

# **Notes about Thesis Project I**

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# 1 Introduction

## 1.1 Numerical Quadrature

(Salgado and Wise, p. 397)

Let  $f \in C([a, b])$ . We seek calculate an approximation of

$$I^{(a,b)}[f] := \int_a^b f(x) \, dx.$$

Suppose that  $g \in C([a, b])$ , whose antiderivative is simply obtained, and  $\|f - g\|_\infty < \varepsilon$ . Then,

$$\left| \int_a^b f(x) \, dx - \int_a^b g(x) \, dx \right| \leq \varepsilon (b - a).$$

**Definition 1.1.1** (Nodal set). Let  $[a, b] \subset \mathbb{R}$ .  $X$  is called a *nodal set* of size  $n + 1 \in \mathbb{N}$  iff  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is a set of distinct elements. The elements of  $X$ ,  $x_i$  are called *nodes*.

**Definition 1.1.2** (Interpolating polynomial). Suppose that  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is a nodal set and  $f: [a, b] \rightarrow \mathbb{R}$  is a function. The function  $I: [a, b] \rightarrow \mathbb{R}$  is called an *interpolant of  $f$*  subordinate to  $X$  iff  $\forall i = 0, \dots, n : I(x_i) = f(x_i)$ , we write  $I(X) = f(X)$ .

**Theorem 1.1.3** (existence and uniqueness). Suppose that  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is a nodal set and  $Y = \{y_i\}_{i=0}^n \subset \mathbb{R}$ . There is a unique polynomial  $p \in \mathbb{P}_n$  with the property that  $p(X) = Y$ .

**Definition 1.1.4** (Lagrange nodal basis). Suppose that  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is a nodal set. The *Lagrange nodal basis* subordinate to  $X$  is the set of polynomials  $\mathcal{L}_X = \{L_\ell\}_{\ell=0}^n \subset \mathbb{P}_n$  defined via

$$L_\ell(x) = \prod_{\substack{i=0 \\ i \neq \ell}}^n \frac{x - x_i}{x_\ell - x_i}.$$

**Definition 1.1.5** (Lagrange interpolating polynomial). Suppose that  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is a nodal set,  $\mathcal{L}_X = \{L_i\}_{i=0}^n \subset \mathbb{P}_n$  is the Lagrange nodal basis subordinate to  $X$ , and  $f: [a, b] \rightarrow \mathbb{R}$ . The *Lagrange interpolating polynomial* of the function  $f$ , subordinate to the nodal set  $X$ , is the polynomial

$$p(x) = \sum_{i=0}^n f(x_i) L_i(x) \in \mathbb{P}_n.$$

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Suppose that  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is a nodal set and  $p \in \mathbb{P}_n$  is the unique Lagrange interpolating polynomial of  $f$  subordinate to  $X$ . Then

$$\forall i = 0, \dots, n : f(x_i) = p(x_i)$$

and

$$\forall x \in [a, b] : f(x) = p(x) + E(x),$$

where  $E$  is an expression of the interpolation error. Then

$$\int_a^b f(x) \, dx = \int_a^b p(x) \, dx + \int_a^b E(x) \, dx.$$

But

$$\int_a^b p(x) \, dx = \int_a^b \sum_{i=0}^n f(x_i) L_i(x) \, dx = \sum_{i=0}^n f(x_i) \int_a^b L_i(x) \, dx = \sum_{i=0}^n f(x_i) \beta_i,$$

where  $L_i \in \mathbb{P}_n$  is the  $i$ th Lagrange nodal basis element and  $\beta_i$  is its definite integral:

$$\beta_i = \int_a^b L_i(x) \, dx.$$

The expression  $\sum_{i=0}^n f(x_i) \beta_i$  is a typical numerical integration formula

$$\left| \int_a^b f(x) \, dx - \sum_{i=0}^n f(x_i) \beta_i \right| = \left| \int_a^b E(x) \, dx \right| \leq \int_a^b |E(x)| \, dx.$$

The *quadrature weights*,  $\beta_i$ , depends only on the positions of the nodes within  $[a, b]$ , as well as the interval  $[a, b]$  itself. We will consider the approximation of a weight integral

$$I_w^{(a,b)}[f] := \int_a^b f(x) w(x) \, dx.$$

**Definition 1.1.6** (Quadrature rule). Suppose that  $n, r \in \mathbb{N}_0$ ,  $w$  is a weight function on  $[a, b] \subset \mathbb{R}$ ,  $h = b - a > 0$ , and  $f \in C^r([a, b])$ . The expression

$$Q_{w,r}^{(a,b)}[f] = \sum_{i=0}^r \sum_{j=0}^n \beta_{i,j} f^{(i)}(x_j) = \sum_{j=0}^n \left( \beta_{0,j} f(x_j) + \beta_{1,j} f'(x_j) + \dots + \beta_{r,j} f^{(r)}(x_j) \right),$$

where

$$\forall i \in \{0, \dots, r\} : \forall j \in \{0, \dots, n\} : \beta_{i,j} = h^{i+1} \widehat{\beta}_{i,j}$$

and

$$\forall j \in \{0, \dots, n\} : x_j = a + h \cdot \widehat{x}_j,$$

is called a *quadrature rule of degree  $r$  with intrinsic nodes  $\widehat{X} = \{\widehat{x}_j\} \subset [0, 1]$  and intrinsic weights  $\{\beta_{i,j}\} \subset \mathbb{R}$  are called the *effective nodes* and *effective weights*, respectively.*

A quadrature rule of degree  $r = 0$  is called a *simple quadrature rule*, and we simplify the notation by writing  $\beta_j = \beta_{0,j}$  and

$$Q_{w,r}^{(a,b)}[f] = \sum_{j=0}^n \beta_j f^{(i)}(x_j).$$

The *quadrature rule error* is defined as

$$E_Q[f] = I_w^{(a,b)}[f] - Q_{w,r}^{(a,b)}[f].$$

**Definition 1.1.7** (consistency). The quadrature rule is *consistent of order at least*  $m \in \mathbb{N}_0$  iff  $E_Q[q] = 0$  for all  $q \in \mathbb{P}_m$ . The quadrature rule is *consistent of order exactly*  $m$  iff  $E_Q[q] = 0$  for all  $q \in \mathbb{P}_m$ ; however, for some  $r \in \mathbb{P}_{m+1}$ ,  $E_Q[r] \neq 0$ .

**Definition 1.1.8** (interpolatory quadrature rule). Assume that  $n \in \mathbb{N}_0$ ,  $w$  is a weight function on  $[a, b] \subset \mathbb{R}$ , and  $f \in C([a, b])$ . Suppose that  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is a nodal set and  $p \in \mathbb{P}_n$  is the unique Lagrange Interpolating polynomial of  $f$  subordinate to  $X$ , with

$$p(x) = \sum_{j=0}^n f(x_j) L_j(x),$$

where  $L_j \in \mathbb{P}_n$  is the  $j$ th Lagrange nodal basis element.

The expression

$$Q_w^{(a,b)}[f] = \sum_{j=0}^n f(x_j) \beta_j,$$

where

$$\beta_j = \int_a^b L_j(x) w(x) \, dx,$$

is called an *interpolatory quadrature rule subordinate to  $X$  of Lagrange type* for approximating  $I_w^{(a,b)}[f]$ .

**Theorem 1.1.9** (existence and uniqueness). Suppose that  $X = \{x_i\}_{i=0}^n \subset \mathbb{R}$  is a nodal set. There exists uniqueness weights  $\{\beta_j\}_{j=0}^n$  such that

$$\forall q \in \mathbb{P}_n : \int_a^b q(x) w(x) \, dx = \sum_{j=0}^n \beta_j q(x_j),$$

or, equivalently,

$$\forall q \in \mathbb{P}_n : E_Q[q] = 0.$$

Moreover, these weights are given by

$$\forall j \in \{0, \dots, n\} : \beta_j = \int_a^b L_j(x) w(x) \, dx,$$

where  $L_j$  is the  $j$ th Lagrange nodal basis polynomial subject to  $X$ .

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**Theorem 1.1.10** (consistency). *The last result shows that the simple quadrature rule,*

$$Q_w^{(a,b)}[f] = \sum_{j=0}^n \beta_j f(x_j),$$

*is consistent of order at least  $n$  iff it is a quadrature rule of Lagrange type.*

**Theorem 1.1.11** (Error estimate). *Suppose that  $n \in \mathbb{N}_0$ ,  $w$  is a weight function on  $[a, b] \subset \mathbb{R}$ .  $f \in C^{n+1}([a, b])$ , and  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is a nodal set. Suppose that  $Q_w^{(a,b)}[f]$  is the interpolatory quadrature rule subordinate to  $X$  of Lagrange type. Then,*

$$|E_Q[f]| \leq \frac{M_{n+1}}{(n+1)!} \int_a^b |\omega_{n+1}(x)| w(x) dx,$$

where

$$\omega_{n+1}(x) = \prod_{j=0}^n (x - x_j)$$

and

$$M_{n+1} = \|f^{(n+1)}\|_{\infty}.$$

Consequently, an interpolatory quadrature rule subordinate to  $X$  of Lagrange type is consistent of order at least  $n$ .

**Definition 1.1.12** (characteristic function). Suppose that  $B \subset \mathbb{R}$ . The characteristic function of  $B$  is the function

$$\chi_B(t) = \begin{cases} 1, & t \in B, \\ 0, & t \in \mathbb{R} \setminus B. \end{cases}$$

**Theorem 1.1.13** (kernel). Suppose that  $r \in \mathbb{N}_0$  and  $m \in \mathbb{N}$ , with  $m > r$ . Define the function  $k_m: [a, b] \times [a, b] \rightarrow \mathbb{R}$  via

$$k_m(x, y) = (x - y)^m \xi_{[a, x]}(y) = \begin{cases} (x - y)^m, & a \leq y \leq x \leq b, \\ 0, & a \leq x < y \leq b. \end{cases}$$

Then, for each  $i \in \{0, \dots, r\}$ ,

$$\frac{\partial^i k_m}{\partial x^i} \in C([a, b] \times [a, b])$$

and

$$\frac{\partial^i k_m(x, y)}{\partial x^i} = \begin{cases} \prod_{k=0}^{i-1} (m - k) (x - y)^{m-i}, & a \leq y \leq x \leq b, \\ 0, & a \leq x < y \leq b. \end{cases}$$

**Theorem 1.1.14** (Peano Kernel Theorem). *Suppose that  $r \in \mathbb{N}_0$ ,  $m \in \mathbb{N}$ , with  $m > n$ ,  $w$  is a weight function on  $[a, b] \subset \mathbb{R}$ , and  $f \in C^{m+1}([a, b])$ . Assume that  $Q_{w,r}^{(a,b)}[f]$  is a quadrature rule of degree  $r$ , that is consistent of order at least  $m$ . Let the function  $k_m: [a, b] \times [a, b] \rightarrow \mathbb{R}$ . Set*

$$K_m(y) = E_Q[k_m(\cdot, y)] = \int_a^b k_m(x, y) w(x) dx - \sum_{j=0}^n \sum_{i=0}^r \beta_{i,j} \frac{\partial^i k_m(x_j, y)}{\partial x^i}.$$

*Then the quadrature error satisfies*

$$E_Q[f] = \frac{1}{m!} \int_a^b f^{(m+1)}(y) K_m(y) dy.$$

*The function  $K_m(y)$  is called the Peano Kernel.*

**Theorem 1.1.15** (quadrature error stability). *We have*

$$|E_Q[f]| \leq \frac{1}{m!} \|f^{(m+1)}\|_\infty \|K_m\|_1.$$

*Since  $\|K_m\|_1 < \infty$ , there is a constant  $C > 0$  that may depend on the size of the interval but is independent of  $f$  such that*

$$|E_Q[f]| \leq C \|f^{(m+1)}\|_\infty.$$

**Theorem 1.1.16** (constant sign). *If  $K_m$  does not change sign in  $[a, b]$ , then*

$$E_Q[f] = \frac{f^{(m+1)}(\xi)}{m!} \int_a^b K_m dy$$

*for some  $\xi \in [a, b]$ . Furthermore, we have the simple representation for the error*

$$E_Q[f] = \frac{E_Q[x^{m+1}]}{(m+1)!} f^{(m+1)}(\xi)$$

*for some  $\xi \in [a, b]$ , where  $E_Q[x^{m+1}]$  is the quadrature error for the function  $x \mapsto x^{m+1}$ .*

**Definition 1.1.17** (Peano Kernel). Let  $M_1 = \chi_{[a,b]}$ . For  $k \in \mathbb{N}$  with  $k \geq 2$ , set

$$M_k(x) = \int_{\mathbb{R}} M_{k-1}(x-y) M_1(y) dy.$$

For  $k \in \mathbb{N}$  and  $h > 0$ , we define

$$M_k(x, h) = \frac{1}{h} M_k\left(\frac{x}{h}\right).$$

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**Definition 1.1.18** (Closed Newton-Cotes quadrature rule). Suppose that  $w$  is a weight function on  $[a, b] \subset \mathbb{R}$  and  $n \in \mathbb{N}$ . Set  $h = b - a > 0$  and  $\hat{h} = \frac{h}{n}$ . Suppose that, for the simple quadrature rule, the nodal set  $X = \{x_i\}_{i=0}^n \subset [a, b]$  is defined by

$$x_j = a + j\hat{h}, \quad j \in \{0, \dots, n\}.$$

The resulting method, denoting  $Q_n[f]$ , is called a *closed Newton-Cotes quadrature rule of order  $n$* .

*Example 1.1.19* (Newton-Cotes quadrature rules of order  $n = 1, 2$  and weight function  $w \equiv 1$  on  $[a, b]$ ).

$$x_j = a + h\hat{x}_j, \quad \beta_j = h\hat{\beta}_j, \quad h = b - a.$$

$n$	rule	$\hat{x}_j$	$\hat{\beta}_j$	Error Formula
1	Trapezoidal	0, 1	$\frac{1}{2}, \frac{1}{2}$	$-\frac{1}{12}h^3 f^{(2)}(\xi)$
2	Simpson's	$0, \frac{1}{2}, 1$	$\frac{1}{6}, \frac{4}{6}, \frac{1}{6}$	$-\frac{1}{90}h^5 f^{(4)}(\xi)$

For  $n = 2$ , we observe the phenomenon of *super-convergence*, i.e., a higher than expected convergence.

For example, consider the case  $n = 3$ , Simpson's  $\frac{3}{8}$  rule, on the reference interval  $[0, 1]$ . The second Lagrange nodal basis element is

$$\hat{L}_1(x) = \frac{x(x - \frac{2}{3})(x - 1)}{\frac{1}{3}(\frac{1}{3} - \frac{2}{3})(\frac{1}{3} - 1)} = \frac{27}{2} \left( x^3 - \frac{5}{3}x^2 + \frac{2}{3}x \right).$$

Then,

$$\hat{\beta}_1 = \int_0^1 \hat{L}_1(x) \, dx = \frac{27}{2} \left( \frac{1}{4}x^4 - \frac{5}{9}x^3 + \frac{1}{3}x^2 \right) \Big|_{x=0}^{x=1} = \frac{3}{8}.$$

**Theorem 1.1.20** (Error estimate). Let  $[a, b] \subset \mathbb{R}$ . Suppose that  $Q_n^{(a,b)}[f]$  is a closed Newton-Cotes quadrature rule of order  $n \in \mathbb{N}$ . Then, the order of quadrature rule is consistent of order at least  $n$ . Moreover, if  $f \in C^{n+1}([a, b])$ , then

$$|E_{Q_n}[f]| \leq Ch^{n+2} \left\| f^{(n+1)} \right\|_{\infty},$$

where  $h = b - a$  and  $C > 0$  is independent of  $h$  and  $f$ .

**Theorem 1.1.21** (Integral Mean Value Theorem). Suppose that  $-\infty < a < b < \infty$ ,  $f \in C([a, b])$ , and  $g \in \mathcal{R}(a, b)$ . Furthermore, suppose that  $\forall x \in [a, b] : g(x) \geq 0$ . Then there exists a point  $\xi \in [a, b]$  such that

$$\int_a^b f(x) g(x) \, dx = f(\xi) \int_a^b g(x) \, dx.$$

Thus, if  $\forall x \in [a, b] : g(x) = 1$ , there exists a point  $\xi \in [a, b]$  such that

$$f(\xi) = \frac{1}{b-a} \int_a^b f(x) \, dx.$$



## 1.2 Heuristic derivation of Advection Differential Equation

**Definition 1.2.1** (Average concentration). Let  $h > 0$ ,  $h \ll 1$ . We define the *average concentration*  $\bar{u}(x, t)$  in a space-time cell  $[x - \frac{1}{2}h, x + \frac{1}{2}h] \times [0, T]$ .

$$\bar{u}(x, t) = \frac{1}{h} \int_{x - \frac{1}{2}h}^{x + \frac{1}{2}h} u(s, t) \, ds.$$

**Definition 1.2.2** (Mass flux). The mass flux is the product of

$$J_x = cv_x.$$

- $J_x$  is the mass flux in  $x$ -direction.
- $c$  is the concentration of the substance.
- $v_x$  is the velocity of the substance in the  $x$ -direction.

**Theorem 1.2.3** (Conservation law). *If the species is carried along by a flowing medium with velocity  $a(x, t)$ , then the mass conservation law implies that the change of  $\bar{u}(x, t)$  per unit of time is the net balance of inflow and outflow over the cell boundaries,*

$$\frac{\partial \bar{u}(x, t)}{\partial t} = \frac{1}{h} \left[ a \left( x - \frac{1}{2}h, t \right) u \left( x - \frac{1}{2}h, t \right) - a \left( x + \frac{1}{2}h, t \right) u \left( x + \frac{1}{2}h, t \right) \right].$$

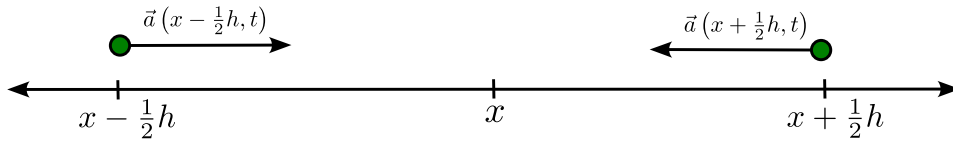
### Numerical Quadrature (Hundsdorfer and Verwer, p. 9)

**Definition 1.2.4** (Advection equation).

$$\frac{\partial u(x, t)}{\partial t} + \frac{\partial a(x, t) u(x, t)}{\partial x} = 0.$$

**Definition 1.2.5** (Diffusion equation).

$$\frac{\partial u(x, t)}{\partial t} = \frac{\partial}{\partial x} \left( d(x, t) \frac{\partial u(x, t)}{\partial x} \right).$$



**Theorem 1.2.6** (Mass conservation law). *If  $u(x, t)$  is a concentration and*

$$M(t) := \int_0^1 u(x, t) \, dx$$

*represents the mass in  $[0, 1]$  at time  $t$ , then  $M$  is a conserved quantity.*

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*Proof.*

$$\begin{aligned}\frac{dM(t)}{dt} &= \int_0^1 u_t(x, t) \, dx = \int_0^1 (-au_x(x, t) + du_{xx}(x, t)) \, dx \\ &= -a(u(1, t) - u(0, t)) + d(u_x(1, t) - u_x(0, t)) = 0.\end{aligned}$$

□

### 1.3 The Advection Problem in One Dimension

**Definition 1.3.1** (space-time grid). Let  $d \in \mathbb{N}$ ,  $\Omega = (0, 1)^d$ , and  $T > 0$ . For  $K, N \in \mathbb{N}$ , we set  $\tau = \frac{T}{K}$  and  $h = \frac{1}{N+1}$ . We define the *space-time grid domain*

$$\overline{\mathcal{C}}_h^\tau = \overline{\Omega}_h \times [0, T]_\tau = \{(\mathbf{x}, t_k) \mid \mathbf{x} \in \overline{\Omega}_h, t_k = k\tau, k \in \{0, \dots, K\}\},$$

where we recall that  $\overline{\Omega}_h = \overline{\Omega} \cap \mathbb{Z}_h^d$ . We define the *discrete interior* of  $\overline{\mathcal{C}}_h^\tau$  to be

$$\mathcal{C}_h^\tau = \Omega_h \times (0, T)_\tau.$$

**Definition 1.3.2** (space-time grid functions). Let  $\mathcal{C}_h^\tau$  be a space-time grid domain. We denote by

$$\mathcal{V}(\overline{\mathcal{C}}_h^\tau) = \{v \mid \overline{\mathcal{C}}_h^\tau \rightarrow \mathbb{R}\}$$

be the space of *space-time grid functions*. The spaces

$$\mathcal{V}(\mathcal{C}_h^\tau), \mathcal{V}(\partial_L \mathcal{C}_h^\tau)$$

**Definition 1.3.3** (space-time discrete norms). Let  $d \in \{1, 2\}$ ,  $p \in [1, \infty]$ , and  $q \in [1, \infty)$ . We define the *space-time norm*

$$\|v\|_{L_\tau^q(L_h^p)} = \left( \tau \sum_{k=1}^K \|v^k\|_{L_h^p}^q \right)^{\frac{1}{q}}$$

and

$$\|v\|_{L_\tau^\infty(L_h^p)} = \max_{k=0}^K \|v^k\|_{L_h^p}$$

**Definition 1.3.4** (Péclet number). Consider the simple constant-coefficient advection-diffusion equation

$$u_t + au_x = du_{xx}, \quad t > 0, \quad 0 < x < L,$$

with the given initial profile  $u(x, 0)$ . If  $d > 0$  we need boundary conditions at  $x = 0$  and  $x = L$ , such as Dirichlet conditions. On the other hand, for the pure advection problem we need only to prescribe the solution at the *inflow* boundary, that is, at  $x = 0$  if  $a > 0$  and  $x = L$  if  $a < 0$ . If  $d > 0$  but  $d \approx 0$ , or more precisely if the *Péclet number*

$$|a| \frac{L}{d}$$

is large, the Dirichlet condition at the outflow boundary will give rise to a *boundary layer*.

If the Péclet number  $|a \frac{L}{d}|$  is large, the problem is called *singularity perturbed*.

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