

Approximation

- We will use our computers to approximate "stuff"
e.g. integrals, derivatives, solutions to differential equation
- We say "approximation" because there is (almost) always some deviation from exactness

In principle:

$$\hat{O}\hat{O} = \hat{O}\hat{O} + \text{error}$$

↑
exact
↑
approximation

error = sum of many terms

$$= \hat{O}\hat{O} + \hat{O}\hat{O} + \hat{O}\hat{O} + \dots$$

↑
largest = leading error term

$$= \text{coefficient} * \Delta x^n$$

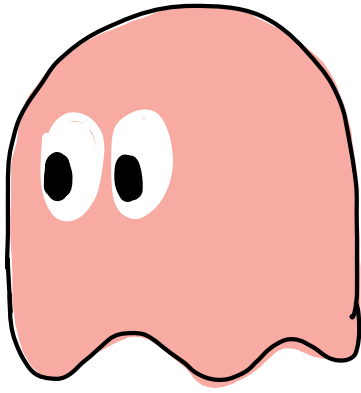
↑
measure of granularity of approximation
← power n

$$\Delta x = \frac{L}{N}$$

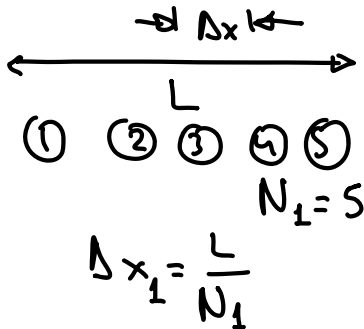
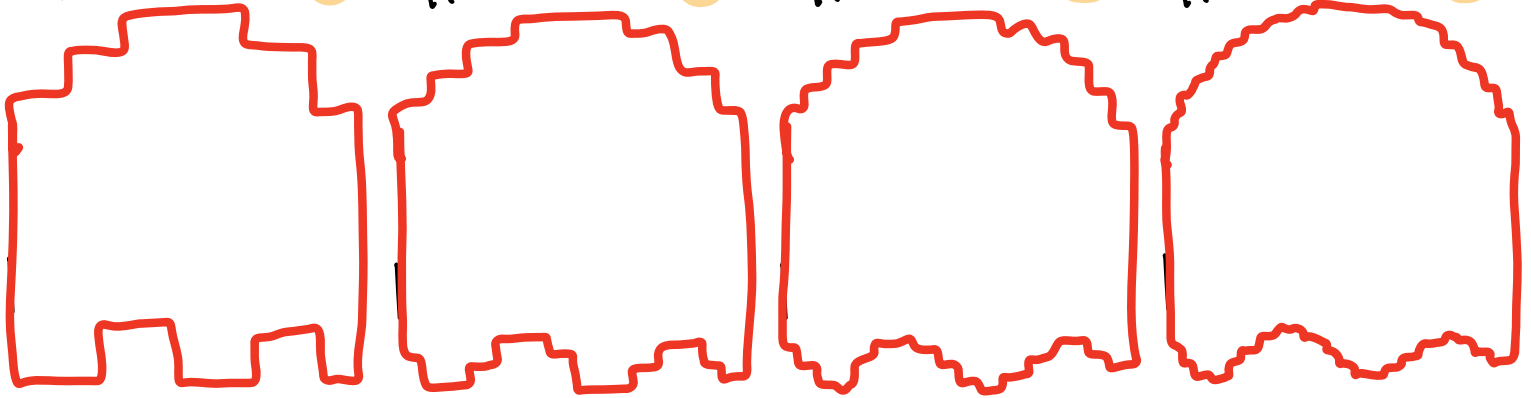
← size (length)
← number of granules

Example:

Pac-Man Red Ghost



Approximation 1 Approximation 2 Approximation 3 Approximation 4



Δx is a measure of the granularity of the approximation

Δx has dimensions (typically distance)

$N = \frac{L}{\Delta x}$ is also a measure of granularity

N has no dimensions

Approximation	Error	Δx
1	error ₁	$\Delta x_1 = \frac{L}{N_1}$
2	error ₂	$\Delta x_2 = \frac{L}{N_2}$
3	error ₃	$\Delta x_3 = \frac{L}{N_3}$
4	error ₄	$\Delta x_4 = \frac{L}{N_4}$

$$\text{error} = C \Delta x^n = C \left(\frac{L}{N} \right)^n = C' N^{-n}$$

we are interested in the coefficient n

n is called the rate of convergence or convergence rate

- How to determine n from computer simulations

- Vary Δx (or N) and compute the error each time

e.g.

Δx_1	error ₁
Δx_2	error ₂
\vdots	\vdots

the relation is $\text{error} = C \Delta x^n$ or $C_2 N^{-n}$

↑ coefficient does not depend on Δx or n

- How to find exponent?

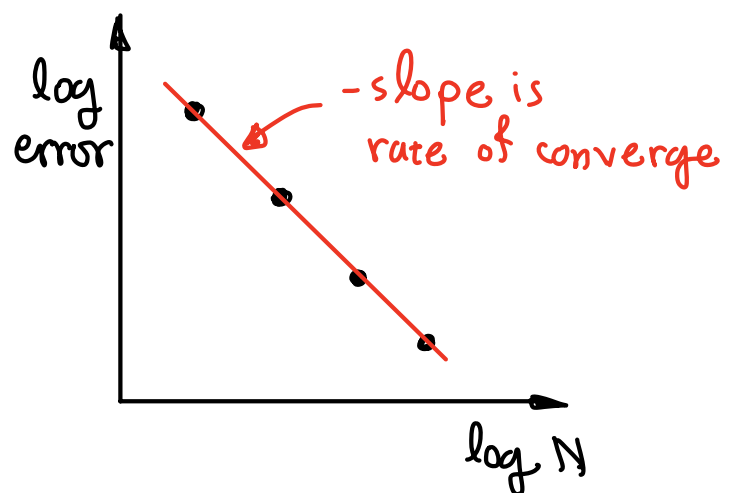
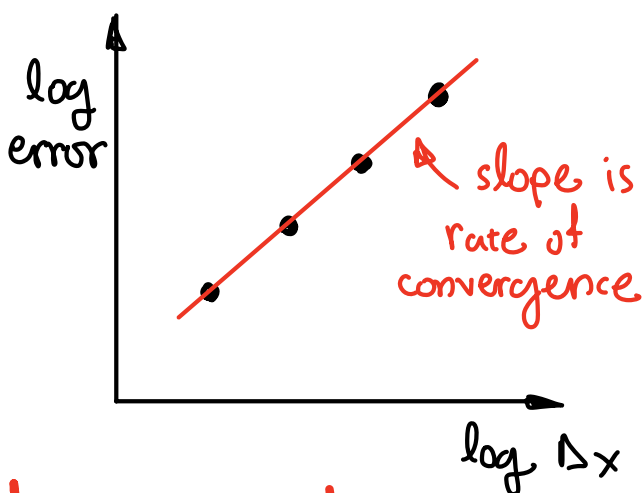
trick: $\text{error} = C \Delta x^n$

$$\log(\text{error}) = n \log(C \Delta x) \leftarrow \text{easier!}$$

↑ slope

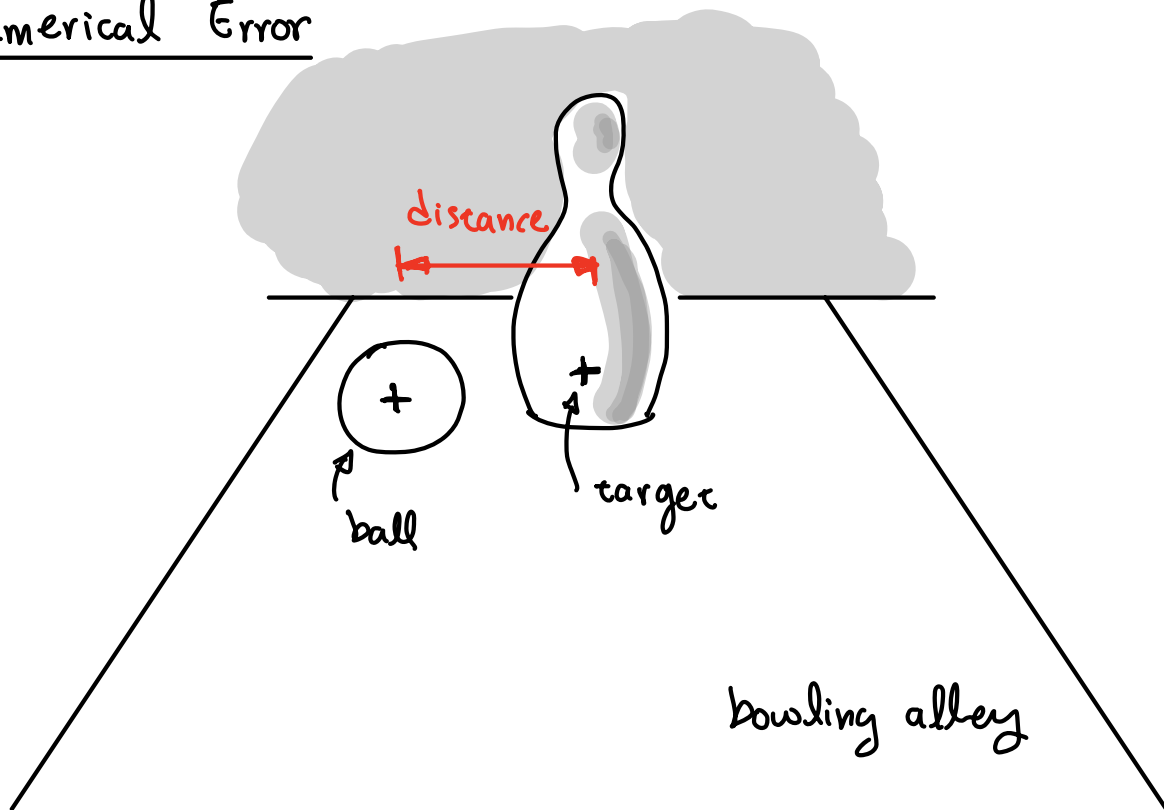
$$\log(\text{error}) = -n \log(C_2 N)$$

In practice:



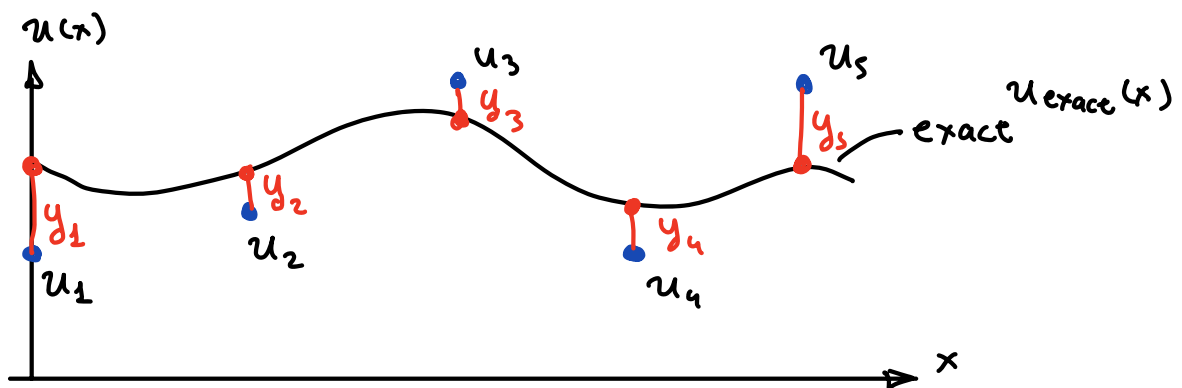
Important: limit is asymptotic: need to take $\Delta x \rightarrow 0$ or $N \rightarrow \infty$

Numerical Error



Definition: error = measure of the "distance" between an exact solution "target" and an approximation

error for scalars: $\text{error} = |u_{\text{approximation}} - u_{\text{exact}}|$
 ↑
 absolute value of difference



error vector: $\vec{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_5 \end{bmatrix} = \begin{bmatrix} u_1 - u_{1,\text{exact}} \\ u_2 - u_{2,\text{exact}} \\ \vdots \\ \end{bmatrix}$

Vector Norms

For a vector y_i we define its p -norm:

$$p\text{-norm} \quad |\vec{y}|_p = \left(\sum_i^n |y_i|^p \right)^{1/p} \quad p = 1, 2, 3, \dots$$

$$2\text{-norm} \quad \text{for 3D vectors } y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

$$|y|_2 = \sqrt{y_1^2 + y_2^2 + y_3^2}$$

$$p = \infty, \text{ infinity norm } |\vec{y}|_\infty = \max_i |y_i|$$

DO NOT
use this
equation
for convergence
studies

Vector norm properties:

$$1. \quad |\vec{y}| > 0 \quad \text{when } \vec{y} \neq 0 \quad \text{and } |\vec{y}| = 0 \quad \text{if and only if } \vec{y} = 0$$

$$2. \quad |\vec{x} + \vec{y}| \leq |\vec{x}| + |\vec{y}|$$

$$3. \quad \underset{\substack{\uparrow \\ \text{scalar}}}{|c \vec{y}|} = |c| |\vec{y}| \quad \text{for any scalar } c$$

Vector Norms for CFD: error norms

Consider two vectors of N elements on a grid x_i , $i = 1, \dots, N$ with equal spacing Δx

u_i is the numerical approximation

v_i is the exact solution

Important: vectors u_i and v_i are at the same location

remember that $u_i = u(x_i)$ and $v_i = v(x_i)$

p -norm

$$L_p = \left[\Delta x \sum_{i=1}^N |u_i - v_i|^p \right]^{\frac{1}{p}}$$

$p = 1, 2, 3, \dots$ integer > 0

Note Δx (pointing to Δx)

exact solution (pointing to v_i)

special case : infinity norm

$$L_\infty = \max_i |u_i - v_i|$$

We need to include Δx because we will be comparing norms of vectors of different lengths.

Similarly for two-dimensional "vectors"

$$L_p = \left[\Delta x \Delta y \sum_{j=1}^N \sum_{i=1}^M |u_{ij} - v_{ij}|^p \right]^{\frac{1}{p}}$$

$$L_\infty = \max_{ij} |u_{ij} - v_{ij}|$$