



筑波大学
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Bayesian and Structure Mechanics ベイズ統計と構造力学

フロンティア工学研究グループ 01
LAFEE

システム情報系 助教

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Bayes Theorem

$$\underset{\substack{\text{posterior distribution} \\ \text{事後分布}}}{p(\boldsymbol{H}|\boldsymbol{D})} = \frac{\overset{\substack{\text{likelihood} \\ \text{尤度}}}{p(\boldsymbol{D}|\boldsymbol{H})} \times \overset{\substack{\text{prior distribution} \\ \text{事前分布}}}{p(\boldsymbol{H})}}{p(\boldsymbol{D})}$$

$\boldsymbol{H}$: Parameters
 $\boldsymbol{D}$: Data

- If we know the relationship between “What we can observe ($\boldsymbol{D}$)” and “What we’d like to know ($\boldsymbol{H}$)”, we can estimate “What we’d like to know ($\boldsymbol{H}$)” from “What we can observe ($\boldsymbol{D}$)” by Bayes Theorem
- もし、「 $\boldsymbol{D}$: 観測しやすいもの」と「 $\boldsymbol{H}$: 知りたいもの」の関係が分かっているならば、ベイズ理論により「 $\boldsymbol{D}$: 観測しやすいもの」から「 $\boldsymbol{H}$: 知りたいもの」を推定できる

Example

- Estimate “**Young Modulus**” of a Beam from observed natural frequency.



- **Material Parameters** (Young Modulus: E and Density: ρ) follow logarithm normal distributions.
- **Member Parameters** (Length: L , Width: b , Height: h) follow logarithm normal distributions.
- **Observation Noise** into the Natural Frequency follows a normal distribution.

Simulating Data

I	A
5.08E-10	0.001508

- ❑ In this time, data is **simulated** instead of **experiment**.
 - ❑ Let's generate Observed Natural Frequency: f by **Monte Carlo** Simulation
- ❑ **Setting** Statistical characteristics of Parameters

	Material		Member		
	E	ρ	b	h	L
	[N/m ²]	[kg/m ³]	[m]	[m]	[m]
Average	2.05×10^{11}	7850	0.7501	0.00201	1.829
S.D.	2.00×10^9	5	0.0015	0.00001	0.001



	Member	
	I	A
	[m]	[m]
Average	5.08E-10	0.001508

Basis of Statistics

□ X follows Logarithm Normal Distribution

□ X 's Average: μ_X , Standard Deviation: σ_X

$$p(X) = \frac{1}{\sqrt{2\pi}\sigma} \exp \left[-\frac{(\ln X - \mu)^2}{2\sigma^2} \right]$$

Average and
Standard Deviation

$$\mu_X = \exp \left[\mu + \frac{\sigma^2}{2} \right]$$

$$\sigma_X = \exp[2\mu + \sigma^2] (\exp \sigma^2 - 1)$$

Parameters of Log. Norm. Dist.

$$\sigma = \sqrt{\ln \left[1 - \left(\frac{\sigma_X}{\mu_X} \right)^2 \right]}$$

$$\mu = \ln \mu_X - \frac{\sigma^2}{2}$$

	E
	[N/m ²]
Average	2.05×10^{11}
S.D.	2.00×10^9

	E
	[N/m ²]
μ	26.046
σ	0.010

Monte Carlo Simulation

□ Generate X

Sample	E	ρ	I	A	b	h	L	Frequency f
1	2.10.E+11	7847.731	5.087E-10	1.509.E-03	7.502.E-01	2.011.E-03	1.829414	1.41
2	2.07.E+11	7856.437	5.151E-10	1.514.E-03	7.492.E-01	2.021.E-03	1.828602	1.41
3	2.04.E+11	7852.566	5.103E-10	1.515.E-03	7.536.E-01	2.010.E-03	1.828302	1.39
4	2.08.E+11	7846.214	4.969E-10	1.498.E-03	7.509.E-01	1.995.E-03	1.828626	1.39
5	2.08.E+11	7850.578	5.101E-10	1.511.E-03	7.504.E-01	2.013.E-03	1.829197	1.40
6	2.06.E+11	7852.279	5.114E-10	1.513.E-03	7.510.E-01	2.014.E-03	1.830040	1.40
7	2.05.E+11	7840.501	5.08E-10	1.509.E-03	7.509.E-01	2.010.E-03	1.827074	1.40
8	2.08.E+11	7851.889	5.013E-10	1.499.E-03	7.483.E-01	2.003.E-03	1.830499	1.40
9	2.05.E+11	7846.629	5.015E-10	1.501.E-03	7.498.E-01	2.002.E-03	1.829832	1.39
10	2.07.E+11	7853.577	5.083E-10	1.508.E-03	7.497.E-01	2.011.E-03	1.827193	1.40

□ Natural Frequency

$$f = \left(\frac{\pi}{L}\right)^2 \sqrt{\frac{EI}{\rho A}}$$

Observing Data

□ Simulating Observed Data including Noise

Freq.	Noise	Observed Freq.
1.41	-0.000575	1.408
1.41	-0.000463	1.407
1.39	-0.001111	1.390
1.39	-0.001155	1.391
1.40	-0.000788	1.404
1.40	-0.000414	1.396
1.40	0.000034	1.395
1.40	-0.002033	1.394
1.39	-0.000094	1.386
1.40	-0.000160	1.402

Bayes Theorem

□ Bayes Theorem

$$p(E|f) = \frac{p(f|E) \times p(E)}{p(f)}$$

□ Likelihood:

$$p(f|E) = \frac{1}{\sqrt{2\pi}s} \exp \left[-\frac{\left(f - \left(\frac{\pi}{L} \right)^2 \sqrt{\frac{EI}{\rho A}} \right)^2}{2s^2} \right]$$

- If $f = \left(\frac{\pi}{L} \right)^2 \sqrt{\frac{EI}{\rho A}}$, Likelihood: $p(f|E)$ returns the maximum value.
- The variance: s^2 can be set freely

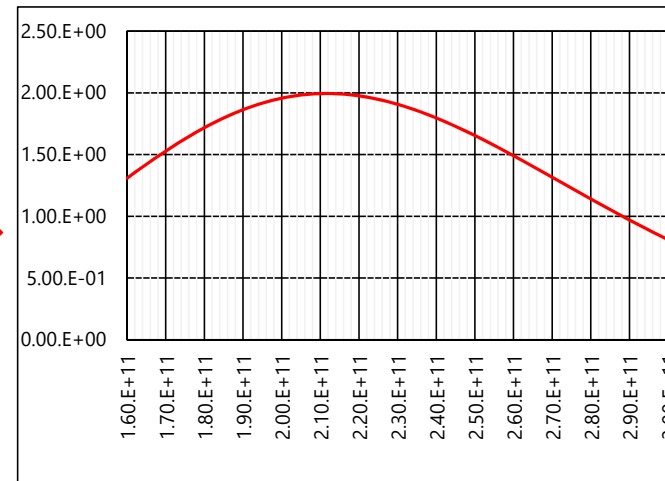
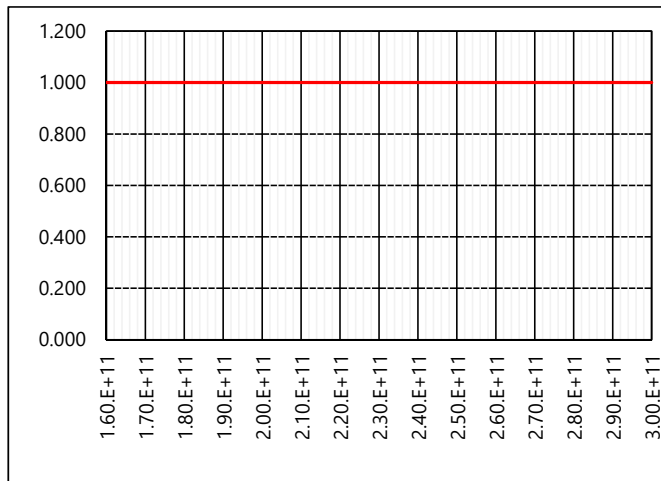
Estimation of Parameter

□ Step One

- From the Prior Distribution, we can estimate the distribution

$$\frac{p(f|E) \times p(E)}{p(f)} = p(E|f)$$

First Observed data:
 $f=1.408$



Estimation of Parameter

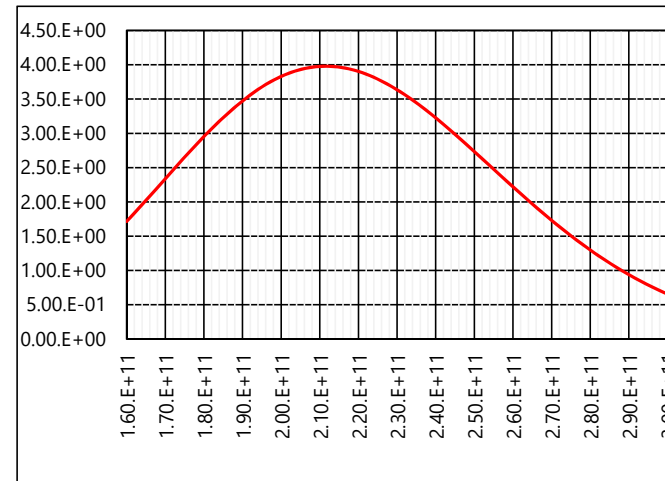
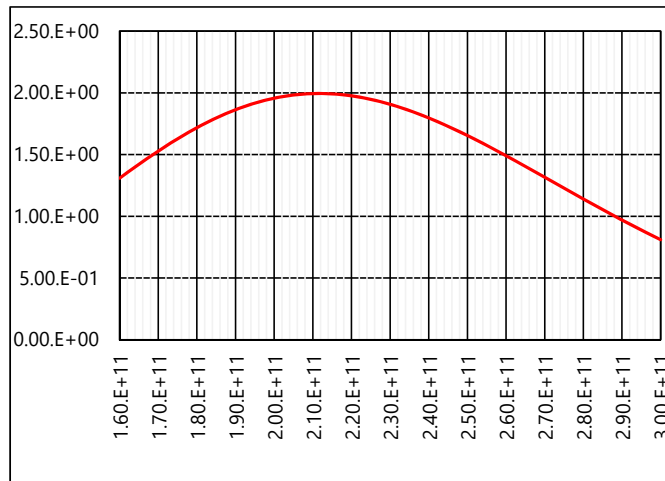
□ Step: 2

- From the Prior Distribution, we can estimate the distribution

The last **posterior** dist.
is taken as the new **prior** dist

$$\frac{p(f|E) \times p(E)}{p(f)} = p(E|f)$$

2nd Observed data:
 $f=1.407$



Estimation of Parameter

□ Repeat the procedure

□ By the **production** of Likelihoods into which data: f are substituted

$$p(E|f) = \frac{p(f_n|E) \times \dots \times p(f_2|E) \times p(f_1|E) \times p(E)}{C}$$

