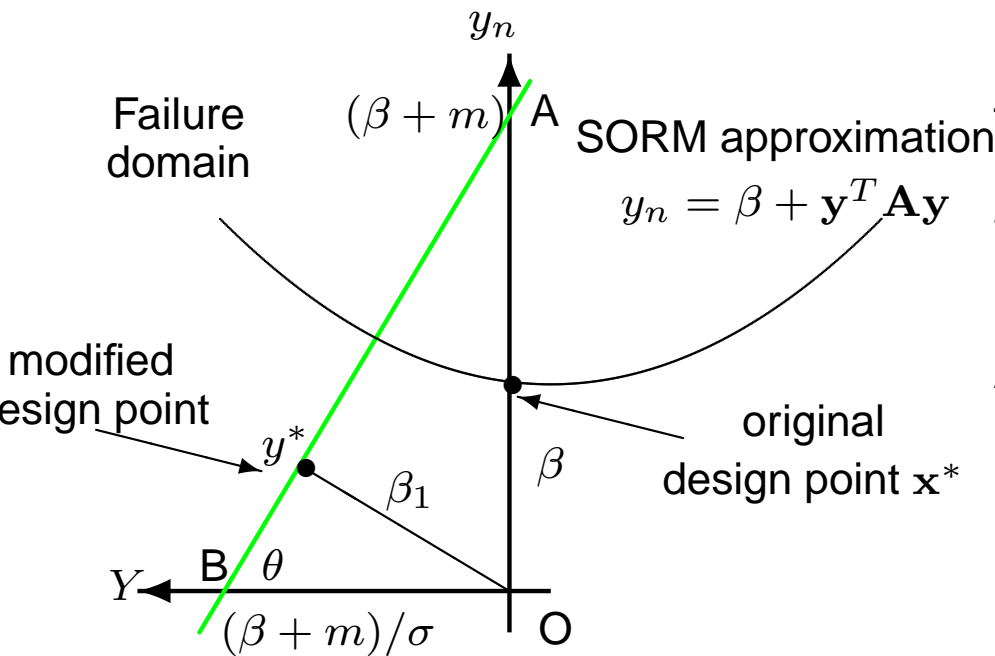
$$P_f \approx \text{Prob} [y_n \geq \beta + U] = \text{Prob} [y_n - U \geq \beta]$$

Since  $U$  is asymptotically Gaussian, the variable  $z = y_n - U$  is also Gaussian with mean  $(-\text{Trace}(\mathbf{A}))$  and variance  $(1 + 2 \text{Trace}(\mathbf{A}^2))$ . Thus,

$$P_{f_{\text{Strict}}} \rightarrow \Phi(-\beta_1), \quad \beta_1 = \frac{\beta + \text{Trace}(\mathbf{A})}{\sqrt{1 + 2 \text{Trace}(\mathbf{A}^2)}}, \quad n \rightarrow \infty$$

# Graphical explanation

$$m = \text{Trace}(\mathbf{A}), \sigma^2 = 2\text{Trace}(\mathbf{A}^2)$$



Failure surface:  $y_n - U \geq \beta$ . Using the standardizing transformation  $Y = (U - m)/\sigma$ , modified

failure surface 
$$\frac{y_n}{\beta+m} + \frac{Y}{-\frac{\beta+m}{\sigma}} \geq 1 .$$

From  $\triangle AOB$ , 
$$\sin \theta = \frac{\tan \theta}{\sqrt{1+\tan^2 \theta}} = \frac{\sigma}{\sqrt{1+\sigma^2}} .$$

Therefore, from  $\triangle OBy^*$ :

$$\beta_1 = \frac{\beta+m}{\sigma} \sin \theta = \frac{\beta+m}{\sqrt{1+\sigma^2}} = \frac{\beta + \text{Trace}(\mathbf{A})}{\sqrt{1+2 \text{Trace}(\mathbf{A}^2)}} .$$

If  $n$  is small,  $m, \sigma$  will be small. When  $m, \sigma \rightarrow 0$ ,  $AB$  rotates clockwise and eventually becomes parallel to the  $Y$ -axis with a shift of  $+\beta$ . In this situation  $y^* \rightarrow x^*$  in the  $y_n$ -axis and  $\beta_1 \rightarrow \beta$  as expected. This explains why classical F/SORM approximations based on the original design point  $x^*$  do not work well when a large number of random variables are considered.

# Weak asymptotic formulation

$$\begin{aligned} P_f &\approx \text{Prob} [y_n \geq \beta + U] \\ &= \int_{\mathbb{R}} \left\{ \int_{\beta+u}^{\infty} \varphi(y_n) dy_n \right\} p_U(u) du = \mathbb{E} [\Phi(-\beta - U)] \end{aligned}$$

Noticing that  $u \in \mathbb{R}^+$  as  $\mathbf{A}$  is positive definite we rewrite

$$P_f \approx \int_{\mathbb{R}^+} e^{\ln[\Phi(-\beta-u)] + \ln[p_U(u)]} du$$

# Weak asymptotic formulation

For the maxima of the integrand (say at point  $u^*$ )

$$\frac{\partial}{\partial u} \{ \ln [\Phi(-\beta - u)] + \ln [p_U(u)] \} = 0$$

Recalling that

$$p_U(u) = (2\pi)^{-1/2} \sigma^{-1} e^{-(u-m)^2/(2\sigma^2)}$$

we have

$$\frac{\varphi(\beta + u)}{\Phi(-(\beta + u))} = \frac{m - u}{\sigma^2}$$

# Weak asymptotic formulation

Because this relationship holds at the optimal point  $u^*$ , define a constant  $\eta$  as

$$\eta = \frac{\varphi(\beta + u^*)}{\Phi(-(\beta + u^*))} = \frac{m - u^*}{\sigma^2}$$

Taking a first-order Taylor series expansion of  $\ln [\Phi(-\beta - u)]$  about  $u = u^*$ :

$$\Phi(-\beta - u) \approx e^{\ln[\Phi(-(\beta+u^*))] - \frac{\varphi(\beta+u^*)}{\Phi(-(\beta+u^*))}(u-u^*)}$$

# Weak asymptotic formulation

Using  $\eta$  we have

$$\Phi(-\beta - u) \approx \Phi(-\beta_2) e^{\eta u^*} e^{-\eta u} \quad (1)$$

where the modified reliability index

$$\beta_2 = \beta + u^*$$

Taking the expectation of (1) and using the expression of the moment generating function:

$$P_f \approx \mathbb{E} [\Phi(-\beta - U)] = \Phi(-\beta_2) e^{\eta u^*} \|\mathbf{I}_{n-1} + 2 \eta \mathbf{A}\|^{-1/2}$$

# Weak asymptotic formulation

Considering the asymptotic expansion of the ratio

$$\eta = \frac{\varphi(\beta + u^*)}{\Phi(-(\beta + u^*))} \approx (\beta + u^*) = \beta_2 \approx \frac{m - u^*}{\sigma^2}$$

We obtain

$$u^* \approx \frac{m - \beta\sigma^2}{1 + \sigma^2}, \quad \beta_2 = \beta + u^* \approx \frac{\beta + m}{1 + \sigma^2} = \frac{\beta + \text{Trace}(\mathbf{A})}{1 + 2 \text{Trace}(\mathbf{A}^2)}$$

Since  $\eta \approx \beta_2$ ,  $u^*$  can be expressed in terms of  $\beta_2$  as

$$u^* \approx -(\beta_2\sigma^2 - m) = -(2\beta_2 \text{Trace}(\mathbf{A}^2) - \text{Trace}(\mathbf{A}))$$

# Weak asymptotic formulation

Using the expression of  $\eta$  and  $u^*$ , the failure probability using weak asymptotic formulation:

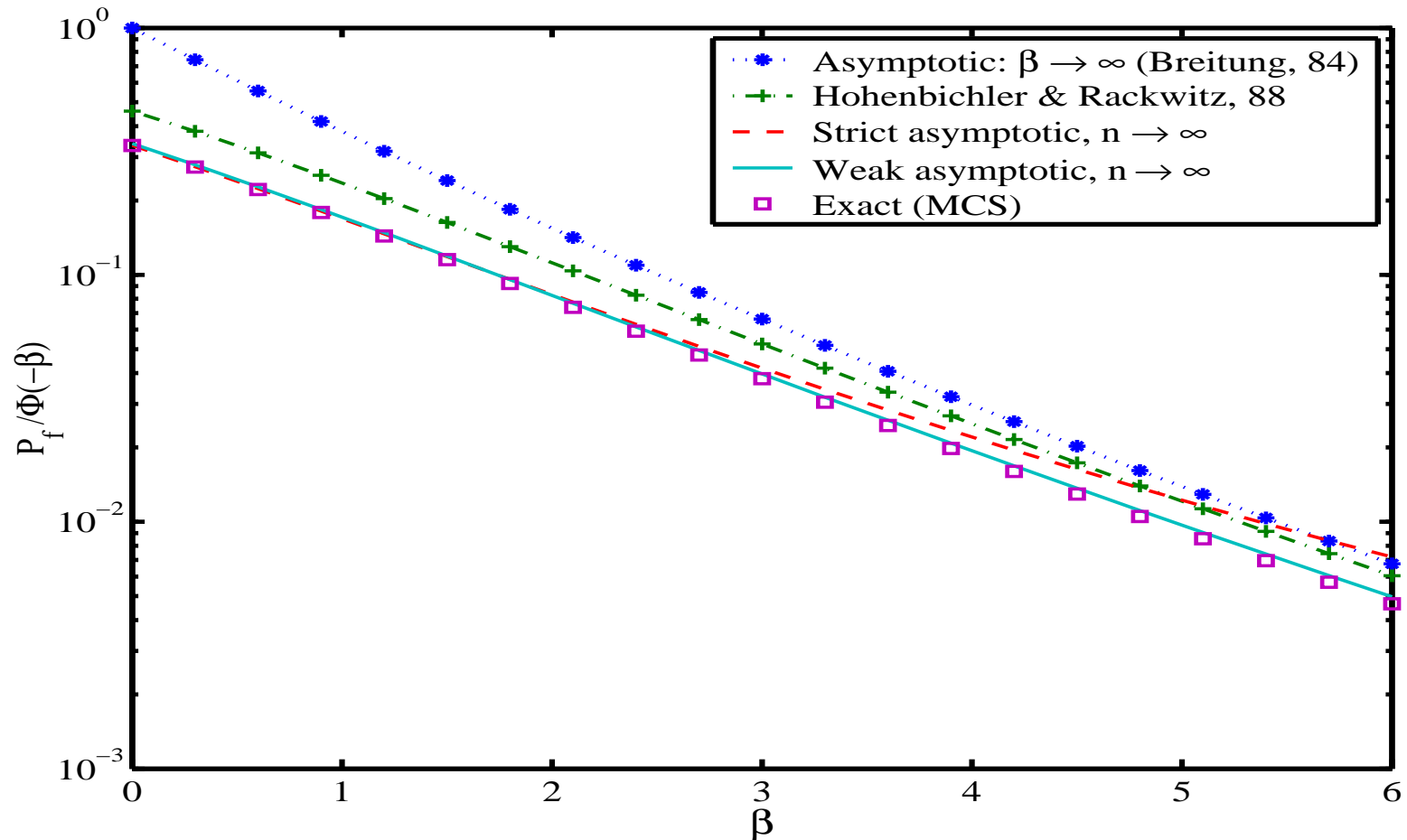
$$P_{f\text{Weak}} \rightarrow \frac{\Phi(-\beta_2) e^{-\left(2\beta_2^2 \text{Trace}(\mathbf{A}^2) - \beta_2 \text{Trace}(\mathbf{A})\right)}}{\sqrt{\|\mathbf{I}_{n-1} + 2\beta_2 \mathbf{A}\|}},$$

$$\text{where } \beta_2 = \frac{\beta + \text{Trace}(\mathbf{A})}{1 + 2 \text{Trace}(\mathbf{A}^2)} \text{ when } n \rightarrow \infty$$

For the small  $n$  case,  $\text{Trace}(\mathbf{A})$ ,  $\text{Trace}(\mathbf{A}^2) \rightarrow 0$  and it can be seen that  $P_{f\text{Weak}}$  approaches to Breitung's formula.

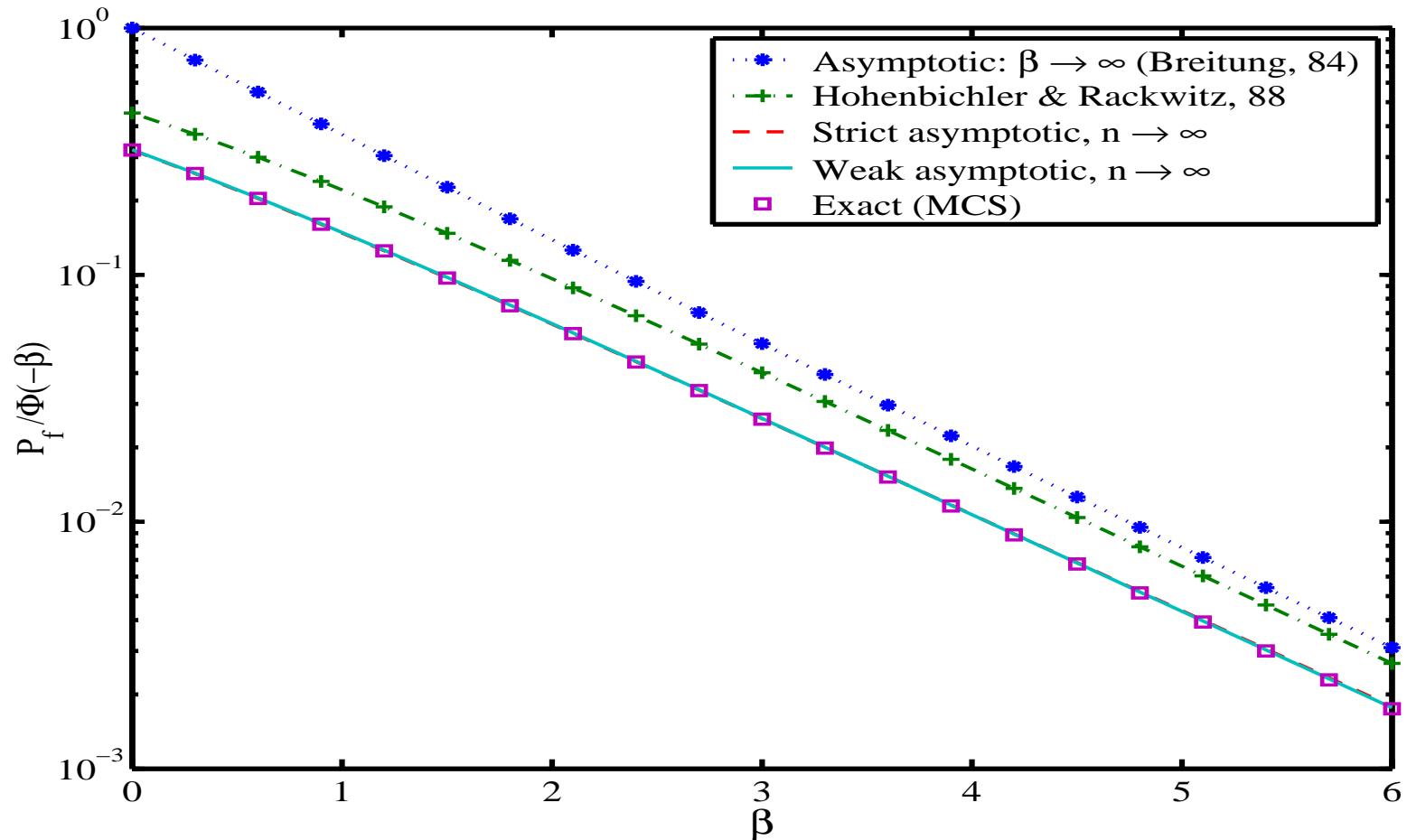


# $P_f$ from asymptotic analysis



Failure probability for  $n - 1 = 35$ ,  $\text{Trace}(\mathbf{A}) = 1$  [ $n_{min} = 176$ ]

# $P_f$ from asymptotic analysis



Failure probability for  $n - 1 = 200$ ,  $\text{Trace}(\mathbf{A}) = 1$

# Summary & conclusions

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- Geometric analysis shows that the classical design point should be modified in high dimension. This also explains why classical FORM/SORM work poorly in high dimension.

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- $$P_{f\text{Strict}} \rightarrow \Phi(-\beta_1), \beta_1 = \frac{\beta + \text{Trace}(\mathbf{A})}{\sqrt{1 + 2 \text{Trace}(\mathbf{A}^2)}}, n \rightarrow \infty$$

The **strict asymptotic formula** can be viewed as the ‘correction’ needed to the existing **FORM** formula in high dimension.

# Summary & conclusions

$$P_{f_{\text{Weak}}} \rightarrow \frac{\Phi(-\beta_2) e^{-\left(2\beta_2^2 \text{Trace}(\mathbf{A}^2) - \beta_2 \text{Trace}(\mathbf{A})\right)}}{\sqrt{\|\mathbf{I}_{n-1} + 2\beta_2 \mathbf{A}\|}},$$

$$\text{where } \beta_2 = \frac{\beta + \text{Trace}(\mathbf{A})}{1 + 2 \text{Trace}(\mathbf{A}^2)} \text{ when } n \rightarrow \infty$$

The **weak asymptotic formula** can be viewed as the correction needed to the existing **SORM** formula in high dimension.

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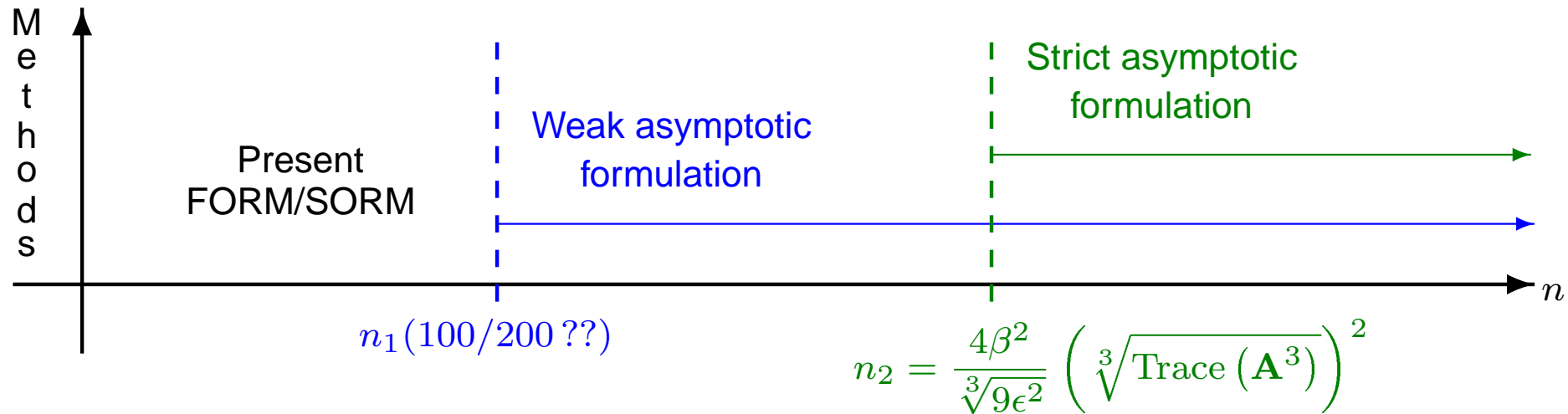


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- Why the design points for the two asymptotic formulations are different?
- Any geometric interpretation for the weak formulation?
- Why these asymptotic results degrade as  $\beta$  becomes high?
- Any expression of  $n_{min}$  for the weak formulation?

# Open Questions

## The broad picture:



$\beta \downarrow, n \downarrow \checkmark$

$\beta \uparrow, n \downarrow \checkmark$  (Asymptotic:  $\beta \rightarrow \infty$ )

$\beta \downarrow, n \uparrow \checkmark$  (Asymptotic:  $n \rightarrow \infty$ )

$\beta \uparrow, n \uparrow \times$  (Joint asymptotic:  $n, \beta \rightarrow \infty$ ?)

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