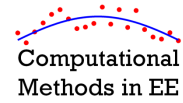




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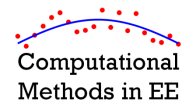


Topic 1 – Errors in Computation

EE 4386/5301 Computational Methods in EE

Outline

- Errors in Computation
- Uncertainty Analysis



Errors in Computation

Types of Errors

Truncation Error – Arises when approximations are used instead of performing exact operations.

$$\frac{\partial f(x)}{\partial x} \approx \frac{f(x_b) - f(x_a)}{b - a} \quad \sqrt{1+x} \approx 1 + \frac{x}{2}$$

Round-Off Error – Arises when limited significant figures are used to represent exact numbers

$$\pi, e, \sqrt{3}$$

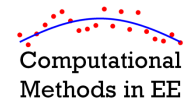
```
>> 0.3/0.1 - 3
ans =
-4.4409e-16
```

Human Error – Arises when people make mistakes.

$$1+1=3$$

Quantifying Error

Significant Figures



Very often quantities are limited to some number of digits.

This can happen because a computer cannot store any more digits or because the measurement is not accurate out to that many digits.

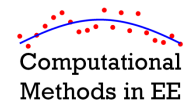
Rules: 1. All non-zero digits are significant
2. Zeros that do nothing but place the decimal point are not significant.

How many significant digits?

1.234	4 significant digits
0.00545	3 significant digits
32100	3 or 5 significant digits (ambiguous)
0.0500	3 significant digits

Quantifying Error

True Error



True Error – Actual error in a measurement.

$$\Delta = (\text{true error}) = (\text{true value}) - (\text{approximate value})$$

Example

The speed of light is exactly 299,792,458 m/s. An experiment was performed that measured the speed of light to be 299,792,445 m/s.

$$(\text{true value}) = 299,792,458 \text{ m/s}$$

$$(\text{approximate value}) = 299,792,445 \text{ m/s}$$

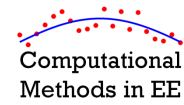
$$\Delta = (\text{true error}) = (299,792,458 \text{ m/s}) - (299,792,445 \text{ m/s}) = 13 \text{ m/s}$$

Notes

- It is very rare to know the true value or true error of a measurement.
- Unrealistic because the true value must be known, but good for analyzing how errors propagate through computations.
- True error does not convey the relative severity of the error.

Quantifying Error

Relative Error



Relative Error – Actual error in a measurement relative to the size of the measurement.

$$\varepsilon_t = (\text{relative error}) = \frac{(\text{true error})}{(\text{true value})}$$

Example

The speed of light is exactly 299,792,458 m/s. An experiment was performed that measured the speed of light to be 299,792,445 m/s.

$$(\text{true error}) = 13 \text{ m/s}$$

$$(\text{true value}) = 299,792,458 \text{ m/s}$$

$$(\text{relative error}) = \frac{13 \text{ m/s}}{299,792,458 \text{ m/s}} = 4.33 \times 10^{-8} = 0.0000000433 = 0.00000433\%$$

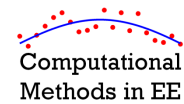
Pretty good!

Notes

This is a much better metric to determining the severity of the error.

Quantifying Error

Approximate Relative Error



Approximate Relative Error – We rarely know the true value of something so we must calculate errors based on our approximation and approximate error.

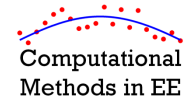
$$\varepsilon_a = (\text{approximate relative error}) = \frac{(\text{approximate error})}{(\text{approximate value})}$$

Note

We can be sure that our number is correct to at least n significant digits if:

$$|\varepsilon_a| = (0.5 \times 10^{2-n}) \%$$

Error Propagation



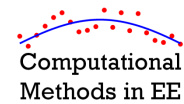
In computation, we are forced to calculate things from quantities with some error. This results in some degree of error in our computations. error.

$$\tilde{x} = x + \Delta x$$

How do we calculate the error Δx of $f(\tilde{x})$?

Error Propagation

Taylor Series



The Taylor series is very often used to determine error associated with truncation.

$$f(x) = f(a) + f'(a) \cdot (x-a) + \frac{f''(a)}{2!} (x-a)^2 + \frac{f'''(a)}{3!} (x-a)^3 + \dots = \sum_{k=0}^{\infty} \frac{f^{(k)}(a)}{k!} (x-a)^k$$

This is an infinite array. Suppose we retain only the first $n+1$ terms?

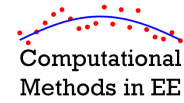
$$f(x) = f(a) + f'(a) \cdot (x-a) + \frac{f''(a)}{2!} (x-a)^2 + \frac{f'''(a)}{3!} (x-a)^3 + R_n$$

We have lumped all remaining terms into R_n .

$$R_n = \int_a^x \frac{(x-t)^n}{n!} f^{(n+1)}(t) dt$$

Error Propagation

Propagating Error



First, we rewrite the Taylor series in terms of our measured value.

$$f(x) = f(\tilde{x}) + f'(\tilde{x})(x - \tilde{x}) + \frac{f''(\tilde{x})}{2!}(x - \tilde{x})^2 + \dots$$

Second, we bring $f(\tilde{x})$ to the left side of the equation.

$$f(x) - f(\tilde{x}) = f'(\tilde{x})(x - \tilde{x}) + \frac{f''(\tilde{x})}{2!}(x - \tilde{x})^2 + \dots$$

This is the definition of error Δf of $f(\tilde{x})$.

These are Δx .

Third, we ignore all higher order terms to get an expression for error.

$$f(x) - f(\tilde{x}) = f'(\tilde{x})(\Delta x) + \frac{f''(\tilde{x})}{2!}(\Delta x)^2 + \dots$$

Generalization to Two Variables

$$\Delta f = \Delta x \left| \frac{\partial f}{\partial x} \right| + \Delta y \left| \frac{\partial f}{\partial y} \right|$$

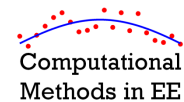
$$\text{error} = \Delta f = f'(\tilde{x}) \Delta x$$

Topic 1 -- Errors in Computation

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Error Propagation

Examples



Example 1

Suppose we have $\tilde{x} = 2.0$ with an error of $\Delta x = 0.05$.

What is the error Δf if $f(\tilde{x}) = \tilde{x}^2$?

$$\Delta f(\tilde{x}) = f'(\tilde{x}) \cdot \Delta x = 2\tilde{x} \cdot \Delta x = 2(2.0) \cdot 0.05 = 4 \cdot 0.05 = \underline{0.2}$$

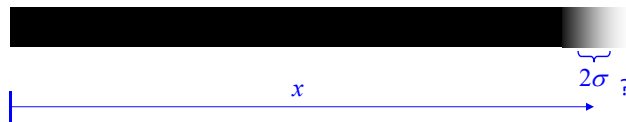
Topic 1 -- Errors in Computation

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Uncertainty Analysis

What is Uncertainty?

How long is this object?



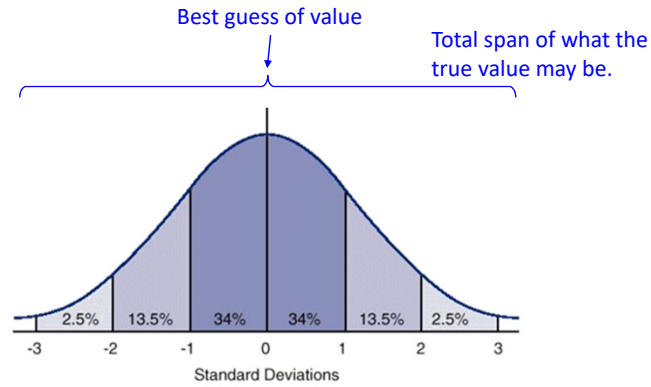
We write our measurement as $x \pm \sigma$

Since the measurement error is “blurred,” we are not entirely certain about the error. The uncertainty σ is a statistically derived quantity.

Notes

- This is a more realistic way to treat error because we do not need to know the true value.
- This is used to predict errors in computations.

How to Interpret σ



σ is one standard deviation.

There is a 68% chance that the error will be less than 1σ .

There is a 32% chance that the error will be greater than 1σ .

Uncertainty Analysis

Suppose we have some number of parameters, each with an associated uncertainty.

$$x \pm \sigma_x \quad y \pm \sigma_y \quad z \pm \sigma_z$$

Now suppose we calculate a new quantity from these.

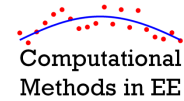
$$f(x, y, z)$$

What is the uncertainty σ_f of $f(x, y, z)$?

$$\sigma_f^2 = \left(\sigma_x \frac{\partial f}{\partial x} \right)^2 + \left(\sigma_y \frac{\partial f}{\partial y} \right)^2 + \left(\sigma_z \frac{\partial f}{\partial z} \right)^2$$

We call this propagating uncertainty through a calculation.

Table of Uncertainty Calculations

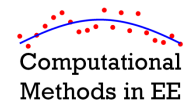


Function	Uncertainty
$f = ax$	$\sigma_f = a\sigma_x$
$f = ax \pm by$	$\sigma_f^2 = (a\sigma_x)^2 + (b\sigma_y)^2$
$f = xy$	$(\sigma_f/f)^2 = (\sigma_x/x)^2 + (\sigma_y/y)^2$
$f = x/y$	$(\sigma_f/f)^2 = (\sigma_x/x)^2 + (\sigma_y/y)^2$
$f = x^{\pm b}$	$\sigma_f/f = b\sigma_x/x$
$f = \ln(\pm bx)$	$\sigma_f = b\sigma_x/x$
$f = \log x$	$\sigma_f = b\sigma_x/(x \ln 10)$
$f = e^{\pm bx}$	$\sigma_f/f = b\sigma_x$
$f = a^{\pm bx}$	$\sigma_f/f = b\sigma_x \ln a$
$f = \sin x$	$\sigma_f = \sigma_x \cos x$
$f = \cos x$	$\sigma_f = \sigma_x \sin x$
$f = \tan x$	$\sigma_f = \sigma_x / \cos^2 x$
$f = \sin^{-1}(x)$	$\sigma_f^2 = \sigma_x^2 / (1-x^2)$
$f = \cos^{-1}(x)$	$\sigma_f^2 = \sigma_x^2 / (1-x^2)$
$f = \tan^{-1}(x)$	$\sigma_f = \sigma_x / (1+x^2)$

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Uncertainty Through Multiple Calculations



Suppose we wish to find the uncertainty σ_f when multiple calculations are involved.

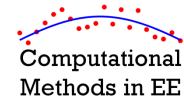
$$f(a, b, c) = \ln(a + bc) \quad \text{Given } \sigma_a, \sigma_b, \text{ and } \sigma_c$$

	Value	Uncertainty
1	bc	$\left(\frac{\sigma_{bc}}{bc}\right)^2 = \left(\frac{\sigma_b}{b}\right)^2 + \left(\frac{\sigma_c}{c}\right)^2$
2	$a + bc$	$\sigma_{a+bc} = \sigma_a^2 + \sigma_{bc}^2$
3	$\ln(a + bc)$	$\sigma_{\ln(a+bc)} = \frac{\sigma_{a+bc}}{a + bc}$

Topic 3b -- Numerical Linear Algebra

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Example 1



What is the uncertainty of a decibel quantity?

$$10 \log_{10}(x \pm \sigma_x)$$

Solution

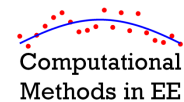
$$\sigma_f^2 = \left(\sigma_x \frac{\partial f}{\partial x} \right)^2 \quad f(x) = 10 \log_{10}(x)$$

$$\frac{\partial f}{\partial x} = \frac{\partial}{\partial x} [10 \log_{10}(x)] = 10 \left[\frac{\partial}{\partial x} \log_{10}(x) \right] = 10 \frac{1}{x \ln 10} = \frac{10}{x \ln 10}$$

$$\sigma_f^2 = \left(\sigma_x \frac{10}{x \ln 10} \right)^2$$

$$\sigma_f = 4.3429 \frac{\sigma_x}{x}$$

Example 2



The signal-to-noise ratio (SNR) of a system is 50 ± 1.2 . What is the SNR in decibels along with the uncertainty?

Solution

The SNR in decibels is

$$10 \log_{10}(50) = 16.99 \text{ dB}$$

The uncertainty is calculated using the equation derived in the previous example.

$$\sigma_f = 4.3429 \frac{\sigma_x}{x} = 4.3429 \frac{1.2}{50} = 0.1$$

The final answer is

$$\text{SNR} = 16.99 \pm 0.1 \text{ dB}$$

Example 3

What is the uncertainty of the sum of two quantities?

$$x + y$$

Solution

$$\sigma_f^2 = \left(\sigma_x \frac{\partial f}{\partial x} \right)^2 + \left(\sigma_y \frac{\partial f}{\partial y} \right)^2 \quad f = x + y$$

$$\sigma_f^2 = \left[\sigma_x \frac{\partial}{\partial x}(x + y) \right]^2 + \left[\sigma_y \frac{\partial}{\partial y}(x + y) \right]^2 = [\sigma_x \cdot 1]^2 + [\sigma_y \cdot 1]^2$$

$$\sigma_f = \sqrt{\sigma_x^2 + \sigma_y^2}$$

Example 4

The height of person 1 was measured to be 6.0 ± 0.1 ft.

The height of person 2 was measured to be 5.5 ± 0.1 ft.

If person 2 stands on the head of person 1, what is the total height and the uncertainty of the total height?



Solution

Total height

$$\begin{aligned} h &= h_1 + h_2 \\ &= (6.0 \text{ ft}) + (5.5 \text{ ft}) \\ &= 11.5 \text{ ft} \end{aligned}$$

Uncertainty

$$\begin{aligned} \sigma_h &= \sqrt{\sigma_1^2 + \sigma_2^2} \\ &= \sqrt{(0.1 \text{ ft})^2 + (0.1 \text{ ft})^2} \\ &= 0.14 \text{ ft} \end{aligned}$$

Final Answer

$$h = 11.5 \pm 0.14 \text{ ft}$$