

# Integration with Gaussian quadratures

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## 1 Theoretical elements

Our goal is to numerically calculate the definite integral

$$I = \int_a^b f(x) dx \quad (1)$$

with  $f$  a well-behaved function (smooth and non-diverging) in the interval  $[a, b]$ . The **mean value theorem for integrals**<sup>1</sup> allows us to write the integral  $I$  as a sum of  $N$  terms

$$I = \sum_{i=1}^N \tilde{w}_i f(x_i) \quad x_i \in [a, b] \quad \forall i \quad (2)$$

where  $\{x\}$  are the **integration nodes** and  $\{\tilde{w}\}$  the **weights**. Solving the above equality would mean finding the exact values of  $\{x\}$  and  $\{\tilde{w}\}$ , for *all*  $f$ . In theory, this could be possible, but then we'd have an analytical solution to the problem, thus no need for a numerical method.

Developing a numerical integration method consists in requiring that the equality holds exactly for *some*  $f$  and finding suitable values for the nodes and weights. For other functions, the sum will then be an *approximation* to the integral, so it's formally more correct to write:  $I \simeq \sum_{i=1}^N \tilde{w}_i f(x_i)$ . Note that we'll need  $2N$  conditions to solve the  $2N$  unknowns ( $N$  nodes and  $N$  weights).

As an example, let's analyze the popular trapezoid rule, in which we impose that  $x_1 = a$ ,  $x_N = b$  and that all nodes be equidistant. These  $N$  conditions for the nodes yield the familiar<sup>2</sup>  $x_i = a + (i-1) \cdot h$ , with  $h = (b-a)/(N-1)$ . On the other hand, we require the method to be *exact* for polynomials of degree zero *and* for polynomials of degree one, which translates to  $N$  additional conditions. These allow us to find the values for the weights:  $\tilde{w}_1 = \tilde{w}_N = h/2$  and  $\tilde{w}_i = h, i \in \{2, 3, \dots, N-1\}$ .

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<sup>1</sup> see appendix A

<sup>2</sup> Beware of the notation!  $N$  is the total number of nodes, not subintervals.

## 1.1 Gaussian quadratures

Before continuing we introduce a small change of notation<sup>3</sup>,  $f = g \cdot v$ . We're also slightly changing the equality between integral and sum<sup>4</sup>

$$\int_a^b g(x)v(x) dx \simeq \sum_{i=1}^N w_i g(x_i) . \quad (3)$$

Remember that our problem consists in solving  $2N$  unknowns ( $N$  nodes and  $N$  weights), so we'll need  $2N$  conditions. In Gaussian quadratures, we demand our solution to be *exact* for any polynomial  $T$ , with  $0 \leq \deg(T) \leq 2N - 1$ . *Habemus* our  $2N$  conditions!

Let  $\{\phi\}$  be an **orthogonal polynomial basis** of  $L_v^2[a, b]$

$$\langle \phi_k | \phi_l \rangle = \int_a^b \phi_k(x) \phi_l(x) v(x) dx = A_k^2 \delta_{k,l} \quad (4)$$

with  $A_k^2$  the normalization constant and  $v(x)$  the weight function<sup>5</sup>. Additionally we demand that  $\phi_k$  has  $k$  **simple roots** in  $[a, b]$ .

Now we take *any* polynomial of degree  $N - 1$ ,  $R_{N-1}$ , which can be expanded (in the interval  $[a, b]$ ) using the chosen orthogonal basis

$$R_{N-1}(x) = \sum_{k=0}^{N-1} \alpha_k \phi_k(x) , \quad (5)$$

and construct the function  $g = \phi_N R_{N-1}$ . Note that  $g$  is a polynomial of order  $2N - 1$ ; if we calculate its integral, the equality in (3) will hold *exactly*

$$\int_a^b \phi_N(x) R_{N-1}(x) v(x) dx = \sum_{i=1}^N w_i \phi_N(x_i) R_{N-1}(x_i) . \quad (6)$$

Writing out the LHS, using (4) and (5)

$$\int_a^b \phi_N(x) R_{N-1}(x) v(x) dx = \sum_{k=0}^{N-1} \alpha_k \int_a^b \phi_N(x) \phi_k(x) v(x) dx = \sum_{k=0}^{N-1} \alpha_k A_k^2 \delta_{k,N} = 0 . \quad (7)$$

The RHS of (6) must also sum zero. In general,  $R_{N-1}(x_i) \neq 0$  and, obviously,  $w_i \neq 0$ ; thus:  $\phi_N(x_i) = 0$ .

**First result:** Our integration nodes will be the roots of the  $N$ -th degree polynomial of the  $L_v^2[a, b]$  orthogonal basis  $\{\phi\}$ .

<sup>3</sup> If  $f$  is well-behaved, there shouldn't be any problem in obtaining this decomposition.

<sup>4</sup> You might be alarmed by the fact that we're messing with the mean value theorem, but we've only changed the weights (there's a map from the previously defined weights to these new ones).

<sup>5</sup> Not to be confused with our weights  $\{w\}$ !

## 1.2 Lagrange interpolation

So far we've solved the  $N$  nodes  $\{x_i\}$ , using an  $L_v^2[a, b]$  orthogonal polynomial basis  $\{\phi\}$ . We still have to find the value for the  $N$  weights  $\{w_i\}$ ; to do so we use **Lagrange interpolation**.

Consider *any* function  $h$  and its Lagrange interpolator  $H_{N-1}$ , a polynomial of degree  $N - 1$ . In the interval  $[a, b]$ , we want them to coincide at  $N$  points, specifically the  $N$  roots of  $\phi_N$ ,  $\{x_j\}$ . We define  $H_{N-1}$  as

$$H_{N-1}(x) = \sum_{j=1}^N h(x_j) \frac{\phi_N(x)}{\phi'_N(x_j)(x - x_j)} \quad (8)$$

with  $\phi'_N$  the derivative of  $\phi_N$ . One might be worried about  $H_{N-1}$  diverging (we're apparently dividing by zero!). Remember that any polynomial can be written as a product of roots, particularly

$$\phi_N = c_N(x - x_1)(x - x_2) \cdots (x - x_N) \quad (9)$$

which makes it easy to check that for each term of the sum, the apparent zero in the denominator cancels out with a term in the numerator<sup>6</sup>. On the other hand, remember that  $\phi_N$  has  $N$  simple roots, which implies<sup>7</sup>  $\phi'_N(x_j) \neq 0, \forall j$ . Let's check if the two functions  $h$  and  $H_{N-1}$  coincide at the  $N$  roots<sup>8</sup>

$$H_{N-1}(x_k) = \lim_{x \rightarrow x_k} H_{N-1}(x) = \sum_{j=1}^N h(x_j) \lim_{x \rightarrow x_k} \frac{\phi_N(x)}{\phi'_N(x_j)(x - x_j)} = \sum_{j=1}^N h(x_j) \delta_{j,k} = h(x_k). \quad (10)$$

Now we particularize this expression for a polynomial of degree  $2N - 1$ ,  $q_{2N-1}$ , whose Lagrange interpolator we denote  $Q_{N-1}$ . Remember that  $q_{2N-1}$  and  $Q_{N-1}$  coincide at  $N$  points, the roots of  $\phi_N$ . We can write the difference  $s = q_{2N-1} - Q_{N-1}$  (note that  $\deg(s) = 2N - 1$ ) as

$$s(x) = q_{2N-1}(x) - Q_{N-1}(x) \quad (11)$$

$$= B(x - x_1)(x - x_2) \cdots (x - x_N)(x - \tilde{x}_1)(x - \tilde{x}_2) \cdots (x - \tilde{x}_{N-1}). \quad (12)$$

The first  $N$  roots of  $s$  coincide with those of  $\phi_N$ , the remaining  $N - 1$  roots isn't relevant. We define one last polynomial

$$r(x) = \frac{s(x)}{\phi_N(x)} = \frac{q_{2N-1}(x) - Q_{N-1}(x)}{\phi_N(x)} = B(x - \tilde{x}_1)(x - \tilde{x}_2) \cdots (x - \tilde{x}_{N-1}), \quad (13)$$

with  $\deg(r) = N - 1$ . Rearranging expression (13)

$$q_{2N-1} = Q_{N-1} + r\phi_N. \quad (14)$$

Finally we identify  $q_{2N-1}$  with the function  $g$  and, as we've done before, calculate its integral. Again, the equality in (3) will hold *exactly*

$$\int_a^b q_{2N-1}(x)v(x) dx = \sum_{i=1}^N w_i q_{2N-1}(x_i). \quad (15)$$

Using (14) we write out the LHS in (15)

$$\int_a^b q_{2N-1}(x)v(x) dx = \int_a^b Q_{N-1}(x)v(x) dx + \int_a^b r(x)\phi_N(x)v(x) dx \quad (16)$$

<sup>6</sup> This also assures that  $\deg(H_{N-1}) = N - 1$ .

<sup>7</sup> You might want to review your calculus notes ☺

<sup>8</sup> The limit is not too difficult to calculate, try it! If you get stuck, see appendix B.

and substitute (8) in the first term

$$\int_a^b Q_{N-1}(x)v(x) dx = \int_a^b \sum_{j=1}^N q_{2N-1}(x_j) \frac{\phi_N(x)}{\phi'_N(x_j)(x-x_j)} v(x) dx. \quad (17)$$

Expanding  $r$  in the orthogonal basis  $\{\phi\}$

$$r(x) = \sum_{k=0}^{N-1} \beta_k \phi_k(x) \quad (18)$$

and applying property (4) to the second term in (16)

$$\int_a^b r(x) \phi_N(x) v(x) dx = \sum_{k=0}^{N-1} \beta_k \int_a^b \phi_N(x) \phi_k(x) v(x) dx = \sum_{k=0}^{N-1} \beta_k A_k^2 \delta_{k,N} = 0. \quad (19)$$

Equating to the RHS in (15)

$$\sum_{j=1}^N q_{2N-1}(x_j) \int_a^b \frac{\phi_N(x)v(x)}{\phi'_N(x_j)(x-x_j)} dx = \sum_{i=1}^N w_i q_{2N-1}(x_i). \quad (20)$$

From this last expression we find the weights<sup>9</sup>, calculating  $N$  integrals, one for each root  $x_i$ .

**Second result:** We'll calculate our weights with the formula  $w_i = \frac{1}{\phi'_N(x_i)} \int_a^b \frac{\phi_N(x)v(x)}{x-x_i} dx$ .

### 1.3 Mathemagics

Virtually we've solved our problem of finding  $\{x_i\}$  and  $\{w_i\}$ , but we've gone from calculating one single integral to calculating  $N$  integrals, one for each of the weights. This is rather counter-intuitive and time-consuming, specially if we choose a large value for  $N$  in order to obtain a *good* approximation.

We start solving this problem with something quite simple, adding and subtracting

$$\frac{1}{x-x_i} = \frac{1}{x-x_i} + \frac{1}{x-x_i} \left(\frac{x}{x_i}\right)^k - \frac{1}{x-x_i} \left(\frac{x}{x_i}\right)^k \quad (21)$$

$$= \frac{1}{x-x_i} \left[ 1 - \left(\frac{x}{x_i}\right)^k \right] + \frac{1}{x-x_i} \left(\frac{x}{x_i}\right)^k \quad (22)$$

$$= \frac{1}{x_i^k} \frac{x_i^k - x^k}{x-x_i} + \frac{1}{x_i^k} \frac{x^k}{x-x_i}. \quad (23)$$

Using equality (23) we calculate the integral

$$\int_a^b \frac{\phi_N(x)v(x)}{x-x_i} dx = \frac{1}{x_i^k} \int_a^b \phi_N(x)v(x) \frac{x_i^k - x^k}{x-x_i} dx + \frac{1}{x_i^k} \int_a^b \frac{x^k \phi_N(x)v(x)}{x-x_i} dx. \quad (24)$$

Note that the fraction  $(x_i^k - x^k)/(x-x_i)$  in the first term is a polynomial of degree  $k-1$  (both numerator and denominator share a common root,  $x_i$ ). If  $k \leq N$ , this polynomial can be expanded in the interval  $[a, b]$  as

$$\frac{x_i^k - x^k}{x-x_i} = \sum_{m=0}^{N-1} \gamma_m \phi_m(x).$$

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<sup>9</sup> Note that both  $i$  and  $j$  are dummy indices.

Applying property (4) in the first term of (24)

$$\frac{1}{x_i^k} \int_a^b \phi_N(x) v(x) \frac{x_i^k - x^k}{x - x_i} dx = \sum_{m=0}^{N-1} \frac{\gamma_m}{x_i^k} \int_a^b \phi_N(x) \phi_m(x) v(x) dx = \sum_{m=0}^{N-1} \frac{\gamma_m}{x_i^k} A_m^2 \delta_{m,N} = 0, \quad (25)$$

and (24) reduces to

$$\int_a^b \frac{\phi_N(x) v(x)}{x - x_i} dx = \frac{1}{x_i^k} \int_a^b \frac{x^k \phi_N(x) v(x)}{x - x_i} dx. \quad (26)$$

This “decomposition” is generalizable to any polynomial  $S$  with  $\deg(S) \leq N$

$$\frac{1}{x - x_i} = \frac{1}{x - x_i} \left[ 1 - \frac{S(x)}{S(x_i)} \right] + \frac{1}{x - x_i} \frac{S(x)}{S(x_i)} \quad (27)$$

$$\int_a^b \frac{\phi_N(x) v(x)}{x - x_i} dx = \frac{1}{S(x_i)} \int_a^b \frac{S(x) \phi_N(x) v(x)}{x - x_i} dx. \quad (28)$$

In particular, we choose  $S = \phi_{N-1}$  and rewrite the expression for our weights

$$w_i = \frac{1}{\phi'_N(x_i) \phi_{N-1}(x_i)} \int_a^b \frac{\phi_N(x) \phi_{N-1}(x) v(x)}{x - x_i} dx. \quad (29)$$

Remember from (9) that  $\phi_N(x) = c_N(x - x_1)(x - x_2) \cdots (x - x_i) \cdots (x - x_N)$  and note that  $\phi_N(x)/(x - x_i)$  is a polynomial of degree  $N - 1$ , which can be written as<sup>10</sup>

$$\frac{\phi_N(x)}{x - x_i} = c_N x^{N-1} + \sum_{k=0}^{N-2} C_k x^k = c_N x^{N-1} + \sum_{k=0}^{N-2} \epsilon_k \phi_k(x). \quad (30)$$

Substituting in the integral of (29) and applying property (4)

$$\int_a^b \frac{\phi_N(x) \phi_{N-1}(x) v(x)}{x - x_i} dx = c_N \int_a^b x^{N-1} \phi_{N-1}(x) v(x) dx + \sum_{k=0}^{N-2} \epsilon_k \int_a^b \phi_k(x) \phi_{N-1}(x) v(x) dx \quad (31)$$

$$= c_N \int_a^b x^{N-1} \phi_{N-1}(x) v(x) dx + \sum_{k=0}^{N-2} \epsilon_k A_k^2 \delta_{k,N-1} \quad (32)$$

$$= c_N \int_a^b x^{N-1} \phi_{N-1}(x) v(x) dx. \quad (33)$$

Finally, to get rid of the  $x^{N-1}$  in (33) we write

$$\phi_{N-1}(x) = c_{N-1} x^{N-1} + \sum_{k=0}^{N-2} D_k x^k = c_{N-1} x^{N-1} + \sum_{k=0}^{N-2} \zeta_k \phi_k \quad (34)$$

$$x^{N-1} = \frac{\phi_{N-1}(x)}{c_{N-1}} - \sum_{k=0}^{N-2} \frac{\zeta_k}{c_{N-1}} \phi_k. \quad (35)$$

Substituting in the integral of (33) and applying property (4)

$$c_N \int_a^b x^{N-1} \phi_{N-1}(x) v(x) dx = \frac{c_N}{c_{N-1}} \int_a^b \phi_{N-1}^2(x) v(x) dx - \sum_{k=0}^{N-2} \frac{c_N \zeta_k}{c_{N-1}} \int_a^b \phi_k(x) \phi_{N-1}(x) v(x) dx \quad (36)$$

$$= \frac{c_N}{c_{N-1}} A_{N-1}^2 - \sum_{k=0}^{N-2} \frac{c_N \zeta_k}{c_{N-1}} \delta_{k,N-1} \quad (37)$$

$$= \frac{c_N}{c_{N-1}} A_{N-1}^2. \quad (38)$$

<sup>10</sup> Once again, we can expand the polynomial  $\sum_{k=0}^{N-2} C_k x^k$  in the orthogonal basis  $\{\phi\}$ .

Tracing these results back to (29) we obtain a very compact expression for the weights

$$w_i = \frac{A_{N-1}^2}{\phi'_N(x_i)\phi_{N-1}(x_i)} \frac{c_N}{c_{N-1}}. \quad (39)$$

**Conclusion:** Choosing an  $L^2_v[a, b]$  orthogonal polynomial basis  $\{\phi\}$ , our integration nodes will be the roots of  $\phi_N$ ,  $\{x_i | \phi_N(x_i) = 0\}$ , and we'll calculate the weights using the formula  $w_i = \frac{A_{N-1}^2}{\phi'_N(x_i)\phi_{N-1}(x_i)} \frac{c_N}{c_{N-1}}$ , with  $A^2$  the normalization constant and  $c$  the coefficient of the largest monomial.

## 2 Gauss-Legendre quadrature

In the previous section we've derived the general form of Gaussian quadratures. Let's remember the relevant formulae

$$\int_a^b g(x)v(x) dx \simeq \sum_{i=1}^N w_i g(x_i) \quad (40)$$

$$x_i \rightarrow \phi_N(x_i) = 0 \quad (41)$$

$$w_i = \frac{A_{N-1}^2}{\phi'_N(x_i)\phi_{N-1}(x_i)} \frac{c_N}{c_{N-1}}. \quad (42)$$

Note that the only variability lies in the orthogonal polynomial basis  $\{\phi\}$ . The question that arises is: what basis do we use? The choice depends on the interval  $[a, b]$ .

If it's semi-infinite ( $a = 0, b = \infty$ ), an appropriate basis are Laguerre polynomials; if the interval is infinite ( $a = -\infty, b = \infty$ ), we'd use Hermite polynomials. If the interval is finite we can use Legendre polynomials or Chebyshev polynomials of the second kind. We choose the former since their weight function is  $v(x) = 1$ , as opposed to  $v(x) = \sqrt{1-x^2}$  for the latter.

### 2.1 Interval shift

Legendre polynomials are defined in the interval  $x \in [-1, 1]$ , i.e. the conclusions of the previous section are valid *only* if  $a = -1$  and  $b = 1$ . If we want to compute an integral with other bounds

$$I = \int_a^b f(y) dy \quad (43)$$

we have to perform a change of variables

$$y = \frac{b-a}{2}x + \frac{a+b}{2} \rightarrow dy = \frac{b-a}{2}dx \quad (44)$$

$$I = \frac{b-a}{2} \int_{-1}^1 f(x) dx. \quad (45)$$

### 2.2 Normalization constant

The normalization constant for Legendre polynomials is

$$A_n^2 = \int_{-1}^1 P_n^2(x) dx = \frac{2}{2n+1}, \quad (46)$$

and particularly of our interest

$$A_{N-1}^2 = \frac{2}{2(N-1)+1} = \frac{2}{2N-1}. \quad (47)$$

### 2.3 Rodrigues' formula

Legendre polynomials may be expressed using Rodrigues' formula

$$P_n(x) = \frac{1}{2^n n!} \frac{d^n}{dx^n} \left( (x^2 - 1)^n \right), \quad (48)$$

which can be manipulated slightly

$$P_n(x) = \frac{1}{2^n n!} \frac{d^n}{dx^n} (x^{2n} + O(x^{2n-1})) = \frac{1}{2^n n!} \left( \frac{(2n)!}{n!} x^n + O(x^{n-1}) \right). \quad (49)$$

From this expression we find

$$c_n = \frac{1}{2^n n!} \frac{(2n)!}{n!} = \frac{(2n)!}{2^n (n!)^2}, \quad (50)$$

and particularly

$$c_N = \frac{(2N)!}{2^N (N!)^2} \quad (51)$$

$$c_{N-1} = \frac{(2N-2)!}{2^{N-1} ((N-1)!)^2}. \quad (52)$$

Now we can compute the fraction

$$\frac{c_N}{c_{N-1}} = \frac{(2N)!}{2^N (N!)^2} \frac{2^{N-1} ((N-1)!)^2}{(2N-2)!} \quad (53)$$

$$= \frac{2N \cdot (2N-1) \cdot (2N-2)!}{2 \cdot 2^{N-1} \cdot N^2 \cdot ((N-1)!)^2} \frac{2^{N-1} ((N-1)!)^2}{(2N-2)!} \quad (54)$$

$$= \frac{2N \cdot (2N-1)}{2 \cdot N^2} \quad (55)$$

$$= \frac{2N-1}{N}. \quad (56)$$

## 2.4 Recurrence relations

There are a few interesting recurrence relations involving Legendre polynomials. Of particular interest are

$$(n+1)P_{n+1}(x) = (2n+1)xP_n(x) - nP_{n-1}(x) \quad (57)$$

and

$$(x^2 - 1)P'_n(x) = nxP_n(x) - nP_{n-1}(x). \quad (58)$$

We'll use the first in our algorithm and from the second

$$(x_i^2 - 1)P'_N(x_i) = Nx_iP_N(x_i) - NP_{N-1}(x_i) \rightarrow P'_N(x_i) = \frac{NP_{N-1}(x_i)}{1 - x_i^2}, \quad (59)$$

where we've used  $P_N(x_i) = 0$ .

Plugging (47), (56) and (59) into (42) we find

$$w_i = \frac{2}{N^2} \frac{1 - x_i^2}{P_{N-1}^2(x_i)}. \quad (60)$$



### 3 Computational algorithm

User input parameters:

- number of integration nodes,  $N$
- integral bounds,  $a$  and  $b$
- function,  $f(y)$

The steps we have to implement are:

1. find the  $N$  roots of  $P_N$ ,  $\{x_i\}$
2. calculate the weights,  $\{w_i\}$
3. perform the sum  $\sum_{i=1}^N w_i f(x_i)$

To find the  $N$  roots of  $P_N$  we'll combine the secant and bisection methods. In order to do so, we'll need to narrow the interval where each root lies.

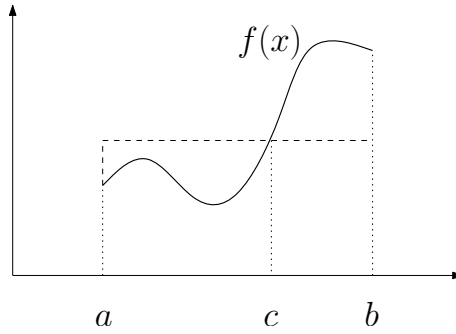
1. start with a step of  $\Delta = 2/N$
2. sweep the interval  $[-1, 1]$  with step  $\Delta$ ,  $x_{j+1} = x_j + \Delta$ 
  - compare the sign of  $P_N(x_j)$  and  $P_N(x_{j+1})$
  - if there's a change of sign, record  $x_j$  and  $x_{j+1}$
3. count the number of records,  $M$ 
  - if  $M < N$  repeat step 2 with  $\Delta = \Delta/10$
  - if  $M = N$  we're finished

Now that we've delimited the interval where each root lies we apply the enhanced secant method to find the integration nodes with a fixed precision,  $\epsilon$ . This enhanced method consists simply in correcting the secant result,  $x_{i+1} = x_i - f(x_i)(x_i - x_{i-1})/(f(x_i) - f(x_{i-1}))$ , if it diverges from the interval, using the bisection result instead.

To compute the values of the Lagrange polynomials we'll need the recurrence relation (57) and the fact that  $P_0 = 1$  and  $P_1 = x$ .

Remember that we'll have to remap the variable  $x$  into the variable  $y$ , using (44), in order to compute  $f(x_i)$ .

## Appendix A



Given a well-behaved function  $f$ , there exists a value  $c$ ,  $a < c < b$ , such that

$$\int_a^b f(x) dx = (b - a) \cdot f(c) . \quad (61)$$

Graphically this means that the area below the curve (between  $a$  and  $c$ ) has the same value as the area above the curve (between  $c$  and  $b$ ).

## Appendix B

We want to show

$$\lim_{x \rightarrow x_k} \frac{\phi_N(x)}{\phi'_N(x_j)(x - x_j)} = \delta_{j,k} . \quad (62)$$

Rewriting (9)

$$\phi_N(x) = c_N(x - x_1)(x - x_2) \cdots (x - x_N) = c_N \prod_{i=1}^N (x - x_i) . \quad (63)$$

The denominator  $x - x_j$  cancels with a term in the numerator, thus

$$\frac{\phi_N(x)}{x - x_j} = c_N \prod_{i \neq j}^N (x - x_i) . \quad (64)$$

Differentiating (63)

$$\begin{aligned} \phi'_N(x) &= c_N [(x - x_2)(x - x_3) \cdots (x - x_N) + (x - x_1)(x - x_3) \cdots (x - x_N) + \cdots \\ &\quad + (x - x_1)(x - x_2) \cdots (x - x_{N-1})] = c_N \sum_{m=1}^N \frac{\prod_{i=1}^N (x - x_i)}{x - x_m} = c_N \sum_{m=1}^N \prod_{i \neq m}^N (x - x_i) . \end{aligned} \quad (65)$$

Once again, the denominator  $x - x_m$  cancels with a term in the numerator. Considering a specific value of  $x = x_j$

$$\phi'_N(x_j) = c_N \sum_{m=1}^N \prod_{i \neq m}^N (x_j - x_i) . \quad (66)$$

Most of the summands in (66) include the term  $i = j \rightarrow x_j - x_i = 0$ . The only summand that survives is  $i \neq j$ , i.e.  $m = j$  and

$$\phi'_N(x_j) = c_N \prod_{i \neq j} (x_j - x_i). \quad (67)$$

Substituting these expressions in the limit

$$\lim_{x \rightarrow x_k} \frac{\phi_N(x)}{\phi'_N(x_j)(x - x_j)} = \lim_{x \rightarrow x_k} \frac{c_N \prod_{i \neq j} (x - x_i)}{c_N \prod_{i \neq j} (x_j - x_i)} = \frac{\prod_{i \neq j} (x_k - x_i)}{\prod_{i \neq j} (x_j - x_i)}. \quad (68)$$

It's clear that in the case  $k = j$  numerator and denominator are equal and the limit's value is 1. On the other hand, if  $k \neq j$  there's a term in the numerator with  $i = k \rightarrow x_k - x_i = 0$  and the limit's value is 0. Therefore

$$\lim_{x \rightarrow x_k} \frac{\phi_N(x)}{\phi'_N(x_j)(x - x_j)} = \delta_{j,k} \quad \text{Q.E.D.} \quad (69)$$