

VALUATION OF AMERICAN LOOKBACK-OPTIONS AND THE DISTRIBUTION OF KRONECKER SEQUENCES

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ABSTRACT. In this paper, we determine the complexity of calculating the fair value of an American lookback-option in a binomial n -step model via backwardation. It will turn out, that the complexity also depends on diophantine properties of a certain parameter of the model. Therefore, we will need some basic tools from the theory of uniform distribution for the determination of the complexity.

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1. Introduction

It is the goal of this paper to explicitly determine the complexity of finding the fair value of an American lookback-option in a binomial stock-model. This topic and this goal should not scare a potential reader who is not so familiar with or not so interested in mathematical finance! Indeed, this goal reduces to a simple counting problem, and the final result shows some connection to irregularities of distributions, more concrete, to discrepancy estimates for Kronecker sequences.

So a reader with no interest in finance can skip the subsequent Chapter 2 in which we will describe the valuation problem and in which we will reduce it to the counting problem. In Chapter 3 we formulate the problem once more, we give the result (Theorem 1) in the so-called “rational case”, and we prove Theorem 1. In Chapter 4 we give the tools from uniform distribution theory which we will need to handle the “irrational case” and to prove the corresponding Theorem 2 in Chapter 5.

2. The valuation of American lookback-options in a binomial stock-model

A *binomial n-step model* is a simple discrete model for the expansion of a stock price. This model gets its importance from the fact, that a geometric Brownian motion (which is a very common continuous stock-price model) can be approximated arbitrarily fine by such binomial models. The parameters which determine a binomial n-step model are (remark for insiders: for simplicity in the following, we restrict to interest rates $r = 0$):

- the number n of steps at times $k \cdot dt$ for $k = 0, 1, 2, \dots, n - 1$
- the starting value S_0 , i.e., the price of the stock at time 0 (a positive real number, for simplicity we set $S_0 = 1$ in all the following)
- the constant factors u (for an up-step) and d (for a down-step) (positive real numbers with $d < 1 < u$ (for insiders: this must hold since we assumed $r = 0$, otherwise the model would not be arbitrage-free))
- the constant probability p for an up-step in any of the time points $k \cdot dt$.

With these parameters within this model a stock price S can develop in the following schematic form from time 0 until time $T := n \cdot dt$.

Here each step up appears with probability p and each step down with probability $1 - p$.

An *American lookback-option* D based on this stock now is a financial product with the following property (we present here a special version of such an option, there also exist other versions):

If we possess the product D , then in every time-point $k \cdot dt$ for $k = 0, 1, 2, \dots, n$ we consider

- the actual price S_k of the stock S , and
- the maximum price $S_k(max)$ attained by the stock between times 0 and $k \cdot dt$,

and we can decide then in every time-point $k \cdot dt$ if we want to receive a payout of quantity $S_k(max) - S_k$ right now (then the product terminates) or to wait to the next time-point.

A basic question in the theory of derivative pricing is: “What is the fair value of this product D at time 0?” This question can be answered with the help of an algorithm, which is called *backwardation*. We will not explain and define this algorithm, an interested reader can find a detailed description and discussion of the algorithm with examples in [6], Volume 2, Chapters 2.10 and 2.11] (see also [2]). For further information concerning valuation of path-dependent American options in general and American lookback options in special see [1], [3] or [7].

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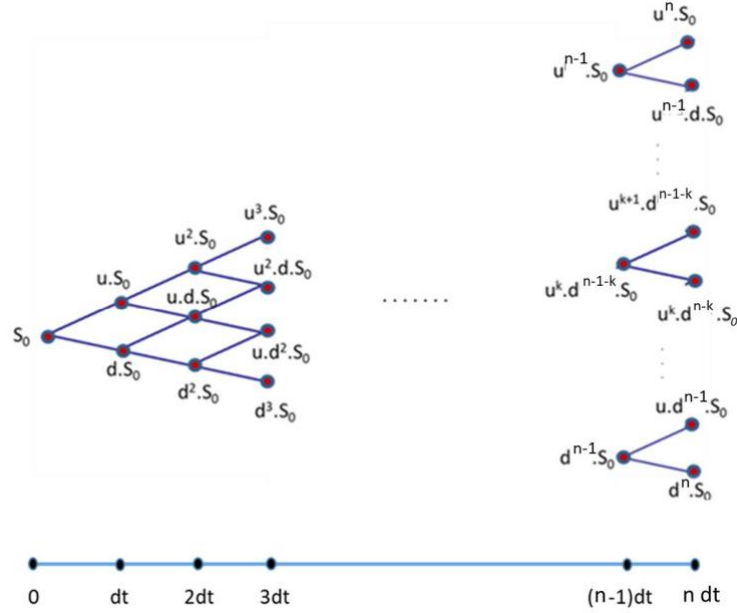


FIGURE 1. Graph of a binomial n -step model.

As is pointed out in [6], Volume 2, Chapter 2.10, the complexity of the backwardation-algorithm for an American lookback-option essentially depends on a certain quantity Q which will be defined in the following:

For an arbitrary price-point, say $u^a d^b S_0$ in the binomial model at time $k \cdot dt$ with $k = a + b$ (for $0 \leq a, b \leq n$ with $a + b = k \leq n$) let

$$M(a, b) := \#\{u^x d^y S_0 : 0 \leq x \leq a, 0 \leq y \leq b, u^x d^y S_0 \geq \max(S_0, u^a d^b S_0)\},$$

i.e., $M(a, b)$ is the number of possible maximum-values on possible paths from S_0 to $u^a d^b S_0$. Then the complexity Q of the algorithm is essentially given by

$$Q = \sum_{\substack{a, b=0 \\ a+b \leq n}}^n M(a, b).$$

A trivial upper estimate for Q obviously is

$$Q \leq \sum_{a=0}^n \sum_{b=0}^{n-a} (a+1)(b+1) = \frac{(n+1)(n+2)(n+3)(n+4)}{24} = \frac{n^4}{24} + \frac{5n^3}{12} + \mathcal{O}(n^2)$$

In the following we want to give the (at least for the leading terms) exact value of Q .

It will turn out, that the value of Q depends on diophantine properties of the following quantity $\alpha := -\frac{\log u}{\log d} = \frac{\log u}{\log e}$, if we set $d = \frac{1}{e}$. Note, that by the assumption $0 < d < 1 < u$ we have $\alpha > 0$.

The first obvious dependence on Diophantine properties is the following: Note, that $M(a, b)$ is not defined as the *number of knots* in the graph, where a possible maximum is attained, but as the *number of different possible maximum values*!

Hence: If $\alpha = \frac{\log u}{\log e}$ is rational, say $\alpha = \frac{A}{B}$ with relative prime integers A and B , then $u^B d^A = 1$, hence $u^{lB} d^{lA} = 1$ for all integers l and therefore $u^{lB+x} d^{lA+y} = u^x d^y$ for all integers l . That means, that many of the expressions $u^x d^y S_0$ in the definition of $M(a, b)$ have the same value. If α is irrational, then $u^x d^y S_0 = u^{x'} d^{y'} S_0$ for $(x, y) \neq (x', y')$ never can happen. So it seems to be obvious, that the complexity of the algorithm is smaller in the case of a rational α than in the case of an irrational α .

As a second dependence it will turn out, that in the case of an irrational α , also the degree of approximability of α by rationals will have an (slight) influence on the complexity. (Note, that rationality or irrationality of the factor α does not affect the speed of convergence of the binomial model to the geometric Brownian motion.)

In the following we will handle the case of rational α first, and the case of irrational α later.

By the way: We can choose, without loss of generality here and in all the following $S_0 = 1$.

3. The complexity of the algorithm for rational α

Before we start the investigation of the quantity Q we state the problem once more, without any reference to mathematical finance, just as an enumeration problem in a simple graph:

Problem: In a graph as illustrated in Figure 1, determine the following quantity

$$Q = \sum_{\substack{a, b=0 \\ a+b \leq n}}^n M(a, b),$$

where:

$$M(a, b) = \#\{u^x d^y S_0 : 0 \leq x \leq a, 0 \leq y \leq b, u^x d^y S_0 \geq \max(S_0, u^a d^b S_0)\}.$$

In this chapter we consider the case $\alpha := -\frac{\log u}{\log d} = \frac{\log u}{\log e} = \frac{A}{B}$ with relative prime integers A and B .

Example 1: As a warm-up – and as a support for the intuition – we first handle the most basic case (which also appears frequently in the concrete application in finance), namely the case, where $d = \frac{1}{u}$, i.e., $\alpha = A = B = 1$. In this case $u^x d^y = u^{x-y}$, hence $u^x d^y = u^{x'} d^{y'}$ iff $x - y = x' - y'$, and the graph of the binomial model then has the following (schematic) form (see Figure 2):

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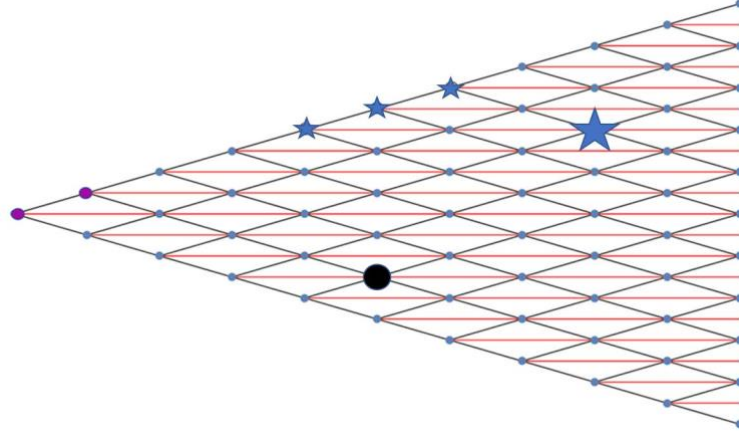


FIGURE 2. Graph of a binomial n -step model

In Figure 2 the horizontal lines indicate the different values that can be attained by $u^x d^y$.

For a “big-star” point $u^a d^b$ which is larger or equal to 1, the “small-star” points in the graphic indicate the possible different values $u^x d^y$ with $0 \leq x \leq a$, $0 \leq y \leq b$, and $u^x d^y \geq u^a d^b$.

For a “big-disk” point $u^a d^b$ which is less than 1, the “small-disc” points (most left two knots on the upper barrier of the model) in the graphic indicate the possible different values $u^x d^y$ with $0 \leq x \leq a$, $0 \leq y \leq b$, and $u^x d^y \geq 1$.

From Figure 2 it is easy to understand that for a, b such that $u^a d^b$ is larger or equal to 1 (e.g., “big-star” point) the value of $M(a, b)$ (e.g., number of “small-star” points) is given by $b + 1$. Also from Figure 2 it is easy to understand that for a, b such that $u^a d^b$ is less than 1 (e.g., “big-disk” point) the value of $M(a, b)$ (e.g., number of “small-disk” points) is given by $a + 1$.

Hence for the complexity Q of the algorithm we have

$$\begin{aligned} Q &= \sum_{\substack{a,b=0 \\ a+b \leq n}}^n M(a, b) = \sum_{a=0}^n \sum_{\substack{b=0 \\ b \leq a}}^{n-a} (b + 1) + \sum_{a=0}^n \sum_{\substack{b=0 \\ a < b}}^{n-a} (a + 1) \\ &= \sum_{a=0}^{\lfloor \frac{n}{2} \rfloor} \sum_{b=0}^a (b + 1) + \sum_{a=\lfloor \frac{n}{2} \rfloor + 1}^n \sum_{b=0}^{n-a} (b + 1) + \sum_{a=0}^{\lfloor \frac{n}{2} \rfloor} \sum_{b=a+1}^{n-a} (a + 1) \\ &= \begin{cases} \frac{n^3}{12} + \frac{5n^2}{8} + \frac{17n}{12} + 1 & \text{if } n \text{ is even,} \\ \frac{n^3}{12} + \frac{5n^2}{8} + \frac{17n}{12} + \frac{7}{8} & \text{if } n \text{ is odd.} \end{cases} \end{aligned}$$

In any case $Q = \frac{n^3}{12} + \mathcal{O}(n^2)$.

Now we move to the general case $\alpha = \frac{A}{B}$ with integers A and B with $\gcd(A, B) = 1$. We start with a Lemma:

LEMMA 1. *If $u^a d^b \geq 1$ then*

$$M(a, b) = \sum_{x=0}^{B-1} \sum_{y=0}^b 1 + \sum_{x=0}^a \sum_{y=0}^{A-1} 1 + \mathcal{O}(1) =: \sum^{\oplus} + \mathcal{O}(1)$$

$$\frac{u^x}{e^y} \geq \frac{u^a}{e^b} \quad \frac{u^x}{e^y} \geq \frac{u^a}{e^b}$$

where the $\mathcal{O}$ -constant may depend on A and B only.

a) If $u^a d^b < 1$ then

$$M(a, b) = \sum_{x=0}^{B-1} \sum_{y=0}^b 1 + \sum_{x=0}^a \sum_{y=0}^{A-1} 1 + \mathcal{O}(1).$$

$$\frac{u^x}{e^y} \geq 1 \quad \frac{u^x}{e^y} \geq 1$$

Proof. a) Since we prove the result with the additional $\mathcal{O}(1)$ -term it suffices to prove the result for all a, b which both are larger than A and B .

We have $u^{lB+x} d^{lA+y} = u^x d^y$ for all integers l and all x and y . Hence, every value $u^x d^y$ with $0 \leq x \leq a$ and $0 \leq y \leq b$ equals $u^{x'} d^{y'}$ with some x' with $0 \leq x' \leq B-1$ and y' with $0 \leq y' \leq b$ or with some y' with $0 \leq y' \leq A-1$ and x' with $0 \leq x' \leq a$. Hence $M(a, b) \leq \sum^{\oplus} + \mathcal{O}(1)$.

On the other hand any two values $u^x d^y$ and $u^{x'} d^{y'}$ with $0 \leq x, x' \leq B-1$ and $0 \leq y, y' \leq b$ or with $0 \leq x, x' \leq a$ and $0 \leq y, y' \leq A-1$ are different unless $(x, y) = (x', y')$, since $\frac{u^x}{e^y} = \frac{u^{x'}}{e^{y'}}$ implies $(x' - x)A = (y' - y)B$. Hence, $x' - x = lB$ and $y' - y = lA$ for some integer l . If both x', x are in $[0, B-1]$ or if both y', y are in $[0, A-1]$, then $l = 0$ and $(x, y) = (x', y')$. If $l \neq 0$ and $x \in [0, B-1]$ and $y' \in [0, A-1]$, then $x' = x + lB$ and $y = y' - lA$, hence x' or y is negative, a contradiction. Hence $\sum^{\oplus} \leq M(a, b) + \mathcal{O}(1)$, and the result follows. (Note that the twice counted x and y in $\sum^{\oplus}$ are absorbed by the $\mathcal{O}(1)$ -term.)

Part b) is shown quite analogously.

THEOREM 1. If u, d are such that $\alpha := -\frac{\log u}{\log d} =: \frac{A}{B}$ is rational, then for the complexity Q of the backwardation algorithm we have

$$Q = \frac{AB}{6(A+B)} n^3 + \mathcal{O}(n^2)$$

REMARK. Note that the result coincides with the result of Example 1, by inserting $A = B = 1$.

Proof. By Lemma 1 it remains to show that

$$\sum_{a=0}^n \sum_{b=0}^{n-a} \sum_{x=0}^{B-1} \sum_{y=0}^b 1 + \sum_{a=0}^n \sum_{b=0}^{n-a} \sum_{x=0}^a \sum_{y=0}^{A-1} 1 +$$

$$\sum_{a=0}^n \sum_{b=0}^{n-a} \sum_{x=0}^{B-1} \sum_{y=0}^b 1 + \sum_{a=0}^n \sum_{b=0}^{n-a} \sum_{x=0}^a \sum_{y=0}^{A-1} 1 =:$$

$$\sum_{\frac{u^a}{e^b} \geq 1} + \sum_{\frac{u^a}{e^b} < 1} + \sum_{\frac{u^x}{e^y} \geq 1} + \sum_{\frac{u^x}{e^y} < 1} = \frac{AB}{6(A+B)} n^3 + \mathcal{O}(n^2).$$

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Consider Σ_1 : For given a, b and x with $0 \leq x \leq B - 1$ the conditions $\frac{u^x}{e^y} \geq \frac{u^a}{e^b}$ and $\frac{u^a}{e^b} \geq 1$ imply $\frac{u^x}{e^y} \geq 1$, hence $y \leq x \frac{A}{B} < A$ and therefore $\Sigma_1 = \mathcal{O}(n^2)$.

Consider Σ_2 : Here for given a, b and x with $0 \leq x \leq B - 1$ the condition $\frac{u^x}{e^y} \geq 1$ implies $y \leq x \frac{A}{B} < A$ and so also $\Sigma_3 = \mathcal{O}(n^2)$.

It remains to study Σ_2 and Σ_4 .

We start with Σ_2 : The conditions $\frac{u^a}{e^b} \geq 1$ and $\frac{u^x}{e^y} \geq \frac{u^a}{e^b}$ are equivalent with $b \leq \left\lfloor a \frac{A}{B} \right\rfloor$ and $y \leq b - (a - x) \frac{A}{B}$. Hence

$$\Sigma_2 = \sum_{a=0}^n \sum_{b=0}^{\min(n-a, \left\lfloor a \frac{A}{B} \right\rfloor)} \sum_{x=0}^a \sum_{\substack{y=0 \\ y < b - (a-x) \frac{A}{B}}}^{A-1} 1.$$

The innermost sum has the value A if $b - (a - x) \frac{A}{B} > A - 1$, i.e., if $x > a + B - (b + 1) \frac{A}{B}$ and it has the value $\left\lfloor b - (a - x) \frac{A}{B} \right\rfloor + 1$ if $0 < b - (a - x) \frac{A}{B} \leq A - 1$, i.e., if $a - \frac{B}{A} b < x \leq B - (b + 1) \frac{B}{A} + a$. Therefore

$$\begin{aligned} \Sigma_2 &= \sum_{a=0}^n \sum_{b=0}^{\min(n-a, \left\lfloor a \frac{A}{B} \right\rfloor)} \sum_{x=\left\lfloor a - \frac{B}{A} b \right\rfloor + 1}^{\left\lfloor B - (b + 1) \frac{B}{A} + a \right\rfloor} \left(\left\lfloor b - (a - x) \frac{A}{B} \right\rfloor + 1 \right) + \\ &+ \sum_{a=0}^n \sum_{b=0}^{\min(n-a, \left\lfloor a \frac{A}{B} \right\rfloor)} \sum_{x=\max(0, \left\lfloor a + B - (b + 1) \frac{B}{A} \right\rfloor + 1)}^a A =: \Sigma_{21} + \Sigma_{22}. \end{aligned}$$

In Σ_{21} the summand always is at most A and the number of x in Σ_x is at most $\left\lfloor B - (b + 1) \frac{B}{A} + a \right\rfloor - \left\lfloor a - \frac{B}{A} b \right\rfloor \leq B + 1$, hence $\Sigma_{21} = \mathcal{O}(n^2)$, and $\Sigma_2 = \Sigma_{22} + \mathcal{O}(n^2)$.

Further

$$\begin{aligned} \Sigma_{22} &= \sum_{a=0}^n \sum_{b=0}^{\min(n-a, \left\lfloor a \frac{A}{B} \right\rfloor)} \sum_{x=\max(0, \left\lfloor a - b \frac{B}{A} \right\rfloor + 1)}^a A + \mathcal{O}(n^2) \\ &= \sum_{a=0}^n \sum_{b=0}^{\min(n-a, \left\lfloor a \frac{A}{B} \right\rfloor)} bB + \mathcal{O}(n^2) \\ &= \sum_{a=0}^{\left\lfloor \frac{n}{1+A/B} \right\rfloor} \sum_{b=0}^{\left\lfloor a \frac{A}{B} \right\rfloor} bB + \sum_{a=\left\lfloor \frac{n}{1+A/B} \right\rfloor + 1}^n \sum_{b=0}^{n-a} bB + \mathcal{O}(n^2) \\ &= \frac{A^2 B}{6(A+B)^2} n^3 + \mathcal{O}(n^2). \end{aligned}$$

The reader may trust us, that by quite analogous elementary calculation we obtain

$$\Sigma_4 = \frac{B^2 A}{6(A+B)^2} n^3 + \mathcal{O}(n^2),$$

and the result follows

4. Some tools from uniform distribution theory

Let us turn now to the case of irrational α . To handle this case we will need two well-known tools from uniform distribution theory, namely the Koksma-Hlawka inequality and an estimate for the discrepancy of the Kronecker-sequence $\{n\alpha\}$ for $n = 1, 2, 3, \dots$. These tools will lead to Lemma 2, which will be needed for the proof of Theorem 2 in Chapter 5.

For an irrational number β with infinite continued fraction expansion $\beta := [b_0, b_1, b_2, \dots]$ and resulting best approximation denominators $q_0 \leq q_1 < q_2 < \dots$, and for an arbitrary positive integer n let s be such that $q_s \leq n < q_{s+1}$. Then we define $KB_n(\beta) := \frac{n}{q_s} + \sum_{i=1}^s b_i$.

Note that $KB_n(\beta) \leq n$ always and that for all β irrational we have $KB_n(\beta) = o(n)$.

Moreover, for almost all irrationals β it is $KB_n(\beta) = o((\log n)^{1+\epsilon})$ for all $\epsilon > 0$ (see [4]).

LEMMA 2. *There is an absolute constant c such that*

- a) $\left| \sum_{l=1}^n \{l\beta\} \left(\frac{l}{n} \right)^2 - \frac{1}{6} \right| \leq cKB_n(\beta)$ for all n ;
- b) $\left| \sum_{l=1}^n \{l\beta\} \left(1 - \frac{l}{n+\delta} \right)^2 - \frac{1}{6} \right| \leq cKB_n(\beta)$ for all n and all δ with $0 \leq \delta < 1$;
- c) $\left| \sum_{l=1}^n \{l\beta\} - \frac{1}{2} \right| \leq cKB_n(\beta)$.

Proof. a) This follows immediately by applying the two-dimensional Koksma-Hlawka-inequality (see [[5], Theorem 5.5 in Part 2]) to the function $f: [0,1]^2 \rightarrow \mathbb{R}$ with $f(x, y) := xy^2$, which is of bounded variation V in the sense of Hardy and Krause, and which satisfies $\int_0^1 \int_0^1 f(x, y) dx dy = 1/6$, i.e.,

$$\left| \int_0^1 \int_0^1 f(x, y) dx dy - \frac{1}{n} \sum_{l=1}^n f\left(\{l\beta\}, \frac{l}{n}\right) \right| \leq V \cdot D_n,$$

where D_n denotes the discrepancy of the point set $\left(\{l\beta\}, \frac{l}{n}\right) l = 1, 2, \dots, n$ in $[0,1]^2$. And finally, it follows from the fact, that for the discrepancy D_n we have $D_n \leq c' \cdot \frac{KB_n(\beta)}{n}$ with an absolute constant c' . This is an immediate consequence of formula (3.18) in the proof of Theorem 3.4 in Part 2 of [5] concerning the discrepancy of the Kronecker-sequence $\{l\beta\}$ for $l = 1, 2, 3, \dots$ and by Example 2.2 in Part 2 of [5].

- a) For the point set $\left(\{l\beta\}, 1 - \frac{l}{n+\delta}\right)$ for $l = 0, \dots, n$ we have the same discrepancy estimate as used for $\left(\{l\beta\}, \frac{l}{n}\right), l = 1, \dots, n$, since it is just a slightly perturbed and mirrored version of the second point set, and so also this result follows.
- b) This follows immediately from the one-dimensional Koksma-Hlawka in-equality for $f(x) = x$ and the discrepancy estimate for Kronecker sequences cited above.

For the proof of Theorem 2 we also will use the following simple fact:

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LEMMA 3. *For every α irrational there is a constant $c = c(\alpha)$ such that for all n we have*

$$KB_n\left(\frac{1}{\alpha}\right) \leq c \cdot KB_n(\alpha)$$

Proof. Let first $\alpha < 1$, $\alpha := [0, b_1, b_2, \dots]$ and $q_s \leq n < q_{s+1}$. By definition $KB_n(\alpha) = b_1 + b_2 + \dots + b_s + \frac{n}{q_s}$. We have $\frac{1}{\alpha} = [b_1, b_2, b_3, \dots]$. Let $Q_0 \leq Q_1 < Q_2 < \dots$ be the best approximation denominations of $\frac{1}{\alpha}$, i.e., $Q_{i+1} = b_{i+2}Q_i + Q_{i-1}$, then

$$\begin{aligned} \begin{pmatrix} p_{s-1} & p_s \\ q_{s-1} & q_s \end{pmatrix} &= \begin{pmatrix} 0 & 1 \\ 1 & b_1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & b_2 \end{pmatrix} \dots \begin{pmatrix} 0 & 1 \\ 1 & b_s \end{pmatrix} \\ &= \begin{pmatrix} 0 & 1 \\ 1 & b_1 \end{pmatrix} \begin{pmatrix} P_{s-2} & P_{s-1} \\ Q_{s-2} & Q_{s-1} \end{pmatrix}, \end{aligned}$$

hence $Q_{s-1} < q_s$ and, more detailed,

$$q_s = P_{s-1} + b_1 Q_{s-1} \leq (\alpha + 1)Q_{s-1} + b_1 Q_{s-1} \leq \left(\alpha + 1 + \frac{1}{\alpha}\right) Q_{s-1} =: \kappa Q_{s-1}$$

With $\kappa := \alpha + 1 + \frac{1}{\alpha}$

We have $KB_n\left(\frac{1}{\alpha}\right) = b_2 + b_3 + \dots + b_s + \dots + b_{s+v+1} + \frac{n}{Q_{s+v}}$ if $v \geq -1$ is such that $Q_{s+v} \leq n < Q_{s+v+1}$. Further

$$\frac{n}{Q_{s+v}} \frac{Q_{s+v}}{Q_{s+v-1}} \dots \frac{Q_s}{Q_{s-1}} = \frac{n}{Q_{s-1}} \leq \frac{n\kappa}{q_s}.$$

Assume now, that v is odd and ≥ 1 , say $v = 2t + 1$ and let $x_i := \frac{Q_{s+2i}}{Q_{s+2i-1}} \cdot \frac{Q_{s+2i+1}}{Q_{s+2i}}$. Note that $x_i = \frac{Q_{s+2i+1}}{Q_{s+2i-1}} \geq 2$ for all i , hence $x_0 x_1 \dots x_t \geq x_0 + x_1 + \dots + x_t$. Further

$$x_i = \frac{Q_{s+2i}}{Q_{s+2i-1}} \frac{Q_{s+2i+1}}{Q_{s+2i}} \geq \frac{1}{2} \left(\frac{Q_{s+2i}}{Q_{s+2i-1}} + \frac{Q_{s+2i+1}}{Q_{s+2i}} \right) \geq \frac{1}{2} (b_{s+2i+1} + b_{s+2i+1}),$$

Hence

$$\begin{aligned} \frac{n\kappa}{q_s} &\geq \frac{n}{2Q_{s+v}} (b_{s+v+1} + b_{s+v} + \dots + b_{s+1}) \\ &\geq \frac{1}{5} \left(\frac{n}{Q_{s+v}} + b_{s+v+1} + b_{s+v} + \dots + b_{s+1} \right) \end{aligned}$$

(note that $\frac{n}{2Q_{s+v}} \geq \frac{1}{2}$ and $b_{s+v+1} + b_{s+v} + \dots + b_{s-1} \geq 2$).

Altogether we have

$$\begin{aligned} KB_n\left(\frac{1}{\alpha}\right) &= b_2 + \dots + b_s + b_{s+1} + \dots + b_{s+v+1} + \frac{n}{Q_{s+v}} \\ &\leq 5\kappa \frac{n}{q_s} + b_2 + b_3 + \dots + b_s \\ &\leq 5\kappa KB_n(\alpha). \end{aligned}$$

The cases $v = -1$, v even and $\alpha > 1$ are dealt with quite analogously with obvious adaptations and will not be carried out here in detail.

5. The complexity of the algorithm for irrational α

In the irrational case we will show:

THEOREM 2. *If u, d are such that $\alpha := -\frac{\log u}{\log d}$ is irrational, then for the complexity Q of the backwardation algorithm we have*

$$Q = \frac{\alpha}{24(1+\alpha)^2} n^4 + \frac{1}{12} \left(1 + \frac{\alpha}{(1+\alpha)^2} \right) n^3 + \mathcal{O}(KB_n(\alpha)n^2),$$

where $KB_n(\alpha)$ is as defined in Chapter 4.

REMARK. *Note again that $KB_n(\alpha) = o(n)$ for all irrational α . Furthermore, $KB_n(\alpha) = o((\log n)^{1+\epsilon})$ for almost all α for all $\epsilon > 0$.*

Proof. Since for α irrational we have $\frac{u^x}{d^y} = \frac{u^{x'}}{d^{y'}}$ for some integers x, x', y, y' if and only if $(x, y) = (x', y')$, we just have to estimate the sum

$$Q = \sum_{\substack{a,b=0 \\ a+b \leq n}}^n \sum_{x=0}^a \sum_{\substack{y=0 \\ \frac{u^x}{e^y} \geq \max\left(1, \frac{u^a}{e^b}\right)}}^b 1.$$

We split the sum according to $\frac{u^a}{e^b} \geq 1$, or $\frac{u^a}{e^b} < 1$, i.e., according to $b \leq a\alpha$, or $b > a\alpha$:

$$Q = \sum_{a=0}^n \sum_{\substack{b=0 \\ b \leq a\alpha}}^{n-a} \sum_{x=0}^a \sum_{\substack{y=0 \\ \frac{u^x}{e^y} \geq \frac{u^a}{e^b}}}^b 1 + \sum_{a=0}^n \sum_{\substack{b=0 \\ b > a\alpha}}^{n-a} \sum_{x=0}^a \sum_{\substack{y=0 \\ \frac{u^x}{e^y} \geq 1}}^b 1 =: \Sigma_1 + \Sigma_2$$

Using the abbreviation $M := \left\lfloor \frac{n}{1+\alpha} \right\rfloor$ we have

$$\begin{aligned} \Sigma_1 &= \sum_{a=0}^M \sum_{b=0}^{\lfloor a\alpha \rfloor} \sum_{x=0}^a \sum_{\substack{y=0 \\ (b-y) \geq (a-x)\alpha}}^b 1 \\ &\quad + \sum_{a=M+1}^n \sum_{x=0}^{n-a} \sum_{x=0}^a \sum_{\substack{y=0 \\ (b-y) \geq (a-x)\alpha}}^b 1 \\ &= \Sigma_{11} + \Sigma_{12}. \end{aligned}$$

We now simplify the inner double sum $\sum_x \sum_y \dots$ in these expressions:

$$\sum_{x=0}^a \sum_{\substack{y=0 \\ (b-y) \geq (a-x)\alpha}}^b 1 = \sum_{x'=0}^a \sum_{\substack{y'=0 \\ y' \geq x'\alpha}}^b 1 = \sum_{y=0}^b \sum_{x=0}^{\lfloor \frac{y}{\alpha} \rfloor} 1 = \sum_{y=0}^b \left(\left\lfloor \frac{y}{\alpha} \right\rfloor + 1 \right).$$

Here the last-but-one equality holds since the conditions $b \leq a\alpha$ made for Σ_1 and $y \geq x\alpha$ imply $x \leq \frac{y}{\alpha} \leq \frac{b}{\alpha} \leq a$, hence x runs from 0 to $\left\lfloor \frac{y}{\alpha} \right\rfloor$. We get

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$$\begin{aligned}\Sigma_{11} &= \sum_{a=0}^M \sum_{b=0}^{\lfloor a\alpha \rfloor} \sum_{y=0}^b \left(\left\lfloor \frac{y}{\alpha} \right\rfloor + 1 \right) \\ &= \sum_{a=0}^M \sum_{b=0}^{\lfloor a\alpha \rfloor} \sum_{y=0}^b \left(\frac{y}{\alpha} + 1 \right) - \sum_{a=0}^M \sum_{b=0}^{\lfloor a\alpha \rfloor} \sum_{y=0}^b \left\{ \frac{y}{\alpha} \right\} =: \Sigma_{111} - \Sigma_{112} .\end{aligned}$$

We have

$$\begin{aligned}\Sigma_{111} &= \sum_{a=0}^M \sum_{b=0}^{\lfloor a\alpha \rfloor} \left(\frac{1}{\alpha} \frac{b(b+1)}{2} + b + 1 \right) \\ &= \frac{1}{2\alpha} \sum_{a=0}^M \sum_{b=0}^{\lfloor a\alpha \rfloor} b^2 + \left(\frac{1}{2\alpha} + 1 \right) \sum_{a=0}^M \sum_{b=0}^{\lfloor a\alpha \rfloor} b + \mathcal{O}(n^2) \\ &= \frac{1}{2\alpha} \sum_{a=0}^M \frac{\lfloor a\alpha \rfloor (\lfloor a\alpha \rfloor + 1) (2\lfloor a\alpha \rfloor + 1)}{6} \\ &\quad + \left(\frac{1}{2\alpha} + 1 \right) \sum_{a=0}^M \frac{\lfloor a\alpha \rfloor (\lfloor a\alpha \rfloor + 1)}{2} + \mathcal{O}(n^2) \\ &= \frac{1}{6\alpha} \sum_{a=0}^M \lfloor a\alpha \rfloor^3 + \left(\frac{1}{2\alpha} + \frac{1}{2} \right) \sum_{a=0}^M \lfloor a\alpha \rfloor^2 + \mathcal{O}(n^2).\end{aligned}$$

Now we see that

$$\begin{aligned}\sum_{a=0}^M \lfloor a\alpha \rfloor^2 &= \sum_{a=0}^M (a\alpha - \{a\alpha\})^2 = \alpha^2 \sum_{a=0}^M a^2 + \mathcal{O}(n^2) \\ &= \alpha^2 \frac{M(M+1)(2M+1)}{6} + \mathcal{O}(n^2) = \frac{\alpha^2}{3} M^3 + \mathcal{O}(n^2)\end{aligned}$$

and

$$\begin{aligned}\sum_{a=0}^M \lfloor a\alpha \rfloor^3 &= \sum_{a=0}^M (a\alpha - \{a\alpha\})^3 \\ &= \alpha^3 \sum_{a=0}^M a^3 - 3\alpha^2 \sum_{a=0}^M a^2 \{a\alpha\} + \mathcal{O}(n^2) \\ &= \frac{\alpha^3}{4} M^2 (M+1)^2 - 3\alpha^2 M^3 \left(\frac{1}{M} \sum_{a=1}^M \left(\frac{a}{M} \right)^2 \{a\alpha\} - \frac{1}{6} \right) - \frac{\alpha^2}{2} M^3 \\ &= \frac{\alpha^3}{4} M^2 (M+1)^2 - \frac{\alpha^2}{2} M^3 - 3\alpha^2 M^3 E_1,\end{aligned}$$

with

$$E_1 := \frac{1}{M} \sum_{a=1}^M \left(\frac{a}{M} \right)^2 \{a\alpha\} - \frac{1}{6}.$$

Hence

$$\begin{aligned}\Sigma_{111} &= \frac{\alpha^2}{24} M^2 (M+1)^2 - \frac{\alpha}{12} M^3 - \frac{\alpha}{2} E_1 + \left(\frac{1}{2\alpha} + \frac{1}{2} \right) \frac{\alpha^2}{3} M^3 + \mathcal{O}(n^2) \\ &= \frac{\alpha^2}{24} M^4 + \left(\frac{\alpha^2}{4} + \frac{\alpha}{12} \right) M^3 - \frac{\alpha}{2} M^3 E_1 + \mathcal{O}(n^2).\end{aligned}$$

Further

$$\begin{aligned}
 \Sigma_{112} &= \sum_{a=0}^M \sum_{b=0}^{\lfloor a\alpha \rfloor} \sum_{y=0}^b \left\{ \frac{y}{\alpha} \right\} = \sum_{y=1}^{\lfloor M\alpha \rfloor} \left\{ \frac{y}{\alpha} \right\} \sum_{\substack{a=0 \\ y \leq \lfloor a\alpha \rfloor}}^M \sum_{b=y}^{\lfloor a\alpha \rfloor} \\
 &= \sum_{y=1}^{\lfloor M\alpha \rfloor} \left\{ \frac{y}{\alpha} \right\} \sum_{a=0}^M ([a\alpha] - y + 1) \\
 &= \sum_{y=1}^{\lfloor M\alpha \rfloor} \left\{ \frac{y}{\alpha} \right\} \sum_{l=1}^{M - \lfloor \frac{y}{\alpha} \rfloor} l\alpha + \mathcal{O}(n^2) \\
 &= \sum_{y=1}^{\lfloor M\alpha \rfloor} \left\{ \frac{y}{\alpha} \right\} \frac{\alpha}{2} \left(M - \frac{y}{\alpha} \right)^2 + \mathcal{O}(n^2) \\
 &= \frac{\alpha}{2} M^2 \lfloor M\alpha \rfloor \left(\frac{1}{\lfloor M\alpha \rfloor} \sum_{y=1}^{\lfloor M\alpha \rfloor} \left\{ \frac{y}{\alpha} \right\} \left(1 - \frac{y}{M\alpha} \right)^2 \right) + \mathcal{O}(n^2) \\
 &= \frac{\alpha^2 M^3}{12} + \frac{\alpha^2 M^3}{2} E_2 + \mathcal{O}(n^2),
 \end{aligned}$$

where

$$E_2 := \frac{1}{\lfloor M\alpha \rfloor} \sum_{y=1}^{\lfloor M\alpha \rfloor} \left\{ \frac{y}{\alpha} \right\} \left(1 - \frac{y}{M\alpha} \right)^2 - \frac{1}{6}.$$

Summing up, we get

$$\begin{aligned}
 \Sigma_{11} &= \Sigma_{111} - \Sigma_{112} \\
 &= \frac{\alpha^2}{24} M^4 + \left(\frac{\alpha^2}{6} + \frac{\alpha}{12} \right) M^3 - \frac{\alpha}{2} M^3 E_1 - \frac{\alpha^2}{2} M^3 E_2 + \mathcal{O}(n^2).
 \end{aligned}$$

We turn now to Σ_{12} :

$$\begin{aligned}
 \Sigma_{12} &= \sum_{a=M+1}^n \sum_{b=0}^{n-a} \sum_{x=0}^a \sum_{\substack{y=0 \\ (b-y) \geq (a-x)\alpha}}^b 1 \\
 &= \sum_{a=M+1}^n \sum_{b=0}^{n-a} \sum_{y=0}^b \left(\left\lfloor \frac{y}{\alpha} \right\rfloor + 1 \right) \\
 &= \sum_{a=M+1}^n \sum_{b=0}^{n-a} \sum_{y=0}^b \left(\frac{y}{\alpha} + 1 \right) - \sum_{a=M+1}^n \sum_{b=0}^{n-a} \sum_{y=0}^b \left\{ \frac{y}{\alpha} \right\} \\
 &=: \Sigma_{121} - \Sigma_{122}.
 \end{aligned}$$

We have

$$\begin{aligned}
 \Sigma_{121} &= \sum_{a=M+1}^n \sum_{b=0}^{n-a} \left(\frac{1}{\alpha} \frac{b(b+1)}{2} + b + 1 \right) \\
 &= \frac{1}{2\alpha} \sum_{a=M+1}^n \sum_{b=0}^{n-a} b^2 + \left(\frac{1}{2\alpha} + 1 \right) \sum_{a=M+1}^n \sum_{b=0}^{n-a} b + \mathcal{O}(n^2) \\
 &= \frac{1}{24\alpha} (n-M)^2 ((n-M)^2 - 1) \\
 &\quad + \left(\frac{1}{12\alpha} + \frac{1}{6} \right) (n-M) ((n-M)^2 - 1) + \mathcal{O}(n^2) \\
 &= \frac{1}{24\alpha} (n-m)^4 + \left(\frac{1}{12\alpha} + \frac{1}{6} \right) (n-M)^3 + \mathcal{O}(n^2).
 \end{aligned}$$

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$$\begin{aligned}
\Sigma_{122} &= \sum_{a=M+1}^n \sum_{b=0}^{n-a} \sum_{y=0}^b \left\{ \frac{y}{\alpha} \right\} \\
&= \sum_{y=1}^{n-M-1} \left\{ \frac{y}{\alpha} \right\} \sum_{a=M+1}^{n-y} \sum_{b=y}^{n-a} 1 \\
&= \sum_{y=1}^{n-M-1} \left\{ \frac{y}{\alpha} \right\} \sum_{a=M+1}^{n-y} (n-a-y+1) \\
&= \sum_{y=1}^{n-M-1} \left\{ \frac{y}{\alpha} \right\} \frac{(n-M-y)(n-M-y+1)}{2} \\
&= \frac{1}{2} \sum_{y=1}^{n-M} \left\{ \frac{y}{\alpha} \right\} (n-M-y)^2 + \mathcal{O}(n^2) \\
&= \frac{(n-M)^3}{2} \left(\frac{1}{n-M} \sum_{y=1}^{n-M} \left\{ \frac{y}{\alpha} \right\} \left(1 - \frac{y}{n-M} \right)^2 \right) + \mathcal{O}(n^2) \\
&= \frac{(n-M)^3}{12} + \frac{(n-M)^3}{2} E_3,
\end{aligned}$$

where

$$E_3 := \frac{1}{n-M} \sum_{y=1}^{n-M} \left\{ \frac{y}{\alpha} \right\} \left(1 - \frac{y}{n-M} \right)^2 - \frac{1}{6}.$$

Summing up we get

$$\begin{aligned}
\Sigma_{12} &= \Sigma_{121} - \Sigma_{122} \\
&= \frac{1}{24\alpha} (n-M)^4 + \left(\frac{1}{12\alpha} + \frac{1}{12} \right) (n-M)^3 - \frac{(n-M)^3}{2} E_3 + \mathcal{O}(n^2).
\end{aligned}$$

Finally we have to deal with Σ_2 :

$$\begin{aligned}
\Sigma_2 &= \sum_{a=0}^n \sum_{\substack{b=0 \\ b > a\alpha}}^{n-a} \sum_{x=0}^a \sum_{\substack{y=0 \\ \frac{u^x}{e^y} \geq 1}}^b 1 \\
&= \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} \sum_{x=0}^a \sum_{\substack{y=0 \\ y < x\alpha}}^b 1 \\
&= \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} \sum_{x=0}^a \sum_{y=0}^{\lfloor x\alpha \rfloor} 1 \\
&= \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} \sum_{x=0}^a (\lfloor x\alpha \rfloor + 1) \\
&= \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} \sum_{x=0}^a (x\alpha + 1) - \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} \sum_{x=0}^a \{x\alpha\} \\
&= \Sigma_{21} - \Sigma_{22}.
\end{aligned}$$

Now

$$\begin{aligned}
\Sigma_{21} &= \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} \left(\alpha \frac{a(a+1)}{2} + (a+1) \right) \\
&= \frac{\alpha}{2} \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} a^2 + \left(\frac{\alpha}{2} + 1 \right) \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} a + \mathcal{O}(n^2) \\
&= \frac{\alpha}{2} \sum_{a=0}^M a^2 (n-a-[a\alpha]) + \left(\frac{\alpha}{2} + 1 \right) \sum_{a=0}^M a (n-a-[a\alpha]) + \mathcal{O}(n^2) \\
&= \sum_{a=0}^M \left(\frac{\alpha}{2} a^2 + \left(\frac{\alpha}{2} + 1 \right) a \right) (n-a-a\alpha) + \frac{\alpha}{2} \sum_{a=0}^M a^2 \{a\alpha\} + \mathcal{O}(n^2) \\
&= \frac{\alpha}{6} M^3 n - M^4 \left(\frac{\alpha}{8} + \frac{\alpha^2}{8} \right) - M^3 \left(\frac{5\alpha^2}{12} + \frac{3\alpha}{4} + \frac{1}{3} \right) \\
&\quad + \left(\frac{\alpha}{2} + \frac{1}{2} \right) M^2 n - \frac{\alpha}{2} \sum_{a=0}^M a^2 \{a\alpha\} + \mathcal{O}(n^2).
\end{aligned}$$

Further

$$\begin{aligned}\Sigma_{21} &= \frac{\alpha}{6} M^3 n - M^4 \left(\frac{\alpha}{8} + \frac{\alpha^2}{8} \right) - M^3 \left(\frac{5\alpha^2}{12} + \frac{2\alpha}{3} + \frac{1}{3} \right) \\ &\quad + \left(\frac{\alpha}{2} + \frac{1}{2} \right) M^2 n + \frac{\alpha}{2} M^3 E_4 + \mathcal{O}(n^2),\end{aligned}$$

where

$$E_4 := \frac{1}{M} \sum_{a=0}^M \left(\frac{a}{M} \right)^2 \{a\alpha\} - \frac{1}{6}.$$

It remains to estimate Σ_{22} :

$$\begin{aligned}\Sigma_{22} &= \sum_{a=0}^M \sum_{b=[a\alpha]+1}^{n-a} \sum_{x=0}^a \{x\alpha\} \\ &= \sum_{x=0}^M \{x\alpha\} \sum_{a=x}^M \sum_{b=[a\alpha]+1}^{n-a} 1 \\ &= \sum_{x=0}^M \{x\alpha\} \sum_{a=x}^M (n - a(\alpha + 1)) + \mathcal{O}(n^2) \\ &= \sum_{x=0}^M \{x\alpha\} \left(-\frac{M^2}{2} + Mn - \frac{M^2\alpha}{2} \right) + \mathcal{O}(n^2) \\ &= E_5 \left(-\frac{M^3}{2} + M^2n - \frac{M^3\alpha}{2} \right) - \frac{M^3}{4} + \frac{M^2n}{2} - \frac{M^3\alpha}{4} + \mathcal{O}(n^2),\end{aligned}$$

Where $E_5 := \frac{1}{M} \sum_{x=0}^M \{x\alpha\} - \frac{1}{2}$.

Now we can sum up $Q = \Sigma_{11} + \Sigma_{12} + \Sigma_{21} - \Sigma_{22}$ and insert $M := \frac{n}{1+\alpha} \tau$ with some τ with $0 \leq \tau < 1$ and we obtain

$$\begin{aligned}Q &= \frac{n^4}{24} \frac{\alpha}{(1+\alpha)^2} + \frac{n^3}{12} \left(1 + \frac{\alpha}{(1+\alpha)^2} \right) - \frac{\alpha}{2} M^3 E_1 - \frac{\alpha^2}{12} M^3 E_2 - \frac{(n-M)^3}{2} E_3 \\ &\quad + \frac{\alpha}{2} M^3 E_4 + \left(M^2n - \frac{M^3}{2} - \frac{M^3\alpha}{2} \right) E_5 + \mathcal{O}(n^2),\end{aligned}$$

and from Lemma 2 and Lemma 3 we obtain the result

$$Q = \frac{n^4}{24} \frac{\alpha}{(1+\alpha)^2} + \frac{n^3}{12} \left(1 + \frac{\alpha}{(1+\alpha)^2} \right) + \mathcal{O}(n^2 KB_n(\alpha)).$$

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