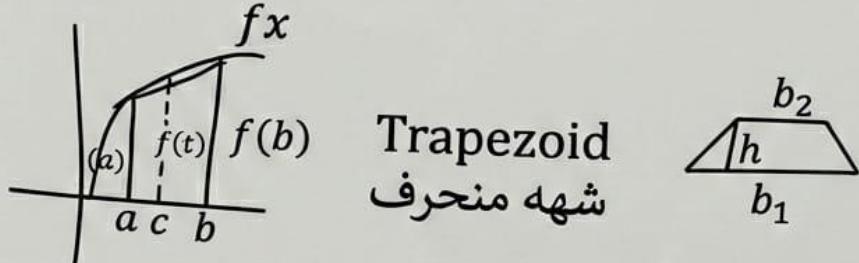


Trapezoid Rule

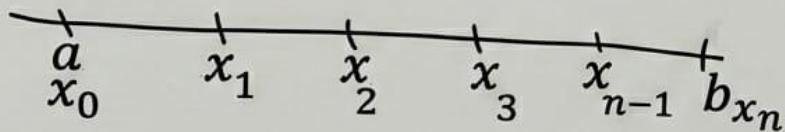
Numerical Integration and Diff

$$I = \int_a^b f(x) dx \quad \text{التكامل}$$



$$A = \frac{(b_1 + b_2)}{2} \cdot h$$

$$T_1 = [f(a) + f(b)] \frac{(b - a)}{2}$$



$$\int_a^b f(x) dx \approx \frac{h}{2} [f(x_0) + 2f(x_1) + 2f(x_2) + \dots + f(x_n)] + F(x_n)$$

$$\int_0^{\pi/2} \sin x dx \quad T_1(1) = \left[\begin{matrix} \sin 0 & + & \sin \frac{\pi}{2} \\ 0 & + & 1 \end{matrix} \right] \frac{\pi}{4}$$

Integration for Error

$$\int_0^{\pi/2} \sin x dx = [-\cos x]_0^{\pi/2} = -[\cos \frac{\pi}{2} - \cos 0] = -[0 - 1] = 1$$

Matrix Normalization & AI Applications

Matrix normalization is the process of scaling the numbers inside a matrix so they fall within a specific, consistent range (usually between 0 and 1, or scaled to have a mean of 0 and a standard deviation of 1).

- **How it works (Min-Max Normalization):** You take every number in the matrix, subtract the smallest number in the entire matrix, and divide by the range (maximum - minimum).

$$X_{norm} = \frac{X - X_{min}}{X_{max} - X_{min}}$$

- **Applications in AI:**

- **Equal Footing:** If an AI model looks at house prices (**\$500,000**) and number of bedrooms (**3**), the giant price numbers will overpower the bedroom numbers. Normalization makes sure all features are treated equally by the algorithm.
- **Faster Training:** Neural networks learn using a process called Gradient Descent. When data is normalized, the network converges on the right answer much faster and more stably.
- **Image Processing:** In Computer Vision, pixel values (0 to 255) are often normalized to a range of 0 to 1 so the neural network processes the image data efficiently.

- **Example:**

Imagine a simple matrix: [2, 4, 8]

1. Min is 2, Max is 8. Range is 6.
2. Normalize 2: $(2 - 2) / 6 = 0$
3. Normalize 4: $(4 - 2) / 6 = 0.333$
4. Normalize 8: $(8 - 2) / 6 = 1$
5. Resulting normalized matrix: [0, 0.333, 1]