

Structural Reliability
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Lecture –173
Capacity Demand Component Reliability (Part 21)

We have solved two variable problems two random variable problems in the previous lecture but now we graduate to three variable problems and then to 4 variable problems in this lecture we stick with the same setup the cable intention that we have been looking at in various ways. So this is the problem with the third random variable defined.

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Capacity Demand example with FORM

Structural Reliability
 Lecture 22
 Capacity demand
 component reliability

Example C1: three RV cable reliability problem
 We now consider Y, A and Q to be random, and mutually independent. The properties are as follows:
 The yield strength $Y \sim$ Weibull (mean 38 ksi, COV 15%)
 The cross-sectional area $A \sim N$ (mean 60 sqin, COV 10%)
 The load $Q \sim$ Gumbel (mean 1200 kip, COV 20%)
 Find the reliability of the cable and the checking point values. Employ full distribution transformation.

$$X_1 = Y(u_1, k_Y), X_2 = A(\mu_A, \sigma_A), X_3 = Q(\alpha_Q, u_Q)$$

$$g(\underline{X}) = X_1 X_2 - X_3$$

$$u_i = \Phi^{-1}[F_i(x_i)]$$

$$h(\underline{u}) = u_1 \left[\ln \left(\frac{1}{1 - \Phi(u_1)} \right) \right]^{1/k_Y} \left[\mu_A + \sigma_A u_2 \right] - u_3 + \frac{1}{\alpha_Q} \ln \ln \left(\frac{1}{\Phi(u_3)} \right)$$

$\min_{\underline{u}} h(\underline{u})$
 such that $h(\underline{u}) = 0$

It will be convenient to have x_1, x_2 and x_3 as the decision variables of the minimization problem.



So, let us take a minute to read through the problem. So, we have three mutually independent random variables Y the yield strength A the cross-sectional area and Q the load they are respectively Y is viable a is normal and Q is gumball. Earlier in the last one or two examples we fixed Q at the mean value 1200 k now Q is a gumball random variable. We no longer wish to use the second moment transformation we lose our probabilistic information if the variables are not normal.

So, here we would like to employ the full distribution transformation. So, let us take that up. So,

X_1 let us call $X_1 = Y$ or $Y = X_1$ a X_2 and $Q = X_3$ our limit state therefore becomes $X_1 X_2 - X_3$. So, it is a simple looking limit state now as required the distribution transformation is u is Φ inverse of F of x . So, that is for each of the x 's. So, i going from 1 to 3 basically you can interpret this as Φ of u is F of x .

So that is a distribution wise equivalence. Now if we wanted to find out h express h in terms of u and then find the minimum distance to that h equals zero. So, this is how you would proceed let us do variable by variable. So, X_1 can be given in terms of u_1 in a rather complicated nonlinear fashion which is what you see this is the this is the result of equating the viable that two parameter viable with the normal.

And once you go through the algebra this is what you end up with the viable distribution looks like $1 - \exp(-X^k / u^k)$. So, once you solve it this is the form it comes to the next one would be much simpler because area is normal. So, X_2 in terms of u_2 is a simple linear transformation and then when we come to the gumball we also have a rather complicated looking transformation and that is what you see on this screen the gumball form uses the double exponential form.

So, if you want to go through the algebra and work through step by step this is what you are going to find. So, in the first one and in this third one we see that the normal cdf is implicitly there in the new limit state function. Obviously this is a complicated limit state function and to find the minimum distance to this from u equals 0 is going to be complicated we could do that but there is not a single way of solving a problem.

So, what I want to show you in this one is in this example is why do not we stick to the basic variable space things look more manageable here more intuitive our limit state function is quite simple looking $X_1 X_2 - X_3$ instead of something as complicated when we go to u space. So, why do not we see what happens if we stick to the X space and try to do the same thing. Obviously we cannot optimize we cannot minimize the distance $X_1^2 + X_2^2$

plus X 3 squared that would not be right.

We still have to do u 1 square plus u 2 squared plus u 3 squared we have to minimize that but we can operate from xbase. So, that is our whole power and let us see how far we can go with that.

(Refer Slide Time: 05:53)

Capacity Demand example with FORM

Structural Reliability
Lecture 22
Capacity demand
component reliability

Example C1: three RV cable reliability problem (contd.)

<p>$X_1 = Y(u_1, k_1) \sim$ Weibull $X_2 = A(u_2, \sigma_2) \sim$ Normal $X_3 = Q(\alpha_3, u_3) \sim$ Gumbel $g(\underline{X}) = X_1 X_2 - X_3$</p> <p>$\min r(x_1, x_2, x_3) = u_1^2 + u_2^2 + u_3^2$ subject to: $g(x_1, x_2, x_3) = x_1 x_2 - x_3 = 0$</p> <p>where $u_1 = \Phi^{-1} \left[1 - \exp \left(- (x_1 / u_1)^{k_1} \right) \right]$ $u_2 = (x_2 - \mu_2) / \sigma_2$ $u_3 = \Phi^{-1} \left[\exp \left\{ - \exp \left(\alpha_3 (x_3 - u_3) \right) \right\} \right]$</p>	<p>Gradients of the objective: $\frac{\partial r}{\partial x_i} = 2u_i f_{X_i}(x_i) / \phi(u_i)$ where f_{X_i} = pdf of X_i ϕ = standard normal pdf</p> <p>Gradients of the constraint: $\frac{\partial g}{\partial x_1} = x_2, \frac{\partial g}{\partial x_2} = x_1, \frac{\partial g}{\partial x_3} = -1$</p> <p>solution: $\beta = \sqrt{r^*}$ design point $= (x_1^*, x_2^*, x_3^*)$</p> <table border="0" style="width: 100%;"> <tr> <td style="width: 50%;">Design point:</td> <td style="width: 50%;">Minimum distance point:</td> </tr> <tr> <td>$Y^* = 27.9 \text{ ksi}$</td> <td>$u_1^* = -1.62$</td> </tr> <tr> <td>$A^* = 56.1 \text{ sqin}$</td> <td>$u_2^* = -0.65$</td> </tr> <tr> <td>$Q^* = 1565 \text{ kip}$</td> <td>$u_3^* = 1.43$</td> </tr> <tr> <td></td> <td>$\beta = 2.26$</td> </tr> </table>	Design point:	Minimum distance point:	$Y^* = 27.9 \text{ ksi}$	$u_1^* = -1.62$	$A^* = 56.1 \text{ sqin}$	$u_2^* = -0.65$	$Q^* = 1565 \text{ kip}$	$u_3^* = 1.43$		$\beta = 2.26$
Design point:	Minimum distance point:										
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For comparison, we will solve the same problem later with MCS. The answer is 0.0158 (equivalent beta 2.15) from 1 million simulations

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So, this is our problem statement we have g of X equals X 1 X 2 minus X 3 and now we want to minimize the function of X 1 X 2 X 3 which is our distance in the standard independent channel normal space u 1 square plus u 2 square plus u 3 square. Obviously there is a relation between the x's and the u's the constraint is also put in the basic variable space. So, we have a nice looking constraint and equality constraint X 1 X 2 minus X 3 is equal to 0.

And here are the function relations between the u's and the x's. So, as I was saying that this is the weibull to normal transformation this is the normal to normal transformation for u 2 and for u 3 this is the gumball to normal transformation. So, this is our problem set up and but we would need if we could provide it would be helpful we would need the gradients of the objective if we go for a gradient-based algorithm we would also need the gradient of the constraint if we can then that will be great the problem becomes more efficient to solve.

So, let us look at it how **how** to work this. So, the objective function r which is the distance to the

the distance to h in the u space. So, that u_1 squared plus u_2 squared plus u_3 squared the first derivative of that the first partial the gradient of that in respect to the axis would be we can use the chain rule of differentiation. So, twice u and then $\frac{\partial u}{\partial x}$. So, $\frac{\partial u}{\partial X}$ for that we will use the distribution to distribution equivalence.

So, that term that you see the density of X divided by the density of u that basically is the product of equating the **the** two cdfs. So, ϕ of u is cdf of x . So, we differentiate the left hand side with respect to X . So ϕ of u with respect to u first then partial u partial X and on the right hand side it is the partial of the cdf with respect to x . So, that gives me the density of x . So, that is how this twice u pdf of X divided by the pdf of u that form comes in the final solution.

And the gradient of the constraint in this particular case it is very simple the first one is X_2 the second one is X_1 and the third one is -1 you just differentiate g and that is what you get. Now our solution would be if we can find the minimum of r our beta would be the square root of that minimum. So, root of r^* and the design point automatically since we are operating in X space would be X_1^* X_2^* X_3^* .

If you want to find u_1^* we have to use the map that you see on the left. So, u_1 is fine inverse of $1 - \exp(-u)$ so on. So, that expression has to be used if we want to find the u_1^* the u_2^* and u_3^* 1 by 1. Here is the answer first and then I will walk you through the steps that I have used in obtaining these answers and that is basically employing a MATLAB program and it is a good way to understand how to set up such problems in a standard platform such as MATLAB.

So, the answer we get the design point is Y^* is about 28 ksi the a^* is about 56 square inch and the design value of the load is about 1565 kilo pounds. The corresponding minimum distance points in u space are as you see minus. So, 1.62 standard deviations below the mean for the first one about 60% standard deviation below the mean for the second one and about one and half standard deviations above the mean for the third one and that gives me a beta of about 2.26.

Now how accurate is this 2.26 beta of 2.26 we have mentioned that FORM is approximate. So, to compare we later have solved this problem using Monte Carlo simulations with a good number of trials and we get the value of P_f about 0.0158 which gives an equivalent beta of about 2.15. So, it seems that FORM is overestimating reliability or underestimating the P_f and this is something we will remember when we take up a discussion of SORM later in this lecture.