Lecture 25 (hrs. 50,51) - November 20, 2025, 8:30 - 10:30 F3

(3.08) <u>Definition</u> (continuous and piecewise linear functions).

Given ordered real numbers $x_0 < x_1 < \dots < x_k$, for $j = 1, \dots, k$ let $I_j = [x_{j-1}, x_j]$. A function $f \colon [x_0, x_k] \to R$ is said to be continuous and piecewise linear on x_0, \dots, x_k if:

- f is a continuous function;
- f is piecewise linear on x_0, \dots, x_k i.e.: defined $p_j(x)$ the restriction of f to I_j , it is: $p_j \in P_1(R)$.

The set of continuous and piecewise linear functions on x_0,\dots,x_k is denoted by C-LAT(x_0,\dots,x_k).

(3.09) Example.

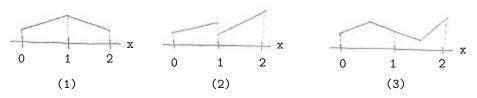


Figure (1) represents the graph of a continuous and piecewise linear function on 0,1,2. Figure (2) represents the graph of a piecewise linear function on 0,1,2 but *not* continuous. Figure (3) represents the graph of a continuous and piecewise linear function but *not* on 0,1,2.La figura (1) rappresenta il grafico di una funzione continua e lineare a tratti su 0,1,2.

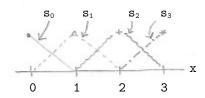
(3.09) Remark.1

(1) The set $C-LAT(x_0,...,x_k)$ is a *vector subspace* of the space of continuous functions on $[x_0,x_k]$, of *dimension* k+1. The k+1 elements $s_0(x),...,s_k(x)$ of $C-LAT(x_0,...,x_k)$ defined by:

$$s_i(x_i) = 1$$
 if $i = j$, $s_i(x_i) = 0$ if $i \neq j$

are the 'canonical basis' of C-LAT($x_0, ..., x_k$).

For example, the graphs of the elements of the canonical basis of C-LAT(0,1,2,3) are:



(2) Given real numbers y_0, \ldots, y_k , the linear combination $y_0 s_0(x) + \cdots + y_k s_k(x)$ is the *unique* element of C-LAT (x_0, \ldots, x_k) that interpolates the data $(x_0, y_0), \ldots, (x_k, y_k)$.

¹ The proof of the statements of this Remark is omitted.

(3.10) Applications.

(1) How we plot graphs in Scilab.

Let $f: [a,b] \rightarrow R$ be a continuous function. The sequence of instructions:

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> x = linspace(a,b,n)';
> plot(x,f(x));
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generates, in a graphics window, the graph of the broken line whose vertices are the points with coordinates $(x(1),f(x(1)),\ldots,(x(n),f(x(n)))$. This broken line is the graph of the unique element $\sigma_n(x)\in C\text{-LAT}(x(1),\ldots,x(n))$ that interpolates the data $(x(1),f(x(1)),\ldots,(x(n),f(x(n)))$. The graph of $\sigma_n(x)$ is used as an approximation to the graph of the function f(x). We will see shortly how accurate this approximation is.

(2) Trapezoid formula.

Let $f: [a,b] \to R$ be a continuous function. We want to know the value (which certainly exists due to the continuity of f):

$$I = \int_{a}^{b} f(x) dx$$

A procedure that provides an approximation of I is the following:

 given an integer number k, the interval [a,b] is divided into k subintervals of equal size:

$$\frac{b-a}{k}$$

identified by the k+1 points $x_0 = a, x_1, ..., x_{k-1}, x_k = b$ (called *nodes*);

- we consider the only element $\sigma_k(x) \in C\text{-LAT}(x_0, \ldots, x_k)$ that interpolates the data $(x_0, f(x_0)), \ldots, (x_k, f(x_k));$
- we approximate I with:

$$J_k = \int_a^b \sigma_k(x) dx$$

The value J_k is easily computed. By introducing the canonical basis $s_0(x), \ldots, s_k(x)$ of C-LAT (x_0, \ldots, x_k) we have:

$$\mathbf{J_k} = \int\limits_0^b \sigma_k(\mathbf{x}) d\mathbf{x} = \int\limits_0^b (\mathbf{f}(\mathbf{x_0}) \mathbf{s_0}(\mathbf{x}) + \cdots + \mathbf{f}(\mathbf{x_k}) \mathbf{s_k}(\mathbf{x})) d\mathbf{x} = \mathbf{f}(\mathbf{x_0}) \int\limits_0^b \mathbf{s_0}(\mathbf{x}) d\mathbf{x} + \cdots + \mathbf{f}(\mathbf{x_k}) \int\limits_0^b \mathbf{s_k}(\mathbf{x}) d\mathbf{x}$$

hence:

$$J_{k} = h \left[\frac{f(x_{0})}{2} + f(x_{1}) + \dots + f(x_{k-1}) + \frac{f(x_{k})}{2} \right]$$

This last expression is called the $trapezoid\ formula$. We'll see shortly how accurate this approximation is.

(3.11) Theorem (error in polynomial interpolation).

Given real numbers $x_0 < x_1 < \cdots < x_k$, and set $I = [x_0, x_k]$, let $f \colon I \to R$ be function with a continuous derivative of order k+1, and let $p_k \in P_k(R)$ be the polynomial that interpolates the data $(x_0, f(x_0)), \ldots, (x_k, f(x_k))$. Then, for each $x \in I$, there exists $\theta \in I$ such that:

$$f(x) - p_k(x) = \frac{f^{(k+1)}(\theta)}{(k+1)!} (x-x_0) \cdots (x-x_k)$$

Moreover, defined:

$$M_{j} = \max_{x \in I} |f^{(j)}(x)|$$

we easily get the limitation:

$$\max_{x \in I} |f(x) - p_k(x)| \le \frac{M_{k+1}}{(k+1)!} (\text{meas } I)^{k+1}$$

Proof: omitted.

(3.12) Remark (approximation with C-LAT elements).

Let $f: [a,b] \rightarrow R$ with continuous second derivative and, for j = 0, ..., k, let:

$$x_j = a + \frac{b-a}{k} j$$

The points x_0, \ldots, x_k divide the interval [a,b] into k intervals of equal width.

Consider the interval $[x_0,x_1]$. Let $p_1 \in P1(R)$ be the polynomial that interpolates the data $(x_0,f(x_0)),(x_1,f(x_1))$. Using the limitation shown in the previous theorem, we have:

$$\max_{\mathbf{x} \in [\mathbf{x}_0, \mathbf{x}, \mathbf{I}]} | \mathbf{f}(\mathbf{x}) - \mathbf{p}(\mathbf{x}) | \leqslant \frac{M_2}{2} \left(\frac{\mathbf{b} - \mathbf{a}}{\mathbf{k}}\right)^2$$

In the same way, we see that the same limitation holds for each of the k subintervals of [a,b]. Therefore, denoted by $\sigma_k(x)$ the element of C-LAT (x_0,\ldots,x_k) that interpolates the data $(x_0,f(x_0)),\ldots,(x_k,f(x_k))$, we have:

$$\max_{x \in [a,b]} |f(x) - \sigma_k(x)| \leq \frac{M_2}{2} \left(\frac{b-a}{k}\right)^2$$

- (3.13) Remark (accuracy of approximations in Applications (3.10)).
- (1) We chose the following quantity as an (absolute) measure of the error made by approximating the graph of f(x) with that of $\sigma_n(x)$:

$$e_n(f) = \max_{x \in [a,b]} |f(x) - \sigma_n(x)|$$

The result of the previous remark shows that: if f has a continuous second derivative, then:

$$\lim_{n\to\infty} e_n(f) = 0$$

and the approximation can be made as accurate as desired by choosing an appropriately large number n.

(2) We choose the quantity $|J_k - I|$ as the (absolute) measure of the error made by approximating I with J_k . The result of the previous remark shows that: if f has a second derivative, then:

$$|J_{k} - I| = \left| \int_{a}^{b} (\sigma_{k}(x) - f(x)) dx \right| \leq \int_{a}^{b} |\sigma_{k}(x) - f(x)| dx \leq \int_{a}^{b} \frac{M_{2}}{2} \left(\frac{b - a}{k}\right)^{2} dx = \frac{M_{2}}{2} \frac{(b - a)^{3}}{k^{2}}$$

In this case too we have:

$$\lim_{k\to\infty} |J_k - I| = 0$$

and the approximation can be made as accurate as desired by choosing an appropriately large k.

(3.14) Remark (reformulation of the least squares problem).

Consider the data (k = 2): (x_0, y_0) , (x_1, y_1) , (x_2, y_2) and the vector space $F = \text{span}\{ f_1(x), f_2(x) \} = \{ a_1 f_1(x) + a_2 f_2(x) \text{ with } a_1, a_2 \in R \}$. The square deviation SQ(f) is rewritten, using the expression for f(x) in terms of the generators $f_1(x), f_2(x)$, as:

$$SQ(f) = (a_1 f_1(x_0) + a_2 f_2(x_0) - y_0)^2 + (a_1 f_1(x_1) + a_2 f_2(x_1) - y_1)^2 + (a_1 f_1(x_2) + a_2 f_2(x_2) - y_2)^2$$

Observing that if v = (v_1, \ldots, v_n) $^t \in R^n$ then we have $(N_2(v))^2 = v_1^2 + \cdots + v_n^2$, the last sum can be rewritten as:

$$\left\| \begin{bmatrix} \mathbf{a}_1 \mathbf{f}_1(\mathbf{x}_0) + \mathbf{a}_2 \mathbf{f}_2(\mathbf{x}_0) - \mathbf{y}_0 \\ \mathbf{a}_1 \mathbf{f}_1(\mathbf{x}_1) + \mathbf{a}_2 \mathbf{f}_2(\mathbf{x}_1) - \mathbf{y}_1 \\ \mathbf{a}_1 \mathbf{f}_1(\mathbf{x}_2) + \mathbf{a}_2 \mathbf{f}_2(\mathbf{x}_2) - \mathbf{y}_2 \end{bmatrix} \right\|^2 = \left\| \begin{bmatrix} \mathbf{f}_1(\mathbf{x}_0) & \mathbf{f}_2(\mathbf{x}_0) \\ \mathbf{f}_1(\mathbf{x}_1) & \mathbf{f}_2(\mathbf{x}_1) \\ \mathbf{f}_1(\mathbf{x}_2) & \mathbf{f}_2(\mathbf{x}_2) \end{bmatrix} \begin{bmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \end{bmatrix} - \begin{bmatrix} \mathbf{y}_0 \\ \mathbf{y}_1 \\ \mathbf{y}_2 \end{bmatrix} \right\|^2$$

hence, defined:

$$\mathbf{A} = \begin{bmatrix} \mathbf{f}_1(\mathbf{x}_0) & \mathbf{f}_2(\mathbf{x}_0) \\ \mathbf{f}_1(\mathbf{x}_1) & \mathbf{f}_2(\mathbf{x}_1) \\ \mathbf{f}_1(\mathbf{x}_2) & \mathbf{f}_2(\mathbf{x}_2) \end{bmatrix} \quad \text{,} \quad \mathbf{x} = \begin{bmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \end{bmatrix} \quad \mathbf{e} \quad \mathbf{b} = \begin{bmatrix} \mathbf{y}_0 \\ \mathbf{y}_1 \\ \mathbf{y}_2 \end{bmatrix}$$

we finally get:

$$SQ(f) = ||Ax - b||^2$$

(3.15) Remark.

The system A x = b obtained in the previous example is the system that translates the interpolation conditions $f(x_0) = y_0$, $f(x_1) = y_1$, $f(x_2) = y_2$.

² To simplify the notation, for each $v \in R^n$, in this part we will indicate by $\|v\|$ the two norm of v.