

Contents lists available at ScienceDirect

European Journal of Combinatorics

journal homepage: www.elsevier.com/locate/ejc



Uniform approximation and Bernstein polynomials with coefficients in the unit interval*

Weikang Qian^a, Marc D. Riedel^a, Ivo Rosenberg^b

- ^a Electrical and Computer Engineering, University of Minnesota, 200 Union St. S.E., Minneapolis, MN 55455, USA
- ^b Mathematics and Statistics, University of Montreal, C.P. 6128, Succ. Centre-Ville, Montreal, Quebec, Canada H3C 3J7

ARTICLE INFO

Article history: Received 27 May 2010 Accepted 21 October 2010

ABSTRACT

This paper presents two main results. The first result pertains to uniform approximation with Bernstein polynomials. We show that, given a power-form polynomial g, we can obtain a Bernstein polynomial of degree m with coefficients that are as close as desired to the corresponding values of g evaluated at the points $0, \frac{1}{m}, \frac{2}{m}, \ldots, 1$, provided that m is sufficiently large. The second result pertains to a subset of Bernstein polynomials: those with coefficients that are all in the unit interval. We show that polynomials in this subset map the open interval (0, 1) into the open interval (0, 1) and map the points 0 and 1 into the closed interval [0, 1]. The motivation for this work is our research on probabilistic computation with digital circuits. Our design methodology, called stochastic logic, is based on Bernstein polynomials with coefficients that correspond to probability values; accordingly, the coefficients must be values in the unit interval. The mathematics presented here provides a necessary and sufficient test for deciding whether polynomial operations can be implemented with stochastic logic.

© 2010 Elsevier Ltd. All rights reserved.

1. Introduction

The Weierstrass Approximation Theorem is a famous theorem in mathematical analysis. It asserts that every continuous function defined on a closed interval can be uniformly approximated as closely as desired by a polynomial function [4].

[†] This work was supported by a grant from the Semiconductor Research Corporation's Focus Center Research Program on Functional Engineered Nano-Architectonics, contract No. 2003-NT-1107.

E-mail addresses: qianx030@umn.edu (W. Qian), mriedel@umn.edu (M.D. Riedel), rosenb@DMS.UMontreal.CA (I. Rosenberg).

The Weierstrass Approximation Theorem. *Let f be a continuous function defined on the closed interval* [a, b]. For any $\epsilon > 0$, there exists a polynomial function p such that for all x in [a, b], we have

$$|f(x) - p(x)| < \epsilon$$
.

The theorem can be proved by a transformation with Bernstein polynomials [8]. By a linear substitution, the interval [a, b] can be transformed into the unit interval [0, 1]. Thus, the original statement of the theorem holds if and only if the theorem holds for every continuous function f defined on the interval [0, 1].

A Bernstein polynomial of degree n is a polynomial expressed in the following form [1]:

$$\sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x), \tag{1}$$

where each $\beta_{k,n}$, k = 0, 1, ..., n, is a real number and

$$b_{k,n}(x) = \binom{n}{k} x^k (1-x)^{n-k}.$$
 (2)

The coefficients $\beta_{k,n}$ are called *Bernstein coefficients* and the polynomials $b_{0,n}(x), b_{1,n}(x), \ldots, b_{n,n}(x)$ are called *Bernstein basis polynomials* of degree n. Define the nth Bernstein polynomial for f to be

$$B_n(f;x) = \sum_{k=0}^n f\left(\frac{k}{n}\right) b_{k,n}(x).$$

In 1912, Bernstein showed the following result [2,7]:

The Bernstein Theorem. Let f be a continuous function defined on the closed interval [0, 1]. For any $\epsilon > 0$, there exists a positive integer M such that for all x in [0, 1] and integer $m \ge M$, we have

$$|f(x) - B_m(f; x)| < \epsilon$$
.

Note that the function $B_m(f;x)$ is a polynomial on x. Thus, on the basis of the Bernstein Theorem, the Weierstrass Approximation Theorem holds. Given a power-form polynomial g of degree n, it is well known that for any $m \ge n$, g can be uniquely converted into a Bernstein polynomial of degree m [5]. Combining this fact with the Bernstein Theorem, we have the following corollary.

Corollary 1. Let g be a polynomial of degree n. For any $\epsilon > 0$, there exists a positive integer $M \ge n$ such that for all x in [0, 1] and integer $m \ge M$, we have

$$\left|\sum_{k=0}^{m} \left(\beta_{k,m} - g\left(\frac{k}{m}\right)\right) b_{k,m}(x)\right| < \epsilon,$$

where $\beta_{0,m}, \beta_{1,m}, ..., \beta_{m,m}$ satisfy $g(x) = \sum_{k=0}^{m} \beta_{k,m} b_{k,m}(x)$.

In the first part of the paper, we prove a stronger result than this:

Theorem 1. Let g be a polynomial of degree $n \ge 0$. For any $\epsilon > 0$, there exists a positive integer $M \ge n$ such that for all integers $m \ge M$ and k = 0, 1, ..., m, we have

$$\left|\beta_{k,m}-g\left(\frac{k}{m}\right)\right|<\epsilon,$$

where $\beta_{0,m}, \beta_{1,m}, ..., \beta_{m,m}$ satisfy $g(x) = \sum_{k=0}^{m} \beta_{k,m} b_{k,m}(x)$.

(Combining Theorem 1 with the fact that $\sum_{k=0}^{m} b_{k,m}(x) = 1$, we can easily prove Corollary 1.) In the second part of the paper, we consider a subset of Bernstein polynomials: those with

In the second part of the paper, we consider a subset of Bernstein polynomials: those with coefficients that are all in the unit interval [0, 1].

Definition 1. Define *U* to be the set of Bernstein polynomials with coefficients that are all in the unit interval [0, 1]:

$$U = \left\{ p(x) \mid \exists \ n \ge 1, \ 0 \le \beta_{0,n}, \ \beta_{1,n}, \dots, \ \beta_{n,n} \le 1, \ \text{such that} \ p(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x) \right\}.$$

The question that we ask is: which polynomials can be converted into Bernstein polynomials in *U*?

Definition 2. Define the set *V* to be the set of polynomials which are either identically equal to 0 or equal to 1, or map the open interval (0, 1) into (0, 1) and the points 0 and 1 into the closed interval [0, 1], i.e.,

$$V = \{p(x) \mid p(x) \equiv 0, \text{ or } p(x) \equiv 1, \text{ or } 0 < p(x) < 1, \forall x \in (0, 1) \text{ and } 0 \le p(0), p(1) \le 1\}.$$

We prove that the above two sets are equivalent:

Theorem 2.

$$V = II$$

In what follows, we will refer to a Bernstein polynomial of degree n converted from a polynomial g as "the Bernstein polynomial of degree n of g". When we say that a polynomial is of degree n, we mean that the power form of the polynomial is of degree n.

Example 1. Consider the polynomial $g(x) = 3x - 8x^2 + 6x^3$. It maps the open interval (0, 1) into (0, 1) with g(0) = 0, g(1) = 1. Thus, g is in the set V. On the basis of Theorem 2, we have that g is in the set U. We verify this by considering Bernstein polynomials of increasing degree.

• The Bernstein polynomial of degree 3 of g is

$$g(x) = b_{1,3}(x) - \frac{2}{3}b_{2,3}(x) + b_{3,3}(x).$$

Note that here the coefficient $\beta_{2,3}=-\frac{2}{3}<0$. • The Bernstein polynomial of degree 4 of g is

$$g(x) = \frac{3}{4}b_{1,4}(x) + \frac{1}{6}b_{2,4}(x) - \frac{1}{4}b_{3,4}(x) + b_{4,4}(x).$$

Note that here the coefficient $\beta_{3,4} = -\frac{1}{4} < 0$. • The Bernstein polynomial of degree 5 of g(x) is

$$g(x) = \frac{3}{5}b_{1,5}(x) + \frac{2}{5}b_{2,5}(x) + b_{5,5}(x).$$

Note that here all the coefficients are in [0, 1].

Since the Bernstein polynomial of degree 5 of g satisfies Definition 1, we conclude that g is in the set U.

Example 2. Consider the polynomial $g(x) = \frac{1}{4} - x + x^2$. Since g(0.5) = 0, thus g is not in the set V. On the basis of Theorem 2, we have that g is not in the set U. We verify this. By contraposition, suppose that there exist $n \ge 1$ and $0 \le \beta_{0,n}, \beta_{1,n}, \dots, \beta_{n,n} \le 1$ such that

$$g(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x).$$

Since g(0.5)=0, therefore, $\sum_{k=0}^n \beta_{k,n} b_{k,n}(0.5)=0$. Note that for all $k=0,1,\ldots,n,b_{k,n}(0.5)>0$. Thus, we have that for all $k=0,1,\ldots,n,\beta_{k,n}=0$. Therefore, $g(x)\equiv 0$, which contradicts the original assumption about g. Thus, g is not in the set U.

The remainder of the paper is organized as follows. In Section 2, we present some mathematical preliminaries pertaining to Bernstein polynomials. In Section 3, we prove Theorem 1. On the basis of this theorem, in Section 4, we prove Theorem 2. Finally, we conclude the paper with a discussion on applications of these theorems to our research in probabilistic computation with digital circuits.

2. Properties of Bernstein polynomials

We list some pertinent properties of Bernstein polynomials.

(a) The positivity property:

For all k = 0, 1, ..., n and all x in [0, 1], we have

$$b_{k,n}(x) > 0. (3)$$

(b) The partition of unity property:

The binomial expansion of the left-hand side of the equality $(x + (1 - x))^n = 1$ shows that the sum of all Bernstein basis polynomials of degree n is the constant 1, i.e.,

$$\sum_{k=0}^{n} b_{k,n}(x) = 1. (4)$$

(c) Converting power-form coefficients to Bernstein coefficients:

The set of Bernstein basis polynomials $b_{0,n}(x)$, $b_{1,n}(x)$, ..., $b_{n,n}(x)$ forms a basis of the vector space of polynomials of real coefficients and degree no more than n [5]. Each power basis function x^j can be uniquely expressed as a linear combination of the n+1 Bernstein basis polynomials:

$$x^{j} = \sum_{k=0}^{n} \sigma_{jk} b_{k,n}(x), \tag{5}$$

for j = 0, 1, ..., n. To determine the elements of the transformation matrix σ , we write

$$x^{j} = x^{j}(x + (1 - x))^{n-j}$$

and perform a binomial expansion on the right-hand side. This gives

$$x^{j} = \sum_{k=j}^{n} \frac{\binom{k}{j}}{\binom{n}{j}} b_{k,n}(x),$$

for j = 0, 1, ..., n. Therefore, we have

$$\sigma_{jk} = \begin{cases} \sigma_{jk} = \binom{k}{j} \binom{n}{j}^{-1}, & \text{for } j \le k \\ 0, & \text{for } j > k. \end{cases}$$
 (6)

Suppose that a power-form polynomial of degree no more than n is

$$g(x) = \sum_{k=0}^{n} a_{k,n} x^{k} \tag{7}$$

and the Bernstein polynomial of degree n of g is

$$g(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x).$$
 (8)

Substituting Eqs. (5) and (6) into Eq. (7) and comparing the Bernstein coefficients, we have

$$\beta_{k,n} = \sum_{j=0}^{n} a_{j,n} \sigma_{jk} = \sum_{j=0}^{k} {k \choose j} {n \choose j}^{-1} a_{j,n}.$$
 (9)

Eq. (9) provides a means for obtaining Bernstein coefficients from power-form coefficients.

(d) Degree elevation:

On the basis of Eq. (2), we have that for all k = 0, 1, ..., m,

$$\frac{1}{\binom{m+1}{k}}b_{k,m+1}(x) + \frac{1}{\binom{m+1}{k+1}}b_{k+1,m+1}(x) = x^{k}(1-x)^{m+1-k} + x^{k+1}(1-x)^{m-k}
= x^{k}(1-x)^{m-k} = \frac{1}{\binom{m}{k}}b_{k,m}(x),$$

or

$$b_{k,m}(x) = \frac{\binom{m}{k}}{\binom{m+1}{k}} b_{k,m+1}(x) + \frac{\binom{m}{k}}{\binom{m+1}{k+1}} b_{k+1,m+1}(x)$$

$$= \frac{m+1-k}{m+1} b_{k,m+1}(x) + \frac{k+1}{m+1} b_{k+1,m+1}(x).$$
(10)

Given a power-form polynomial g of degree n, for any $m \ge n$, g can be uniquely converted into a Bernstein polynomial of degree m. Suppose that the Bernstein polynomials of degree m and degree

$$m+1$$
 of g are $\sum_{k=0}^{m} \beta_{k,m} b_{k,m}(x)$ and $\sum_{k=0}^{m+1} \beta_{k,m+1} b_{k,m+1}(x)$, respectively. We have

$$\sum_{k=0}^{m} \beta_{k,m} b_{k,m}(x) = \sum_{k=0}^{m+1} \beta_{k,m+1} b_{k,m+1}(x).$$
(11)

Substituting Eq. (10) into the left-hand side of Eq. (11) and comparing the Bernstein coefficients, we have

$$\beta_{k,m+1} = \begin{cases} \beta_{0,m}, & \text{for } k = 0\\ \frac{k}{m+1} \beta_{k-1,m} + \left(1 - \frac{k}{m+1}\right) \beta_{k,m}, & \text{for } 1 \le k \le m\\ \beta_{m,m}, & \text{for } k = m+1. \end{cases}$$
(12)

Eq. (12) provides a means for obtaining the coefficients of the Bernstein polynomial of degree m+1 of g from the coefficients of the Bernstein polynomial of degree m of g. We will call this procedure degree elevation.

For convenience, given a Bernstein polynomial $g(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x)$, we can also express it as

$$g(x) = \sum_{k=0}^{n} c_{k,n} x^{k} (1-x)^{n-k}, \tag{13}$$

where

$$c_{k,n} = \binom{n}{k} \beta_{k,n},\tag{14}$$

for k = 0, 1, ..., n. Substituting Eq. (14) into Eq. (12), we have

$$c_{k,m+1} = \begin{cases} c_{0,m}, & \text{for } k = 0\\ c_{k-1,m} + c_{k,m}, & \text{for } 1 \le k \le m\\ c_{m,m}, & \text{for } k = m + 1. \end{cases}$$
 (15)

3. A proof of Theorem 1

Suppose that the polynomial g is of degree n. Applying Eq. (15) recursively, we can express $c_{k,m}$ as a linear combination of $c_{0,n}, c_{1,n}, \ldots, c_{n,n}$.

Lemma 1. Let g be a polynomial of degree n. For any $m \ge n$, suppose that the Bernstein polynomial of degree m of g is $g(x) = \sum_{k=0}^{m} c_{k,m} x^k (1-x)^{m-k}$. Let $c_{k,m} = 0$ for all k < 0 and all k > m. Then for all $k = 0, 1, \ldots, m$, we have

$$c_{k,m} = \sum_{i=0}^{m-n} {m-n \choose i} c_{k-m+n+i,n}.$$
 (16)

Proof. We prove the lemma by induction on m-n.

Base case: For m - n = 0, the right-hand side of Eq. (16) reduces to $\binom{0}{0} c_{k,n} = c_{k,m}$, so the equation holds.

Inductive step: Suppose that Eq. (16) holds for some $m \ge n$ and all k = 0, 1, ..., m. Consider m + 1. Since we assume that $c_{-1,m} = c_{m+1,m} = 0$, Eq. (15) can be written as

$$c_{k,m+1} = c_{k-1,m} + c_{k,m}, (17)$$

for all $k=0,\ldots,m+1$. With our convention that $c_{i,n}=0$ for all i<0 and i>n, it is easily seen that

$$c_{-1,m} = 0 = \sum_{i=0}^{m-n} {m-n \choose i} c_{-1-m+n+i,n}, \qquad c_{m+1,m} = 0 = \sum_{i=0}^{m-n} {m-n \choose i} c_{m+1-m+n+i,n}.$$

Combining this with the induction hypothesis, we conclude that for all $k = -1, 0, \dots, m, m + 1$,

$$c_{k,m} = \sum_{i=0}^{m-n} {m-n \choose i} c_{k-m+n+i,n}.$$
 (18)

On the basis of Eqs. (17) and (18), for all k = 0, 1, ..., m + 1, we have

$$c_{k,m+1} = \sum_{i=0}^{m-n} {m-n \choose i} c_{k-1-m+n+i,n} + \sum_{i=0}^{m-n} {m-n \choose j} c_{k-m+n+j,n}.$$

In the first sum, we change the summation index to j = i - 1. We obtain

$$c_{k,m+1} = \sum_{j=-1}^{m-n-1} {m-n \choose j+1} c_{k-m+n+j,n} + \sum_{j=0}^{m-n} {m-n \choose j} c_{k-m+n+j,n}$$

$$= {m-n \choose 0} c_{k-m+n-1,n} + \sum_{j=0}^{m-n-1} \left[{m-n \choose j+1} + {m-n \choose j} \right] c_{k-m+n+j,n} + {m-n \choose m-n} c_{k,n}.$$

Applying the basic formula $\binom{r}{q} = \binom{r-1}{q-1} + \binom{r-1}{q}$, we obtain

$$c_{k,m+1} = c_{k-m+n-1,n} + \sum_{j=0}^{m-n-1} {m+1-n \choose j+1} c_{k-m+n+j,n} + c_{k,n}$$
$$= \sum_{i=0}^{m+1-n} {m+1-n \choose i} c_{k-m-1+n+i,n}.$$

Thus Eq. (16) holds for m+1. By induction, it holds for all $m \ge k$. \square

Remark. Eq. (16) can be formulated as

$$c_{k,m} = \sum_{i=\max\{0, k-m+n\}}^{\min\{k,n\}} {m-n \choose k-i} c_{i,n},$$
(19)

for all $m \ge n$ and k = 0, 1, ..., m. Indeed, in Eq. (16), first use the basic formula $\binom{r}{q} = \binom{r}{r-q}$ and then change the summation index to j = k - m + n + i to obtain

$$c_{k,m} = \sum_{i=0}^{m-n} {m-n \choose m-n-i} c_{k-m+n+i,n} = \sum_{i=k-m+n}^{k} {m-n \choose k-j} c_{j,n}.$$

Note that $c_{i,n} \neq 0$ implies $0 \leq j \leq n$. This yields Eq. (19).

Lemma 2. Let n be a positive integer. For all integer m, k and i such that

$$m > n$$
, $0 \le k \le m$, $\max\{0, k - m + n\} \le i \le \min\{k, n\}$, (20)

we have

$$\left| \left(\frac{k}{m} \right)^i \left(1 - \frac{k}{m} \right)^{n-i} - \frac{\binom{m-n}{k-i}}{\binom{m}{k}} \right| \le \frac{n^2}{m}. \tag{21}$$

Proof. For simplicity, we define $\delta = (\frac{k}{m})^i (1 - \frac{k}{m})^{n-i} - \frac{\binom{m-n}{k-i}}{\binom{m}{k}}$. Now

$$\frac{\binom{m-n}{k-i}}{\binom{m}{k}} = \frac{(m-n)!}{(k-i)!(m-n-k+i)!} \cdot \frac{k!(m-k)!}{m!}$$

$$= \frac{k(k-1)\cdots(k-i+1)(m-k)(m-k-1)\cdots(m-n-k+i+1)}{m(m-1)\cdots(m-n+1)}$$

$$= \prod_{j=0}^{i-1} \frac{k-j}{m-j} \cdot \prod_{j=0}^{n-i-1} \frac{m-k-j}{m-i-j} = \prod_{j=0}^{i-1} \left(1 - \frac{m-k}{m-j}\right) \cdot \prod_{j=0}^{n-i-1} \left(1 - \frac{k-i}{m-i-j}\right). \tag{22}$$

We obtain an upper bound for $\frac{\binom{m-n}{k-i}}{\binom{m}{k}}$ by replacing j in Eq. (22) with its least value, 0:

$$\frac{\binom{m-n}{k-i}}{\binom{m}{k}} \leq \prod_{i=0}^{i-1} \left(1 - \frac{m-k}{m}\right) \cdot \prod_{i=0}^{n-i-1} \left(1 - \frac{k-i}{m-i}\right) = \left(\frac{k}{m}\right)^i \left(\frac{m-k}{m-i}\right)^{n-i}.$$

We need the following simple inequality: for real numbers $0 \le x \le y \le 1$ and a non-negative integer l,

$$y^{l} - x^{l} = (y - x) \sum_{i=0}^{l-1} y^{i} x^{l-1-i} \le (y - x)l.$$
 (23)

From Eq. (20), we obtain $0 \le i \le \min\{k, n\} \le k \le m$ and so we can use Eq. (23) for

$$0 \le x = \frac{m-k}{m} \le \frac{m-k}{m-i} = y \le 1, \quad l = n-i \ge 0.$$

We obtain

$$\begin{split} \delta &= \left(\frac{k}{m}\right)^i \left(1 - \frac{k}{m}\right)^{n-i} - \frac{\binom{m-n}{k-i}}{\binom{m}{k}} \ge \left(\frac{k}{m}\right)^i \left(\left(\frac{m-k}{m}\right)^{n-i} - \left(\frac{m-k}{m-i}\right)^{n-i}\right) \\ &= -\left(\frac{k}{m}\right)^i \left(\left(\frac{m-k}{m-i}\right)^{n-i} - \left(\frac{m-k}{m}\right)^{n-i}\right) \\ &\ge -\left(\frac{k}{m}\right)^i \left(\frac{m-k}{m-i} - \frac{m-k}{m}\right) (n-i) = -\left(\frac{k}{m}\right)^i \frac{(m-k)i(n-i)}{(m-i)m}. \end{split}$$

Since $0 \le \frac{k}{m} \le 1$, $0 \le \frac{m-k}{m-i} \le 1$, and $0 \le i \le n$, we obtain

$$-\left(\frac{k}{m}\right)^{i}\frac{(m-k)i(n-i)}{(m-i)m}\geq -\frac{i(n-i)}{m}>-\frac{n^{2}}{m}.$$

Therefore,

$$\delta = \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} - \frac{\binom{m-n}{k-i}}{\binom{m}{k}} > -\frac{n^2}{m}.$$
 (24)

Similarly, we obtain a lower bound for $\frac{\binom{m-n}{k-i}}{\binom{m}{k}}$ by replacing the index j in Eq. (22) with i in the first product and with n-i in the second product, obtaining

$$\begin{split} \frac{\binom{m-n}{k-i}}{\binom{m}{k}} &= \prod_{j=0}^{i-1} \left(1 - \frac{m-k}{m-j}\right) \cdot \prod_{j=0}^{n-i-1} \left(1 - \frac{k-i}{m-i-j}\right) \\ &\geq \prod_{j=0}^{i-1} \left(1 - \frac{m-k}{m-i}\right) \cdot \prod_{j=0}^{n-i-1} \left(1 - \frac{k-i}{m-n}\right) \\ &= \left(\frac{k-i}{m-i}\right)^i \left(\frac{m-n-k+i}{m-n}\right)^{n-i} \geq \left(\frac{k-i}{m-i}\right)^i \left(\frac{m-n-k+i}{m-n+i}\right)^{n-i}. \end{split}$$

Thus, proceeding as above, we have

$$\begin{split} \delta &= \left(\frac{k}{m}\right)^i \left(1 - \frac{k}{m}\right)^{n-i} - \frac{\binom{m-n}{k-i}}{\binom{m}{k}} \leq \left(\frac{k}{m}\right)^i \left(\frac{m-k}{m}\right)^{n-i} - \left(\frac{k-i}{m-i}\right)^i \left(\frac{m-n-k+i}{m-n+i}\right)^{n-i} \\ &= \left[\left(\frac{k}{m}\right)^i - \left(\frac{k-i}{m-i}\right)^i\right] \left(\frac{m-k}{m}\right)^{n-i} \\ &+ \left[\left(\frac{m-k}{m}\right)^{n-i} - \left(\frac{m-n-k+i}{m-n+i}\right)^{n-i}\right] \left(\frac{k-i}{m-i}\right)^i. \end{split}$$

Due to Eq. (20), we have

$$0 \le \frac{k-i}{m-i} \le \frac{k}{m} \le 1, \quad 0 \le \frac{m-n-k+i}{m-n+i} \le \frac{m-k}{m} \le 1,$$

and so we obtain

$$\delta \le \left(\frac{k}{m}\right)^i - \left(\frac{k-i}{m-i}\right)^i + \left(\frac{m-k}{m}\right)^{n-i} - \left(\frac{m-n-k+i}{m-n+i}\right)^{n-i}. \tag{25}$$

Applying Eq. (23) twice to the right-hand side of Eq. (25), we obtain

$$\delta \leq i \left(\frac{k}{m} - \frac{k-i}{m-i} \right) + (n-i) \left(\frac{m-k}{m} - \frac{m-n-k+i}{m-n+i} \right)$$
$$= \frac{i^2}{m} \cdot \frac{m-k}{m-i} + \frac{(n-i)^2}{m} \cdot \frac{k}{m-n+i}.$$

From Eq. (20), we have

$$0 \le \frac{m-k}{m-i} \le 1, \quad 0 \le \frac{k}{m-n+i} \le 1.$$

Therefore,

$$\delta = \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} - \frac{\binom{m-n}{k-i}}{\binom{m}{k}} \le \frac{i^2 + (n-i)^2}{m} \le \frac{ni + n(n-i)}{m} = \frac{n^2}{m}.$$
 (26)

Eqs. (24) and (26) together yield Eq. (21). \Box

Now we give a proof of Theorem 1.

Theorem 1. Let g be a polynomial of degree $n \ge 0$. For any $\epsilon > 0$, there exists a positive integer $M \ge n$ such that for all integer $m \ge M$ and k = 0, 1, ..., m, we have

$$\left|\beta_{k,m}-g\left(\frac{k}{m}\right)\right|<\epsilon,$$

where $\beta_{0,m}, \beta_{1,m}, \ldots, \beta_{m,m}$ satisfy that $g(x) = \sum_{k=0}^{m} \beta_{k,m} b_{k,m}(x)$.

Proof. For n=0, g is a constant polynomial. Suppose that g(x)=y, where y is a constant value. We select M=1. Then, for all integers $m\geq M$ and all integers $k=0,1,\ldots,m$, we have $\beta_{k,m}=y=g\left(\frac{k}{m}\right)$. Thus, the theorem holds.

For n > 0, we select M such that $M > \max\{\frac{n^2}{\epsilon} \sum_{i=0}^n |c_{i,n}|, 2n\}$, where the real numbers $c_{0,n}, c_{1,n}, \ldots, c_{n,n}$ satisfy

$$g(x) = \sum_{i=0}^{n} c_{i,n} x^{i} (1-x)^{n-i}.$$
 (27)

Now consider any m > M. Since

$$2n \leq \max \left\{ \frac{n^2}{\epsilon} \sum_{i=0}^n |c_{i,n}|, 2n \right\} < M \leq m,$$

we have m - n > n. Consider the following three cases for k.

1. The case where $n \le k \le m - n$. Here $\max\{0, k - m + n\} = 0$ and $\min\{k, n\} = n$. Thus, the summation indices in Eq. (19) range from 0 to n. Therefore,

$$\beta_{k,m} = \frac{c_{k,m}}{\binom{m}{k}} = \sum_{i=0}^{n} \frac{\binom{m-n}{k-i}}{\binom{m}{k}} c_{i,n}.$$
 (28)

Substituting *x* with $\frac{k}{m}$ in Eq. (27), we have

$$g\left(\frac{k}{m}\right) = \sum_{i=0}^{n} c_{i,n} \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i}.$$
 (29)

By Lemma 2, since 0 < n < m and $0 \le k \le m$, Eq. (21) holds for all $0 = \max\{0, k - m + n\} \le i \le \min\{k, n\} = n$. Thus, by Eqs. (21), (28) and (29) and the well-known inequality $|\sum x_i| \le \sum |x_i|$, we have

$$\left| \beta_{k,m} - g\left(\frac{k}{m}\right) \right| = \left| \sum_{i=0}^{n} \left[\frac{\binom{m-n}{k-i}}{\binom{m}{k}} - \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} \right] c_{i,n} \right|$$

$$\leq \sum_{i=0}^{n} \left| \frac{\binom{m-n}{k-i}}{\binom{m}{k}} - \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} \right| |c_{i,n}| \leq \frac{n^{2}}{m} \sum_{i=0}^{n} |c_{i,n}|.$$

Since $\frac{n^2}{\epsilon} \sum_{i=0}^n |c_{i,n}| < M \le m$, we have

$$\frac{n^2}{m} \sum_{i=0}^n |c_{i,n}| < \epsilon. \tag{30}$$

Therefore, for all $n \le k \le m-n$, we have $|\beta_{k,m} - g(\frac{k}{m})| < \epsilon$.

2. The case where $0 \le k < n$. Since m > 2n, we have k - m + n < k - n < 0. Thus, $\max\{0, k - m + n\} = 0$ and $\min\{k, n\} = k$. Thus, the summation indices in Eq. (19) range from 0 to k. Therefore,

$$\beta_{k,m} = \frac{c_{k,m}}{\binom{m}{k}} = \sum_{i=0}^{k} \frac{\binom{m-n}{k-i}}{\binom{m}{k}} c_{i,n}.$$
(31)

When $k + 1 \le i \le n$, we have that $1 \le k + 1 \le i$ and so

$$\left| \left(\frac{k}{m} \right)^i \left(1 - \frac{k}{m} \right)^{n-i} \right| = \left(\frac{k}{m} \right) \left| \left(\frac{k}{m} \right)^{i-1} \left(1 - \frac{k}{m} \right)^{n-i} \right| \le \frac{k}{m} < \frac{n}{m} \le \frac{n^2}{m}. \tag{32}$$

By Lemma 2, since 0 < n < m and $0 \le k \le m$, Eq. (21) holds for all $0 = \max\{0, k - m + n\} \le i \le \min\{k, n\} = k$. Thus, by Eqs. (21) and (29)–(32) and the inequality $|\sum x_i| \le \sum |x_i|$, we have

$$\begin{aligned} \left| \beta_{k,m} - g\left(\frac{k}{m}\right) \right| &= \left| \sum_{i=0}^{k} \frac{\binom{m-n}{k-i}}{\binom{m}{k}} c_{i,n} - \sum_{i=0}^{n} \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} c_{i,n} \right| \\ &= \left| \sum_{i=0}^{k} \left[\frac{\binom{m-n}{k-i}}{\binom{m}{k}} - \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} \right] c_{i,n} - \sum_{i=k+1}^{n} \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} c_{i,n} \right| \\ &\leq \sum_{i=0}^{k} \left| \frac{\binom{m-n}{k-i}}{\binom{m}{k}} - \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} \right| |c_{i,n}| + \sum_{i=k+1}^{n} \left| \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} \right| |c_{i,n}| \\ &\leq \frac{n^{2}}{m} \sum_{i=0}^{n} |c_{i,n}| < \epsilon. \end{aligned}$$

3. The case where $m-n < k \le m$. Since m > 2n, we have n < m-n < k. Thus, $\max\{0, k-m+n\} = k-m+n$ and $\min\{k,n\} = n$. Now, the summation indices in Eq. (19) range from k-m+n to n. Therefore,

$$\beta_{k,m} = \frac{c_{k,m}}{\binom{m}{k}} = \sum_{i=k-m+n}^{n} \frac{\binom{m-n}{k-i}}{\binom{m}{k}} c_{i,n}.$$
 (33)

When $0 \le i \le k - m + n - 1$, we have that $1 \le m + 1 - k \le n - i$. Thus,

$$\left| \left(\frac{k}{m} \right)^i \left(1 - \frac{k}{m} \right)^{n-i} \right| = \left(1 - \frac{k}{m} \right) \left| \left(\frac{k}{m} \right)^i \left(1 - \frac{k}{m} \right)^{n-i-1} \right| \le \frac{m-k}{m} < \frac{n}{m} \le \frac{n^2}{m}. \tag{34}$$

By Lemma 2, since 0 < n < m and $0 \le k \le m$, Eq. (21) holds for all $k-m+n = \max\{0, k-m+n\} \le i \le \min\{k, n\} = n$. Thus, by Eqs. (21), (29), (30), (33) and (34) and the inequality $|\sum x_i| \le \sum |x_i|$, we have

$$\begin{split} \left| \beta_{k,m} - g\left(\frac{k}{m}\right) \right| &= \left| \sum_{i=k-m+n}^{n} \frac{\binom{m-n}{k-i}}{\binom{m}{k}} c_{i,n} - \sum_{i=0}^{n} \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} c_{i,n} \right| \\ &= \left| \sum_{i=k-m+n}^{n} \left[\frac{\binom{m-n}{k-i}}{\binom{m}{k}} - \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} \right] c_{i,n} - \sum_{i=0}^{k-m+n-1} \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} c_{i,n} \right| \\ &\leq \sum_{i=k-m+n}^{n} \left| \frac{\binom{m-n}{k-i}}{\binom{m}{k}} - \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} \right| |c_{i,n}| + \sum_{i=0}^{k-m+n-1} \left| \left(\frac{k}{m}\right)^{i} \left(1 - \frac{k}{m}\right)^{n-i} \right| |c_{i,n}| \\ &\leq \frac{n^{2}}{m} \sum_{i=0}^{n} |c_{i,n}| < \epsilon. \end{split}$$

In conclusion, if m > M, then for all k = 0, 1, ..., m, we have

$$\left|\beta_{k,m}-g\left(\frac{k}{m}\right)\right|<\epsilon.$$

4. A proof of Theorem 2

We demonstrate that the sets U and V defined in the introduction – see Definitions 1 and 2 – are one and the same. We demonstrate that $U \subseteq V$ and $V \subseteq U$ separately. First, we prove the former – the easier one. Then we use Theorem 1 to prove the latter.

Theorem 3.

$$U \subseteq V$$
.

Proof. Let $n \ge 1$ and $\beta_{k,n} = 0$, for all $0 \le k \le n$. Then the polynomial $p(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x) = 0$. Let $n \ge 1$ and $\beta_{k,n} = 1$, for all $0 \le k \le n$. Then, by Eq. (4), the polynomial $p(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x) = 1$. Thus $0 \in U$ and $1 \in U$. From the definition of V, $0 \in V$ and $1 \in V$.

Now consider any polynomial $p \in U$ such that $p \not\equiv 0$ and $p \not\equiv 1$. There exist $n \geq 1$ and $0 \leq \beta_{0,n}, \beta_{1,n}, \ldots, \beta_{n,n} \leq 1$ such that

$$p(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x).$$

From Eqs. (3) and (4) and the fact that $0 \le \beta_{0,n}, \beta_{1,n}, \dots, \beta_{n,n} \le 1$, for all x in [0, 1], we have

$$0 \le p(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x) \le \sum_{k=0}^{n} b_{k,n}(x) = 1.$$

We further claim that for all x in (0, 1), we must have 0 < p(x) < 1. By contraposition, we assume that there exists a $0 < x_0 < 1$ such that $p(x_0) \le 0$ or $p(x_0) \ge 1$. Since for $0 < x_0 < 1$, we have $0 \le p(x_0) \le 1$, thus $p(x_0) = 0$ or 1.

We first consider the case where $p(x_0) = 0$. Since $0 < x_0 < 1$, it is not hard to see that for all k = 0, 1, ..., n, $b_{k,n}(x_0) > 0$. Thus, $p(x_0) = 0$ implies that for all k = 0, 1, ..., n, $\beta_{k,n} = 0$. In this case, for any real number x, $p(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x) = 0$, which contradicts the assumption that $p(x) \neq 0$

Similarly, in the case where $p(x_0) = 1$, we can show that $p(x) \equiv 1$, which contradicts the assumption that $p(x) \not\equiv 1$. In both cases, we get a contradiction; this proves the claim that for all x in (0, 1), 0 < p(x) < 1.

Therefore, for any polynomial $p \in U$ such that $p \not\equiv 0$ and $p \not\equiv 1$, we have $p \in V$. Since we showed at the outset that $0 \in U$, $1 \in U$, $0 \in V$ and $1 \in V$, thus, for any polynomial $p \in U$, we have $p \in V$. Therefore, $U \subseteq V$. \square

Next we prove the claim that $V \subseteq U$. We will first show that each of four possible different categories of polynomials in the set V are in the set U. The different categories are tackled in Theorems 4 and 5 and Corollaries 2 and 3.

Theorem 4. Let g be a polynomial of degree n mapping the open interval (0, 1) into (0, 1) with $0 \le g(0), g(1) < 1$. Then $g \in U$.

Proof. Since g is continuous on the closed interval [0, 1], it attains its maximum value M_g on [0, 1]. Since g(x) < 1, for all $x \in [0, 1]$, we have $M_g < 1$.

Let $\epsilon_1 = 1 - M_g > 0$. By Theorem 1, there exists a positive integer $M_1 \ge n$ such that for all integers $m \ge M_1$ and k = 0, 1, ..., m, we have $|\beta_{k,m} - g(\frac{k}{m})| < \epsilon_1$, where $\beta_{0,m}, \beta_{1,m}, ..., \beta_{m,m}$ satisfy that $g(x) = \sum_{k=0}^m \beta_{k,m} b_{k,m}(x)$. Thus, for all $m \ge M_1$ and all k = 0, 1, ..., m,

$$\beta_{k,m} < g\left(\frac{k}{m}\right) + \epsilon_1 \le M_g + 1 - M_g = 1. \tag{35}$$

Denote by r the multiplicity of 0 as a root of g(x) (where r = 0 if $g(0) \neq 0$) and by s the multiplicity of 1 as a root of g(x) (where s = 0 if $g(1) \neq 0$). We can factorize g(x) as

$$g(x) = x^{r}(1-x)^{s}h(x),$$
 (36)

where h(x) is a polynomial, satisfying that $h(0) \neq 0$ and $h(1) \neq 0$.

We show that h(0) > 0. By contraposition, suppose that $h(0) \le 0$. Since $h(0) \ne 0$, we have h(0) < 0. By the continuity of the polynomial h(x), there exists some $0 < x^* < 1$ such that $h(x^*) < 0$. Thus, $g(x^*) = x^{*r}(1-x^*)^s h(x^*) < 0$. However, g(x) > 0, for all $x \in (0, 1)$. Therefore, h(0) > 0. Similarly, we have h(1) > 0.

Since g(x) > 0 for all x in (0, 1), we have $h(x) = \frac{g(x)}{x^r(1-x)^s} > 0$ for all x in (0, 1). In view of the fact that h(0) > 0 and h(1) > 0, we have h(x) > 0, for all x in [0, 1]. Since h(x) is continuous on the closed interval [0, 1], it attains its minimum value m_h on [0, 1]. Clearly, $m_h > 0$.

Let $\epsilon_2 = m_h > 0$. By Theorem 1, there exists a positive integer $M_2 \ge n - r - s$, such that for all integers $d \ge M_2$ and $k = 0, 1, \ldots, d$, we have $|\gamma_{k,d} - h(\frac{k}{d})| < \epsilon_2$, where $\gamma_{0,d}, \gamma_{1,d}, \ldots, \gamma_{d,d}$ satisfy that

$$h(x) = \sum_{k=0}^{d} \gamma_{k,d} b_{k,d}(x). \tag{37}$$

Thus, for all $d \ge M_2$ and all $k = 0, 1, \ldots, d$,

$$\gamma_{k,d} > h\left(\frac{k}{d}\right) - \epsilon_2 \ge m_h - m_h = 0.$$

Combining Eqs. (36) and (37), we have

$$g(x) = x^{r} (1 - x)^{s} h(x) = x^{r} (1 - x)^{s} \sum_{k=0}^{d} \gamma_{k,d} b_{k,d}(x)$$

$$= x^{r} (1 - x)^{s} \sum_{k=0}^{d} \gamma_{k,d} {d \choose k} x^{k} (1 - x)^{d-k}$$

$$= \sum_{k=0}^{d} \frac{\gamma_{k,d} {d \choose k}}{{d+r+s \choose k+r}} {d+r+s \choose k+r} x^{k+r} (1 - x)^{d+s-k} = \sum_{k=r}^{d+r} \frac{\gamma_{k-r,d} {d \choose k-r}}{{d+r+s \choose k}} b_{k,d+r+s}(x)$$

$$= \sum_{k=0}^{d+r+s} \beta_{k,d+r+s} b_{k,d+r+s}(x),$$

where $\beta_{k,d+r+s}$ are the coefficients of the Bernstein polynomial of degree d+r+s of g and

$$\beta_{k,d+r+s} = \begin{cases} 0, & \text{for } 0 \le k < r \text{ and } d+r < k \le d+r+s \\ \frac{\gamma_{k-r,d} \binom{d}{k-r}}{\binom{d+r+s}{k}} > 0, & \text{for } r \le k \le d+r. \end{cases}$$

Thus, when $m = d + r + s \ge M_2 + r + s$, we have for all k = 0, 1, ..., m,

$$\beta_{k,m} \ge 0. \tag{38}$$

According to Eqs. (35) and (38), if we select an $m_0 \ge \max\{M_1, M_2 + r + s\}$, then g(x) can be expressed as a Bernstein polynomial of degree m_0 :

$$g(x) = \sum_{k=0}^{m_0} \beta_{k,m_0} b_{k,m_0}(x),$$

with $0 \le \beta_{k,m_0} \le 1$, for all $k = 0, 1, ..., m_0$. Therefore, $g \in U$. \square

Theorem 5. Let g be a polynomial of degree n mapping the open interval (0, 1) into (0, 1) with g(0) = 0 and g(1) = 1. Then $g \in U$.

Proof. Denote by r the multiplicity of 0 as a root of g(x). We can factorize g(x) as

$$g(x) = x^r h(x), (39)$$

where h(x) is a polynomial satisfying $h(0) \neq 0$. By a reasoning similar to that in the proof of Theorem 4, we obtain h(0) > 0. Since for all x in (0, 1], $h(x) = \frac{g(x)}{x^T} > 0$, we have for all x in [0, 1], h(x) > 0. Since h(x) is continuous on the closed interval [0, 1], it attains its minimum value m_h on [0, 1]. Clearly, $m_h > 0$.

Let $\epsilon_1=m_h>0$. By Theorem 1, there exists a positive integer $M_1\geq n-r$ such that for all integers $d\geq M_1$ and $k=0,1,\ldots,d$, we have $|\gamma_{k,d}-h(\frac{k}{d})|<\epsilon_1$, where $\gamma_{0,d},\gamma_{1,d},\ldots,\gamma_{d,d}$ satisfy

$$h(x) = \sum_{k=0}^{d} \gamma_{k,d} b_{k,d}(x). \tag{40}$$

Thus, for all $d \ge M_1$ and all k = 0, 1, ..., d,

$$\gamma_{k,d} > h\left(\frac{k}{d}\right) - \epsilon_1 \ge m_h - m_h = 0.$$

Combining Eqs. (39) and (40), we have

$$\begin{split} g(x) &= x^r h(x) = x^r \sum_{k=0}^d \gamma_{k,d} b_{k,d}(x) = x^r \sum_{k=0}^d \gamma_{k,d} \binom{d}{k} x^k (1-x)^{d-k} \\ &= \sum_{k=0}^d \frac{\gamma_{k,d} \binom{d}{k}}{\binom{d+r}{k+r}} \binom{d+r}{k+r} x^{k+r} (1-x)^{d-k} = \sum_{k=r}^{d+r} \frac{\gamma_{k-r,d} \binom{d}{k-r}}{\binom{d+r}{k}} b_{k,d+r}(x) \\ &= \sum_{k=0}^{d+r} \beta_{k,d+r} b_{k,d+r}(x), \end{split}$$

where $\beta_{k,d+r}$ are the coefficients of the Bernstein polynomial of degree d+r of g and

$$\beta_{k,d+r} = \begin{cases} 0, & \text{for } 0 \le k < r \\ \frac{\gamma_{k-r,d} \binom{d}{k-r}}{\binom{d+r}{k}} > 0, & \text{for } r \le k \le d+r. \end{cases}$$

Thus, when $m = d + r \ge M_1 + r$, we have for all k = 0, 1, ..., m

$$\beta_{k,m} \ge 0. \tag{41}$$

Let

$$g^* = 1 - g(x). (42)$$

Then g^* maps the open interval (0, 1) into (0, 1) with $g^*(0) = 1$, $g^*(1) = 0$. Denote by s the multiplicity of 1 as a root of $g^*(x)$. Thus, we can factorize $g^*(x)$ as

$$g^*(x) = (1-x)^s h^*(x),$$
 (43)

where $h^*(x)$ is a polynomial satisfying that $h^*(1) \neq 0$. As in the proof of Theorem 4, we obtain $h^*(1) > 0$. Since for all x in [0, 1), $h^*(x) = \frac{g^*(x)}{(1-x)^s} > 0$, we have for all $x \in [0, 1]$, $h^*(x) > 0$. Since $h^*(x)$ is continuous on the closed interval [0, 1], it attains its minimum value m_h^* on [0, 1]. Clearly, $m_h^* > 0$.

Let $\epsilon_2=m_h^*>0$. By Theorem 1, there exists a positive integer $M_2\geq n-s$ such that for all integers $q\geq M_2$ and $k=0,1,\ldots,q$, we have $|\gamma_{k,q}^*-h^*(\frac{k}{q})|<\epsilon_2$, where $\gamma_{0,q}^*,\gamma_{1,q}^*,\ldots,\gamma_{q,q}^*$ satisfy

$$h^*(x) = \sum_{k=0}^{q} \gamma_{k,q}^* b_{k,q}(x). \tag{44}$$

Thus, for all $q \ge M_2$ and all k = 0, 1, ..., q,

$$\gamma_{k,q}^* > h^*\left(\frac{k}{q}\right) - \epsilon_2 \ge m_h^* - m_h^* = 0.$$

Combining Eqs. (42)-(44), we have

$$g(x) = 1 - g^*(x) = 1 - (1 - x)^s h^*(x) = 1 - (1 - x)^s \sum_{k=0}^q \gamma_{k,q}^* b_{k,q}(x)$$

$$= 1 - (1 - x)^s \sum_{k=0}^q \gamma_{k,q}^* {q \choose k} x^k (1 - x)^{q-k} = 1 - \sum_{k=0}^q \frac{\gamma_{k,q}^* {q \choose k}}{{q+s \choose k}} {q+s \choose k} x^k (1 - x)^{q+s-k}.$$

Further using Eq. (4), we obtain

$$g(x) = \sum_{k=0}^{q+s} b_{k,q+s}(x) - \sum_{k=0}^{q} \frac{\gamma_{k,q}^* \binom{q}{k}}{\binom{q+s}{k}} b_{k,q+s}(x) = \sum_{k=0}^{q+s} \beta_{k,q+s} b_{k,q+s}(x),$$

where the $\beta_{k,q+s}$'s are the coefficients of the Bernstein polynomial of degree q + s of g:

$$\beta_{k,q+s} = \begin{cases} 1 - \frac{\gamma_{k,q}^*\binom{q}{k}}{\binom{q+s}{k}} < 1, & \text{for } 0 \le k \le q\\ 1, & \text{for } q < k \le q+s. \end{cases}$$

Thus, when $m = q + s \ge M_2 + s$, we have for all k = 0, 1, ..., m,

$$\beta_{k,m} < 1. \tag{45}$$

According to Eqs. (41) and (45), if we select an $m_0 \ge \max\{M_1 + r, M_2 + s\}$, then g(x) can be expressed as a Bernstein polynomial of degree m_0 :

$$g(x) = \sum_{k=0}^{m_0} \beta_{k,m_0} b_{k,m_0}(x),$$

with $0 \le \beta_{k,m_0} \le 1$, for all $k = 0, 1, ..., m_0$, Therefore, $g \in U$. \square

Lemma 3. If a polynomial p is in the set U, then the polynomial 1 - p is also in the set U.

Proof. Since p is in the set U, there exist $n \ge 1$ and $0 \le \beta_{0,n}, \beta_{1,n}, \ldots, \beta_{n,n} \le 1$ such that

$$p(x) = \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x).$$

By Eq. (4), we have

$$1 - p(x) = \sum_{k=0}^{n} b_{k,n}(x) - \sum_{k=0}^{n} \beta_{k,n} b_{k,n}(x) = \sum_{k=0}^{n} (1 - \beta_{k,n}) b_{k,n}(x) = \sum_{k=0}^{n} \gamma_{k,n} b_{k,n}(x),$$

where $\gamma_{k,n}=1-\beta_{k,n}$ satisfying $0\leq \gamma_{k,n}\leq 1$, for all $k=0,1,\ldots,n$. Therefore, 1-p is in the set U. \square

Corollary 2. Let g be a polynomial of degree n mapping the open interval (0, 1) into (0, 1) with 0 < g(0), g(1) < 1. Then $g \in U$.

Proof. Let polynomial h=1-g. Then h maps (0,1) into (0,1) with $0 \le h(0), h(1) < 1$. By Theorem 4, $h \in U$. By Lemma 3, g=1-h is also in the set U. \square

Corollary 3. Let g be a polynomial of degree n mapping the open interval (0, 1) into (0, 1) with g(0) = 1 and g(1) = 0. Then $g \in U$.

Proof. Let the polynomial h = 1 - g. Then h maps (0, 1) into (0, 1) with h(0) = 0, h(1) = 1. By Theorem 5, $h \in U$. By Lemma 3, g = 1 - h is also in the set U. \Box

Combining Theorems 4 and 5, Corollaries 2 and 3, we show that $V \subseteq U$.

Theorem 6.

 $V \subset U$.

Proof. On the basis of the definition of V, for any polynomial $p \in V$, we have one of the following five cases.

- 1. The case where $p \equiv 0$ or $p \equiv 1$. In the proof of Theorem 3, we have shown that $0 \in U$ and $1 \in U$. Thus $p \in U$.
- 2. The case where p maps the open interval (0, 1) into (0, 1) with $0 \le p(0), p(1) < 1$. By Theorem 4, $p \in U$.
- 3. The case where p maps the open interval (0, 1) into (0, 1) with $0 < p(0), p(1) \le 1$. By Corollary 2, $p \in U$.
- 4. The case where p maps the open interval (0, 1) into (0, 1) with p(0) = 0 and p(1) = 1. By Theorem 5, $p \in U$.
- 5. The case where p maps the open interval (0, 1) into (0, 1) with p(0) = 1 and p(1) = 0. By Corollary $3, p \in U$.

In summary, for any polynomial $p \in V$, we have $p \in U$. Thus, $V \subseteq U$. On the basis of Theorems 3 and 6, we have proved Theorem 2:

$$V = U$$
. \square

5. Discussion

We are interested in Bernstein polynomials with coefficients in the unit interval because this concept has applications in the area of digital circuit design. Specifically, the concept is a mathematical prerequisite for a design methodology that we have been advocating called *stochastic logic* [6,3,9]. We provide a brief overview of this application and point the reader to further sources.

Stochastic logic implements Boolean functions with inputs that are random Boolean variables. A Boolean function f on n variables x_1, x_2, \ldots, x_n is a mapping

$$f: \{0, 1\}^n \to \{0, 1\}.$$

With stochastic logic, the variables x_1, x_2, \ldots, x_n are a set of independent random Boolean variables, i.e., for $1 \le i \le n$, x_i has a certain probability p_i ($0 \le p_i \le 1$) of being 1 and a probability $1 - p_i$ of being 0. With random Boolean variables as inputs, the output is also a random Boolean variable: the function f has a certain probability p_0 of being 1 and a probability $1 - p_0$ of being 0.

If implemented by digital circuitry, stochastic logic can be viewed as computation that transforms input probabilities into output probabilities [3]. Given an arbitrary Boolean function f and a set of input probabilities p_1, p_2, \ldots, p_n that correspond to the probabilities of the input random Boolean variables being 1, the output probability p_0 is a function on p_1, p_2, \ldots, p_n . In fact, we have shown that the general form of the function is a multivariate polynomial on variables p_1, \ldots, p_n with integer coefficients and with the degree of each variable no more than 1 [9].

Example 3. Consider stochastic logic based on the Boolean function

$$f(x_1, x_2, x_3) = (x_1 \wedge x_2) \vee (\neg x_1 \wedge x_3),$$

where \land means logical AND (conjunction), \lor means logical OR (disjunction), and \neg means logical negation.

The Boolean function f evaluates to 1 if and only if the 3-tuple (x_1, x_2, x_3) takes values from the set $\{(0, 0, 1), (0, 1, 1), (1, 1, 0), (1, 1, 1)\}$. The probability of the output being 1 is

$$p_0 = \Pr(f = 1) = \Pr(x_1, x_2, x_3 : (x_1, x_2, x_3) \in \{(0, 0, 1), (0, 1, 1), (1, 1, 0), (1, 1, 1)\})$$

$$= \Pr(x_1 = 0, x_2 = 0, x_3 = 1) + \Pr(x_1 = 0, x_2 = 1, x_3 = 1)$$

$$+ \Pr(x_1 = 1, x_2 = 1, x_3 = 0) + \Pr(x_1 = 1, x_2 = 1, x_3 = 1).$$

If x_1 , x_2 , and x_3 are independent random Boolean variables with probability p_1 , p_2 , and p_3 of being 1, respectively, then we obtain

$$p_{0} = (1 - p_{1})(1 - p_{2})p_{3} + (1 - p_{1})p_{2}p_{3} + p_{1}p_{2}(1 - p_{3}) + p_{1}p_{2}p_{3}$$

$$= (1 - p_{1})p_{3} + p_{1}p_{2}$$

$$= p_{1}p_{2} + p_{3} - p_{1}p_{3},$$
(46)

which confirms that the function computed by stochastic logic is a multivariate polynomial on arguments p_1 , p_2 , and p_3 with integer coefficients and with the degree of each variable no more than 1.

In design problems, we encounter univariate polynomials that have real coefficients and degree greater than 1. Sometimes it is possible to implement these by setting some of the probabilities p_i to be a common variable x and the others to be constants. For example, if we set $p_1 = p_3 = x$ and $p_2 = 0.75$ in Eq. (46), then we obtain the polynomial $g(x) = 1.75x - x^2$. With different underlying Boolean functions and different assignments of probability values, we can implement many different univariate polynomials.

An interesting and yet practical question is: which univariate polynomials can be implemented by stochastic logic? Define the set W to be the set of (univariate) polynomials that can be implemented. We are interested in characterizing the set W.

In [9] we showed that $U \subseteq W$, i.e., if a polynomial can be expressed as a Bernstein polynomial with all coefficients in the unit interval, then the polynomial can be implemented by stochastic logic. In this paper, we proved that V = U. Thus, we have $V \subseteq W$.

Further, in [9] we showed that $W \subseteq V$, i.e., if a polynomial can be implemented by stochastic logic, then it is either identically equal to 0 or equal to 1, or it maps the open interval (0, 1) into the open interval (0, 1) and maps the points 0 and 1 into the closed interval [0, 1]. Therefore, we conclude that W = V, i.e., a polynomial can be implemented by stochastic logic if and only if it is either identically equal to 0 or equal to 1, or it maps the open interval (0, 1) into the open interval (0, 1) and maps the points 0 and 1 into the closed interval [0, 1].

This necessary and sufficient conditions allows us to answer the question of whether any given polynomial can be implemented by stochastic logic. On the basis of the mathematics, we have proposed a constructive design method [9]. An overview of the method and its applications in circuit design will appear in a forthcoming "Research Highlights" article in Communications of the ACM [10].

References

- [1] J. Berchtold, A. Bowyer, Robust arithmetic for multivariate Bernstein-form polynomials, Computer-Aided Design 32 (11) (2000) 681–689.
- [2] S.N. Bérnstein, Démonstration du théorème de Weierstrass fondée sur le calcul des probabilités, Communications of the Kharkov Mathematical Society 13 (1912) 1–2.
- [3] B. Brown, H. Card, Stochastic neural computation I: computational elements, IEEE Transactions on Computers 50 (9) (2001) 891–905.
- [4] P.J. Davis, Interpolation and Approximation, Blaisdell Publishing Company, 1963.
- [5] R. Farouki, V. Rajan, On the numerical condition of polynomials in Bernstein form, Computer Aided Geometric Design 4 (3) (1987) 191–216.
- [6] B. Gaines, Stochastic computing systems, in: Advances in Information Systems Science, vol. 2, Plenum, 1969, pp. 37–172. ch. 2.
- [7] K.M. Levasseur, A probabilistic proof of the Weierstrass approximation theorem, The American Mathematical Monthly 91 (4) (1984) 249–250.
- [8] G. Lorentz, Bernstein Polynomials, University of Toronto Press, 1953.
- [9] W. Qian, M.D. Riedel, The synthesis of robust polynomial arithmetic with stochastic logic, in: Design Automation Conference, 2008, pp. 648–653.
- [10] W. Qian, M.D. Riedel, Synthesizing logical computation on stochastic bit streams, Communications of the ACM, 2010 (submitted for publication).