



# Modeling and Simulation

From observation via implementation  
to computation and back...

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# Part Seven

## Error Propagation



# Sources of Errors

- Unlike symbolic calculations: Errors are *inherent* in numerical calculations
  - Modeling
  - Data
  - Computational
    - Round-off
    - Truncation
    - Bugs
    - Hardware errors



# Computational error is a necessary evil!

- Round-off      $9.999 \cdot 10^9$  and  $9.999 \cdot 10^{-9}$ 
  - Sum: 22 digits: 9999000000,0000000009999
  - Prod: 8 digits: 99.980001

- Truncation

$$e^x \approx \sum_{n=0}^{100} \frac{x^n}{n!}$$

or for example:

$$f'(x) \approx \frac{f(x+h) - f(x-h)}{2h}$$



# Definitions

- $\text{Error} := \text{Approximation} - \text{RealValue}$
- $\text{Relative Error} := |\text{Error}/\text{RealValue}|$
- $\Rightarrow \text{Approximation} := \text{RealValue} (1 + \text{RE})$

$$\tilde{x} = x(1 + \varepsilon)$$

# More definitions

$f(x)$  : Maps Input to Output

Propagated data error :  $f(\tilde{x}) - f(x)$

Even worse since we have  $\tilde{f}(\tilde{x})$

So we can calculate the *total* error:



# Total Error

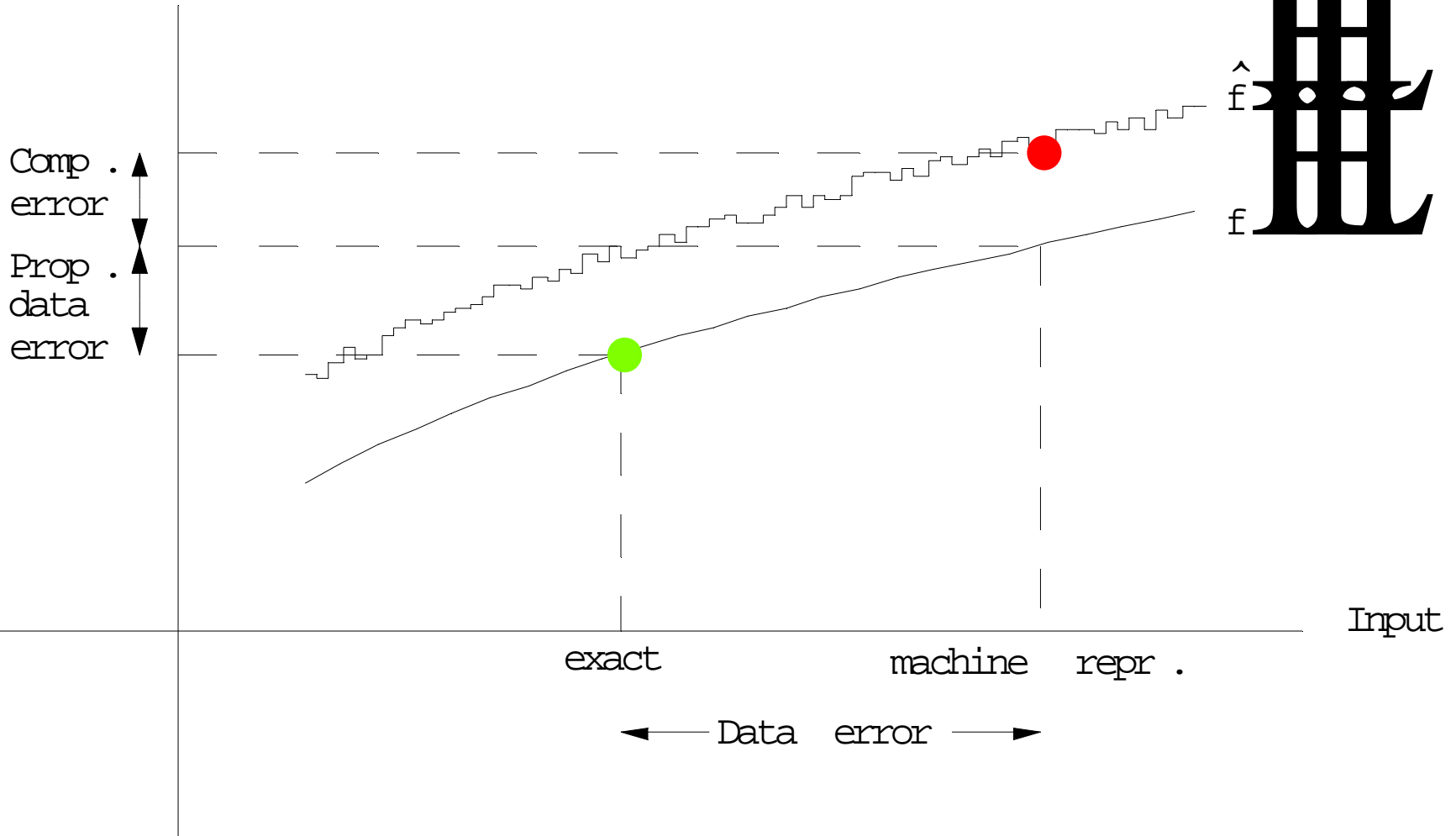
$$\begin{aligned} \text{Total error} &\approx |\tilde{f}(\tilde{x}) - f(x)| = |\tilde{f}(\tilde{x}) - f(\tilde{x}) + f(\tilde{x}) - f(x)| \\ &\leq |\tilde{f}(\tilde{x}) - f(\tilde{x})| + |f(\tilde{x}) - f(x)| \end{aligned}$$

$\Uparrow$  *Computational Error (Implementation)*

$\Uparrow$  *Propagated data error (Condition)*



Output



# Conditioning

- Example: Chaotic systems: Extreme sensitive to initial conditions
  - Gilpin's model of Predator/Prey
  - Weather predictions (< 15 days...)

$$\text{Sensitivity} \equiv \text{Condition Number} \equiv \frac{|\text{Relative error in result}|}{|\text{Relative error in data}|} \gg 1$$



- Ill-conditioned : Sensitive
- Well-Conditioned : in-Sensitive

## Calculate Condition Number of Unary Operation:

- Assume  $f(x)$  is linear (else use Taylor Expansion)
- Assume *Exact* calculation
- Assume approximation result of  $f(x)$  is due to use of approximated  $x$
- Use:  $\text{Appr} = \text{True} (1 + \text{RelError})$



Consequence:

$$\tilde{f}(x) \equiv f(x) (1 + \varepsilon_{f(x)}) \equiv f(x(1 + \varepsilon_x))$$

Therefore we have:

$$f(x)\varepsilon_{f(x)} = f(x + x\varepsilon_x) - f(x)$$

$$\Rightarrow \frac{\varepsilon_{f(x)}}{\varepsilon_x} \equiv \text{Sensitivity} = \frac{f(x + x\varepsilon_x) - f(x)}{f(x)\varepsilon_x}$$

So if derivative of  $f(x)$  in  $x$  exists than:



$$\frac{f(x + x\varepsilon_x) - f(x)}{x\varepsilon_x} \xrightarrow{\varepsilon_x \rightarrow 0} f'(x)$$

Substitute:  $\frac{\varepsilon_{f(x)}}{\varepsilon_x} \rightarrow \frac{xf'(x)}{f(x)}$  for  $\varepsilon_x \rightarrow 0$

Then the Condition Number is a measure of the relative sensitivity of  $f(x)$  for small variations in  $x$ :

$$\left| \frac{\mathcal{E}_{f(x)}}{\mathcal{E}_x} \right| \equiv \text{Condition number of } f(x) \text{ in } x$$

Example I: Condition Number of  $e^x$

Relative derivative is  $\frac{x \cdot e^x}{e^x} = x \Rightarrow \text{CN} = |x|$

Consequence I: Exponential function is

sensitive for  $|x| \gg 1$

insensitive for  $|x| \ll 1$



## Example II : Condition Number of $\ln(x)$

$$\text{Relative derivative of } \ln(x) \equiv \frac{x(\ln(x))'}{\ln(x)} = \frac{1}{\ln(x)}$$

$$\Rightarrow \text{CN} = \left| \frac{1}{\ln(x)} \right|$$

Consequence :  $\ln(x)$  is sensitive for  $x \approx 1$

