

## Project #4 Model

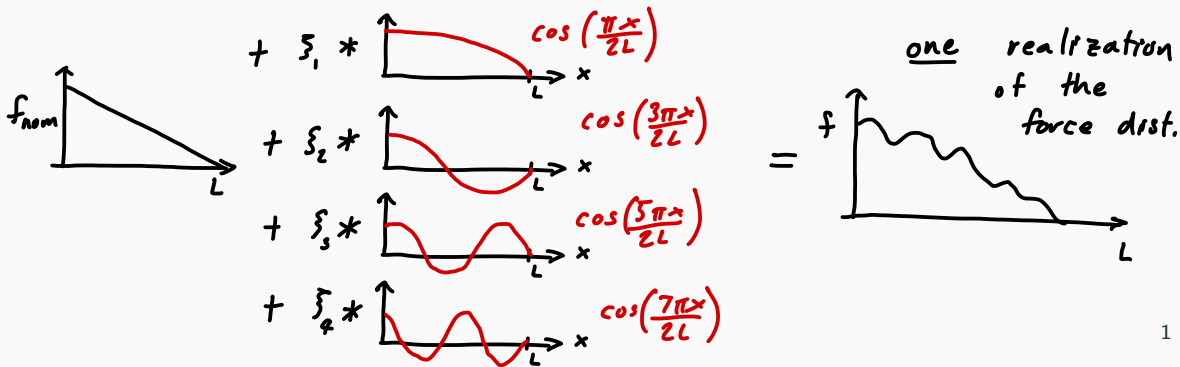
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\* Use Gauss-Hermite quadrature,  
not Gauss-Legendre (as in Assign. 21)

## First, Let's Discuss the Model Adopted For the Uncertain Load

The loading on the wing is modelled as  $f(x, \xi) = f_{\text{nom}}(x) + \delta_f(x, \xi)$  where the probabilistic perturbation has the form

$$\delta_f(x, \xi) = \sum_{n=1}^4 \xi_n \cos\left(\frac{(2n-1)\pi x}{2L}\right), \quad \text{with } \xi_n \sim \mathcal{N}\left(0, \frac{f_{\text{nom}}(0)}{10n}\right).$$



## Example: Statistics For A Function Of Multivariate Random Variables

Suppose we have two random variables,  $\xi_1$  and  $\xi_2$ , that are normally distributed as follows:

$$\xi_1 \sim \mathcal{N}(2, 1), \quad \text{and} \quad \xi_2 \sim \mathcal{N}(-1, 4).$$

Consider the function

$$f(\xi_1, \xi_2) = \frac{(\xi_1 - 2)^2}{1 + \xi_2^2} \quad \leftarrow \text{like stress at one node}$$

Let's estimate the mean, or expected value, of  $f$ .

## We Will Need the Gauss-Hermite Quadrature Rules

$$\int_{-\infty}^{\infty} f(u) e^{-u^2} du \approx \sum_{i=1}^m w_i f(u_i).$$

Recall that we need to adjust the weights and quadrature locations, because  $e^{-u^2}$  is not quite the same as a Gaussian probability density.

$$u = \frac{\xi - \mu}{\sqrt{2} \sigma} \Rightarrow \xi = \sqrt{2} \sigma u + \mu, \quad \frac{d\xi}{du} = \sqrt{2} \sigma$$


$$\begin{aligned} \frac{1}{\sigma \sqrt{2\pi}} \int_{-\infty}^{\infty} f(\xi) e^{-\frac{(\xi - \mu)^2}{2\sigma^2}} d\xi &= \frac{1}{\cancel{\sigma} \cancel{\sqrt{2\pi}}} \int_{-\infty}^{\infty} f(\sqrt{2} \sigma u + \mu) e^{-u^2} \cancel{\sqrt{2} \sigma} du \\ &= \frac{1}{\sqrt{\pi}} \int_{-\infty}^{\infty} f(\sqrt{2} \sigma u + \mu) e^{-u^2} du \end{aligned}$$

## Make the Substitution Into the Integral For the Mean of $f$

Adjusting the weights and quadrature locations we get

$$\begin{aligned}\mu_f &= E(f) = \int_{-\infty}^{\infty} f(\xi) P(\xi; \mu, \sigma) d\xi \\ &= \int_{-\infty}^{\infty} f(\xi) \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(\xi-\mu)^2}{2\sigma^2}} d\xi = \frac{1}{\sqrt{\pi}} \int_{-\infty}^{\infty} f(\sqrt{2}\sigma u + \mu) e^{-u^2} du \\ &\approx \frac{1}{\sqrt{\pi}} \sum_{i=1}^m w_i f(\sqrt{2}\sigma \xi_i + \mu).\end{aligned}$$

## The 2D Integral for $E(f)$ Can Be Written As An Iterated Integral

$$\begin{aligned} E(f) &= \int_{\xi_2=-\infty}^{\infty} \int_{\xi_1=-\infty}^{\infty} f(\xi_1, \xi_2) P(\xi_1; \mu_1, \sigma_1) P(\xi_2; \mu_2, \sigma_2) d\xi_1 d\xi_2 \\ &= \int_{\xi_2=-\infty}^{\infty} g(\xi_2) P(\xi_2; \mu_2, \sigma_2) d\xi_2 \end{aligned}$$


where

$$g(\xi_2) = \int_{\xi_1=-\infty}^{\infty} f(\xi_1, \xi_2) P(\xi_1; \mu_1, \sigma_1) d\xi_1$$

## In the Implementation, Each Dimension Requires A Loop

```
1  % define function and Gaussian variables
2  f = @(xi1,xi2) (xi1-2.0).^2./(1.0 + xi2.^2);
3  mu1 = 2.0; sigma1 = 1; mu2 = -1.0; sigma2 = 4.0;
4  % using a 3 point Gauss-Hermite quadrature
5  xi = [-1.22474487139; 0.0; 1.22474487139];
6  wts = [0.295408975151; 1.1816359006; ...
7         0.295408975151]./sqrt(pi); % adjusted weights!
8  mean_f = 0.0;
9  for i1 = 1:size(xi,1)
10     pt1 = sqrt(2)*sigma1*xi(i1) + mu1;
11     for i2 = 1:size(xi,1)
12         pt2 = sqrt(2)*sigma2*xi(i2) + mu2;
13         mean_f = mean_f + wts(i1)*wts(i2)*f(pt1,pt2);
14     end
15 end
```

Handwritten annotations illustrating the nested loop structure for the 2D Gaussian quadrature:

- For  $i_1$  (loop 1)
- For  $i_2$  (loop 2)
- Calculation of  $pt1 = \sqrt{2} \cdot \sigma_1 \cdot xi(i1) + \mu_1 = \xi_1^{(i1)}$
- Calculation of  $pt2 = \sqrt{2} \cdot \sigma_2 \cdot xi(i2) + \mu_2 = \xi_2^{(i2)}$
- Update of  $mean\_f$  using the function value  $f(pt1, pt2)$  and weights  $wts(i1)$  and  $wts(i2)$ .
- Final calculation:  $mean\_f2 = mean\_f2 + wts(i1) * wts(i2) * f(pt1, pt2)^2$

## What About Computing the Standard Deviation?

Here is a useful formula for computing the standard deviation

$$\sigma = \sqrt{E(f^2) - E(f)^2}$$

- During the quadrature loop, compute the expected value of the function squared at the same time you are computing the mean
- After the loop, use the above formula to get the standard deviation.



