

Numerical Analysis

Derivation of Gauss-Quadrature Formula using Legendre Polynomials







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- 1 Introduction to Legendre Polynomials
- 2 Lagrange Polynomial Interpolation Using Gauss Quadrature Nodes
- **3** Computer Model to Compute Quadrature Nodes and Weights



Introduction to Legendre Polynomials

Legendre polynomials: Eigen functions of the singular Sturm-Lioville problem

N-th Order
$$\frac{d}{d\tau} \left\{ (1 - \tau^2) \frac{dL_N(\tau)}{d\tau} \right\} + N(N+1)L_N(\tau) = 0 \qquad L_N(1) = 1$$

Derivation of Legendre polynomials using Recursive Relations

$$L_{0}(\tau) = 1$$

$$L_{1}(\tau) = \tau$$

$$L_{2}(\tau) = \frac{3}{2}\tau L_{1}(\tau) - \frac{1}{2}L_{0}(\tau) = \frac{1}{2}(3\tau^{2} - 1)$$

$$L_{3}(\tau) = \frac{5}{3}\tau L_{2}(\tau) - \frac{3}{2}L_{1}(\tau) - \frac{5}{6}(3\tau^{3} - \tau) - \frac{2}{3}\tau = \frac{1}{2}(5\tau^{3} - 3\tau)$$

$$L_{4}(\tau) = \frac{7}{4}\tau L_{3}(\tau) - \frac{3}{4}L_{2}(\tau) = \frac{7}{8}(5\tau^{4} - 3\tau^{2}) - \frac{3}{8}(3\tau^{2} - 1) = \frac{1}{8}(35\tau^{4} - 30\tau^{2} + 3)$$

$$L_{5}(\tau) = \frac{9}{5}\tau L_{4}(\tau) - \frac{4}{5}L_{3}(\tau) = \frac{9}{40}(35\tau^{5} - 30\tau^{3} + 3\tau) - \frac{2}{5}(5\tau^{3} - 3\tau) = \frac{1}{8}(63\tau^{5} - 70\tau^{3} + 15\tau)$$

$$L_{6}(\tau) = \frac{11}{6}\tau L_{5}(\tau) - \frac{5}{6}L_{4}(\tau) = \frac{11}{48}(63\tau^{6} - 70\tau^{4} + 15\tau^{2}) - \frac{5}{48}(35\tau^{4} - 30\tau^{2} + 3)$$

$$= \frac{1}{16}(231\tau^{6} - 315\tau^{4} + 105\tau^{2} - 5)$$

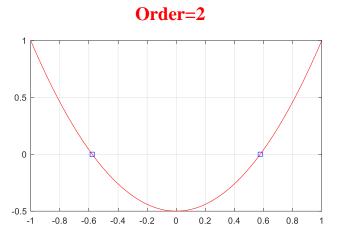
$$\vdots$$

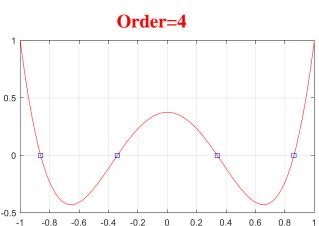
 $(N = 1, 2, 3, \cdots)$

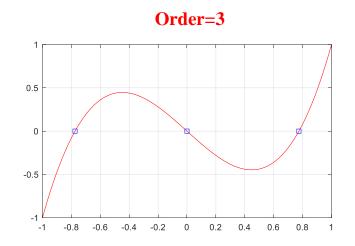


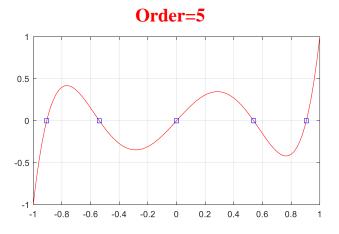
Introduction to Legendre Polynomials

Shapes of Legendre polynomials: Program=A2_Legendre_Polynomial_Root_Plot.m









Orthogonal Polynomials Meeting the orthogonality condition of

$$(L_j, L_k)_{\varpi} = \int_{-1}^{1} L_j(\tau) L_k(\tau) d\tau = \frac{2}{2j+1} \delta_{jk}$$

0

-0.6

-0.4

-0.2

0

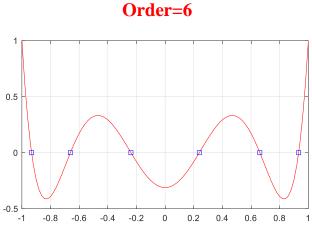
0.2

0.4

-0.8

Introduction to Legendre Polynomials

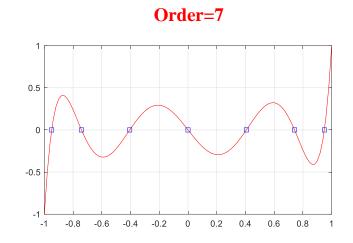
Shapes of Legendre polynomials: Program=A2_Legendre_Polynomial_Root_Plot.m

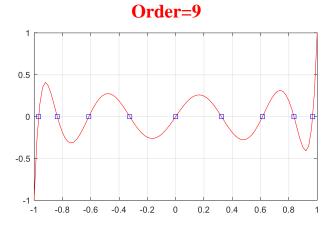


Order=8



0.6





$$(L_j, L_k)_{\varpi} = \int_{-1}^{1} L_j(\tau) L_k(\tau) d\tau = \frac{2}{2j+1} \delta_{jk}$$

The roots of Legendre Polynomials corresponds to the Quadrature Nodes for the Gauss Integration Formula

Introduction to Legendre Polynomials

Useful Recursive Formula to Compute the Gauss Quadrature Nodes

$$(N+1)L_{N+1}(\tau) = (2N+1)\tau L_N(\tau) - NL_{N-1}(\tau), \quad \text{with } L_0(\tau) = 1, L_1(\tau) = \tau$$

$$(2N+1)L_N(\tau) = L'_{N+1}(\tau) - L'_{N-1}(\tau), \quad N \ge 1$$

$$(1-\tau^2)L'_N(\tau) = \frac{N(N+1)}{2N+1} \left(L_{N-1}(\tau) - L_{N+1}(\tau) \right)$$

Computational tips to Compute the Gauss Quadrature Nodes: L2_Legendre_Poly_Root_one.m

- (i) There exists only one root of the (k+1)-th order Legendre polynomial between two adjacent roots of the k-th order polynomial.
- (ii) Thus, the average of such tow roots may become a good approximation for one root of the (k+1)-th order Legendre polynomial and can be used initial solution of the root.
- (iii) The rest one root is less than the smallest root of the k-th order polynomial and another one root is greater than the biggest root of the k-th order polynomial.
- (iv) The Newton-Raphson method can be adopted using the recursive estimation of the gradient of the (k+1)-th order Legendre polynomial. Only a few iterations are required to get a fully converged solution for each root.

$$(2N+1)L_{N}(\tau) = L'_{N+1}(\tau) - L'_{N-1}(\tau), \quad N \ge 1$$

$$(1-\tau^{2})L'_{N}(\tau) = \frac{N(N+1)}{2N+1} \left(L_{N-1}(\tau) - L_{N+1}(\tau) \right)$$



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Lagrange Polynomial Interpolation Using Gauss Quadrature Nodes

Gauss Quadrature Nodes $\left\{\tau_{j}\right\}_{j=0}^{j=N} \leftarrow \tau_{j} = 2\frac{x_{j} - x_{0}}{x_{N} - x_{0}} - 1 \in [-1, 1]$

Functions at Gauss Quadrature Nodes

$$\left\{f_{j}\right\}_{j=0}^{j=N} \leftarrow f_{j} = f_{j}(x_{j}) \leftarrow x_{j} = \frac{x_{N} - x_{0}}{2} \tau_{j} + \frac{x_{N} + x_{0}}{2}$$

Data set for Lagrange Interpolation $\left\{ (\tau_j, f_j) \right\}_{j=0}^{j=N}$

Lagrange Interpolation using Gauss Quadrature Nodes

$$f(\tau) = \sum_{j=0}^{j=N} \phi_{j}(\tau) f_{j}, \leftarrow \phi_{j}(\tau) = \prod_{k=0}^{k=N} \frac{(\tau - \tau_{k})}{(\tau_{j} - \tau_{k})}$$

$$\tau = 2 \frac{x - x_{0}}{x_{N} - x_{0}} - 1 \in [-1, 1]$$

$$d\tau = \frac{2}{x_{N} - x_{0}} dx$$



Gauss Quadrature Weights (1)

Integration Formula with
$$f(\tau) = \sum_{j=0}^{j=N} \phi_j(\tau) f_j, \leftarrow \phi_j(\tau) = \prod_{\substack{k=0 \ k \neq j}}^{k=N} \frac{(\tau - \tau_k)}{(\tau_j - \tau_k)} \qquad \tau = 2 \frac{x - x_0}{x_N - x_0} - 1 \in [-1, 1]$$

$$d\tau = \frac{2}{x_N - x_0} dx$$

$$I = \int_{x_0}^{x_N} f(x)dx = \frac{x_N - x_0}{2} \int_{-1}^{1} f(\tau)d\tau = \frac{x_N - x_0}{2} \sum_{j=0}^{j=N} w_j f_j$$

$$\tau = 2 \frac{x - x_0}{x_N - x_0} - 1 \in [-1, 1]$$

$$d\tau = \frac{2}{x_N - x_0} dx$$

Gauss Quadrature Weights

$$\int_{-1}^{1} f(\tau) d\tau = \int_{-1}^{1} \sum_{j=0}^{j=N} \phi_{j}(\tau) f_{j} d\tau = \sum_{j=0}^{j=N} \left(\int_{-1}^{1} \phi_{j}(\tau) d\tau \right) f_{j} = \sum_{j=0}^{j=N} w_{j} f_{j}$$

$$w_j = \int_{-1}^1 \phi_j(\tau) d\tau$$

Gauss Quadrature Weights (2)

Weight Computation
$$w_j = \int_{-1}^1 \phi_j(\tau) d\tau \leftarrow \phi_j(\tau) = \prod_{\substack{k=0 \ k \neq j}}^{k=N} \frac{(\tau - \tau_k)}{(\tau_j - \tau_k)}$$

Expansion of Lagrange Polynomial

$$\phi_{j}(\tau) = \prod_{\substack{k=0\\k\neq j}}^{k=N} \frac{(\tau - \tau_{k})}{(\tau_{j} - \tau_{k})} = \alpha_{N-1} \tau^{N-1} + \alpha_{N-2} \tau^{N-2} + \dots + \alpha_{2} \tau^{2} + \alpha_{1} \tau + \alpha_{0}$$

$$w_{j} = \int_{-1}^{1} \phi_{j}(\tau) d\tau = \left(\frac{\alpha_{N-1}}{N} \tau^{N} + \frac{\alpha_{N-1}}{N-1} \tau^{N-1} + \dots + \frac{\alpha_{2}}{3} \tau^{3} + \frac{\alpha_{1}}{2} \tau^{2} + \alpha_{0} \tau \right) \Big|_{\tau=-1}^{t-1}$$

Computational Tip for Polynomial Expansion

$$p(\tau) = a_{k-1}\tau^{k-1} + \dots + a_2\tau^2 + a_1\tau + a_0$$

$$q(\tau) = p(\tau)\frac{(\tau - \tau_k)}{(\tau_j - \tau_k)} = p(\tau)(b_1\tau + b_0) \leftarrow b_1 = \frac{1}{(\tau_j - \tau_k)}, b_1 = \frac{(\tau - \tau_k)}{(\tau_j - \tau_k)}$$



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Computer Model to Compute Quadrature Nodes and Weights (1)

Program Structure

```
Legendre polynomial and its derivative
function [tau_vec, Weight_vec] = B1_Legedre_Gauss_Quadrature(N)
   (1) Quadrature Nodes
  (2) Quadrature Weights
   (2-1) Polynomial Expansion of Lagrange Polymonials
  [c_mat] = L5_Lagrange_Poly_Expansion(N,tau_vec) ; Polynomial Expansion
  (2-2) Integration of Lagrange Polymonials
  for J=1:N
      cvec(1:N) = c mat(1:N,J);
                                                Integration of Lagrange Polynomials
      [cintq vec] = L6 Poly Intq Coef(N-1, cvec) ; % (N)-th order
      taup = 1.0;
      [funp] = L6 Poly Computing(N,cintg vec,taup);
응
      taum = -1.0 ;
      [funm] = L6 Poly Computing(N, cintg vec, taum);
응
                                                Quadrature Weights
      Weight vec(J) = funp - funm;
  end
end
```



Computer Model to Compute Quadrature Nodes and Weights (1)

Applications

```
N=2; >> N=2; [tau vec, Weight vec] = B1 Legedre Gauss Quadrature(N)
         tau vec
                   = -0.5774
                                0.5774
         Weight vec = 1
N=3; >> N=3; [tau vec, Weight vec] = B1 Legedre Gauss Quadrature(N)
                    = -0.7746
                                    0
                                         0.7746
          tau vec
          Weight_vec = 0.5556 0.8889 0.5556
N=4; >> N=4; [tau vec, Weight vec] = B1 Legedre Gauss Quadrature(N)
                    = -0.8611 -0.3400 0.3400
                                                 0.8611
          tau vec
          Weight_vec = 0.3479 0.6521 0.6521
                                                 0.3479
N=5; >> N=5; [tau_vec, Weight_vec] = B1_Legedre_Gauss_Quadrature(N)
          tau_vec
                    = -0.9062 -0.5385
                                            0
                                                 0.5385
                                                         0.9062
         Weight_vec = 0.2369 0.4786 0.5689 0.4786
                                                         0.2369
N=6; >> N=6; [tau vec, Weight vec] = B1 Legedre Gauss Quadrature(N)
                    = -0.9325 -0.6612 -0.2386
                                                 0.2386
                                                          0.6612
                                                                 0.9325
          tau_vec
          Weight vec = 0.1713 0.3608 0.4679 0.4679
                                                          0.3608
                                                                 0.1713
```



End of Lecture