

## 12. LECTURE 12

## NUMERICAL INTEGRATION

We use polynomial interpolation techniques to derive numerical integration schemes to approximate

$$I(f) = \int_{\alpha}^{\beta} f(x) \, dx,$$

for  $\alpha < \beta$ . Let  $\{x_0, \dots, x_n\} \subset [a, b]$  be distinct, where  $a < b$  are such that  $[\alpha, \beta] \subseteq [a, b]$ . Let  $p \in \mathbb{P}^n$  interpolates  $f$  at  $\{x_0, \dots, x_n\}$ . We propose to approximate  $I(f)$  by

$$Q(f) = \int_{\alpha}^{\beta} p(x) \, dx.$$

Using the Lagrange polynomials  $\{l_i(x)\}_{i=0}^n$  associated with the interpolations points  $\{x_i\}_{i=0}^n$ , we write

$$p(x) = \sum_{i=0}^n f(x_i) l_i(x)$$

so that

$$\begin{aligned} Q(f) &= \int_{\alpha}^{\beta} \left( \sum_{i=0}^n f(x_i) l_i(x) \right) dx = \sum_{i=0}^n f(x_i) \int_{\alpha}^{\beta} l_i(x) \, dx \\ &= \sum_{i=0}^n w_i f(x_i), \end{aligned}$$

where we defined

$$w_i := \int_{\alpha}^{\beta} l_i(x) \, dx.$$

This leads to the following definition of quadrature.

**Definition 12.1** (Quadrature). *An integral approximation of the form*

$$I(f) \approx Q(f) = \sum_{i=0}^n w_i f(x_i)$$

*is called a quadrature. The real numbers  $\{w_i\}$  are the weights and  $\{x_i\}$  are the nodes.*

**Example 12.1** (Rectangle quadrature). *Let  $x_0 \in [a, b]$ . Find the quadrature approximating*

$$I(f) = \int_a^b f(x) \, dx$$

*based on polynomial interpolation using  $x_0$  and  $\mathbb{P}^0$ . In this case,  $p(x) = f(x_0)l_0(x) = f(x_0)$  and*

$$Q(f) = \int_a^b f(x_0) \, dx = (b - a)f(x_0),$$

*which is the area of the shaded region in Figure 4.*

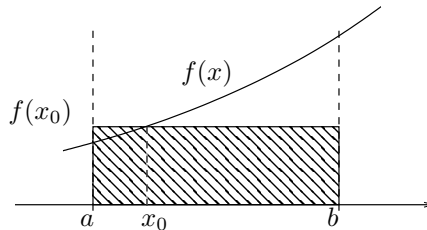


FIGURE 4. Rectangle quadrature. The approximation  $Q(f)$  corresponds to the area of the shaded region.

**Example 12.2** (Trapezoidal quadrature). Consider  $p \in \mathbb{P}^1$  interpolating  $f$  at  $x_0 = a$  and  $x_1 = b$  to approximate

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

In that case,

$$p(x) = l_0(x)f(a) + l_1(x)f(b) = \frac{b-x}{b-a}f(a) + \frac{x-a}{b-a}f(b)$$

so that

$$Q(f) = \int_a^b p(x) \, dx = \frac{f(a)}{b-a} \int_a^b (b-x) \, dx + \frac{f(b)}{b-a} \int_a^b (x-a) \, dx.$$

Both integrals equal  $\frac{1}{2}(b-a)^2$  so

$$Q(f) = \frac{b-a}{2} (f(a) + f(b)).$$

See Figure 5 for an illustration.

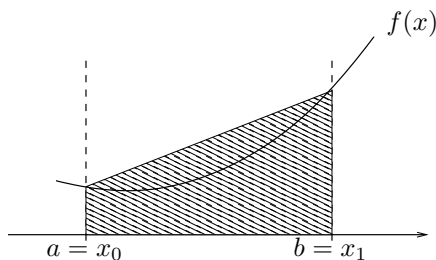


FIGURE 5. Trapezoidal quadrature. The approximation  $Q(f)$  corresponds to the area of the shaded region.

**Example 12.3** (3 Points Quadrature). Consider the interpolation nodes  $\{x, x-h, x-2h\}$  for some  $h > 0$  and  $x \in \mathbb{R}$  and use a quadratic polynomial interpolant to derive a quadrature scheme to approximate

$$I(f) = \int_{x-h}^x f(t) \, dt.$$

Note that this will use interpolation points outside the integration region. In Example 11.1, we have already computed the corresponding lagrange polynomials  $l_0$ ,

$l_1$  and  $l_2$ . To derive such quadrature scheme, we need to compute

$$w_i = \int_{x-h}^x l_i(t) dt,$$

which seems like a lot of work... perhaps there is a better way (later).

We now discuss how well  $Q(f)$  approximate  $I(f)$ .

**Theorem 12.1** (Interpolation error). *Let  $f \in C^{(n+1)}[a, b]$ ,  $\{x_0, \dots, x_n\}$  distinct in  $[a, b]$  and  $a \leq \alpha < \beta \leq b$ . If  $p \in \mathbb{P}^n$  interpolates  $f$  at  $x_i$ ,  $i = 0, \dots, n$ , and  $Q(f) = \sum_{i=0}^n w_i f(x_i)$ , with  $w_i = \int_{\alpha}^{\beta} l_i(x) dx$ , then we have*

$$I(f) - Q(f) = \frac{1}{(n+1)!} \int_{\alpha}^{\beta} f^{(n+1)}(\xi_x) \prod_{i=0}^n (x - x_i) dx,$$

where for every  $x \in [\alpha, \beta]$ ,  $\xi_x \in [a, b]$ . Moreover, if  $\prod_{i=0}^n (x - x_i)$  does not change sign on  $[a, b]$ , then there exists  $\xi \in [a, b]$  with

$$I(f) - Q(f) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \int_{\alpha}^{\beta} \prod_{i=0}^n (x - x_i) dx.$$

*Proof.* In view of the interpolation error provided by Theorem 4.1, for  $x \in (\alpha, \beta)$  there exists  $\xi_x \in (a, b)$  such that

$$I(f) - Q(f) = \int_{\alpha}^{\beta} (f(x) - p(x)) dx = \frac{1}{(n+1)!} \int_{\alpha}^{\beta} f^{(n+1)}(\xi_x) \prod_{i=0}^n (x - x_i) dx.$$

This is the first claim. To continue further, it suffices to note that  $f^{(n+1)}(\xi_x)$  is continuous (see Remark 11.1) and invoke the mean value theorem for integrals to write

$$\frac{1}{(n+1)!} \int_{\alpha}^{\beta} f^{(n+1)}(\xi_x) \prod_{i=0}^n (x - x_i) dx = \frac{f^{(n+1)}(\xi)}{(n+1)!} \int_{\alpha}^{\beta} \prod_{i=0}^n (x - x_i) dx,$$

for some  $\xi \in (a, b)$ . This implies the second claim.  $\square$

**Example 12.4** (Error for the Trapezoidal quadrature). *For some  $\xi \in (a, b)$  the above theorem guarantees that*

$$\int_a^b f(x) dx - \frac{b-a}{2} (f(a) + f(b)) = \frac{f''(\xi)}{2} \int_a^b (x-a)(x-b) dx$$

using the fact that  $(x-a)(x-b)$  does not change sign. Hence, computing the integral leads to

$$\int_a^b f(x) dx - \frac{b-a}{2} (f(a) + f(b)) \stackrel{y=x-a}{=} \frac{f''(\xi)}{2} \int_0^{b-a} y(y-(b-a)) dy = -\frac{f''(\xi)}{12} (b-a)^3.$$

**Example 12.5** (Simpson quadrature). *We want to approximate*

$$I(f) = \int_{-1}^1 f(x) dx$$

using a polynomial of degree 2 interpolating  $f$  at  $x_0 = -1$ ,  $x_1 = 0$  and  $x_2 = 1$ . We first compute the lagrange polynomials

$$\begin{aligned} l_0(x) &= \frac{(x-0)(x-1)}{(-1-0)(-1-1)} = \frac{x^2-x}{2} \\ l_1(x) &= \frac{(x+1)(x-1)}{(0+1)(0-1)} = 1-x^2 \\ l_2(x) &= \frac{(x-0)(x+1)}{(1-0)(1+1)} = \frac{x^2+x}{2} \end{aligned}$$

Therefore,

$$\begin{aligned} w_0 &= \int_{-1}^1 l_0(x) \, dx = \int_{-1}^1 \frac{1}{2}x^2 \, dx = \frac{1}{3} \\ w_1 &= \int_{-1}^1 l_1(x) \, dx = \int_{-1}^1 (1-x^2) \, dx = \frac{4}{3} \\ w_2 &= \int_{-1}^1 l_2(x) \, dx = \int_{-1}^1 \frac{1}{2}x^2 \, dx = \frac{1}{3}. \end{aligned}$$

and the quadrature rule reads

$$I(f) \approx Q(f) = \frac{1}{3}f(-1) + \frac{4}{3}f(0) + \frac{1}{3}f(1).$$

However,

$$\prod_{i=0}^2 (x-x_i) = (x+1)x(x-1) = (x^2-1)x$$

changes sign on  $[-1, 1]$  so we cannot get a formula involving  $f'''(\xi)$  for some (fixed)  $\xi \in (-1, 1)$ .

We now discuss a possibly simpler way to compute the weights  $w_i$ . This relies on the observation that when  $f \in \mathbb{P}^n$  is a polynomial of degree  $n$ , then its interpolant  $p \in \mathbb{P}^n$  is  $f$  itself (since interpolant are unique). This means

$$I(f) = \int_a^b f(x) \, dx = \int_a^b p(x) \, dx = Q(f)$$

for all  $f \in \mathbb{P}^n$ . In other words, the quadrature is *exact* for  $p \in \mathbb{P}^n$ . In particular, this means that

$$\begin{aligned} J_0 &:= \int_a^b 1 \, dx = \sum_{i=0}^n w_i 1 = Q(1) \\ J_1 &:= \int_a^b x \, dx = \sum_{i=0}^n w_i x_i = Q(x) \\ &\vdots \\ J_n &:= \int_a^b x^n \, dx = \sum_{i=0}^n w_i x_i^n = Q(x^n). \end{aligned}$$

Hence, the weights  $\{w_i\}_{i=0}^n$  satisfy the linear system

$$A^t w := \begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ x_0 & x_1 & x_2 & \dots & x_n \\ x_0^2 & x_1^2 & x_2^2 & \dots & x_n^2 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x_0^n & x_1^n & x_2^n & \dots & x_n^n \end{pmatrix} \begin{pmatrix} w_0 \\ w_1 \\ \vdots \\ w_n \end{pmatrix} = \begin{pmatrix} J_0 \\ J_1 \\ \vdots \\ J_n \end{pmatrix}.$$

Recall that you get the linear system

$$Aw := \begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ 1 & x_2 & x_2^2 & \dots & x_2^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{pmatrix} \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_n \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_n \end{pmatrix},$$

when solving the interpolation problem  $p(x_i) = y_i$ , where

$$p(x) = c_0 + c_1 x + \dots + c_n x^n.$$

We already know that  $A$  is non-singular and it follows that  $A^t$  is also non-singular. As a consequence, we realize that

*Remark 12.1 (Uniqueness of Weights).* The weights  $w_i$  making the quadrature  $Q$  exact on  $\mathbb{P}^n$  are uniquely determined from the exactness conditions

$$Q(x^j) = \sum_{i=0}^n w_i x_i^j = \int_a^b x^j dx, \quad j = 0, \dots, n.$$

**Example 12.6** (Simpson's quadrature).

$$Q(f) = w_0 f(-1) + w_1 f(0) + w_2 f(1) \approx \int_{-1}^1 f(x) dx.$$

The exactness conditions (for  $\mathbb{P}^2$ ) are

$$\begin{aligned} 2 &= \int_{-1}^1 1 dx = w_0(1) + w_1(1) + w_2(1) \\ 0 &= \int_{-1}^1 x dx = w_0(-1) + w_1(0) + w_2(1) \\ \frac{2}{3} &= \int_{-1}^1 x^2 dx = w_0(-1)^2 + w_1(0)^2 + w_2(1)^2, \end{aligned}$$

i.e.

$$w_0 = w_2 = \frac{1}{3} \quad \text{and} \quad w_1 = \frac{4}{3}$$

as previously obtained.

*Remark 12.2* (Higher order exactness for Simpson). Note that in addition of being exact for any polynomial of degree 2, the Simpson's quadrature rule also satisfy

$$0 = \int_{-1}^1 x^3 dx = \frac{1}{3}(-1)^3 + 0 + \frac{1}{3}(1)^3$$

but

$$\frac{2}{5} = \int_{-1}^1 x^4 dx \neq \frac{1}{3}(-1)^4 + 0 + \frac{1}{3}(1)^4 = \frac{2}{3}.$$

Hence, the Simpson's quadrature rule is exact for  $\mathbb{P}^3$  but not  $\mathbb{P}^4$ .

## 13. LECTURE 13

Last lecture introduced the Simpson's rule

$$I(f) := \int_{-1}^1 f(x) dx \approx \frac{1}{3}f(-1) + \frac{4}{3}f(0) + \frac{1}{3}f(1) =: Q(f).$$

We found that  $Q$  was exact for cubics.

Consider instead the quadrature scheme based on the nodes  $\{x_0, x_1, x_2, x_3\} := \{-1, 0, 1/2, 1\}$ , i.e.

$$I(f) \approx Q(f) = \sum_{i=0}^3 w_i f(x_i).$$

We also saw during the last lecture that there exists a unique set of weights  $w_i$ ,  $i = 0, 1, 2, 3$  making  $Q$  exact for cubics (see Remark 12.1). Note that the Simpson's scheme can be interpreted as

$$Q(f) = \frac{1}{3}f(-1) + \frac{4}{3}f(0) + 0f(1/2) + \frac{1}{3}f(1)$$

and hence is the unique scheme. Applying the error formula provided by Theorem 12.1 gives

$$I(f) - Q(f) = \frac{1}{4!} \int_{-1}^1 f^{(4)}(\xi_x) \prod_{i=0}^3 (x - x_i) dx.$$

The quantity on the right side of the above relation is usually quite large. To get quadrature to approximate integrals, we need *composite schemes*.

**Composite Schemes.** Suppose you have a scheme

$$Q(f) = \sum_{i=0}^n w_i f(x_i) \approx \int_a^b f(x) dx = I(f),$$

exact on  $\mathbb{P}^n$ , where  $\{x_0, \dots, x_n\}$  are distinct in  $[a, b]$ .

We want to deduce from a scheme on  $[\alpha, \beta]$ . Let  $\lambda$  be the linear mapping taking  $[a, b]$  onto  $[\alpha, \beta]$ , i.e.

$$\lambda(x) = \alpha + \frac{x - a}{b - a}(\beta - \alpha)$$

( $\lambda(a) = \alpha$  and  $\lambda(b) = \beta$ ). As we saw in an early homework, if  $p \in \mathbb{P}^n$ , then so is  $q(x) = p(\lambda(x))$ . Also,

$$\lambda^{-1}(x) = a + \frac{x - \alpha}{\beta - \alpha}(b - a)$$

is a linear mapping from  $[\alpha, \beta]$  to  $[a, b]$ . Now, for  $q \in \mathbb{P}^n$

$$\tilde{I}(q) := \int_{\alpha}^{\beta} q(t) dt = \frac{\beta - \alpha}{b - a} \int_a^b q(\lambda(x)) dx,$$

where we have used the change of variable  $x = \lambda^{-1}(t)$  or  $t = \lambda(x)$  so that  $\frac{dx}{dt} = \frac{b-a}{\beta-\alpha}$  and  $dt = \frac{\beta-\alpha}{b-a} dx$ . Since the composition  $q \circ \lambda$  is in  $\mathbb{P}^n$  and the quadrature scheme is exact on  $\mathbb{P}^n$

$$\tilde{I}(q) := \frac{\beta - \alpha}{b - a} \sum_{i=0}^n w_i q(\lambda(x_i)).$$

We set

$$(8) \quad \tilde{w}_i = \frac{\beta - \alpha}{b - a} w_i \quad \text{and} \quad \tilde{x}_i = \lambda(x_i)$$

to deduce that

$$\int_{\alpha}^{\beta} q(t) dt = \sum_{i=0}^n \tilde{w}_i q(\tilde{x}_i) =: \tilde{Q}(q).$$

In conclusion, given a scheme

$$I(f) = \int_a^b f(x) dx \approx Q(f) = \sum_{i=0}^n w_i f(x_i)$$

which is exact on  $\mathbb{P}^n$ , we get a *translated* scheme

$$\tilde{I}(f) = \int_{\alpha}^{\beta} f(t) dt \approx \tilde{Q}(f) = \sum_{i=0}^n \tilde{w}_i f(\tilde{x}_i)$$

which is also exact on  $\mathbb{P}^n$  using the notation (8).

*Remark 13.1* (Property of the translated scheme). The map  $\lambda$  is a linear map of  $[a, b]$  onto  $[\alpha, \beta]$  so it maps points in a proportional way:  $a \rightarrow \alpha$  and  $b \rightarrow \beta$  implies  $(a + b)/2 \rightarrow (\alpha + \beta)/2$ . More generally, for any  $t \in [0, 1]$

$$[a, b] \ni ta + (1 - t)b \rightarrow t\alpha + (1 - t)\beta \in [\alpha, \beta].$$

**Composite Quadrature.** We want to approximate

$$I(f) = \int_a^b f(x) dx$$

and introduce  $N + 1$  distinct points

$$a = x_0 < x_1 < \dots < x_N = b$$

and set  $h = \max_{i=1, \dots, N} (x_i - x_{i-1})$ . Hence, we split the integral over  $[a, b]$  onto  $N$  pieces

$$I(f) = \sum_{i=1}^N \int_{x_{i-1}}^{x_i} f(x) dx$$

and use a base (translated) quadrature scheme over each sub-interval.

**13.1. Simpson's Composite Quadrature Rule.** If we use the Simpson's rule

$$\int_{-1}^1 g(t) dt \approx \frac{1}{3}g(-1) + \frac{4}{3}g(0) + \frac{1}{3}g(1)$$

to approximate

$$\int_{x_{i-1}}^{x_i} f(x) dx,$$

we have

$$\int_{x_{i-1}}^{x_i} f(x) dx \approx \sum_{i=0}^2 \tilde{w}_i f(\tilde{x}_i),$$

where

$$\tilde{w}_2 = \tilde{w}_0 = \frac{x_i - x_{i-1}}{2} \frac{1}{3} \quad \text{and} \quad \tilde{w}_1 = \frac{x_i - x_{i-1}}{2} \frac{4}{3}$$

and the nodes are moved proportionally

$$-1 \rightarrow x_{i-1}, \quad 1 \rightarrow x_i \quad \text{and} \quad 0 \rightarrow \frac{x_{i-1} + x_i}{2}.$$



This implies

$$\int_{x_{i-1}}^{x_i} f(x) \, dx \approx \frac{x_i - x_{i-1}}{6} (f(x_{i-1}) + 4f(x_{i-1/2}) + f(x_i)),$$

where

$$x_{i-1/2} := \frac{x_{i-1} + x_i}{2}.$$

Gathering all the approximations in all subinterval, we arrive at the *composite Simpson's rule* approximation

$$\int_a^b f(x) \, dx \approx \sum_{i=1}^N \frac{x_i - x_{i-1}}{6} (f(x_{i-1}) + 4f(x_{i-1/2}) + f(x_i)) =: \sum_{i=1}^N \tilde{Q}_i(f).$$

Regarding the integration error, the Simpson's rule is the rule based on  $\{-1, 0, 1/2, 1\}$ , which is exact for cubics. Since  $\frac{1}{2} = \frac{3}{4}(1) + \frac{1}{4}(-1)$ , the translated rule is the rule on

$$\{x_{i-1}, \frac{1}{2}(x_{i-1} + x_i), \frac{3}{4}x_i + \frac{1}{4}x_{i-1}, x_i\} := \{\tilde{x}_0, \tilde{x}_1, \tilde{x}_2, \tilde{x}_3\},$$

which is exact on cubics. The error formula (Theorem 12.1) gives

$$\int_{x_{i-1}}^{x_i} f(x) \, dx - \tilde{Q}_i(f) = \frac{1}{24} \int_{x_{i-1}}^{x_i} f^{(4)}(\xi_x) \prod_{j=0}^3 (x - \tilde{x}_j) \, dx,$$

provided that  $f \in C^4[a, b]$ . Let  $\|f^{(4)}\|_\infty = \max_{t \in [a, b]} |f^{(4)}(t)|$ , then

$$\left| \int_{x_{i-1}}^{x_i} f(x) \, dx - \tilde{Q}_i(f) \right| \leq \frac{1}{24} \|f^{(4)}\|_\infty h^4 \int_{x_{i-1}}^{x_i} dx$$

(since  $|x - \tilde{x}_j| \leq h$  as  $x, \tilde{x}_j \in [x_{i-1}, x_i]$ ). Therefore, we obtain that

$$\left| \int_{x_{i-1}}^{x_i} f(x) \, dx - \tilde{Q}_i(f) \right| \leq \frac{1}{24} \|f^{(4)}\|_\infty h^4 (x_i - x_{i-1}).$$

Summing over all subintervals gives an estimate for the error

$$\left| \int_a^b f(x) \, dx - \sum_{i=1}^N \tilde{Q}_i(f) \right| \leq \sum_{i=1}^N \left| \int_{x_{i-1}}^{x_i} f(x) \, dx - \tilde{Q}_i(f) \right| \leq \frac{b-a}{24} \|f^{(4)}\|_\infty h^4.$$

In general, a quadrature rule which is exact on  $\mathbb{P}^n$  translated to an interval of size  $h$  has a local accuracy of  $h^{n+2}$  and usually results in a global accuracy of  $h^{n+1}$  when used in a composite quadrature.

## 14. LECTURE 14

**Gaussian Quadrature.** We noted last lecture that the order of a quadrature is determined by exactness on  $\mathbb{P}^n$ . It is natural to optimize the order by allowing the nodes to move.

We start with examples.

**Example 14.1** (One point).

$$I(f) = \int_a^b f(x) dx \approx (b-a)f(x_i).$$

Note that the weight  $(b-a)$  makes the quadrature exact on constants. For the quadrature to be exact on linears, we need that

$$\frac{b^2 - a^2}{2} = \int_a^b x dx = (b-x)x_i,$$

or

$$x_i = \frac{b+a}{2},$$

which implies that

$$Q(f) = (b-a)f\left(\frac{b+a}{2}\right).$$

This is called the mid-point rule. It is exact for linears but not quadratics since

$$\frac{b^3 - a^3}{3} = \int_a^b x^2 dx \neq (b-a)\left(\frac{a+b}{2}\right)^2.$$

**Example 14.2** (Two points formula). The two-points formula has 4 unknowns (2 weights and 2 interpolation points). We show that we can determine these unknowns for the quadrature to be exact on cubics. For this, the following 4 exactness conditions (see Remark 12.1)

$$\begin{aligned} 2 &= \int_{-1}^1 dx = w_1 + w_2 \\ 0 &= \int_{-1}^1 x dx = w_1 x_1 + w_2 x_2 \\ \frac{2}{3} &= \int_{-1}^1 x^2 dx = w_1 x_1^2 + w_2 x_2^2 \\ 0 &= \int_{-1}^1 x^3 dx = w_1 x_1^3 + w_2 x_2^3. \end{aligned}$$

From the second condition, we deduce that  $w_1 x_1 = -w_2 x_2$ . This into the 4th condition yield

$$0 = w_2 x_2 (x_1^2 - x_2^2).$$

Note that  $w_1 \neq 0$ , for otherwise it would be a one-point rule which cannot be exact for cubics, and  $x_2 \neq 0$  for otherwise the second condition would imply that  $x_1 = x_2 = 0$  and the interpolation points would not be distinct. Therefore, we must have  $x_1^2 = x_2^2$ , i.e.

$$x_1 = -x_2$$

(again, we want distinct interpolation points). Using the second constraint again, this implies that  $w_2x_1 - w_1x_1 = 0$  or

$$w_1 = w_2.$$

Now the second and fourth constraints hold. The first condition requires

$$w_1 + w_2 = 2 \quad \implies \quad w_1 = w_2 = 1.$$

Finally, the third condition implies

$$\frac{2}{3} = 2w_1x_1^2 \quad \text{or} \quad x_1 = \pm\sqrt{\frac{1}{3}}.$$

Finally, the scheme is

$$\int_{-1}^1 f(x) dx \approx f\left(-\frac{1}{\sqrt{3}}\right) + f\left(\frac{1}{\sqrt{3}}\right) := Q(f)$$

and is exact on cubics. It is not exact on quartics since

$$\frac{2}{5} = \int_{-1}^1 x^4 dx \neq Q(x^4) = \frac{1}{9} + \frac{1}{9}.$$

**Example 14.3** (Three points). Can we make a three points quadrature rule exact on  $\mathbb{P}^5$ ? Here the unknowns are  $\{w_i, x_i\}_{i=0}^2$ . Assume the scheme is symmetric about the origin, i.e.

$$Q(f) = w_1f(-x_1) + w_0f(0) + w_1f(x_1).$$

Notice that the symmetry implies that for all odd degree conditions:

$$0 = \int_{-1}^1 x^{2j+1} dx = -w_1x_1^{2j+1} + w_0 \cdot 0 + w_1x_1^{2j+1} = Q(x^{2j+1}), \quad j \geq 0.$$

We now check the even degree conditions:

$$2 = \int_{-1}^1 1 dx \stackrel{?}{=} 2w_1 + w_0$$

$$\frac{2}{3} = \int_{-1}^1 x^2 dx \stackrel{?}{=} 2w_1x_1^2$$

$$\frac{2}{5} = \int_{-1}^1 x^4 dx \stackrel{?}{=} 2w_1x_1^4.$$

Divide the third relation by the second to get

$$\frac{3}{5} = x_1^2 \quad \implies \quad x_1 = \sqrt{\frac{3}{5}}.$$

From the second relation we compute  $w_1$ :

$$\frac{1}{3} = w_1 \frac{3}{5} \quad \implies \quad w_1 = \frac{5}{9}.$$

This in the first constraint implies that

$$\frac{10}{9} + w_0 = 2 \quad \implies \quad w_0 = \frac{8}{9}$$

and the scheme reads

$$Q(f) = \frac{5}{9}f(-\sqrt{3/5}) + \frac{8}{9}f(0) + \frac{5}{9}f(\sqrt{3/5}).$$

This scheme is exact on  $\mathbb{P}^5$  but not on  $\mathbb{P}^6$ .

**Definition 14.1** (Gaussian Quadrature). *A quadrature involving  $n$  points, which is exact on  $\mathbb{P}^{2n+1}$  is called a Gaussian quadrature.*

**Generalization: Weighted Gaussian Quadrature.** Given a non-negative weight functions  $w(x)$  only vanishing at a discrete set of points, we want to derive Gaussian quadrature schemes such that

$$\int_a^b w(x)f(x) dx \approx \sum_{i=0}^n w_i f(x_i).$$

Notice that the assumption on the weight function implies that when  $[\alpha, \beta] \subseteq [a, b]$  with  $\alpha < \beta$  then

$$\int_{\alpha}^{\beta} w(x) dx > 0.$$

We define

$$\langle f, g \rangle_w := \int_a^b w(x)f(x)g(x) dx.$$

The mapping  $\langle \cdot, \cdot \rangle_w$  provides an inner product on  $C[a, b]$ , i.e.

(1)  $\langle \cdot, \cdot \rangle_w$  is bilinear, i.e.

$$\langle \alpha f + \beta g, h \rangle_w = \alpha \langle f, h \rangle_w + \beta \langle g, h \rangle_w,$$

and

$$\langle h, \alpha f + \beta g \rangle_w = \alpha \langle h, f \rangle_w + \beta \langle h, g \rangle_w,$$

for  $f, g, h \in C[a, b]$  and  $\alpha, \beta \in \mathbb{R}$ .

(2)  $\langle \cdot, \cdot \rangle_w$  is symmetric, i.e.

$$\langle f, g \rangle_w = \langle g, f \rangle_w, \quad f, g \in C[a, b].$$

(3)  $\langle \cdot, \cdot \rangle_w$  is positive definite, i.e.

$$\langle f, f \rangle_w \geq 0, \quad f \in C[a, b]$$

and equals 0 only if  $f$  is the zero function, i.e.  $f(x) = 0$ .

The above three properties implies that

$$\|f\|_w := (\langle f, f \rangle_w)^{1/2}$$

is a norm on  $C[a, b]$  and

$$|\langle f, g \rangle_w| \leq \|f\|_w \|g\|_w$$

(Cauchy-Schwartz inequality).

**Definition 14.2** (w-orthogonality). *We say that  $f$  is  $w$ -orthogonal to  $\mathbb{P}^k$  is*

$$\langle f, p \rangle_w = 0 \quad \text{for all } p \in \mathbb{P}^k.$$

**Theorem 14.1** (Gaussian Quadrature). *Suppose there is a nonzero  $q_{k+1} \in \mathbb{P}^{k+1}$  which is  $w$ -orthogonal to  $\mathbb{P}^k$ . If  $q_{k+1}$  has  $k+1$  distinct roots  $\{x_0, \dots, x_k\}$ , then quadrature based on the nodes  $\{x_0, \dots, x_k\}$  approximating*

$$I(f) = \int_a^b w(x)f(x) dx$$

*which is exact on  $\mathbb{P}^k$  is in fact exact on  $\mathbb{P}^{2k+1}$ , i.e. a Gaussian quadrature.*

Note that the quadrature  $Q$  is exact on  $\mathbb{P}^k$  (or  $\mathbb{P}^{2k+1}$ ) means that

$$I(p) = \int_a^b w(x)p(x) dx = Q(p)$$

for every  $p \in \mathbb{P}^k$  (or  $\mathbb{P}^k$ ). As in the case  $w(x) = 1$ , the exactness conditions uniquely determine the quadrature weights.

We postpone the proof of this theorem for later. For now, we make the following remark.

*Remark 14.1* (Leading Coefficient of  $w$ -orthogonal polynomial). If  $q_{k+1} \in \mathbb{P}^{k+1}$  is  $w$ -orthogonal to  $\mathbb{P}^k$  and nonzero then

$$q_{k+1}(x) = a_{k+1}x^{k+1} + a_kx^k + \dots + a_0$$

and  $a_{k+1} \neq 0$ . Indeed, if  $a_{k+1} = 0$  then  $q_{k+1} \in \mathbb{P}^k$  and

$$0 = \langle q_{k+1}, q_{k+1} \rangle_w = \int_a^b w(x)q_{k+1}^2(x) dx,$$

which implies that  $q_{k+1} = 0$  and contradicts our assumption. Moreover, since we are only interested in the roots of  $q_{k+1}$ , we may assume that  $q_{k+1}$  is monic, i.e.  $a_{k+1} = 1$  and

$$q_{k+1}(x) = x^{k+1} + a_kx^k + \dots + a_0.$$

**Example 14.4** ( $k = 1$  and  $w(x) = 1$ ). Find a monic  $q \in \mathbb{P}^2$  with

$$\langle f, g \rangle_w = \int_{-1}^1 f(x)g(x) dx.$$

We are looking for  $\alpha$  and  $\beta$  such that

$$0 = \langle q, 1 \rangle_w = \int_{-1}^1 (x^2 + \alpha x + \beta) dx = \frac{2}{3} + \alpha 0 + 2\beta,$$

i.e.  $2\beta = -\frac{2}{3}$  or  $\beta = -\frac{1}{3}$ . In addition, we want

$$0 = \langle q, x \rangle_w = \int_{-1}^1 (x^3 + \alpha x^2 + \beta x) dx = \frac{2}{3}\alpha,$$

and so  $\alpha = 0$ . This implies that the desired polynomial is

$$q(x) = x^3 - \frac{1}{3},$$

which has two roots, namely

$$\pm \frac{1}{\sqrt{3}}.$$

There are the quadrature nodes derived in Example 14.2.

**Example 14.5** ( $k = 2$  and  $w(x) = 1$ ). Find  $q \in \mathbb{P}^3$ ,

$$q(x) = x^3 + \alpha x^2 + \beta x + \gamma,$$

which is  $w$ -orthogonal to  $\mathbb{P}^2$  with  $w(x) = 1$ . The desired polynomial must satisfy the following 3 constraints

$$\begin{aligned}\alpha \frac{2}{3} + 2\gamma &= \langle q, 1 \rangle_w = 0 \\ \frac{2}{5} + \frac{2}{3}\beta &= \langle q, x \rangle_w = 0 \\ \alpha \frac{2}{5} + \frac{2}{3}\gamma &= \langle q, x^2 \rangle_w = 0.\end{aligned}$$

The first and last constraints hold only if

$$A := \begin{pmatrix} \frac{2}{3} & 2 \\ \frac{2}{5} & \frac{2}{3} \end{pmatrix} \begin{pmatrix} \alpha \\ \gamma \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

Note that  $\det(A) = \frac{4}{9} - \frac{4}{5} \neq 0$ , so the only solution is  $\alpha = \gamma = 0$ . From the second constraint, we find that

$$\frac{1}{5} + \frac{\beta}{3} = 0 \quad \implies \quad \beta = -\frac{3}{5}.$$

The desired polynomial reads

$$q(x) = x^3 - \frac{3}{5}x = (x^2 - \frac{3}{5})x$$

and has roots  $-\sqrt{3/5}$ ,  $0$ ,  $\sqrt{3/5}$ .

## 15. LECTURE 15

We start with the proof of the quadrature theorem (Theorem 14.1).

*Proof of Theorem 14.1.* Let  $\{x_0, \dots, x_k\}$  be distinct roots of  $q_{k+1}$  and let  $Q$  be the associated exact quadrature scheme exact on  $\mathbb{P}^k$ . The latter is assumed to be non zero, belongs to  $\mathbb{P}^{k+1}$  and is  $w$ -orthogonal to  $\mathbb{P}^k$ . Let  $p \in \mathbb{P}^{2k+1}$  and factor (using polynomial division with remainder)

$$p = qs + r,$$

with  $r, s \in \mathbb{P}^k$ . Then,

$$I(p) = \int_a^b w(x)p(x) dx = \underbrace{\int_a^b w(x)q(x)s(x) dx}_{=0} + \int_a^b w(x)r(x) dx$$

using the fact that  $q$  is  $w$ -orthogonal to  $\mathbb{P}^k$  and  $s \in \mathbb{P}^k$ . Now, since  $Q$  is exact on  $\mathbb{P}^k$ , then

$$I(p) = \sum_{j=0}^k w_j r(x_j).$$

Moreover, the nodes  $\{x_0, \dots, x_k\}$  are the roots of  $p$ , so that computing further

$$I(p) = \sum_{j=0}^k w_j r(x_j) = \sum_{j=0}^k w_j \left( \underbrace{q(x_j)}_{=0} s(x_j) + r(x_j) \right) = Q(p),$$

which proves the quadrature is exact on  $\mathbb{P}^{2k+1}$ .  $\square$

**Lemma 15.1** (Roots of  $w$ -orthogonal polynomial). *If  $q \in \mathbb{P}^{k+1}$  is non zero and is  $w$ -orthogonal to  $\mathbb{P}^k$ , then all of the roots of  $q$  are distinct and in  $(a, b)$ .*

*Proof.* We will see in the next lemma (Lemma 15.2) that  $q$  has real coefficients. If  $q$  does not have any root in  $(a, b)$  then  $q > 0$  or  $q < 0$  in  $(a, b)$  and so

$$\int_a^b w(x)q(x) dx > 0 \quad \text{and} \quad \int_a^b w(x)q(x) dx < 0,$$

either contradicting the  $w$ -orthogonality of  $q$  in  $\mathbb{P}^0 \subset \mathbb{P}^k$ .

Suppose now that  $q$  has  $1 \leq l < k+1$  roots in  $(a, b)$ , denoted  $y_1, y_2, \dots, y_l$  (repeated according to their multiplicity), and set

$$r(x) = \prod_{y_j \text{ root of odd multiplicity}} (x - y_j).$$

As the polynomial  $q$  changes sign across a root of odd multiplicity (as does  $r(x)$ ), the product  $q(x)r(x)$  has the same sign except for  $x = y_j$ , where  $y_j$  is a root of odd multiplicity. This implies that

$$\int_a^b w(x)q(x)r(x) dx \neq 0.$$

As  $r \in \mathbb{P}^k$ , this is a contradiction with the  $w$ -orthogonality in  $\mathbb{P}^k$ . As a consequence, there must be  $k+1$  roots of  $q$  in  $(a, b)$  and so they must be distinct for the resulting  $r$  to belong to  $\mathbb{P}^{k+1}$ .  $\square$

**Lemma 15.2** (Real coefficients). *There is a unique monic real polynomial  $q \in \mathbb{P}^{k+1}$ , which is  $w$ -orthogonal to  $\mathbb{P}^k$ .*

*Proof.* Let  $q = x^{k+1} + \alpha_k x^k + \dots + \alpha_0$ . Then  $q$  is a  $w$ -orthogonal to  $\mathbb{P}^k$  if and only if

$$\langle q, x^j \rangle_w = 0, \quad j = 0, \dots, k,$$

i.e.

$$\langle x^{k+1}, x^j \rangle_w + \sum_{l=0}^k \alpha_l \langle x^l, x^j \rangle_w = 0, \quad j = 0, \dots, k.$$

This is equivalent to

$$A \begin{pmatrix} \alpha_0 \\ \dots \\ \alpha_k \end{pmatrix} = F,$$

where the coefficients of the matrix  $A$  are given by

$$A_{j,l} = \langle x^l, x^j \rangle_w \quad j, l = 0, \dots, k$$

and

$$F_j = -\langle x^{k+1}, x^j \rangle_w, \quad j = 0, \dots, k.$$

Suppose that  $A\beta = 0$  for some  $\beta \in \mathbb{R}^{k+1}$ . Set

$$r(x) = \beta_k x^k + \beta_{k-1} x^{k-1} + \dots + \beta_0.$$

The  $j$ th equation of  $A\beta = 0$  is

$$0 = \sum_{l=0}^k A_{j,l} \beta_l = \sum_{l=0}^k \langle x^l, x^j \rangle_w \beta_l = \sum_{l=0}^k \langle \beta_l x^l, x^j \rangle_w = \langle r(x), x^j \rangle_w, \quad j = 0, 1, \dots, k,$$

i.e.  $r(x)$  is  $w$ -orthogonal to  $\mathbb{P}^k$ . As  $r \in \mathbb{P}^k$

$$0 = \langle r(x), r(x) \rangle_w \implies r(x) = 0, \quad \text{i.e.} \quad \beta = 0.$$

This proves that  $A$  is nonsingular. As  $A$  is real valued, so is  $A^{-1}$ . (For instance, the inverse can be computed by row reducing  $(A : I) \rightarrow (I : A^{-1})$ .)  $\square$

Every weighted quadrature problem gives rise to a sequence of orthogonal polynomial. The sequence follows a 3 term recurrence.

Start with  $\tilde{p}_0 = 1 \in \mathbb{P}^0$  (nothing to be orthogonal to). Then  $\tilde{p}_1 \in \mathbb{P}^1$  must be orthogonal to 1. If  $\tilde{p}_1(x) = x + \alpha$  then  $\alpha$  must satisfy

$$0 = \langle x + \alpha, 1 \rangle_w, \quad \text{or} \quad \alpha = -\langle x, 1 \rangle_w.$$

Suppose we have computed  $\tilde{p}_{j-1}$  and  $\tilde{p}_j$ . Write

$$\tilde{p}_{j+1} = (x + \alpha)\tilde{p}_j + \beta\tilde{p}_{j-1}.$$

Then for  $\theta \in \mathbb{P}^{j-2}$

$$\langle \tilde{p}_{j+1}, \theta \rangle_w = \langle (x + \alpha)\tilde{p}_j, \theta \rangle_w + \beta \underbrace{\langle \tilde{p}_{j-1}, \theta \rangle_w}_{=0} = \underbrace{\langle \tilde{p}_j, (x + \alpha)\theta \rangle_w}_{=0} = 0.$$

We also need

$$0 = \langle \tilde{p}_{j+1}, \tilde{p}_{j-1} \rangle_w = \langle (x + \alpha)\tilde{p}_j, \tilde{p}_{j-1} \rangle_w + \beta \langle \tilde{p}_{j-1}, \tilde{p}_{j-1} \rangle_w.$$

The  $\alpha$  term goes away so

$$\beta = -\frac{\langle x\tilde{p}_j, \tilde{p}_{j-1} \rangle_w}{\langle \tilde{p}_{j-1}, \tilde{p}_{j-1} \rangle_w}.$$



Also

$$0 = \langle \tilde{p}_{j+1}, \tilde{p}_j \rangle_w = \langle (x + \alpha) \tilde{p}_j, \tilde{p}_j \rangle_w + \underbrace{\beta \langle \tilde{p}_{j-1}, \tilde{p}_j \rangle_w}_{=0},$$

and so we find

$$\alpha = -\frac{\langle x \tilde{p}_j, \tilde{p}_j \rangle_w}{\langle \tilde{p}_j, \tilde{p}_j \rangle_w}.$$

The values of  $\alpha$  and  $\beta$  determines  $\tilde{p}_{j+1}$ . Note that the orthogonal polynomials always satisfy 3 term recurrence relations!

### Rodrigues Formula for Legendre Polynomials.

**Example 15.1.**  $w(x) = 1$ ,  $a = -1$ ,  $b = 1$ ] Consider the approximation  $I(f) = \int_{-1}^1 f(x) dx$ . Then

$$\tilde{p}_n(x) = \frac{1}{(2n)(2n-1) \cdot \dots \cdot (n+1)} \frac{d^n}{dx^n} [(x^2 - 1)^n].$$

To see this, we check that it is a monic polynomial of degree  $n$  and  $w$ -orthogonal to  $\mathbb{P}^{n-1}$ . We leave the first part as an exercise (Exercise 15.1). Now, if  $p \in \mathbb{P}^{n-1}$

$$\int_{-1}^1 \frac{d^n}{dx^n} [(x^2 - 1)^n] p(x) dx = \underbrace{\frac{d^n}{dx^n} [(x^2 - 1)^n] \Big|_{-1}^1}_{=0} - \int_{-1}^1 \frac{d^{n-1}}{dx^{n-1}} [(x^2 - 1)^n] p'(x) dx.$$

Repeating this by moving all derivatives over to  $p$ , we arrive at

$$\int_{-1}^1 \frac{d^n}{dx^n} [(x^2 - 1)^n] p(x) dx = (-1)^n \int_{-1}^1 (x^2 - 1)^n \underbrace{p^{(n)}(x)}_{=0} dx = 0$$

because  $p \in \mathbb{P}^{n-1}$ .

**Definition 15.1** (Legendre Polynomials). The polynomial

$$\frac{1}{2^n n!} \frac{d^n}{dx^n} [(x^2 - 1)^n].$$

is called the Legendre polynomial (different normalization) and satisfies

$$(n+1)p_{n+1}(x) = (2n+1)xp_n(x) - np_n(x).$$

**Exercise 15.1.** Show that

$$\tilde{p}_n(x) = \frac{1}{(2n)(2n-1) \cdot \dots \cdot (n+1)} \frac{d^n}{dx^n} [(x^2 - 1)^n]$$

is a polynomial of degree  $n$  and is monic (i.e. the leading coefficient is 1).

### Rodrigues Formula for Chebyshev Polynomials.

**Example 15.2** (Chebyshev polynomials). We recall that the Chebyshev are given by

$$T_n(x) = \cos(n \cos^{-1}(x)).$$

Note that for  $n \neq j$

$$\int_0^\pi \cos(n\theta) \cos(j\theta) d\theta = 0.$$

We leave the above claim as exercise.

Set  $\theta = \cos^{-1}(x)$ , then  $x = \cos(\theta)$  and

$$dx = -\sin(\theta)d\theta = -\sqrt{1 - \cos^2(\theta)} d\theta = -\sqrt{1 - x^2} d\theta.$$

Using the orthogonality above

$$\begin{aligned} 0 &= \int_0^\pi \cos(n\theta) \cos(j\theta) d\theta = \int_{-1}^1 \cos(n \cos^{-1}(x)) \cos(j \cos^{-1}(x)) \frac{dx}{\sqrt{1-x^2}} \\ &= \int_{-1}^1 \frac{1}{\sqrt{1-x^2}} T_n(x) T_j(x) dx, \end{aligned}$$

i.e. the Chebyshev polynomial  $T_n$  are orthogonal polynomials on  $-1, 1$  with weights  $w(x) = \frac{1}{\sqrt{1-x^2}}$  and satisfy the recurrence

$$T_{n+1}(x) = 2xT_n(x) - T_{n-1}(x)$$

as we have seen already in Section 5.1.

## 16. LECTURE 16

We saw last lecture that the Chebyshev polynomials are orthogonal with respect to the scalar product

$$\langle f, g \rangle_w = \int_{-1}^1 \underbrace{\frac{1}{\sqrt{1-x^2}}}_{=:w(x)} f(x)g(x) dx,$$

i.e.  $T_{n+1}$  satisfies  $T_{n+1} \in \mathbb{P}^{n+1}$  and

$$\langle T_{n+1}, T_j \rangle_w = 0, \quad 0 \leq j \leq n.$$

As  $\{T_j\}_{j=0}^n$  is a basis for  $\mathbb{P}^n$ ,  $T_{n+1}$  is  $w$ -orthogonal to  $\mathbb{P}^n$ .

The Rodrigues formula for the Chebyshev polynomials reads

$$\tilde{T}_n = w(x) \frac{d^n}{dx^n} (w(x)(1-x^2)^n) = \frac{1}{(1-x^2)^{1/2}} \frac{d^n}{dx^n} ((1-x^2)^{n-1/2})$$

using the definition of the weight  $w(x) = (1-x^2)^{-1/2}$ . Note that

$$\frac{d^n}{dx^n} (fg) = \sum_{j=0}^n \binom{n}{j} f^{(j)} g^{(n-j)}$$

(you can prove this by induction). Therefore,

$$\begin{aligned} \tilde{T}_n(x) &= \frac{1}{(1-x^2)^{1/2}} \frac{d^n}{dx^n} ((1-x)^{n-1/2}(1+x)^{n-1/2}) \\ &= \frac{1}{(1-x^2)^{1/2}} \sum_{j=0}^n \binom{n}{j} c_{j,n} (1-x)^{n-j-1/2} (1+x)^{j-1/2} \\ &= \sum_{j=0}^n \binom{n}{j} c_{j,n} (1-x)^{n-j} (1+x)^j. \end{aligned}$$

This proves that  $\tilde{T}_n \in \mathbb{P}^n$ . We now check that it is  $w$ -orthogonal. We use integration by parts again:

$$\begin{aligned} I &:= \int_{-1}^1 (1-x^2)^{-1/2} \tilde{T}_n(x) p(x) dx \\ &= \int_{-1}^1 \frac{d^n}{dx^n} ((1-x^2)^{-1/2} (1-x^2)^n) p(x) dx \\ &= p(x) \frac{d^{n-1}}{dx^{n-1}} ((1-x)^{n-1/2} (1+x)^{n-1/2}) \Big|_{x=-1}^{x=1} - \int_{-1}^1 \frac{d^{n-1}}{dx^{n-1}} ((1-x^2)^{-1/2} (1-x^2)^n) p'(x) dx \\ &= - \sum_{j=0}^{n-1} \binom{n-1}{j} c_{n-1,j} (1-x)^{n-j-1/2} (1+x)^{j+1-1/2}. \end{aligned}$$

Note that all terms evaluated at either  $x = -1$  or  $x = 1$  are zero because they have positive powers of  $(1-x)$  and  $(1+x)$ . Repeating the argument gives

$$I = (-1)^n \int_{-1}^1 (1-x^2)^{-1/2} (1-x^2)^n p^{(n)}(x) dx = 0$$

for  $p \in \mathbb{P}^{n-1}$ . Since,  $T_n$  and  $\tilde{T}_n$  differ at most by a normalization constant,  $T_n$  is also a polynomial of degree  $n$ ,  $w$ -orthogonal to  $\mathbb{P}^{n-1}$ .