

Legendre-Bernstein basis transformations  
and their applications

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## Agenda

$$\begin{array}{ccc} \mathbf{c} & \xrightarrow{\mathbf{M}_n^{-1}} & \mathbf{l} \\ \downarrow & & \downarrow \tilde{\mathbf{I}}_{n,-r} \\ \mathbf{c}^{(-r)} & \xleftarrow{\mathbf{M}_m} & \mathbf{l}^{(-r)} \end{array}$$

1. Bernstein and Legendre polynomials
2. Basis Transformations
3. Relationships
4. Degree elevation and reduction

## Bernstein polynomials

$$B_i^n(t) = \binom{n}{i} (1-t)^{n-i} t^i$$

## Legendre polynomials

$$L_n(t) = \sqrt{2n+1} \sum_{i=0}^{\lfloor n/2 \rfloor} \binom{n}{i, i} (t^2 - t)^i (2t - 1)^{n-2i}$$

$$L_0(t) = 1,$$

$$L_1(t) = \sqrt{3}(2t - 1),$$

$$L_2(t) = \sqrt{5}(6t^2 - 6t + 1),$$

$$L_3(t) = \sqrt{7}(20t^3 - 30t^2 + 12t - 1).$$

## Polynomials of degree $n$ and Basis Transformations

Consider a polynomial  $P_n(t)$  of degree  $n$ , expressed in the degree  $n$  Bernstein and Legendre bases on  $t \in [0, 1]$ :

$$P_n(t) = \sum_{j=0}^n c_j B_j^n(t) = \sum_{k=0}^n l_k L_k(t).$$

We are interested in the linear transformation

$$c_j = \sum_{k=0}^n M_n(j, k) l_k, \quad j = 0, \dots, n,$$

that maps the Legendre coefficients  $l_0, l_1, \dots, l_n$  into the Bernstein coefficients  $c_0, c_1, \dots, c_n$ , and its inverse.

**THEOREM 1** *The Legendre polynomial  $L_k(t)$  can be expressed in the Bernstein basis  $B_0^n(t), B_1^n(t), \dots, B_n^n(t)$  of degree  $n$  as*

$$\begin{aligned}
 L_k(t) &= \sqrt{2k+1} \sum_{i=0}^k (-1)^{k+i} \binom{k}{i} B_i^k(t) \\
 &= \sqrt{2k+1} \sum_{i=0}^k (-1)^{k+i} \binom{k}{i} \sum_{j=i}^{n-k+i} \frac{\binom{k}{i} \binom{n-k}{j-i}}{\binom{n}{j}} B_j^n(t) \\
 &= \sum_{j=0}^n \frac{\sqrt{2k+1}}{\binom{n}{j}} \sum_{i=\max(0, j+k-n)}^{\min(j, k)} (-1)^{k+i} \binom{k}{i} \binom{k}{i} \binom{n-k}{j-i} B_j^n(t).
 \end{aligned}$$

by Farouki, R.T.(2000), Legendre-Bernstein basis transformations, J. Comput. Appl. Math., 119, 145-160.

THEOREM 2 *The elements of the inverse  $M_n^{-1}$  are given for  $0 \leq j, k \leq n$  by*

$$M_n^{-1}(j, k) = \frac{\sqrt{2j+1}}{n+j+1} \binom{n}{k} \sum_{i=0}^j (-1)^{j+i} \frac{\binom{j}{i} \binom{j}{i}}{\binom{n+j}{k+i}}$$

$$= \frac{\sqrt{2j+1}}{n+j+1} \frac{1}{\binom{n+j}{n}} \sum_{i=0}^j (-1)^{j+i} \binom{j}{i} \binom{k+i}{k} \binom{n-k+j-i}{n-k}.$$

by Farouki, R.T.(2000), Legendre-Bernstein basis transformations, J. Comput. Appl. Math., 119, 145-160.

## Example

$$M_1 = \begin{bmatrix} 1 & -\sqrt{3} \\ 1 & \sqrt{3} \end{bmatrix}, M_1^{-1} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{\sqrt{3}}{6} & \frac{\sqrt{3}}{6} \end{bmatrix}$$

$$M_2 = \begin{bmatrix} 1 & -\sqrt{3} & \sqrt{5} \\ 1 & 0 & -2\sqrt{5} \\ 1 & \sqrt{3} & \sqrt{5} \end{bmatrix}, M_2^{-1} = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ -\frac{\sqrt{3}}{6} & 0 & \frac{\sqrt{3}}{6} \\ \frac{\sqrt{5}}{30} & -\frac{2\sqrt{5}}{30} & \frac{\sqrt{5}}{30} \end{bmatrix}$$

$$M_3 = \begin{bmatrix} 1 & -\sqrt{3} & \sqrt{5} & -\sqrt{7} \\ 1 & -\frac{\sqrt{3}}{3} & -\sqrt{5} & 3\sqrt{7} \\ 1 & \frac{\sqrt{3}}{3} & -\sqrt{5} & -3\sqrt{7} \\ 1 & \sqrt{3} & \sqrt{5} & \sqrt{7} \end{bmatrix}, M_3^{-1} = \begin{bmatrix} \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ -\frac{3\sqrt{3}}{20} & -\frac{\sqrt{3}}{20} & \frac{\sqrt{3}}{20} & \frac{3\sqrt{3}}{20} \\ \frac{\sqrt{5}}{20} & -\frac{\sqrt{5}}{20} & -\frac{\sqrt{5}}{20} & \frac{\sqrt{5}}{20} \\ -\frac{\sqrt{7}}{140} & \frac{3\sqrt{7}}{140} & -\frac{3\sqrt{7}}{140} & \frac{\sqrt{7}}{140} \end{bmatrix}$$

## $L_2$ -norm of the $P_n$ with Bernstein bases

$$\begin{aligned}\|P_n\|_2^2 &= \int_0^1 \left| \sum_{i=0}^n c_i B_i^n(t) \right|^2 dt \\ &= \sum_{i,j} c_i c_j \int_0^1 B_i^n(t) B_j^n(t) dt \\ &= \frac{1}{2n+1} \sum_{i,j} c_i c_j \frac{\binom{n}{i} \binom{n}{j}}{\binom{2n}{i+j}} \\ &= c^t Q_n c.\end{aligned}$$

$L_2$ -norm of the polynomial  $P_n$  is

$$\|P_n\|_2^2 = c^t Q_n c$$

where the elements of the Gram matrix  $Q_n$  of Bernstein basis are given by

$$Q_n(i, j) = \frac{1}{2n+1} \frac{\binom{n}{i} \binom{n}{j}}{\binom{2n}{i+j}}, \quad i, j = 0, 1, \dots, n.$$

$Q_n$  is real symmetric matrix, so it can be diagonalized by a unitary matrix  $U_n$  (i.e.,  $U_n^{-1} = U_n^t$ ) whose column vectors are orthonormal eigenvectors of  $Q_n$ , that is,

$$Q_n = U_n D_n U_n^t$$

where  $D_n$  is the diagonal matrix with positive eigenvalues of the matrix  $Q_n$ .

THEOREM 3

$$Q_n M_n = M_n D_n$$

where  $\lambda_k = \frac{1}{2n+1} \frac{\binom{2n+1}{n-k}}{\binom{2n}{n}}$  ( $k = 0, 1, \dots, n$ ) are eigenvalues of the Gram matrix  $Q_n$ .

by Lyche, T. and Scherer, K.(2000), On the p-norm condition number of the multivariate triangular Bernstein basis, J. Comput. Appl. Math., 119, 259-273.

## $L_2$ norm of the $P_n$ with Legendre bases

$$\begin{aligned}\|P_n\|_2^2 &= \int_0^1 \left| \sum_{i=0}^n l_i L_i(t) \right|^2 dt \\ &= \sum_{i,j} l_i l_j \int_0^1 L_i(t) L_j(t) dt \\ &= \sum_i l_i^2 \\ &= \|l\|.\end{aligned}$$

THEOREM 4

$$M_n^{-1} = D_n M_n^t.$$

*proof.*

$$\begin{aligned} c^t Q_n c &= l^t l = c^t (M_n^{-1})^t M_n^{-1} c \\ Q_n &= (M_n^{-1})^t M_n^{-1} = M_n D_n M_n^{-1} \end{aligned}$$

THEOREM 5

$$U_n = M_n \sqrt{D_n}.$$

*proof.*

$$\begin{aligned} Q_n &= M_n D_n D_n M_n^t = M_n \sqrt{D_n} D_n (M_n \sqrt{D_n})^t \\ (M_n \sqrt{D_n})^{-1} &= \sqrt{D_n}^{-1} M_n^{-1} = \sqrt{D_n} M_n^t = (M_n \sqrt{D_n})^t \end{aligned}$$

### Example

$$U_1 = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}, D_1 = \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{6} \end{bmatrix}$$

$$U_2 = \begin{bmatrix} \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & -\frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \end{bmatrix}, D_2 = \begin{bmatrix} \frac{1}{3} & 0 & 0 \\ 0 & \frac{1}{6} & 0 \\ 0 & 0 & \frac{1}{30} \end{bmatrix}$$

$$U_3 = \begin{bmatrix} \frac{1}{2} & -\frac{3}{\sqrt{20}} & \frac{1}{2} & -\frac{1}{\sqrt{20}} \\ \frac{1}{2} & -\frac{1}{\sqrt{20}} & -\frac{1}{2} & \frac{3}{\sqrt{20}} \\ \frac{1}{2} & \frac{1}{\sqrt{20}} & -\frac{1}{2} & -\frac{3}{\sqrt{20}} \\ \frac{1}{2} & \frac{3}{\sqrt{20}} & \frac{1}{2} & \frac{1}{\sqrt{20}} \end{bmatrix}, D_3 = \begin{bmatrix} \frac{1}{4} & 0 & 0 & 0 \\ 0 & \frac{3}{20} & 0 & 0 \\ 0 & 0 & \frac{1}{20} & 0 \\ 0 & 0 & 0 & \frac{1}{140} \end{bmatrix}$$

## Degree elevation with Bernstein coefficients

$$c^{(1)} = T_n c$$

$$T_n = \frac{1}{n+1} \begin{pmatrix} n+1 & 0 & 0 & \dots & 0 & 0 & 0 \\ 1 & n & 0 & \dots & 0 & 0 & 0 \\ 0 & 2 & n-1 & \dots & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & n-1 & 2 & 0 \\ 0 & 0 & 0 & \dots & 0 & n & 1 \\ 0 & 0 & 0 & \dots & 0 & 0 & n+1 \end{pmatrix}$$

$$c^{(r)} = T_{n,r} c$$

$$T_{n,r} = T_{n+r-1} T_{n+r-2} \dots T_{n+1} T_n$$

$$T_{n,r}(i, j) = \frac{\binom{n}{j} \binom{r}{i-j}}{\binom{n+r}{i}}$$

## Degree elevation with Legendre coefficients

By the orthogonality of Legendre basis, the degree elevation of a polynomial with Legendre basis is given by

$$l = (l_0, l_1, \dots, l_n)^t,$$
$$l^{(r)} = (l_0, l_1, \dots, l_n, \underbrace{0, \dots, 0}_r)^t.$$

After  $r$  degree elevations, we have a linear system  $\tilde{I}_{n,r}l = l^{(r)}$ , where the  $(n + r + 1) \times (n + 1)$   $\tilde{I}_{n,r}$  matrix has elements

$$\tilde{I}_{n,r}(i, j) = \begin{cases} 1 & \text{if } i = j, \\ 0 & \text{if } i \neq j. \end{cases}$$

$$\begin{array}{ccc}
 \mathbf{c} & \xrightarrow{\mathbf{M}_n^{-1}} & \mathbf{l} \\
 \mathbf{T}_{n,r} \downarrow & & \downarrow \tilde{\mathbf{I}}_{n,r} \\
 \mathbf{c}^{(r)} & \xleftarrow{\mathbf{M}_{n+r}} & \mathbf{l}^{(r)}
 \end{array}$$

THEOREM 6 *The degree elevation matrix  $T_{n,r}$  can be expressed in  $M_n^{-1}$ ,  $\tilde{I}_{n,r}$  and  $M_{n+r}$  as*

$$T_{n,r} = M_{n+r} \tilde{I}_{n,r} M_n^{-1}.$$

## Degree Reduction with Bernstein coefficients

PROBLEM 1 ( $L_2$  DEGREE REDUCTION) *Let  $\{c_i\}_{i=0}^n$  be a given set of control points which define the Bézier curve*

$$c^n(t) = \sum_{i=0}^n c_i B_i^n(t)$$

*of degree  $n$ . Then find another point set  $\{b_i\}_{i=0}^m$  defining the approximative Bézier curve*

$$b^m(t) = \sum_{i=0}^m b_i B_i^m(t)$$

*of lower degree  $m < n$  so that an  $L_2$ -distance function  $d_2(b^m, c^n)$  between  $b^m$  and  $c^n$  is minimized.*

THEOREM 7 *The  $L_2$ -distance between the two Bézier curves  $b^m$  and  $c^n$  is*

$$d_2^2(b^m, c^n) = d_2^2(b^{(r)}, c^n) = A^t Q_n A$$

where  $A = c - T_{m,r}b$ ,  $b = (b_0, b_1, \dots, b_m)^t$  and  $c = (c_0, c_1, \dots, c_n)^t$ .

For developing the method, rewrite  $d_2^2(b^m, c^n)$ .

$$\begin{aligned} d_2^2(b^m, c^n) &= A^t Q_n A \\ &= [c - T_{m,r}b]^t Q_n [c - T_{m,r}b] \\ &= c^t Q_n c - 2b^t T_{m,r}^t Q_n c + b^t T_{m,r}^t Q_n T_{m,r} b. \end{aligned}$$

One method of obtaining the vector  $b$  is so-called the method of least squares. This method consists of minimizing  $A^t Q_n A$  with respect to  $b$ . We choose the vector  $\hat{b}$  as that the value of  $b$  minimizes  $A^t Q_n A$ . Equating  $\partial(A^t Q_n A)/\partial b$  to zero and writing the resulting equations in terms of  $\hat{b}$ , we find that these equations are

$$T_{m,r}^t Q_n T_{m,r} \hat{b} = T_{m,r}^t Q_n c.$$

**THEOREM 8** *The  $(n + 1) \times (n + 1)$  matrix  $T_{n-1}^t Q_n T_{n-1}$  has the following property:*

$$T_{n-1}^t Q_n T_{n-1} = Q_{n-1}.$$

Hence, the real symmetric positive definite matrix  $T_{m,r}^t Q_n T_{m,r} = Q_m$  is invertible. Provided  $(T_{m,r}^t Q_n T_{m,r})^{-1}$  exists, we have the unique solution for  $\hat{b}$ ,

$$\hat{b} = (T_{m,r}^t Q_n T_{m,r})^{-1} T_{m,r}^t Q_n c.$$

This approximate curve is the best approximation with respect to the  $L_2$ -norm.

Lee, B.G. and Park, Y.(1997), Distance for Bézier curves and degree reduction, Bull. Australian Math. Soc., 56, 507-515.

Lutterkort, D., Peters, J. and Reif, U.(1999), Polynomial degree reduction in the  $L_2$ -norm equals best Euclidean approximation of Bézier coefficients, Computer Aided Geometric Design, 16, 607-612.

## Degree Reduction with Legendre coefficients

By the orthogonality of Legendre basis, the degree reduction of a polynomial with Legendre basis is given by

$$\begin{aligned}l &= (l_0, l_1, \dots, l_n)^t \\ l^{(-1)} &= (l_0, l_1, \dots, l_{n-1})^t.\end{aligned}$$

After  $r$  degree reductions, we have a linear system

$$l^{(-r)} = \tilde{I}_{n,-r} l.$$

$$\begin{array}{ccc}
 \mathbf{c} & \xrightarrow{\mathbf{M}_n^{-1}} & \mathbf{l} \\
 \downarrow (\mathbf{T}_{m,r}^t \mathbf{Q}_n \mathbf{T}_{m,r})^{-1} \mathbf{T}_{m,r}^t \mathbf{Q}_n & & \downarrow \tilde{\mathbf{I}}_{n,-r} \\
 \mathbf{c}^{(-r)} & \xleftarrow{\mathbf{M}_m} & \mathbf{l}^{(-r)}
 \end{array}$$

THEOREM 9 *The degree reduction matrix can be expressed in  $M_m$  and  $M_n^{-1}$  as*

$$(\mathbf{T}_{m,r}^t \mathbf{Q}_n \mathbf{T}_{m,r})^{-1} \mathbf{T}_{m,r}^t \mathbf{Q}_n = M_m \tilde{\mathbf{I}}_{n,-r} M_n^{-1}.$$

### Example [Parametric case, n=4]

$$c = [0, 1, 2, 1, 0]^t,$$

$$l = M_4^{-1}c = \left[\frac{4}{5}, 0, -\frac{6\sqrt{5}}{35}, 0, \frac{2}{105}\right]^t,$$

$$\|c^4\|_2^2 = \left(\frac{4}{5}\right)^2 + \left(\frac{6\sqrt{5}}{35}\right)^2 + \left(\frac{2}{105}\right)^2,$$

$$M_2 \tilde{I}_{4,-2} M_4^{-1} = \frac{1}{35} \begin{bmatrix} 31 & 9 & -3 & -5 & 3 \\ -13 & 17 & 27 & 17 & -13 \\ 3 & -5 & -3 & 9 & 31 \end{bmatrix},$$

$$l^{(-2)} = \tilde{I}_{4,-2} M_4^{-1} c = \left[\frac{4}{5}, 0, -\frac{6\sqrt{5}}{35}\right]^t,$$

$$c^{(-2)} = M_2 \tilde{I}_{4,-2} M_4^{-1} c = \left[\frac{-2}{35}, \frac{88}{35}, \frac{-2}{35}\right]^t,$$

$$\|c^{(-2)}\|_2^2 = \left(\frac{4}{5}\right)^2 + \left(\frac{6\sqrt{5}}{35}\right)^2.$$

1. For the degree reduction with Bernstein basis, we can use the explicit matrix form of  $M_m \tilde{I}_{n,-r} M_n^{-1}$  instead of finding the inverse matrix of  $T_{m,r}^t Q_n T_{m,r}$ .
2. Our method using the relationship of transformations between Legendre and Bernstein basis is a simple and efficient method for optimal multiple degree reductions with respect to the  $L_2$ -norm.
3. This best approximation does not interpolate the given curve at its endpoints. Thus we have to consider the smoothness of our method for the practical use.

*Thank you.*