# What are some tools to solve more complicated reliability problems?



## Lecture 2B – Key Messages

- Monte Carlo Simulation (MCS) is a statistical simulation modelling approach, which can be applied easily (e.g. in MS Excel) to determine  $P_f$  for more complex design equations.
- Why do we care about this stuff in a course on assessment / rehabilitation of structures?

Answer: understanding these concepts, we can identify parameters to measure to reduce uncertainty and "gain" capacity to possibly avoid rehabilitation.

## Monte Carlo Simulation (MCS)

 R and S are not necessarily normally distributed, and much more complicated limit state functions also exist. MCS is a method to solve these more difficult problems.

The basic idea:

$$p_f = \frac{n(G(ar{Z}) < 0)}{N}$$
 — Number of failures



### Monte Carlo Simulation (MCS) Steps

- Step 1: Generate random trial values for the input parameters (e.g., Strength of material, live load, dead load, etc.)
- Step 2: Check if  $G(\bar{Z}) < 0$  (i.e., check if the trial values lead to a failure.)
- Step 3: Steps 1 and 2 should be repeated N times. (required number of trials (N) will be discussed in the following slides)
- Step 4: The probability of failure is equal to the number of failures  $(n_f)$  divided by the total number of trials (N).

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

#### Generating a random trial value for each input parameter

• A random trial value for each input parameter can be determined using a random number between 0 and 1 ( $0 \le r_i \le 1$ ) and the inverse CDF of the input parameter:

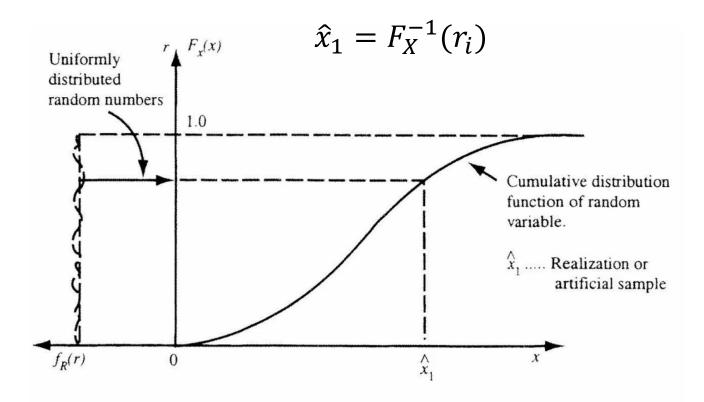
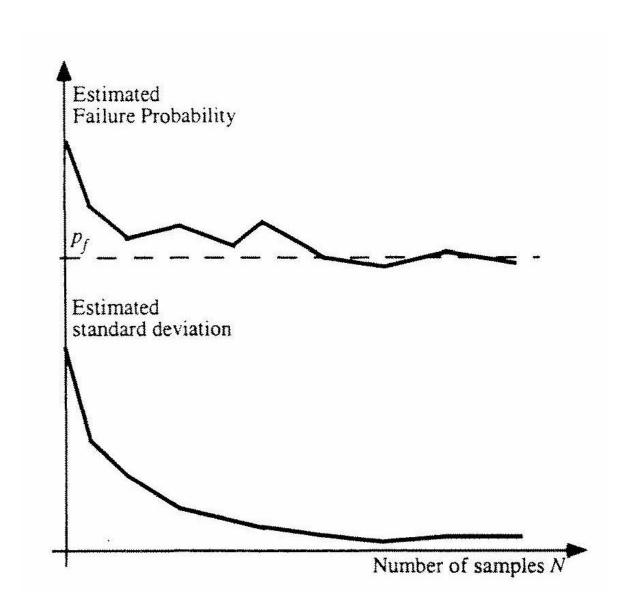


Figure 3.1 Inverse transform method for generation of random variates.

[Melchers 1999]

#### The required number of trials, N



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The number of trials, N, depends on how small  $p_f$  is and the level of confidence desired in the results. Various researchers have proposed formulas.

$$N > \frac{-\ln(1-C)}{p_f}$$
 Level of Confidence

For example:  $p_f = 10^{-3}$ ,  $C = 95\% \rightarrow N \approx 3000$ 

[Broding 1964]



### Bias factors

- characterize différence between design parameter à actual parameter value, ie.

$$\lambda = \frac{Z}{Z_n}$$

- can also be treated as statistical input parameters in limit state functions, ie:

$$\mu_{\lambda} = \frac{Z}{Z_{n}}$$
,  $V_{\lambda} = \frac{\overline{\sigma_{z}}}{Z}$ 

Various references including the CSA-SG Commentary & Calibration Report give statistics for bias factors for various parameters associated with the landing & resistance of structures

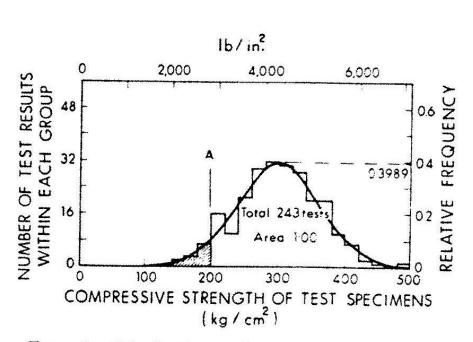


Fig. 1. Distribution of concrete compressive strength.

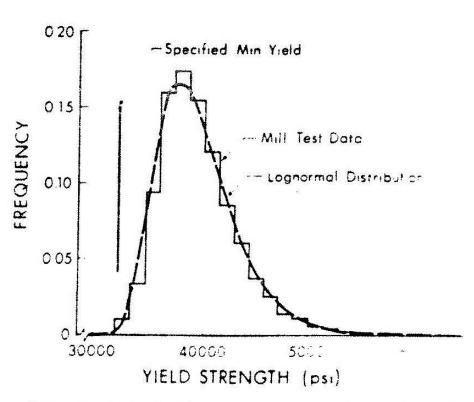


Fig. 3. Distribution of structural steel well strengths (Alpsten 1972).

 Table 3.1
 Statistical Data for Resistance of Steel Components

Category	Source	Bias Factor, δ	Coefficient of
			Variation, V
Welded Sections:	Kennedy &		
Fy	Baker (1984)	1.101	0.0915
S		1.020	0.010
Test/predicted		1.090	0.045
Rolled Sections:	Kennedy &		
Fy	Baker (1984)	1.060	0.051
S		0.990	0.021
Test/predicted		1.090	0.045
Welded Sections:	Kennedy &		
plastic moment	Baker (1984),	1.133	0.096
yield moment	-	1.221	0.100
inelastic moment		1.155	0.084
elastic moment		1.090	0.092
composite, fully plastic moment		1.098	0.096
composite negative yield moment*	*Kennedy	1.221	0.100
compression, $\lambda = 0.8$	(1996)	1.058	0.079
compression, $\lambda = 1.0$		1.048	0.068
compression, $\lambda = 1.2$		1.096	0.075
shear		1.178	0.103

[CSA S6-06 Calibration Report]

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Baker (1984),	1.126	0.081
	1.210	0.077
*Kennedy	1.095	0.075
(1996)	1.210	0.077
	1.143	0.116
	1.187	0.132
	1.185	0.117
	1.102	0.071
Chernenko &		
Kennedy		
(1991)	1.132	0.063
	1.087	0.065
	1.010	0.082
Fisher et al.		
(1978)	1.12	0.090
	1.16	0.100
Fisher et al.	1.22	0.172
1978)		
Lesik &	1.137	0.189
Kennedy		
1990)		
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 Table 3.2
 Statistical Data for Resistance of Prestressed Concrete Components

Component Type	Response	Bias Factor	Coefficient of
		δ	Variation, V
Simple Span Bridges:			
PC Girder	Moment	1.05	0.075
	Shear	1.15	0.140
Continuous Bridges:			And Andrews
PC Girder	Positive Moment	1.05	0.075
continuous for live load only	Negative Moment	1.14	0.130
	Shear	1.15	0.140
PSC Post-tensioned Voided	Positive Moment	1.07	0.065
Slab	Negative Moment	1.07	0.065
	Shear	1.15	0.140
PSC Post-tensioned Box	Positive Moment	1.07	0.065
Girder	Negative Moment	1.07	0.065
	Shear	1.15	0.140

Table 4.1/ Statistical Parameters for Dead Loads

DEAD LOAD	DESCRIPTION	BIAS FACTOR	COEFFICIENT OF
TYPE		$\delta_{ m D}$	VARIATION, VD
D1	factory-produced components, excluding wood	1.03	0.08
D2	cast-in-place concrete, wood and all non-structural components	1.05	0.10
D3	asphalt wearing surface	1.03	0.30

 Table 4.2.3
 Statistical Parameters for Live Load Analysis

ANALYSIS TYPE	BIAS FACTOR, $\delta_{AL}$	COV, V <sub>AL</sub>
Simplified Code Methods	0.93	0.12
Sophisticated Analysis	0.98	0.07

 Table 4.3
 Statistical Parameters for Dynamic Loads

	MEAN VALUE μι	STANDARD DEVIATION $\sigma_{l}$	COV V <sub>I</sub>	DYNAMIC AMPLIFI- CATION FACTOR	COV FOR DYNAMIC AMPLIFI- CATION
Span ≤ 10 m, and Local Components	0.20 μ <sub>L</sub>	0.12 μ <sub>L</sub>	0.60	1.20	0.100
Span > 10 m: One Lane Loaded Two Lanes Loaded	0.15 μ <sub>L</sub> 0.10 μ <sub>L</sub>	0.12 μ <sub>L</sub> 0.08 μ <sub>L</sub>	0.80 0.80	1.15 1.10	0.104 0.072

Example 2 "dynamic load allowance" (aka "impact factor")
$$M_{\Gamma} \geq M_{f}$$

$$\phi M_{P} \geq \chi_{L} \cdot M_{L} \cdot (1+DLA) + \chi_{D} \cdot M_{D}$$

#### Let:

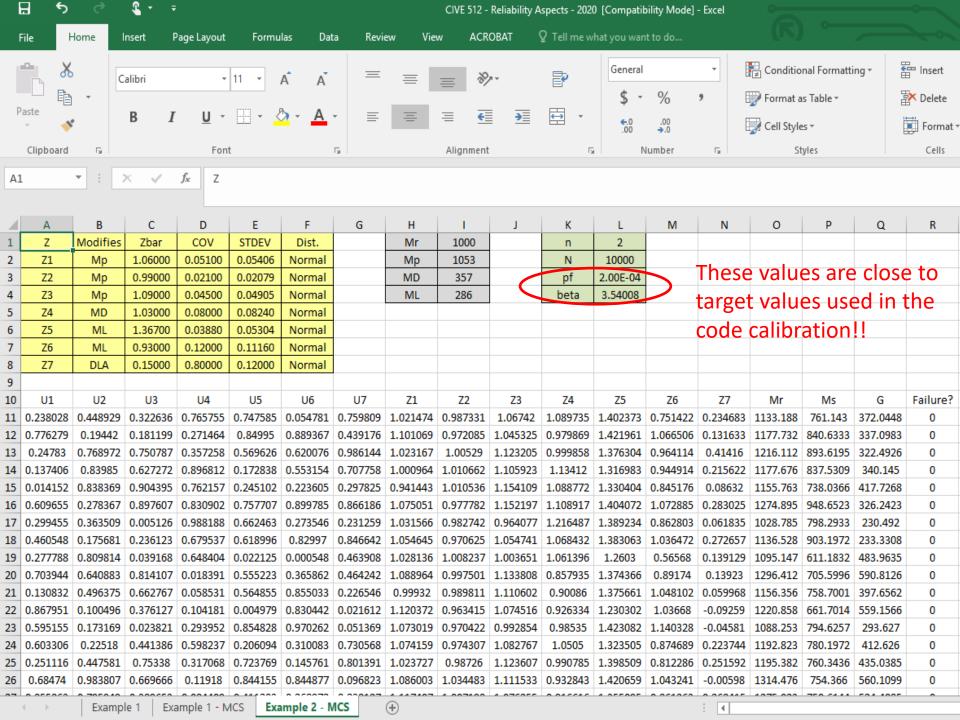
- $M_r = 1000 \text{ kN} \cdot \text{m}$
- $\phi = 0.95$
- $\alpha_L = 1.7$
- DLA = 0.25
- $\alpha_D = 1.1$
- $M_1 = 286 \text{ kN} \cdot \text{m}$
- $M_D = 357 \text{ kN} \cdot \text{m}$

Note: To create this example,  $M_r$ ,  $M_L$ , and  $M_D$  were selected to create a problem where  $M_r$  and  $M_f$  are exactly equal, to show what probability of failure results in this case.

#### \*annual extreme load event

#### Bias factors:

facto-	豆 C	ovz C	ncertainty in:
71	1.06	0:051	Fy
72	0.99	0.021	S
73	1.09	0.045	resistance model
74	1.03	0.08	Mp
75	1,367	0.0388	M L*
76	0.93	0.12	analysis model
77	0.15	08	DLA



## What has all of this got to do with assessment of structures?



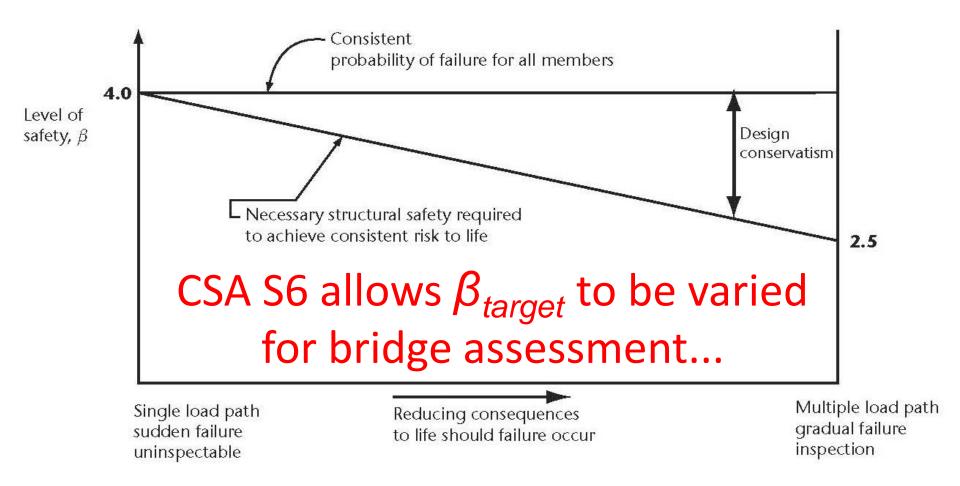


Figure C14.1
Relationship between risk and probability of failure
(See Clause C14.12.1.)

## ...this means load factors can be adjusted favourably in some cases:

Table 14.7 Maximum dead load factors,  $\alpha_D$ 

(See Clause 14.13.2.1.)

Dead load category	Target Reliability Index, $\beta$								
	2.00	2.25	2.50	2.75	3.00	3.25	3.50	3.75	4.00
D1	1.03	1.04	1.05	1.06	1.07	1.08	1.09	1.10	1.11
D2	1.06	1.08	1.10	1.12	1.14	1.16	1.18	1.20	1.22
D3	1.15	1.20	1.25	1.30	1.35	1.40	1.45	1.50	1.55

Table 14.8 Live load factors,  $\alpha_L$ , for normal traffic (Evaluation Levels 1, 2, and 3) for all types of analysis

(See Clause 14.13.3.1.)

	Target reliability index, $\beta$						
Spans	2.50	2.75	3.00	3.25	3.50	3.75	4.00
All Spans	1.35	1.42	1.49	1.56	1.63	1.70	1.77