Open Questions on the Bernoulli Factory Problem

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1 Background

Suppose there is a coin that shows heads with an unknown probability, \$\lambda\$. The goal is to use that coin (and possibly also a fair coin) to build a "new" coin that shows heads with a probability that depends on \$\lambda\$, call it \$f(\lambda)\$. This is the *Bernoulli factory problem*, and it can be solved only for certain functions \$f\$. (For example, flipping the coin twice and taking heads only if exactly one coin shows heads, the probability \$2\lambda(1-\lambda)\$ can be simulated.)

Specifically, the only functions that can be simulated this way **are continuous and polynomially bounded on their domain, and map** $\{0, 1\}$ \$ or a **subset thereof to** $\{0, 1\}$ \$, as well as $\{0, 1\}$ \$ and $\{0, 1\}$ \$. These functions are called *factory functions* in this page. (A function $\{0, 1\}$ \$ is *polynomially bounded* if both $\{0, 1\}$ \$ are greater than or equal to $\{0, 1\}$ \$ for some integer $\{0, 1\}$ \$ (Keane and O'Brien 1994). This implies that $\{0, 1\}$ \$ admits no roots on $\{0, 1\}$ \$ and can't take on the value 0 or 1 except possibly at 0, 1, or both.)

This page contains several questions about the <u>Bernoulli factory</u> problem. Answers to them will greatly improve my pages on this site about Bernoulli factories. If you can answer any of them, post an issue in the <u>GitHub issues page</u>.

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3 Polynomials that approach a factory function "fast"

 $\frac{https://math.stackexchange.com/questions/3904732/what-are-ways-to-compute-polynomials-that-converge-from-above-and-below-to-a-con$

https://mathoverflow.net/questions/442057/explicit-and-fasterror-bounds-for-approximating-continuous-functions

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 $\frac{https://mathoverflow.net/questions/429037/bounds-on-the-expectation-of-a-function-of-a-hypergeometric-random-variable}{}$

 $\frac{https://mathoverflow.net/questions/407179/using-the-holtz-method-to-build-polynomials-that-converge-to-a-continuous-functi$

https://mathoverflow.net/questions/447064/explicit-bounds-onderivatives-of-moments-related-to-bernstein-polynomials

 $\frac{https://mathoverflow.net/questions/409174/concave-functions-series-representation-and-converging-polynomials}{}$

This question involves solving the Bernoulli factory problem with polynomials.

In this question, a polynomial P(x) is written in *Bernstein form of degree* n if it is written as— $p(x)=\sum_{k=0}^n a_k \{n \in k\} x^k (1-x)^{n-k}$, where $a_0, ..., a_n$ are the polynomial's *Bernstein coefficients*.

The degree-n\$ Bernstein polynomial of an arbitrary function f(x)\$ has Bernstein coefficients $a_k = f(k/n)$ \$. In general, this Bernstein polynomial differs from f\$ even if f\$ is a polynomial.

3.1 Main Question

Suppose \$f:[0,1]\to [0,1]\$ is continuous and belongs to a large class of functions (for example, the \$k\$-th derivative, \$k\ge 0\$, is continuous, Lipschitz continuous, concave, strictly increasing, or bounded variation, or \$f\$ is real analytic).

- 1. (Exact Bernoulli factory): Compute the Bernstein coefficients of a sequence of polynomials (\$g_n\$) of degree 2, 4, 8, ..., \$2^i\$, ... that converge to \$f\$ from below and satisfy: \$(g_{2n}-g_{n})\$ is a polynomial with non-negative Bernstein coefficients once it's rewritten to a polynomial in Bernstein form of degree exactly \$2n\$. (See note 5 in "End Notes".) Assume \$0\lt f(\lambda)\lt 1\$ or \$f\$ is polynomially bounded.
- 2. (*Approximate Bernoulli factory*): Given \$\epsilon > 0\$, compute the Bernstein coefficients of a polynomial or rational function (of some degree \$n\$) that is within \$\epsilon\$ of \$f\$.
- 3. (Series expansion of simple functions): Find a non-negative random variable \$X\$ and a series \$f(\lambda)=\sum_{a\ge 0}\gamma_a(\lambda)\$ such that \$\gamma_a(\lambda)/\mathbb{P}(X=a)\$ (letting 0/0 equal 0) is a polynomial or rational function with rational Bernstein coefficients lying in \$[0, 1]\$. (See note 1 in "End Notes".)

The convergence rate must be $O(1/n^{r/2})$ if the class has only functions with Lipschitz-continuous (r-1)-th derivative. The method may not introduce transcendental or trigonometric functions (as with Chebyshev interpolants).

3.2 Solving the Bernoulli factory problem with polynomials

An <u>algorithm</u> (Łatuszyński et al. 2009/2011) simulates a factory function \$f(\lambda)\$ via two sequences of polynomials that converge from above and below to that function. Roughly speaking, the algorithm works as follows:

- 1. Generate U, a uniform random variate in \$[0, 1]\$.
- 2. Flip the input coin (with a probability of heads of \$\lambda\$), then build an upper and lower bound for \$f(\lambda)\$, based on the outcomes of the flips so far. In this case, these bounds come from two degree-\$n\$ polynomials that approach \$f\$ as \$n\$ gets large, where \$n\$ is the number of coin flips so far in the algorithm.
- 3. If U is less than or equal to the lower bound, return 1. If U is greater than the upper bound, return 0. Otherwise, go to step 2.

The result of the algorithm is 1 with probability *exactly* equal to \$f(\lambda)\$, or 0 otherwise.

However, the algorithm requires the polynomial sequences to meet certain requirements; among them, the sequences must be of Bernstein-form polynomials that converge from above and below to a factory function. Specifically:

For $f(\lambda)$ there must be a sequence of polynomials (g_n) in Bernstein form of degree 1, 2, 3, ... that converge to f from below and satisfy: g_n is a polynomial with non-negative Bernstein coefficients once it's rewritten to a polynomial in Bernstein form of degree exactly n+1 (see note 5 in "End Notes"; Nacu and Peres 2005; Holtz et al. 2011). For $f(\lambda)=1-f(\lambda)$ there must likewise be a sequence of this kind.

3.3 A Matter of Efficiency

However, ordinary Bernstein polynomials converge to a function at the rate \$\Omega(1/n)\$ in general, a result known since Voronovskaya (1932) and a rate that will lead to an **infinite expected number of coin flips in general**. (See also my **supplemental notes**.)

But Lorentz (1966) showed that if the function is positive and has a continuous k-th derivative, there are polynomials with nonnegative Bernstein coefficients that converge at the rate $O(1/n^{k/2})$ (and thus can enable a **finite expected number of coin flips** if the function is "smooth" enough).

Thus, people have developed alternatives, including linear combinations and iterated Boolean sums of Bernstein polynomials, to improve the convergence rate. These include Micchelli (1973), Guan (2009), Güntürk and Li (2021a, 2021b), the "Lorentz operator" in Holtz et al. (2011) (see also "New coins from old, smoothly"), Draganov (2014), and Tachev (2022).

These alternative polynomials usually include results where the error bound is the desired $O(1/n^{k/2})$, but most of those results (e.g., Theorem 4.4 in Micchelli; Theorem 5 in Güntürk and Li) have hidden constants with no upper bounds given, making them unimplementable (that is, it can't be known beforehand whether a given polynomial will come close to the target function within a user-specified error tolerance).

3.4 A Conjecture on Polynomial Approximation

The following is a **conjecture** that could help reduce this problem to the problem of finding explicit error bounds when approximating a function by polynomials.

Let $f(\lambda):[0,1]\to(0,1)$ have $r\le 1$ continuous derivatives, let M be the maximum of the absolute value of f and its derivatives up to the f-th derivative, and denote the Bernstein polynomial of degree f-1 of a function f-1 (\lambda), ..., f-1 (\lambda)

For each integer $n\ge 1$ that's a power of 2, suppose that there is D>0 such that— $\|f(\lambda)-B_n(W_n(\lambda))\| \le DM/n^{r/2}$, whenever $\|c\| \le 1$. Then there is $\|c\| \le DM/n^{r/2}$, whenever $\|c\| \le C_0$, the polynomials $\|c\| \le C_0$ in Bernstein form of degree 2, 4, 8, ..., $\|c\| \le C_0$, ..., defined as $\|c\| \le C_0$, converge from below to $\|c\| \le C_0$ and satisfy: $\|c\| \le C_0$ is a polynomial with non-negative Bernstein coefficients once it's rewritten to a polynomial in Bernstein form of degree exactly $\|c\| \le C_0$.

Equivalently (see also Nacu and Peres 2005), there is \$C_1>0\$ such that the inequality \$(PB)\$ (see below) holds true for each integer \$n\ge 1\$ that's a power of 2 (see "Strategies", below).

My goal is to see not just whether this conjecture is true, but also which value of C_0 (or C_1) suffices for the conjecture, especially for any combination of the special cases mentioned at the end of "Main Question", above.

3.5 Strategies

The following are some strategies for answering these questions:

- For iterated Boolean sums (linear combinations of iterates) of
 Bernstein polynomials (\$U_{n,k}\$ in <u>Micchelli 1973</u>; see also
 <u>Güntürk and Li</u>), find an explicit bound, with no hidden
 constants, on the approximation error for functions with
 continuous \$r\$-th derivative, or verify my <u>proofs of these bounds</u>
 <u>in Propositions B10C and B10D</u>.
- For linear combinations of Bernstein polynomials (Butzer 1953, <u>Tachev 2022</u>), verify my proof of those error bounds in <u>my</u> <u>Proposition B10</u>.
- For the "Lorentz operator" (Holtz et al. 2011), find an explicit bound, with no hidden constants, on the approximation error for the operator \$Q_{n,r}(f)\$ and for the polynomials \$(f_n)\$ and \$(g_n)\$ formed with it, and find the hidden constants \$\theta_\alpha\$, \$s\$, and \$D\$ as well as those in Lemmas 15, 17 to 22, 24, and 25 in the paper. Or verify my proof of the order-2 operator's error bounds in my Proposition B10A. The bounds should have the form \$C\cdot\max((\lambda(1-\lambda)/n)^{1/2}, 1/n)^r\$, where \$C\$ is an explicitly given constant depending only on \$f\$ and \$r\$.
- Let \$f:[-1,1]\to [0,1]\$ be continuous. Find explicit bounds, with no hidden constants, on the error in approximating \$f\$ with the following polynomials: The polynomials are similar to Chebyshev interpolants, but evaluate \$f\$ at *rational* values of \$\lambda\$ that converge to Chebyshev points (that is, converging to \$\cos(j\pi/n)\$ with increasing \$n\$). The error bounds must be close to those of Chebyshev interpolants (see, e.g., chapters 7, 8, and 12 of Trefethen, *Approximation Theory and Approximation Practice*, 2013).
- Find other polynomial operators meeting the requirements of the main question (see "Main Question", above) and having explicit error bounds, with no hidden constants, especially operators that preserve polynomials of a higher degree than linear functions.

• Find a sequence of functions $\$(W_n(f))\$$ and an explicit and tight upper bound on $\$C_1>0\$$ such that, for each integer $\$n\ge 1\$$ that's a power of $2-\$\{left|\ell(\sum_{i=0}^k W_n\ell(\frac{i}{n}\right)^k W_n\ell(\frac{i}{n}\right)^k W_n\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}n\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}{n}\ell(\frac{i}n\ell(\frac{i}{n}\ell(\frac{i}n\ell(\frac{i}{n}\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n\ell(\frac{i}n$

4 Other Questions

- Let \$f(\lambda):[0,1]\to [0,1]\$ be writable as \$f(\lambda)=\sum_{n\ge 0} a_n \lambda^n,\$ where \$a_n\ge 0\$ is rational, \$a_n\$ is nonzero infinitely often, and \$f(1)\$ is irrational. Then what are simple criteria to determine whether there is \$0\lt p\lt 1\$ such that \$0\le a_n\le p(1-p)^n\$ and, if so, to find such \$p\$? Obviously, if \$(a_n)\$ is nowhere increasing then \$1\gt p\ge a_0\$.
- For each \$r>0\$, characterize the functions \$f(\lambda)\$ that admit a Bernoulli factory where the expected number of coin flips, raised to the power of \$r\$, is finite.
- Multiple-output Bernoulli factories: Let \$f(\lambda):[a, b] \to (0, 1)\$ be continuous, where \$0\lt a\lt b\lt 1\$. Define the entropy bound as \$h(f(\lambda))/h(\lambda),\$ where \$h(x)=-x \ln(x)-(1-x) \ln(1-x)\$ is related to the Shannon entropy function. Then there is an algorithm that tosses heads with probability \$f(\lambda)\$ given a coin that shows heads with probability \$\lambda\$ and no other source of randomness (Keane and O'Brien 1994).

But, is there an algorithm for \$f\$ that produces *multiple* outputs rather than one and has an expected number of coin flips per output that is arbitrarily close to the entropy bound, uniformly for every \$\lambda\$ in \$f\$'s domain? Call

such an algorithm an *optimal factory*. (See Nacu and Peres 2005, Question 1.) And, does the answer change if the algorithm has access to a fair coin in addition to the biased coin?

So far, constants as well as \$\lambda\$ and \$1-\lambda\$ do admit an optimal factory (see same work), and, as Yuval Peres (Jun. 24, 2021) told me, there is an efficient multiple-output algorithm for \$f(\lambda) = \lambda/2\$. But are there others? See an **appendix** in one of my articles for more information on my progress on the problem.

• Pushdown automata and algebraic functions: A pushdown automaton is a finite state machine with an unbounded stack, driven by a biased coin with an unknown probability of heads, \$\lambda\$. Its stack starts with a single symbol. On each step, the machine flips the coin, then, based on the coin flip, the current state, and the top stack symbol, it moves to a new state (or keeps it unchanged) and replaces the top stack symbol with zero or more symbols. When the stack is empty, the machine stops and returns either 0 or 1 depending on the state it ends up at.

Let $f(\lambda)$ be continuous and map the open interval (0, 1) to itself. Mossel and Peres (2005) showed that a pushdown automaton can output 1 with probability $f(\lambda)$ only if $f(\lambda)$ is algebraic over the rational numbers (there is a nonzero polynomial P(x, y) in two variables and whose coefficients are rational numbers, such that P(x, f(x)) = 0 for every x in the domain of f(x). See an **[appendix in one of my articles]** (https://peteroupc.github.io/bernsupp.html#Pushdown_Automata_a_nd_Algebraic_Functions for more information on my progress on the problem.

Prove or disprove:

- 1. If \$f\$ is algebraic over rational numbers it can be simulated by a pushdown automaton.
- 2. min(\$\lambda\$, \$1-\lambda\$) and \$\lambda^{1/p}\$, for every prime \$p\ge 3\$, can be simulated by a pushdown automaton.
- 3. Given that \$f\$ is algebraic over rational numbers, it can be simulated by a pushdown automaton if and only if its "critical exponent" is a dyadic number greater than −1 or has the form \$-1-1/2^k\$ for some integer \$k\ge 1\$. (See note 2 in "End Notes"****.)

- <u>Simple simulation algorithms</u>: References are sought to papers and books that describe irrational constants or Bernoulli factory functions (continuous functions mapping (0,1) to itself) in any of the following ways. Ideally they should involve only rational numbers and should not compute *p*-adic digit expansions.
 - Simulation experiments that succeed with an irrational probability.
 - Simple **continued fraction** expansions of irrational constants.
 - Functions written as infinite power series with rational coefficients (see "Certain Power Series").
 - Irrational numbers written as series expansions with rational coefficients (see "Certain Converging Series").
 - Functions whose integral is an irrational number.
 - Closed shapes inside the unit square whose area is an irrational number. (Includes algorithms that tell whether a box lies inside, outside, or partly inside or outside the shape.) Example.
 - Generate a uniform (*x*, *y*) point inside a closed shape, then return 1 with probability *x*. For what shapes is the expected value of *x* an irrational number? **Example**.
- Given integer $m \ge 0$, rational number $0 < k \le \exp(1)$, and unknown heads probability $0 \le \lambda \le 1$, find a **Bernoulli factory** for $f(\lambda) = \exp(-(\exp(m+\lambda)-(k(m+\lambda)))) =$ $\frac{\exp(-\exp(m+\lambda))}{\exp(-\exp(m+\lambda))}$ $(k(m+\lambda)))$, $tag{PD}$ that, as much as possible, avoids calculating $h(\lambda) = \exp(m+\lambda)-k(m+\lambda)$; in this sense, the more implicitly the Bernoulli factory works with irrational or transcendental functions, the better. A solution is sought especially when *k* is 1 or 2. Note that the right-hand side of (PD) can be implemented by **ExpMinus** and division Bernoulli factories, but is inefficient and heavyweight due to the need to calculate \$\epsilon\$ for the division factory. In addition there is a Bernoulli factory that first calculates \$h(\lambda)\$ and \$floor(h(\lambda))\$ using constructive reals and then runs **ExpMinus**, but this is likewise far from lightweight. (Calculating exp(.) with floating-point operations is not acceptable for this question.)

Prove or disprove:

- Given that \$f:[0,1]\to (0,1]\$ is convex, the polynomials \$(g_n) = (B_n(f) \max_{0\le\lambda\le 1}|B_n(f)(\lambda)-f(\lambda)|)\$ (where \$n\ge 1\$ is an integer power of 2) are in Bernstein form of degree \$n\$, converge to \$f\$ from below, and satisfy: \$(g_{2n}-g_{n})\$ is a polynomial with non-negative Bernstein coefficients once it's rewritten to a polynomial in Bernstein form of degree exactly \$2n\$. The same is true for the polynomials \$(g_n) = (B_n(f) |B_n(f)(1/2)-f(1/2)|)\$, if \$f\$ is also symmetric about 1/2.
- Let \$f:(D\subseteq [0, 1])\to [0,1]\$. Given a coin that shows heads with probability \$\lambda\$ (which can be 0 or 1), it is possible to toss heads with probability \$f(\lambda)\$ using the coin and no other sources of randomness (and, thus, \$f\$ is strongly simulable) if and only if—
 - \$f\$ is constant on its domain, or is continuous and polynomially bounded on its domain (polynomially bounded means, both \$f\$ and \$1-f\$ are bounded below by min(\$x^n\$, \$(1-x)^n\$) for some integer \$n\$ [Keane and O'Brien 1994]), and
 - \circ \$f(0)\$ is 0 or 1 if 0 is in \$f\$'s domain and \$f(1)\$ is 0 or 1 whenever 1 is in \$f\$'s domain, and
 - \circ if f(0) = 0 or f(1) = 0 or both, then there is a polynomial g(x):[0,1] with computable coefficients, such that g(0) = f(0) and g(1) = f(1) whenever 0 or 1, respectively, is in the domain of f, and such that g(x) for every x in the domain of f, except at 0 and 1, and
 - if f(0) = 1 or f(1) = 1 or both, then there is a polynomial h(x):[0,1] with computable coefficients, such that h(0) = f(0) and h(1) = f(1) whenever 0 or 1, respectively, is in the domain of f(0) and such that g(x) for every f(0) in the domain of f(0) and 1.

A condition such as "0 is not in the domain of \$f\$, or \$f\$ can be extended to a Lipschitz continuous function on $[0, \epsilon]$ for some ϵ 0, 'epsilon's for some \$\epsilon>0\$" does not work. A counterexample is $f(x) = (\sin(1/x)/4+1/2)\cdot(1-(1-x)^n)$ for $\inf 1$ 0, 'f(0) = 01, which is strongly simulable at 0 despite not being Lipschitz at 0. ($f(1-x)^n$ 1 is the probability of the biased coin showing zero $\inf 1$ 1 times in a row.) Keane and O'Brien already showed strong simulability when \$D\$ contains neither 0 nor 1.

5 End Notes

Note 1: An example of X is $\mathbb{P}(X=a) = p (1-p)^a$ where $0 is a known rational. This question's requirements imply that <math>\sum_{a \le p} 1$ is a known rational. This question's requirements imply that $\sum_{a \le p} 1$ is a known rational. This question's requirements imply that $\sum_{a \le p} 1$ is a known rational. This question of 1 is a known rational. The proof of Keane and O'Brien (1994) produces a convex combination of polynomials with 0 and 1 as Bernstein coefficients, but the combination is difficult to construct (it requires finding maximums, for example) and so this proof does not appropriately answer this question.

Note 2: On pushdown automata: Etessami and Yannakakis (2009) showed that pushdown automata with rational probabilities are equivalent to recursive Markov chains (with rational transition probabilities), and that for every recursive Markov chain, the system of polynomial equations has nonnegative coefficients. But this paper doesn't deal with the case of recursive Markov chains where the transition probabilities cannot just be rational, but can also be \$\lambda\$ and \$1-\lambda\$ where \$\lambda\$ is an unknown rational or irrational probability of heads. Also, Banderier and Drmota (2014) showed the asymptotic behavior of power series solutions \$f(\lambda)\$ of a polynomial system, where both the series and the system have nonnegative real coefficients. Notably, functions of the form $\alpha^{1/p}$ where $p \ge 3$ is not a power of 2, are not possible solutions, because their so-called "critical exponent" is not dyadic. But the result seems not to apply to *piecewise* power series such as \$\min(\lambda,1-\lambda)\$, which are likewise algebraic functions.

Note 5: This condition is also known as a "consistency requirement"; it ensures that not only the polynomials "increase" to $f(\lambda)$, but also their Bernstein coefficients do as well. This condition is equivalent in practice to the following statement (Nacu & Peres 2005). For every integer $n\le 1$ that's a power of 2, $a(2n, k)\geq mathb\{E\}[a(n, X_{n,k})]=\left(\sum_{i=0}^k a(n,i) \{n\choose n} + a$

at random, all at once, from a bag containing \$2n\$ balls, \$k\$ of which are "good". See also my <u>MathOverflow question</u> on finding bounds for hypergeometric variables.

Note 6: If $W_n(0)=f(0)$ and $W_n(1)=f(1)$ for every n, then the inequality PB is automatically true when k=0 and k=2n, so that the statement has to be checked only for $0\le 1$ in addition, W_n is symmetric about 1/2, so that $W_n(\lambda)=W_n(1-\lambda)$ whenever $0\le \lambda$ (since the values S=0 in that the statement has to be checked only for $0\le \lambda$ (since the values S=0 in that they satisfy S=0 in that they satisfy S=0 in that they satisfy S=0 in the S=0 in this question is a problem of finding the S=0 in this one and find results that take advantage of S=0 in this one and find results that take advantage of S=0 in this question.

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