Towards a General Theory of Simultaneous Diophantine Approximation of Formal Power Series:
Linear Complexity of Multisequences

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Abstract

We model the development of the linear complexity of multisequences by a stochastic infinite state machine, the Battery–Discharge–Model, BDM. The states \( s \in S \) of the BDM have asymptotic probabilities or mass \( \mu_\infty(s) = \mathcal{P}(q,M)^{-1} \cdot q^{-K(s)} \), where \( K(s) \in \mathbb{N}_0 \) is the class of the state \( s \), and \( \mathcal{P}(q,M) = \sum_{K \in \mathbb{N}_0} P_M(K)q^{-K} = \prod_{i=1}^M q^i/(q^i - 1) \) is the generating function of the number of partitions into at most \( M \) parts. We have (for each timestep modulo \( M + 1 \)) just \( P_M(K) \) states of class \( K \).

We obtain a closed formula for the asymptotic probability for the linear complexity deviation \( d(n) := L(n) - \lceil n \cdot M/(M + 1) \rceil \) with

\[
\gamma(d) = \Theta\left(q^{-|d|(M+1)}\right), \forall M \in \mathbb{N}, \forall d \in \mathbb{Z}.
\]

The precise formula is given in the text. It has been verified numerically for \( M = 1, \ldots, 8 \), and is conjectured to hold for all \( M \in \mathbb{N} \).

From the asymptotic growth (proven for all \( M \in \mathbb{N} \)), we infer the Law of the Logarithm for the linear complexity deviation,

\[
-\lim \inf_{n \to \infty} \frac{d_a(n)}{\log n} = \frac{1}{(M + 1) \log q} = \lim \sup_{n \to \infty} \frac{d_a(n)}{\log n},
\]

which immediately yields \( \frac{L_a(n)}{n} \to \frac{M}{M + 1} \) with measure one, \( \forall M \in \mathbb{N} \), a result recently shown already by Niederreiter and Wang.

**Keywords:** Linear complexity, linear complexity deviation, multisequence, Battery Discharge Model, isometry.

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1. Linear Complexity of Multisequences

The *linear complexity* of a finite string $a \in \mathbb{F}_q^n$, $L_a(n)$, is the least length of an LFSR (Linear Feedback Shift Register), which produces $a_1, \ldots, a_n$ starting with an initial content $a_1, \ldots, a_{L_a(n)}$. If all symbols are zero, we set $L_{(0,\ldots,0)}(n) = 0$. Also, we put $L_a(0) = 0$ for all $a$.

An alternative and equivalent definition defines $L_a(n)$ as the length of the shortest recurrence within the $a_i$, i.e.

$$L_a(n) := \min_{1 \leq l \leq n} \left( \exists \alpha_1, \ldots, \alpha_{l-1} \in \mathbb{F}_q, \forall 1 \leq k \leq n-l: a_{k+l} = \sum_{i=1}^{l-1} \alpha_i \cdot a_{k+i} \right).$$

Given an infinite sequence $a \in \mathbb{F}_q^\infty$, we define $L_a(n)$ as before, taking into account only the finite prefix $a_1, \ldots, a_n$. The sequence $(L_a(n))_{n \in \mathbb{N}_0}$ is called the *linear complexity profile* of $a$. The diophantine approximation of the generating function $G(a) := \sum_{n=1}^\infty a_n x^{-n} \in \mathbb{F}_q[[x^{-1}]]$ by a polynomial function with precision at least $k$ that is

$$G(a) = \frac{u(x)}{v(x)} + o(x^{-k}),$$

requires a polynomial $v(x)$ of degree at least $L_a(k)$, and this length is also sufficient, since $v(x)$ may be chosen as the feedback polynomial of the LFSR producing $a_1, \ldots, a_k$.

Turning to multisequences $(a_{n,m})_{n \in \mathbb{N}, 1 \leq m \leq M} \in \left(\mathbb{F}_q^M\right)^\infty$, we ask for simultaneously approximating all $M$ formal power series

$$\mathbb{F}_q[[x^{-1}]] \ni G_m(a) := \sum_{n=1}^\infty a_{n,m} x^{-n} = \frac{u_m(x)}{v(x)} + o(x^{-k}), 1 \leq m \leq M$$

with the same denominator polynomial $v(x)$, equivalently we search a single LFSR which produces all $M$ sequences with suitable initial contents. The linear complexity profile of $a$ now is defined by $(L_a(n,m))_{n \in \mathbb{N}_0, 1 \leq m \leq M}$ (symbol by symbol), and we set $L_a(n) := L_a(n, M)$, when considering only complete columns of all $M$ sequences at the same place $n$, with profile $(L_a(n))_{n \in \mathbb{N}_0}$.

The goal of this paper is to characterize the behaviour of $L$ as a probability distribution over all multisequences from $\left(\mathbb{F}_q^M\right)^\infty$. 

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2. Continued Fraction Expansion:
Diophantine Approximation of Multisequences

The task of determining the linear complexity profile of one multisequence from \(\left(\mathbb{F}_q^M\right)_{n=0}^{\infty}\) has been resolved by Dai and Feng \[2\]. Their mSCFA (multi-Strict Continued Fraction Algorithm) computes a sequence

\[
\left(\frac{u_k^{(n,m)}}{v_k^{(n,m)}}\right)_{k=1}^{M} \quad (n,m) \in \{1, \ldots, M\} \times \mathbb{N}_0
\]

of best simultaneous approximations to \((G_k)\). The order of timesteps is \((n, m) = (0, M), (1, 1), (1, 2), \ldots, (1, M), (2, 1), (2, 2), \ldots\) with

\[
G_m(a) = \sum_{t \in \mathbb{N}} a_{m,t} \cdot x^{-t} = \frac{u_m^{(n,m)}(x)}{v_m^{(n,m)}(x)} + o(x^{-n}), \quad n \in \mathbb{N}_0, \forall \ 1 \leq m \leq M.
\]

We will denote the degree of \(v_m^{(n,m)}(x)\) by \(\text{deg}(n,m) \in \mathbb{N}_0\), thus the linear complexity profile is \((\text{deg}(n,M))_{n \in \mathbb{N}_0} = (L_{(G_m,1 \leq m \leq M)}(n))_{n \in \mathbb{N}_0}\).

The mSCFA uses \(M\) auxiliary degrees \(w_1, \ldots, w_M \in \mathbb{N}_0\). The update of these values (and \(\text{deg}\)) depends on a so-called “discrepancy” \(\delta(n,m) \in \mathbb{F}_q\). \(\delta(n,m)\) is zero if the current approximation predicts correctly the value \(a_{n,m}\), and \(\delta(n,m)\) is nonzero otherwise.

Furthermore, the polynomials \(u_m(x)\) and \(v(x)\) are updated, crucial for the mSCFA, but of no importance for our concern, and we omit the respective part of the mSCFA in the program listing:

**Algorithm 1. mSCFA**

\[
\text{deg} := 0; w_m := 0, 1 \leq m \leq M
\]

FOR \(n := 1, 2, \ldots\)

FOR \(m := 1, \ldots, M\) 

compute \(\delta(n,m)\) //discrepancy

IF \(\delta(n,m) = 0\): \{\} // do nothing, [2] Thm. 2, Case 2a

IF \(\delta(n,m) \neq 0\) AND \(n - \deg - w_m \leq 0\): \{\} // [2] Thm. 2, Case 2c

IF \(\delta(n,m) \neq 0\) AND \(n - \deg - w_m > 0\): // [2] Thm. 2, Case 2b

\(\text{deg_copy} := \text{deg}\)

\(\text{deg} := n - w_m\)

\(w_m := n - \text{deg_copy}\)

ENDFOR

ENDFOR
3. The Battery–Discharge–Model

This section introduces the Battery–Discharge–Model (BDM), a stochastic infinite state machine or Markov chain, which will serve as a container to memorize the behaviour of deg in the mSCFA for all inputs \( a \in (\mathbb{F}_q^M)^\infty \) simultaneously.

Since the linear complexity grows approximately like \( L_a(n) = \deg(n, M) \approx \left\lfloor n \cdot \frac{M}{M+1} \right\rfloor \) (exactly, if we had always \( \delta(n, m) \neq 0 \)), and the auxiliary degrees \( w_m \) of the mSCFA grow like \( w_m \approx \left\lfloor \frac{n}{M+1} \right\rfloor \), we extract the deviation from this average behaviour as follows:

**Definition.** The *linear complexity deviation* or degree deviation is

\[
d := d_a(n) := \deg - \left\lfloor n \cdot \frac{M}{M+1} \right\rfloor \in \mathbb{Z},
\]

(1)

which we call the “drain” value, and the deviation of the auxiliary degrees is

\[
b_m := \left\lfloor n \cdot \frac{1}{M+1} \right\rfloor - w_m \in \mathbb{Z}, \quad 1 \leq m \leq M,
\]

(2)

which we call the “battery charges”.

The BDM will assemble all necessary information about the development of \( d_a(n) \) with \( n \to \infty \) and \( a \) running through all of \( (\mathbb{F}_q^M)^\infty \).

We establish the behaviour of \( d \) and \( b_m \) in two steps. First we treat the change of \( d, b_m \) when increasing \( n \) to \( n + 1 \) in the mSCFA (keeping \( \deg, w_m \) fixed for the moment):

\[
\deg - \left\lfloor (n + 1) \cdot \frac{M}{M+1} \right\rfloor = \begin{cases} 
\deg - \left\lfloor n \cdot \frac{M}{M+1} \right\rfloor - 1, & n \equiv M \mod M + 1, \\
\deg - \left\lfloor n \cdot \frac{M}{M+1} \right\rfloor, & n \neq M \mod M + 1,
\end{cases}
\]

(3)

and

\[
\left\lfloor (n + 1) \cdot \frac{1}{M+1} \right\rfloor - w_m = \begin{cases} 
\left\lfloor n \cdot \frac{1}{M+1} \right\rfloor - w_m, & n \equiv M \mod M + 1, \\
\left\lfloor n \cdot \frac{1}{M+1} \right\rfloor - w_m + 1, & n \neq M \mod M + 1.
\end{cases}
\]

(4)

Hence, by (3) we have to decrease \( d \) in all steps (we call this an action \( d_- \)), except when \( n \equiv M \to n \equiv 0 \mod(M + 1) \), and only here we increase all \( M \) battery values \( b_m \), by (4) (action \( b_+ \)).
With \( d(M, 0) = b_m(M, 0) := 0, \forall m \), we obtain the invariant
\[
d(n, m) + \left( \sum_{k=1}^{M} b_k(n, m) \right) + n \mod(M + 1) = 0, \quad \forall n \in \mathbb{N}_0, 1 \leq m \leq M \quad (5)
\]
for the initial timestep \((n, m) = (M, 0)\). Also, by (3) and (4), the actions \( d_- \) (increase \( n \), decrease \( d \)) and \( b_+ \) (decrease \( n \mod (M + 1) \) by \( M \), increase \( M \) batteries by 1 each) do not change the invariant.

Now, for \( n \) fixed, the \( M \) steps of the inner loop of the mSCF A change \( w_m \) and \( \text{deg} \) only in the case of \( \delta(n, m) \neq 0 \) and \( n - \text{deg} - w_m > 0 \) that is
\[
n - \text{deg} - w_m > 0 \iff n - \left( d + \left[ n \cdot \frac{M}{M + 1} \right] \right) - \left( \left[ n \cdot \frac{1}{M + 1} \right] - b_m \right) > 0
\]
\[\iff b_m > d. \]
In the case \( \delta \neq 0 \) and \( b_m > d \), the new values are (see mSCF A)
\[
\text{deg}^+ = n - w_m \quad \text{and} \quad w_m^+ = n - \text{deg}
\]
and thus in terms of the BDM variables:
\[
d^+ \overset{(1;6)}{=} (n - w_m) - \left[ n \cdot \frac{M}{M + 1} \right] \overset{(2)}{=} \left[ n \cdot \frac{M}{M + 1} \right] + b_m - \left[ n \cdot \frac{1}{M + 1} \right] = b_m
\]
and
\[
b_m^+ \overset{(2;6)}{=} \left[ n \cdot \frac{1}{M + 1} \right] - (n - \text{deg}) \overset{(1)}{=} \left[ n \cdot \frac{M}{M + 1} \right] + \left( d + \left[ n \cdot \frac{1}{M + 1} \right] \right) = d,
\]
an interchange of the values \( d \) and \( b_m \). We say in this case that “battery \( b_m \) discharges the excess charge into the drain”, and call this behaviour an action “D” of battery \( b_m \), corresponding to case 2b of \([2, \text{Thm. 2}]\). A discharge does not affect the invariant (5), which is thus valid for all timesteps \((n, m)\).

The remaining cases are \( b_m > d \), but \( \delta = 0 \), an inhibition of \( b_m \), action “I” (case 2a of \([2, \text{Thm. 2}]\)), and two actions of do nothing, “\( \text{N}=\)” and “\( \text{N}_<\)”, distinguishing between \( b_m = d \) and \( b_m < d \) (case 2c and part of case 2a).

Since we do not actually compute the discrepancy \( \delta \) (in fact, we do not even have a sequence \( a \)), we have to model the distinction between \( \delta = 0 \) and \( \delta \neq 0 \) probabilistically.

**Proposition 2.** In any given position \((n, m), n \in \mathbb{N}, 1 \leq m \leq M\) of the formal power series, exactly one choice for the next symbol \( a_{n,m} \) will yield a discrepancy \( \delta = 0 \), all other \( q - 1 \) symbols from \( \mathbb{F}_q \) result in some \( \delta \neq 0 \).
Proof. The current approximation \( u_{m}^{(n,m)}(x) / v_{m}^{(n,m)}(x) \) determines exactly one approximating coefficient sequence for the \( m \)-th formal power series \( G_m \). The (only) corresponding symbol belongs to \( \delta = 0 \).

In fact, for every position \((n, m)\), each discrepancy value \( \delta \in \mathbb{F}_q \) occurs exactly once for some \( a_{n,m} \in \mathbb{F}_q \), in other words (see \[10\] for \( M = 1 \)):

**Fact** The mSCFA induces an isometry on \((\mathbb{F}_q^M)^\infty\).

Hence, we can model \( \delta = 0 \) as occurring with probability \( 1/q \), and \( \delta \neq 0 \) as having probability \( (q-1)/q \).

To keep track of the variables \( d, b_m \), we define the following state set for the BDM:

**Definition.** The augmented state set is

\[
\overline{S} := \{ s = (b_1, \ldots, b_M, d; T, t) \mid b_m \in \mathbb{Z}, 1 \leq m \leq M; d \in \mathbb{Z}; T \in \mathbb{Z}; 1 \leq t \leq M + 1; d + T + \sum_{m=1}^{M} b_m = 0 \},
\]

where the last condition is the invariant (5). For the BDM, we only use the timesteps \( 0 \leq T \leq M \), and the BDM thus has the state set

\[
S := \{ s \in \overline{S} \mid 0 \leq T \leq M \}
\]

with initial state \( s_0 := (0, \ldots, 0; 0, M + 1) \).

To facilitate notation, we also define \( S(T_0, t_0, d_0) = \{ s \in S \mid T(s) = T_0, t(s) = t_0, d(s) = d_0 \} \), and similarly \( S(T_0), S(T_0), \overline{S}(T_0, t_0, d_0) \).

A state stores the values of the batteries and the drain in \( b_1, \ldots, b_M, d \), the value \( T \) corresponds to the time modulo \( M + 1 \) that is \( T \equiv n \mod M + 1 \), and the “ministeps” \( t = 1, \ldots, M \) correspond to the update of battery \( b_m \) between \( t = m \) and \( t = m + 1 \), while \( t = M + 1 \rightarrow 1 \) corresponds to the updates \( d_{-}, b_{+} \).

The allowed transitions \( \alpha \) (action) from a state \( s = (b_1, \ldots, b_M, d; T, t) \) are \( \alpha = d_{-} \) or \( b_{+} \) for \( t = M + 1 \), and otherwise depend on the relative size of \( b_t \) and \( d \) (\( \alpha \in \{ D, I, N_{=}, N_{<} \} \)). We have \( s \xrightarrow{\alpha} s^{+} \) with the following actions, conditions, nextstates \( s^{+} \), and probabilities:
\[ \alpha \text{ condition} \quad s^+ \quad \text{prob.} \\
D \quad b_t > d, t \leq M \quad (b_1, \ldots, b_{t-1}, d, b_{t+1}, \ldots, b_M, b_t; T, t + 1) \quad q - 1/q \\
I \quad b_t > d, t \leq M \quad (b_1, \ldots, b_M, d; T, t + 1) \quad 1/q \\
N_\geq \quad b_t = d, t \leq M \quad (b_1, \ldots, b_M, d; T, t + 1) \quad 1 \\
N_\leq \quad b_t < d, t \leq M \quad (b_1, \ldots, b_M, d; T, t + 1) \quad 1 \\
d_- \quad T < M, t = M + 1 \quad (b_1, \ldots, b_M, d - 1; T + 1, 1) \quad 1 \\
b_+ \quad T = M, t = M + 1 \quad (b_1 + 1, b_2 + 1, \ldots, b_M + 1, d; 0, 1) \quad 1 \\
\]

Whenever \( b_t > d \), both \( D \) and \( I \) may occur, leading to two feasible transitions from a given state \( s \), whose probabilities sum up to 1.

Recall that from \((T, M + 1)\) to \((T + 1, 1)\), the drain \( d \) is decremented according to (3) for \( T < M \), action “\( d_\_ \)”, and from \((M, M + 1)\) to \((0, 1)\), the batteries \( b_m \) are incremented according to (4), action “\( b_+ \)”. 

**Definition.** The state transition matrix \( T \) of the BDM is an infinite stochastic matrix indexed by \( s, s' \in S \), and where 

\[
T(s, s') = \begin{cases} 
1, & \text{if } s \xrightarrow{N_\leq} s', s \xrightarrow{N_\geq} s', s \xrightarrow{d_-} s', \text{ or } s \xrightarrow{b_+} s', \\
(q - 1)/q, & s \xrightarrow{I} s', \\
1/q, & s \xrightarrow{D} s', \\
0, & \text{otherwise.} 
\end{cases}
\]

Every row either includes an “I” and a “D”, or else one of “\( N_\geq \)”, “\( N_\leq \)”, “\( d_\_ \)”, or “\( b_+ \)”. Reading the feasible transitions backwards, one obtains that a state with \( b_t < d \) (at \((T, t + 1)\)) is reached either by a discharge, or by a “\( N_\leq \)”, hence the corresponding column of \( s' \) sums up to \( \frac{q-1}{q} + 1 \). Likewise, if \( s' \) has \( b_t > d \), this may only be the result of an inhibition, hence column sum \( 1/q \). The cases “\( N_\leq \)”, “\( d_\_ \)”, and “\( b_+ \)” all are by themselves the only nonzero entry within a column, which has thus sum 1.

In terms of \( d, b_m \), we have the following equivalent probabilistic formulation of the mSCFA (timestep \( t = M + 1 \) comes after the \texttt{FOR} \( m \equiv t \) loop):

**Algorithm 3. BatteryDischargeModel**

\[
d := 0; b_m := 0, 1 \leq m \leq M \\
d := d - 1 \quad // \quad d_- \\
\text{FOR } n := 1, 2, \ldots \\
\text{FOR } m := 1, \ldots, M \\
\quad \text{IF } b_m > d: \\
\qquad \text{WITH prob. } (q - 1)/q: \\
\qquad \quad \text{swap}(b_m, d) \quad // \quad \text{action } D 
\]
WITH prob. \(1/q\):
\[
\{\} \quad \text{// action } I
\]
ELSE
\[
\{\} \quad \text{// action } N_+, N_<
\]
ENDIF
ENDFOR
IF \(n \neq M \mod M + 1\):
\(d := d - 1 \quad \text{// } d_-
\)
ELSE
\(b_m := b_m + 1, 1 \leq m \leq M \quad \text{// } b_+
\)
ENDIF
ENDFOR

4. Classes of BDM States

The Markov chain BDM will turn out to be strongly concentrated on few states. We define a family of measures \(\mu_{\tau}\) on \(S\), indexed by \(\tau \in \mathbb{N}_0\). We start for \(\tau = 0\) with all mass concentrated on the initial state \(s_0\):

\[
\mu_0(s) = \begin{cases} 
1, & s = s_0 = (0, \ldots, 0; 0, M + 1) \\
0, & s \neq s_0 
\end{cases}
\]

For successive timesteps \(\tau\), we then put \(\mu_{\tau+1}(s') = \mu_{\tau}(s)\), if \(s \xrightarrow{N_}= s'\), \(s \xrightarrow{d_-=} s'\), or \(s \xrightarrow{b_+} s'\). Also, \(\mu_{\tau+1}(s') = \frac{1}{q}\mu_{\tau}(s)\), if \(s \xrightarrow{I} s'\). Finally \(\mu_{\tau+1}(s') = \frac{q-1}{q}\mu_{\tau}(s_1) + \mu_{\tau}(s_2)\), if \(s_1 \xrightarrow{D} s'\) and \(s_2 \xrightarrow{N_<} s'\).

Put in other words, \((\mu_{\tau+1}(s))_{s \in S} = T \cdot (\mu_{\tau}(s))_{s \in S}\).

Be aware that from \(\mu_{\tau}\) to \(\mu_{\tau+1}\), we only deal with one input symbol (or \(d_-, b_+\)), hence the distribution after reading all \(M\) inputs of column \(n\) is in fact \(\mu_{(M+1)n}(s)\).

**Definition.** We will use repeatedly the “timesteps” \((T, t) \in \{0, \ldots, M\} \times \{1, \ldots, M + 1\}\) of the BDM, comparing them with linear time \(\tau \in \mathbb{N}_0\). We define:

\[
(T, t) \equiv \tau \iff (T - 1) \cdot (M + 1) + t \equiv \tau \mod (M + 1)^2
\]

When dealing with the \(m\)-th symbol in column \(n\), the \(\tau\)-th input symbol, we are in a state with \(T(s) \equiv n \mod (M + 1), t = m\), and \((T, t) \equiv \tau\).
Proposition 4. For every $\tau \in \mathbb{N}_0$,

$$\sum_{s \in S(T_0,t_0)} \mu_\tau(s) = \begin{cases} 1, & (T_0,t_0) \equiv \tau \\ 0, & (T_0,t_0) \not\equiv \tau \end{cases}$$

Proof. By induction on $(T,t)$: Initially $(T = 0, t = M + 1)$, all mass is on $s_0$. Also, every transition goes from states with $(T,t) \equiv \tau$ to states with $(T',t') \equiv \tau + 1$, carrying over the mass to the new $S(T',t')$. \qed

Definition. Denote the number of sequence prefixes in $(\mathbb{F}_q^M)^n$ with linear complexity deviation $d \in \mathbb{Z}$ as $N(n,d;q)$.

Since the BDM has been derived from the behaviour of the mSCFA, we obtain

Theorem 5. Assume that exactly $N$ of the $q^{M \cdot n}$ sequence prefixes of length $n$ lead to a certain configuration $(\deg, w_1, \ldots, w_M)$ of the mSCFA, and let $b_1, \ldots, b_M, d, T$ be derived from $n, \deg, w_m$ according to (1), (2), then

$$\mu_{(M+1)\cdot n}(b_1, \ldots, b_M, d; T, M+1) = \frac{N}{q^{M\cdot n}}$$

with $(T, M+1) \equiv (M+1) \cdot n$ that is $T \equiv n \mod (M+1)$.

Proof. The theorem is true for $n = 0$, $(T,t) = (0, M+1)$ with $N = 1$, $\deg = d = b_m = w_m = 0, \forall m$, starting with the (only) prefix $\varepsilon$, the empty string.

From then on, by the construction of the BDM, for $t(s) \leq M$ a transition $s \to s^*$ takes place with probability $\frac{b}{q}$, with $b$ from $\{1, q-1, q\}$, if and only if the mSCFA goes to the state corresponding to $s^*$ for $b$ out of the $q$ possible next symbols $a_{n,m} \in \mathbb{F}_q$, or, for $t(s) = M + 1$, $t(s^*) = 1$, corresponding to actions $d_-, b_+$, with probability one, while the mSCFA increases $n$. \qed

From this theorem now follows as a corollary the description of $N(n,d; q)$ by the mass distribution on the BDM states (more on finite $n$ in Section 9):

Theorem 6. For $n \in \mathbb{N}_0$ with $(T_0, M+1) \equiv (M+1) \cdot n$, for $d \in \mathbb{Z}$,

$$N(n,d; q) = q^{M\cdot n} \times \sum_{s \in S(T_0,M+1,d)} \mu_{(M+1)\cdot n}(s)$$

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**Definition.** For a given state $s \in S$, we define its asymptotic measure as
\[
\mu_\infty(s) := \limsup_{n \to \infty} \mu_n(s) = \lim_{n \to \infty} \mu_n(s).
\]
We have $\sum_{s \in S} \mu_\infty(s) = (M + 1)^2$, since each $S(T, t)$ sums up to 1.

We will see that all states satisfy $\mu_\infty(s) = \mu_\infty(s_0 \cdot q^{-K(s)})$ for some $K(s) \in \mathbb{N}_0$. We call this value $K(s)$ the class of state $s$ and define it algorithmically, generalizing to $s \in \overline{S}$:

**Definition.** The class of a state $s = (b_1, \ldots, b_M, d; T, t) \in \overline{S}$ is
\[
K(s) = -\pi_s + M \cdot T + 2 \cdot \sum_{m=1}^{M+1} \tilde{b}_m \cdot (M + 1 - m),
\]
where $\pi_s$ is minimum number of transpositions between neighbours necessary to sort $(b_1, \ldots, b_{t-1}, d, b_t, \ldots, b_M)$ into decreasing order as $(\tilde{b}_1, \ldots, \tilde{b}_{M+1})$, $\tilde{b}_i \geq \tilde{b}_{i+1}, 1 \leq i \leq M)$. Observe that the place of $d$ in the initial sequence depends on $t$.

**Example.** The state $s = (-5, 4, -4, 2; 1, 2)$ with $M = 3, d = 2, T = 1$ and $t = 2$ requires the sorting of $(-5, 2, 4, -4)$ into $(4, 2, -4, -5)$, using $\pi_s = 4$ transpositions, and thus $K(s) = -4+3 \cdot 1+2(4 \cdot 3+2 \cdot 2+-4 \cdot 1+-5 \cdot 0) = 23$.

This static way of determining $K(s)$ is compatible with the following dynamic consideration of transitions. First we need a technical lemma:

**Lemma 7.** For all $s = (b_1, \ldots, b_M, d; T, t) \in \overline{S}$, we have
\[
K(s) = K(b_1, \ldots, b_M, d; T, t) = K(b_1 + 1, \ldots, b_{m+1}, 1, d + 1; T - M - 1, t).
\]

**Proof.** Let $s = (b_1, \ldots, b_M, d; T, t)$ and $s' = (b'_1, \ldots, b'_{m+1}, d'; T', t) := (b_1 + 1, \ldots, b_{m+1}, 1, d + 1; T - M - 1, t)$. Then $\pi_s = \pi'_{s'}$, since the relative order within $s$ and $s'$ are the same, also $\tilde{b}_i' = \tilde{b}_i + 1, 1 \leq i \leq M + 1$. Using
\[
2 \cdot \sum_{m=1}^{M+1} 1 \cdot (M + 1 - m) = 2 \left( \frac{M+1}{2} \right) = M \cdot (M + 1),
\]
we have...
we thus have

\[
K(s) = -\pi_s + M \cdot T + 2 \cdot \sum_{m=1}^{M+1} \tilde{b}_m \cdot (M + 1 - m)
\]

\[
= -\pi'_s + MT - M(M + 1) + 2 \cdot \sum_{m=1}^{M+1} (\tilde{b}_m + 1) \cdot (M + 1 - m)
\]

\[
= -\pi'_s + M \cdot T' + 2 \cdot \sum_{m=1}^{M+1} \tilde{b}_m' \cdot (M + 1 - m) = K(s').
\]

We now obtain the change in class by counting actions \(I\) and \(N_\prec\):

**Theorem 8.**

(i) For every feasible transition \(s \xrightarrow{\alpha} s'\) between states \(s, s' \in S\) with \(\alpha \in \{D, I, N_\prec, N_\succ, d_-, b_+\}\), we have

\[
K(s') = K(s) + \begin{cases} 
1, & \alpha = I \\
-1, & \alpha = N_\prec \\
0, & \alpha \in \{D, N_\prec, d_-, b_+\}
\end{cases}
\]

(ii) Let \(s_0 \xrightarrow{\alpha_1 \ldots \alpha_k} s\) be some path from the initial state \(s_0\) to \(s\).

Let \(\#I = \#\{1 \leq i \leq k \mid \alpha_i = I\}\) and \(\#N_\prec = \#\{1 \leq i \leq k \mid \alpha_i = N_\prec\}\). Then

\[
K(s') - K(s) = \#I - \#N_\prec.
\]

**Proof.** (i) We first deal with transitions from \(D, I, N_\prec, N_\succ\) that is \(t \neq M + 1\). Since the values of the multiset \(\{b_1, \ldots, b_M, d\}\) only get swapped (in the case of a discharge \(D\)), the sum \(2 \sum_{m=1}^{M+1} \tilde{b}_m \cdot (M + 1 - m)\) as well as the term \(M \cdot T\) stay the same.

It suffices thus to compare \(\pi_s\) with \(\pi_{s'}\). Let \((b_1, \ldots, b_{t-1}, d, b_t, \ldots, b_M)\) be the sequence to be ordered for \(s\) and similarly for \(s'\), after the discharge, \((b'_1, \ldots, b'_t, d', b'_{t+1}, \ldots, b'_M)\). Since \(b_m = b'_m\) for \(m \neq t\), they need the same number of transpositions, and we may in fact restrict our comparison to the sorting of \((d, b_t)\) for \(s\) and \((b'_t, d')\) for \(s'\).

Case \(\alpha = D\): We had \(b_t > d\) (to be able to apply \(D\)), \(d' := b_t, b'_t := d\) and thus \(d' > b'_t\). Both before and after the discharge, one transposition is necessary and thus \(\pi_s = \pi_{s'}, K(s) = K(s')\).

Case \(\alpha = I\): Again \(b_t > d\), \(b'_t := b_t, d' := d\). We sort \((d, b_t)\) with one transposition, \((b'_t, d')\) is already sorted. Hence \(\pi_{s'} = \pi_s - 1\), \(K(s') = K(s) + 1\).

Case \(\alpha = N_\prec\): Now, \(b_t = d = b'_t = d'\) and so \(\pi_{s'} = \pi_s\), \(K(s') = K(s)\).
Case $\alpha = N_\prec$: Here $b_i < d$ and $b'_i < d'$. $(d, b_i)$ is already in order, $(b'_i, d')$ requires a transposition and so $\pi_{s'} = \pi_s + 1$, $K(s') = K(s) - 1$.

Finally, for $d_-$ we have $s' = (b_1, \ldots, b_M, d - 1; T + 1, 1)$ from $t = M + 1$ for $s$. We have to sort $(b_1, \ldots, b_M, d)$ for $s$ (with $t = M + 1$) and $(d - 1, b_1, \ldots, b_M)$ for $s'$ (with $t = 1$). Sorting first the part $b_1, \ldots, b_M$ with equal effort in both $s$ and $s'$, $\pi_s = \pi_{s'}$, we then introduce $d$ from the right, respectively $d - 1$ from the left to the same place: $(b_1, \ldots, b_{k-1}, d)$ or $(d - 1, b_{k+1}, \ldots, b_M)$, where $b_{k-1} \geq d > d - 1$ and $b_{k+1} \leq d - 1 < d$. The total number of transpositions is then $\pi_s = \pi_s + (M + 1 - k)$ and $\pi_{s'} = \pi_{s'} + (k - 1)$.

The class now is

$$K(s) = -\pi(s) + M \cdot T + 2 \sum_{m=1, m \neq k}^{M+1} b_m(M + 1 - m) + 2d(M + 1 - k)$$

$$\quad = -\pi(s) + (M - (k - 1)) - (k - 1) + M \cdot (T + 1)$$

$$\quad + 2 \sum_{m=1, m \neq k}^{M+1} b_m(M + 1 - m) + 2(d - 1)(M - (k - 1)) = K(s').$$

The case $\alpha = b_+$ is equivalent to $d_-$ followed by incrementing all the $b_m$ and $d$, hence follows from the case $\alpha = d_-$ and Lemma 7.

$(ii)$ This follows by applying $(i)$ to the $k$ transitions leading to $s$, starting in $s_0$ with $K(s_0) = 0$. \hfill $\square$

We will now show that the limit mass distribution $\mu_\infty$ follows in fact (up to a constant) from the state classes as $\mu_\infty(s) = C_0 \cdot q^{-K(s)}$. First, we state a theorem by Rosenblatt (an infinite matrix version of Perron–Frobenius):

**Theorem 9.** (Rosenblatt, \cite{Rosenblatt}) Let $T$ be a Markov chain, finite or infinite. “If the chain is irreducible and nonperiodic, there is an invariant instantaneous distribution if and only if the states are persistent, in which case the distribution is unique and given by $\{u_k\}$” \cite[p. 56]{Rosenblatt}, where $u_j = \lim_{n \to \infty} p_{jj}^{(n)}$, and $p_{jj}^{(n)}$ is the probability to return to state $j$ after $n$ steps.

**Proof.** See \cite[p. 56]{Rosenblatt}. \hfill $\square$

Here $T$ certainly is periodic, with period $(M + 1)^2$. The $(M + 1)^2$-th power of $T$ has the property that transitions occur only within the sets $S(T, t)$, so it can be ordered into a block diagonal matrix. We use only the block with $(T, t) = (0, M + 1)$, including $s_0$, as $\hat{T} := T^{(M + 1)^2}_{s \in S(0, M + 1)}$.

$T$ and thus $\hat{T}$ is irreducible, since we get from $s_0$ to every state and back by the following theorem:
Theorem 10. (i) For every state $s \in S$, there is exactly one sequence of transitions $\alpha = \alpha_1 \cdots \alpha_k$ with $s_0 \xrightarrow{\alpha(s)} s$ and $\alpha(s) \in \{D, I, N_\prec, d_-, b_+\}^*$ (avoiding actions of the type $N_\prec$), which touches the state $s_0$ only initially.

(ii) Also, there is exactly one path from $s$ to $s_0$ avoiding actions of type $I$, which touches the state $s_0$ only finally.

Proof. (i) Unicity: There is at most one such transition: When going backwards from $s$ to $s_0$, running through the batteries in reverse order $M, M-1, \ldots, 1$ for each transition, we have:

For $b_m < d$ this results either from a discharge $D$, or a do nothing $N_\prec$.

For $b_m = d$ this results from a do nothing $N_\prec$.

For $b_m > d$, only an inhibition $I$ is possible.

Existence: There is an infinite chain of predecessors, all of class less than or equal to $K(s)$. Since for each $K$, there are only finitely many states with this class, in particular, there is some state $s^*$ with $(T, t) = (0, M+1)$, which is reached repeatedly. If this state is $s^* = s_0$, we are done. If not, $s^* = (b_1, \ldots, b_M, d; 0, M+1)$ has $mx := \max\{b_1, \ldots, b_M, d\} \geq 1$ and $mn := \min\{b_1, \ldots, b_M, d\} \leq -1$ (by the invariant (5), with $T$ being zero).

However, a cycle $s^* \xrightarrow{(D, N_\prec, d_-, b_+)^+} s^*$ without $I$ or $N_\prec$ is impossible: Either $d = mx$ at $(T, t) = (0, M+1)$, or else some battery $b_t = mx$ has to discharge ($I$ prohibited). At $(T, t) = (1, M+1)$, we have $d = mx \geq 1$ in any case, thus at $(T, t) = (2, 1)$, we get $d \geq 0$. Now, since $mn \leq -1$ is the value of one of the batteries, say $b_t^*$, at time $(2, t^*)$ we have $b_t^* < d$ and thus $N_\prec$ is the only possible action. So, no return to $s^*$ avoiding $I$ and $N_\prec$ (having reached $K = 0$, there is no further decrement) is possible, unless $s^* = s_0$. Since the only cycle to avoid passes repeatedly through $s_0$, $\alpha(s)$ is well-defined by excluding this case.

(ii) To get back, just choose $D$, whenever $b_t > d$. In this way, the class can never increase, and thus eventually, we must hit a cycle. But we have already seen that the only cycle avoiding both $I$ and $N_\prec$ passes through the states with class 0, including $s_0$. $\square$

Theorem 11. For any two states $s, s' \in S$,

$$\frac{\mu_\infty(s)}{\mu_\infty(s')} = q^{K(s')-K(s)}.$$

Proof. Let a mass distribution $\mu(s) := q^{-K(s)}$ be given. We show that $\mu$ is invariant under the transition matrix of the BDM, i.e. $(\mu(s))_{s \in S}$ is an
eigenvector of eigenvalue one, and unique with this property up to a constant factor. We consider all states leading to a fixed state \( s \). We have three cases:

1. \( b_t < d \) after the action, coming from \( s_1 \xrightarrow{D} s \) or \( s_2 \xrightarrow{N_s} s \), and thus \( \mu_\infty(s) = \frac{q-1}{q} \mu_\infty(s_1) + \mu_\infty(s_2) \). Since \( N_s \) decrements the class (but \( D \) not), we have
\[
\frac{q-1}{q} p_0 \cdot q^{-K(s_1)} + p_0 \cdot q^{-K(s_2)} = \frac{q-1}{q} p_0 \cdot q^{-K(s)} + p_0 \cdot q^{-K(s+1)} = \left( \frac{q-1}{q} + \frac{1}{q} \right) \cdot p_0 \cdot q^{-K(s)} = p_0 \cdot q^{-K(s)}.
\]

2. \( b_t = d \) after the action, which must be a do nothing, \( \alpha = N_s \), and thus \( K(s) = K(s') \), \( \mu_\infty(s) = \mu_\infty(s') \).

3. \( b_t > d \) afterwards (and before), from an inhibition, \( \alpha = I \) which increments the class, hence \( q^{-K(s)} \cdot \frac{1}{q} = q^{-K(s+1)} \).

This shows consistency of \( \mu(s) = c \cdot q^{-K(s)} \) with the behaviour of the BDM, or stated otherwise: \((\mu(s))_{s \in S} = (q^{-K(s)})_{s \in S} \) is an eigenvector of the infinite state transition matrix of the BDM. Furthermore its eigenvalue 1 is the largest eigenvalue of \( T \), since \( T \) is stochastic.

Now, \( T \) inherits the eigenvector \( \mu \), restricted to states from \( S(0, M + 1) \), with eigenvalue \( 1^{(M+1)^2} = 1 \). This matrix is aperiodic and irreducible by Theorem 10, and by Theorem 9 (Rosenblatt), \( \mu \) is already the only such eigenvector up to a constant factor, and it remains to normalize it.

Returning from \( T \) to \( T \), we obtain the statement, since \( \mu(s) = q^{-K(s)} \) for all \( s \in S(0, M + 1) \) forces all other states in \( S \) also into this eigenvector. \( \square \)

5. Antisymmetry

In this section, we consider only the configurations with \( t = M + 1 \), at the end of a complete column from the input \( a \).

**Proposition 12.** For \( M \in \mathbb{N} \), \( T \in \mathbb{Z} \), \( d \in \mathbb{Z} \), \( k \in \mathbb{N}_0 \), and \( 2 \leq q \in \mathbb{N} \), let
\[
A = \{ s \in \overline{S}(T, M + 1, d) \mid K(s) = k \} \quad \text{and} \quad \overline{A} = \{ s \in \overline{S}(M - T, M + 1, -d) \mid K(s) = k \}.
\]

Then \( |A| = |\overline{A}| \).

**Proof.** We show that states \( \tilde{s} := (b_1, \ldots, b_M, -T - X; T, M + 1) \) and \( \tilde{s} := (-b_M - 1, \ldots, -b_1 - 1, T + X; M - T, M + 1) \), where \( X := \sum_{m=1}^M b_m \), satisfy:

1. \( T(s) = T \) and \( T(\tilde{s}) = M - T \),
2. \( d(s) = d(\tilde{s}) \), and
3. \( K(s) = K(\tilde{s}) \).

Then \( s \in A \iff \tilde{s} \in \overline{A} \), and we have a bijection between \( A \) and \( \overline{A} \), hence \( |A| = |\overline{A}| \).

(i) and (ii) are obvious by inspection. To show (iii), we first sort \( b_1, \ldots, b_M \) of \( s \) into decreasing order as \( \tilde{b}_1 \geq \tilde{b}_2 \geq \cdots \geq \tilde{b}_M \) by \( \pi'_s \) permutations of neigh-
bours. Then \(-b_M - 1, \ldots, -b_1 - 1\) of \(\overline{s}\) can be sorted into decreasing order by \(\pi'_s\) permutations at the same places into \(-b_M - 1, \ldots, -b_1 - 1\).

We now introduce \(d = -T - X\) and \(-d = T + X\), resp., into the ordered \(\tilde{b}\)’s as \(\tilde{b}_1 \geq \ldots \tilde{b}_k \geq -T - X > \tilde{b}_{k+1} \geq \ldots \tilde{b}_M\) and \(-\tilde{b}_M - 1 \geq \ldots -\tilde{b}_{k+1} - 1 \geq T + X > -\tilde{b}_k - 1 \geq \ldots \tilde{b}_1\) (observe the \(>\) inequality in both cases to the right of \(\pm(T + X)\)). We have a total of \(\pi_s = \pi'_s + M - k\) and \(\pi_{\overline{s}} = \pi'_s + k\), resp., permutations, thus \(K(s) = -\pi_s - (M - k) + M \cdot T + 2 \sum_{i=1}^{M} (M + 1 - i)\tilde{b}_i - 2 \sum_{i=k+1}^{M} \tilde{b}_i + 2(M - k)(-T - X)\) and \(K(\overline{s}) = -\pi_{\overline{s}} - k + M \cdot (T - T) + 2 \sum_{i=1}^{M} (M + 1 - i)\tilde{b}_{M+1-i} - 1 - 2 \sum_{i=1}^{k} (-\tilde{b}_i - 1) + 2k(T + X)\), where the first sum treats the \(\tilde{b}\)’s in their place before introducing \(\pm(T + X)\), the second sum adjusts the \(\tilde{b}\)’s, which are shifted while introducing \(d\), by 2, and the last term belongs to the drain \(\pm(T + X)\). The difference is then

\[
K(\overline{s}) - K(s) = M^2 - 2MT + M - 2k + 2 \sum_{i=1}^{M} (-i - M - 1 + i)\tilde{b}_i - \sum_{i=1}^{M} 1 + 2 \sum_{i=1}^{M} \tilde{b}_i + 2 \sum_{i=1}^{k} 1 + 2(k + M - k)(X + T)
\]

\[
= M(M + 1) - 2MT - 2k - 2(M + 1) \sum_{i=1}^{M} \tilde{b}_i - M(M + 1) + 2 \sum_{i=1}^{M} \tilde{b}_i + 2k + 2MX + 2MT = -2(M + 1)X + 2X + 2MX = 0,
\]

and we obtain (iii). \(\square\)

**Theorem 13.** (Antisymmetry)

For all \(M \in \mathbb{N}\), \(T \in \mathbb{Z}\), and \(d \in \mathbb{Z}\),

\[
\sum_{s \in \mathcal{S}(T, M+1, d)} q^{-K(s)} = \sum_{s' \in \mathcal{S}(M-T, M+1, -d)} q^{-K(s')}.
\]

**Proof.** As in the proof of the preceding proposition, we can match the states in the first sum with those in the second one. From property (iii) in 12, we conclude that the classes, and thus the sum terms, are the same in each case. \(\square\)
**Definition.** For all $d \in \mathbb{Z}$, $q = |\mathbb{F}_q|$, $M \in \mathbb{N}$, and $0 \leq T \leq M$, let

$$\gamma(d, T, M + 1) := \sum_{s \in S(T, M + 1, d)} \mu_{\infty}(s)$$

be the asymptotic mass on all states with a given drain value $d$, at times $\equiv (T, M + 1)$.

**Theorem 14.** For all $d \in \mathbb{Z}$, $M \in \mathbb{N}$, and $0 \leq T \leq M$, we have

$$\gamma(d, T, M + 1) = \gamma(-d, M - T, M + 1).$$

**Proof.** This follows immediately from Proposition 12 and Theorem 11. \qed

**Definition.** For $0 \leq T \leq M$ and $1 \leq t \leq M + 1$, let

$$\overline{d}(T, t) := \sum_{s \in S(T, t)} \mu_{\infty}(s) \cdot d(s).$$

Also, let

$$\overline{\overline{d}} := \frac{1}{M + 1} \sum_{T=0}^{M} \overline{d}(T, M + 1).$$

**Proposition 15.**

(i) For $0 \leq T \leq M$, $\overline{d}(T, M + 1) = -\overline{d}(M - T, M + 1)$.

(ii) For even $M$, we have $\overline{d}(M/2, M + 1) = 0$.

**Proof.** (i) follows from Theorem 14, since $\overline{d}(T, M + 1) = \sum_{d \in \mathbb{Z}} d \cdot \gamma(d, T, M + 1)$.

(ii) follows from (i) with $\overline{d}(M/2, M + 1) = -\overline{d}(M/2, M + 1) \Rightarrow \overline{d}(M/2, M + 1) = 0$. \qed

**Theorem 16.** For every $M \in \mathbb{N}$, $\overline{\overline{d}} = 0$.

**Proof.** Using 15(i) and 15(ii) in case of even $M$, we have

$$\overline{\overline{d}} = \frac{1}{M + 1} \sum_{T=0}^{M} \overline{d}(T, M + 1) = \frac{1}{M + 1} \left( \sum_{T=0}^{\lfloor M/2 \rfloor} \overline{d}(T, M + 1) + \overline{d}(M - T, M + 1) \right)$$

$$= 0. \quad \Box$$

**Remark.** In particular, Theorem 16 is an (aesthetical) reason to choose $L_\alpha(n) \approx \lfloor n \cdot \frac{M}{M + 1} \rfloor$ (and not $L_\alpha(n) \approx n \cdot \frac{M}{M + 1}$) as “typical” average behaviour, another reason is that for $q \to \infty$ this same $\lceil \ldots \rceil$ value is the limit behaviour.
6. The Partition Model

**Definition.** Let \( P_M(K) \in \mathbb{N} \), for \( M \in \mathbb{N}, K \in \mathbb{N}_0 \), be the number of partitions of \( K \) into at most \( M \) parts (equivalently, into parts of size at most \( M \)).

**Definition.** Let \( P(M, q) = \sum_{K=0}^{\infty} P_M(K) \cdot q^{-K} \).

**Proposition 17.**  
1. The following initial values and recursion formulae hold:
   - \( P_M(0) = 1 \), \( \forall M \in \mathbb{N} \), \( P_M(1) = 1 \), \( \forall K \leq 0, \forall M \in \mathbb{N} \), and
   - \( P_M(K) = P_M(K - M) + P_{M-1}(K) \).

2. The generating function of \( P_M(K) \) in powers of \( q^{-1} \) is
   \[
   P(M, q) = \sum_{K=0}^{\infty} P_M(K) \cdot q^{-K} = \prod_{m=1}^{M} \frac{q^m}{q^m - 1}.
   \]

3. \( P_M(K) \approx \frac{K^{M-1}}{M!(M-1)!} \) for fixed \( M \) and \( K \to \infty \).

**Proof.** See [12], Sections 2.5.10, 2.5.12 and 2.5.11. \( \square \)

**Remark.** Observe that by c), for every \( K \in \mathbb{N}_0 \), we have only polynomially many states of class \( K \), each with exponentially small probability \( q^{-K} \cdot \mu(s_0) \). This leads to the concentration of mass on the states with small \( K \).

**Definition.** Let \( I_m(s), 1 \leq m \leq M \), count the number of actions \( I \) at battery \( m \) during \( \alpha(s) \) (see Theorem 10). If \( K(s) = 0 \), put \( I_m(s) = 0, 1 \leq m \leq M \).

Let \( (\tilde{I}_1, \ldots, \tilde{I}_M) \) be the ordered \( (\tilde{I}_i \geq \tilde{I}_{i+1}, 1 \leq i < M) \) version of \( \{I_m\} \).

**Corollary 18.** Let \( \#I \) be the number of inhibitions during all of the transitions in \( \alpha(s) \), similarly \( \#N_\prec \). Then \( \sum_{m=1}^{M} I_m = \#I = K(s) \).

**Proof.** The \( I_m \) sum up to \( \#I \) by definition. By Theorem 8, we have \( K(s) - K(s_0) = \#I - \#N_\prec \). With \( \#N_\prec = 0 \) and \( K(s_0) = 0 \), \( \#I = K(s) \) follows. \( \square \)

**Theorem 19.**  
1. For \( 1 \leq M \leq 8, 0 \leq T \leq M, 1 \leq t \leq M + 1 \), and \( 0 \leq K \leq 1200 - 100M \), the state set \( S(T, t) \) contains exactly \( P_M(K) \) states with \( K(s) = K \).

2. For \( 1 \leq M \leq 8 \) and \( 0 \leq K \leq 600 - 50M \), fix a time \( (T_0, t_0) \). Then the \( \tilde{I}(1, \ldots, \tilde{I}_M) \) of all the \( P_M(K) \) states in \( S(T_0, t_0) \) with \( K(s) = K \) give the \( P_M(K) \) different partitions of \( K \) into \( M \) parts (including those of size 0).

**Proof.** By numerical simulation over the mentioned ranges. \( \square \)

**Conjecture 20.** The previous theorem holds for every \( M, T, t, K \).
A graph for $M = 2$, showing states $(b_1, b_2, d)$ with $K(s) \leq 4$ and their associated partitions:

| Class $K(s)$ | 0   | 1   | 2   | 2   | 3   | 3   | 4   | 4   | 4   | >4   |
|--------------|-----|-----|-----|-----|-----|-----|-----|-----|-----|------|
| $T_1$ Part.  | [0,0]| [1,0]| [2,0]| [1,1]| [3,0]| [2,1]| [4,0]| [3,1]| [2,2]| [*,*]|

| 2,3          | $(-1,1,0)$ | $(-1,0,-1)$ | $(0,-1,-1)$ | $(-2,0,0)$ | $(-2,-1,1)$ | $(0,-2,0)$ | $(-2,1,-1)$ | $(-1,-2,1)$ | $(0,0,-2)$ | $(-,)$ | 2,3 |
|--------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|------|
| 0,1          | $(0,0,0)$   | $(0,-1,1)$  | $(1,0,-1)$  | $(-1,1,0)$  | $(-1,0,1)$  | $(1,-1,0)$  | $(-1,2,-1)$ | $(0,-1,1)$  | $(1,-2,1)$  | $(-,)$ | 0,1 |
| 0,2          | $(0,0,0)$   | $(-1,0,1)$  | $(-1,0,1)$  | $(0,1,1)$   | $(1,0,1)$   | $(-1,1,1)$  | $(-2,1,1)$  | $(1,-1,1)$  | $(0,2,1)$   | $(-,)$ | 0,2 |
| 0,3          | $(0,0,0)$   | $(-1,0,1)$  | $(-1,1,0)$  | $(0,-1,1)$  | $(1,-1,1)$  | $(0,1,-1)$  | $(-1,-1,2)$ | $(1,0,-1)$  | $(-2,1,1)$  | $(-,)$ | 0,3 |
| 1,1          | $(0,0,-1)$  | $(-1,0,0)$  | $(-1,1,1)$  | $(0,-1,0)$  | $(1,1,1)$   | $(0,1,-2)$  | $(-1,-1,1)$ | $(1,0,-2)$  | $(-2,1,0)$  | $(-,)$ | 1,1 |
| 1,2          | $(-1,0,0)$  | $(0,-1,1)$  | $(-1,1,1)$  | $(0,-1,0)$  | $(-1,1,1)$  | $(-2,0,1)$  | $(1,-1,1)$  | $(0,-2,1)$  | $(-,)$ | 1,2 |
| 1,3          | $(-1,0,0)$  | $(0,-1,0)$  | $(-1,1,1)$  | $(0,-1,1)$  | $(-2,0,1)$  | $(1,-1,1)$  | $(-2,1,0)$  | $(0,-2,1)$  | $(-,)$ | 1,3 |
| 2,1          | $(-1,0,-1)$ | $(-1,-1,0)$ | $(-1,-1,0)$ | $(0,-2,0)$  | $(-1,1,-2)$ | $(-2,0,0)$  | $(1,-1,-2)$ | $(-2,1,-1)$ | $(0,-2,0)$  | $(-,)$ | 2,1 |
| 2,2          | $(-1,0,-1)$ | $(-1,-1,0)$ | $(-1,-1,0)$ | $(-2,0,0)$  | $(-2,1,-1)$ | $(0,0,-2)$  | $(-2,-1,2)$ | $(1,0,-2)$  | $(-,)$ | 2,2 |
| 2,3          | $(-1,0,-1)$ | $(-1,-1,0)$ | $(-1,-1,0)$ | $(-2,0,0)$  | $(-2,-1,1)$ | $(0,0,-2)$  | $(-2,-1,2)$ | $(1,0,-2)$  | $(-,)$ | 2,3 |

Accumulated Mass up to this column for

$q=2$ 0.375 0.5625 0.6562 0.75 0.7968 0.8437 0.8671 0.8906 0.9140 1.0
$q=3$ 0.5925 0.7901 0.8559 0.9218 0.9437 0.9657 0.9730 0.9803 0.9876 1.0

Legend:

- **D**
- **I**
- **N**, N<
- **d**, b+
- *
**Conjecture 21.** (Theorem for $M \leq 8$) For every state $s \in S$,

$$
\mu_\infty(s) = \frac{q^{-K(s)}}{\mathcal{P}(M, q)}.
$$

**Proof.** We assume the previous Theorem 19 or Conjecture 20. To normalize, we want to have

$$
1 = \sum_{s \in S} c_0 \cdot q^{-K(s)} = c_0 \sum_{K \in \mathbb{N}_0} P_M(K) \cdot q^{-K} = c_0 \mathcal{P}(M, q).
$$

With $c_0 := \mathcal{P}(M, q)^{-1} = \mu_\infty(s_0)$, $\mu_\infty(s) := q^{-K(s)}/\mathcal{P}(M, q)$ is a probability distribution (with $\sum_{s \in S(T, t)} \mu_\infty(s) = 1$ for all $0 \leq T \leq M, 1 \leq t \leq M + 1$), which is invariant under $T$.

**7. Asymptotic ($n \to \infty$) Measure for the Linear Complexity Deviation**

**Definition.** Let the mass on states with drain (deviation) $d$ be $\gamma(d, T, t) = \sum_{s \in S(T, t, d)} \mu_\infty(s)$, distinguished according to the timesteps $(T, t)$.

Numerical results indicate that $\gamma$ indeed depends only on the difference $t - T$:

**Theorem 22.** For $1 \leq M \leq 8$, $0 \leq T \leq M$, $1 \leq t \leq M + 1$, and any finite field $\mathbb{F}_q$, let $\Delta := t - T$. Then for every linear complexity deviation $d \in \mathbb{Z}$,

$$
\gamma(d, T, t) = \frac{1}{\mathcal{P}(M, q)} \sum_{h=1}^{M} (-1)^{h+1} \sum_{k=0}^{h-1} q^{(M+1)(h-1)} \cdot \frac{q^{-h(M+1)\cdot|d|} \cdot q^{\varepsilon_{\operatorname{sgn}(d)}(\Delta, h)} \cdot \prod_{k=1}^{M-h}(q^k - 1) \cdot \prod_{k=M+2}^{M+h}(q^k - 1)}{\prod_{k=1}^{M-h}(q^k - 1) \cdot \prod_{k=M+2}^{M+h}(q^k - 1)}, \quad (7)
$$

where

$$
\varepsilon_-(\Delta, h) = h(M - 1 + \Delta) - \binom{h}{2}
$$
$$
\varepsilon_+(\Delta, h) = h(\Delta - h) + \binom{h}{2}
$$
$$
\varepsilon_0(\Delta, h) = \min\{\varepsilon_+(\Delta, h), \varepsilon_-(\Delta, h)\}
$$

depends only on the sign of $d$, and $\simeq$ means equality with precision at least $q^{-(1200 - 100 \cdot M)}$.

**Proof.** By verifying all states with class up to $1200 - 100 \cdot M$ in the partition model. The left and right side coincide up to precision $q^{-1200+100M}$.

**Remark.** This involved about $2^{39}$ or half a trillion states for $M = 8$. We used Victor Shoup’s library NTL [9] (Thank you!).
Conjecture 23. For every $M \in \mathbb{N}$, $0 \leq T \leq M$, $1 \leq t \leq M + 1$, and every finite field $\mathbb{F}_q$, with $\Delta$ and $\varepsilon(\Delta, h)$ as before, for every $d \in \mathbb{Z}$, we have exactly

$$
\gamma(d, T, t) = \sum_{h=1}^{M} \frac{(-1)^{h+1} \left( \sum_{k=0}^{h-1} q^{(M+1)k} \right) \left( \prod_{k=M-h+1}^{M} (q^k - 1) \right) \cdot q^{\varepsilon \text{sgn}(d)(\Delta, h)}}{q^{(M+1)(h+M/2)} \left( \prod_{k=M+2}^{M+h} (q^k - 1) \right) q^{h(M+1) |d|}} ,
$$

the same formula as in Theorem 22, rearranged.

Remark: The resulting values $\gamma(d, T, M + 1)$ for $M = 2$ and $M = 3$, and $d(T, M + 1)$ for $M = 2$, correspond with Niederreiter’s and Wang’s results in [11, Thm. 3], [11, Thm. 4], and [7, Thm. 11], resp., for $n \to \infty$, see also [6]. Observe that we use $d = L - \lceil n \cdot \frac{2}{3} \rceil$, not $L - n \cdot \frac{2}{3}$.

8. The Law of the Logarithm

We follow the approach by Niederreiter in [5] for the case $M = 1$.

Theorem 24. For all $M \in \mathbb{N}$, for all $0 \leq T \leq M$, and $1 \leq t \leq M + 1$, there exists a constant $C(M, T, t) > 0$ (independent of $d$) such that:

$$
\frac{1}{\mathcal{P}(M, q)} \cdot q^{-d|d|} \cdot (M+1) \leq \gamma(d, T, t) \leq C(M, T, t) \cdot q^{-d|d|} \cdot (M+1),
$$

that is

$$
\gamma(d, T, t) = \Theta(q^{-d|d|(M+1)}).
$$

Proof. Lower bound:

We distinguish cases $d < 0$, $d > 0$, and $d = 0$:

a) $d < 0$

Let $b := \lfloor \frac{-d-T}{M} \rfloor$ and $a := -d - T - M \cdot b$ that is $b = -(a + d + T)/M$. Then $s^* := (b, \ldots, b, b+1, \ldots, b+1, d; T, t)$ with $b_1 = \cdots = b_{M-a} = b$ and $b_{M-a+1} = \cdots = b_M = b + 1$ is in $S(T, t)$.

The class of $s^*$ is $K(s^*) = -\pi s + M \cdot T + 2 \sum_{k=1}^{a} (b+1)(M + 1 - k) + 2 \sum_{k=a+1}^{M} b \cdot (M + 1 - k)$, since after the sorting, the $a$ batteries with value $b + 1$ will be the largest, while $d$ is the smallest value.

Now, sorting starts with $b, \ldots, (d), \ldots, b, b + 1, \ldots, (d), \ldots, b + 1$, where $d$ occupies the $t$–th place from the left. $d$ moves to the right by $M+1 - t$ moves, then all the $b$’s interchange with all the $(b+1)$’s in $a(M-a)$ transpositions,
yielding \( \pi_{s^*} = M + 1 - t + a(M - a) \) and class
\[
K(s^*)
= - (M + 1 - t + aM - a^2) + MT + 2b(M + 1)M/2
+ 2[(M + 1)M/2 - (M - a + 1)(M - a)/2]
= - M - 1 + t - aM + a^2 + MT - (d + T + a)(M + 1)
+ (M + 1)M - (M - a + 1)(M - a)
= |d|(M + 1) - [M - t] + 1 + T
\leq |d|(M + 1)
\]
and already \( s^* \) alone accounts for the lower bound.

b) \( d > 0 \)

With \( b, a, \) and \( s^* \) as before, sorting now leads to \( a + a(M - a) \) transpositions, since \( d \) goes to the left. As before,
\[
K(s^*) = - a(M + 1 - a) + MT + 2dM + 2 \sum_{k=2}^{a+1} (b+1)(M+1-k) + 2 \sum_{k=a+2}^{M} b(M+1-k)
= - a(M + 1 - a) + MT + 2dM + 2b(M + 1)M/2 - 2bM + 2(M(M - 1)/2 - (M - a)(M - a - 1)/2)
= - a(M + 1 - a) + MT + 2dM - (d + T + a)(M + 1) + 2(d + T + a) + M(M - 1) - (M - a)(M - a - 1)
= - aM - a + a^2 + MT + 2dM - dM - MT - aM - d - T - a + 2d + 2T + 2a
+ M^2 - M - M^2 + 2aM - a^2 - M + a
= dM + d + T + a - 2M = |d|(M + 1) - (M - T) - (M - a) \leq |d|(M + 1)
\]
c) \( d = 0 \): Let \( s^* \) be the (only) state in \( S(T, t) \) with \( K(s^*) = 0 = |d|(M + 1) \).

Upper bound:
We have
\[
\gamma(d, T, t) = \sum_{s \in S(T, t, d)} q^{-K(s)} \cdot C_0
\]
(use \( C_0 := \mathcal{P}(M, q)^{-1} \), if you trust Conjecture 21), and
\[
K(s) = - \pi_s + M \cdot T + 2 \cdot \sum_{m=1}^{M+1} \delta_m \cdot (M + 1 - m).
\]
a) $d < 0$:

Let $\{b_1, \ldots, b_M\}$ be ordered nonincreasingly as $\tilde{b}_1' \geq \cdots \geq \tilde{b}_M'$ (where $'$ indicates that $d$ does not enter into the sort).

We write the battery values as sum of their differences $\Delta_k := \tilde{b}_{M-k}' - \tilde{b}_{M-k+1}' \geq 0, 1 \leq k \leq M - 1$

$$\tilde{b}_m' = \tilde{b}_M' + \sum_{k=1}^{M-m} \Delta_k$$  \hfill (8)

(where for $m = M$ the empty sum is zero).

By the invariant $d + T + \sum_m b_m = 0$ we must have

$$d + T + \sum_{m=1}^M (\tilde{b}_M' + \sum_{k=1}^{M-m} \Delta_k) = 0$$  \hfill (9)

$$\iff M \cdot \tilde{b}_M' = -d - T - \sum_{k=1}^{M-1} \Delta_k(M - k)$$  \hfill (10)

$$\iff \tilde{b}_M' = -\frac{d}{M} - T + \frac{\sum_{k=1}^{M-1} \Delta_k(M - k)}{M}$$  \hfill (11)

When running $\Delta_1, \ldots, \Delta_{M-1}$ through all values from $\mathbb{N}_0$ and setting $\tilde{b}_M'$ by (9–11) and then $\tilde{b}_m'$ by (8), we obtain all possible values for $(\tilde{b}_1', \ldots, \tilde{b}_M')$ (and a lot more, since $\tilde{b}_M'$ is taken from $\frac{\mathbb{Z}}{M} \supset \mathbb{Z}$).

Furthermore, each $(\tilde{b}_1', \ldots, \tilde{b}_M')$ corresponds to up to $M!$ states (with different permutation of the values) with $\{\tilde{b}_m'\} = \{b_m\}$.

$\pi_s$ can be bounded in general by $0 \leq \pi_s \leq \binom{M+1}{2}$ (the maximum being attained in the case of $b_1 < b_2 < \cdots < b_m$).

The transition from $(\tilde{b}_1', \ldots, \tilde{b}_M'), d$ to $(\tilde{b}_1', \ldots, \tilde{b}_{M+1}')$ \textit{i.e.} including $d$ in the sort order, gives the inequality

$$2 \sum_{m=1}^M \tilde{b}_m'(M + 1 - m) + 2 \cdot 0 \cdot d \leq 2 \sum_{m=1}^{M+1} \tilde{b}_m'(M + 1 - m)$$

since at any rate a smaller $\tilde{b}_m'$ will be replaced by a larger $d$ or $\tilde{b}_{m-1}'$.  \hfill (22)
Putting things together, we have the upper bound

\[
\gamma(d, T, t) \leq C_0 \sum_{\Delta_1 \in \mathbb{N}_0} \cdots \sum_{\Delta_{M-1} \in \mathbb{N}_0} M! \cdot q^{\binom{M+1}{2}} \cdot q^{-MT} \times
\]

\[
\times q^{-2\sum_{m=1}^{M} \frac{d}{2}(M+1-m) - 2\sum_{m=1}^{M} \frac{T - \sum_{k=1}^{M-1} \Delta_k(M-k)}{M}}
\]

\[
= -q^{-2\sum_{m=1}^{M} \frac{d}{2}(M+1-m)} \cdot C(M, T)
\]

\[
= -q^{-d(M+1)} \cdot C(M, T)
\]

where

\[
C(M, T) = C_0 \sum_{\Delta_1 \in \mathbb{N}_0} \cdots \sum_{\Delta_{M-1} \in \mathbb{N}_0} M! \times
\]

\[
\times q^{\binom{M+1}{2}} \cdot q^{-MT - 2\sum_{m=1}^{M} \frac{T - \sum_{k=1}^{M-1} \Delta_k(M-k)}{M}}
\]

is independent of \(d\).

b) In the case \(d > 0\) we follow the same idea, however we put \(d\) as first (largest) value of the sort order. We obtain

\[
\tilde{b}'_m = \tilde{b}'_2 - \sum_{k=1}^{m-2} \Delta_k, 2 \leq k \leq M + 1
\]

with \(\Delta_k \in \mathbb{N}_0\).

The invariant \(d + T + \sum_m b_m = 0\) requires

\[
d + \sum_{m=2}^{M+1} (\tilde{b}'_2 - \sum_{k=1}^{m-2} \Delta_k) = 0 \iff M\tilde{b}'_2 = -d - T + \sum_{k=1}^{M-1} \Delta_k(M-k)
\]

\[
\iff \tilde{b}'_2 = -\frac{d}{M} - \frac{T}{M} + \frac{\sum_{k=1}^{M-1} \Delta_k(M-k)}{M}.
\]

Again up to \(M!\) states can be attached to one sorted tuple \((\tilde{b}'_2, \ldots, \tilde{b}'_{M+1})\), again \(\pi_s \leq \binom{M+1}{2}\), and introducing \(d\) (from the left) increases (if at all) the
values, i.e. \( \tilde{b}_m \geq \tilde{b}_m' \). We obtain

\[
-2 \sum_{m=1}^{M+1} \tilde{b}_m (M + 1 - m) 
\]

\[
\leq -2 \cdot d \cdot M - 2 \sum_{m=2}^{M+1} \tilde{b}_m' (M + 1 - m) 
\]

\[
= -2Md - 2 \sum_{m=2}^{M+1} \left( -\frac{d}{M} - \frac{T}{M} + \sum_{k=1}^{m-2} \frac{\Delta_k(M-k)}{M} - \sum_{k=1}^{m-2} \frac{\Delta_k}{M} \right) (M + 1 - m) 
\]

\[
= -2Md + 2 \sum_{m=2}^{M+1} -\frac{d}{M} + C_1 
\]

\[
= d(2M - 2 + \frac{(M-1)M}{2}) + C_1 
\]

\[
= -d(M + 1) + C_1, 
\]

where \( C_1 \) does not depend on \( d \), and thus

\[
\gamma(d, T, t) \leq C_0 \sum_{\Delta_1 \in \mathbb{N}_0} \cdots \sum_{\Delta_{M-1} \in \mathbb{N}_0} M! \cdot q^{\binom{M+1}{2}} \cdot q^{-MT} q^{-d(M+1) + C_1} 
\]

\[
= q^{-d(M+1)} \cdot C(M, T). 
\]

\[\square\]

**Lemma 25.** (Borel–Cantelli)

(i) Let \( A_1, A_2, \ldots \) be events which happen with probability \( a_1, a_2, \ldots \), resp. If now \( \sum_{k \in \mathbb{N}} a_k < \infty \), then with probability one only finitely many of the events \( A_k \) occur simultaneously.

(ii) Let \( A_1, A_2, \ldots \) be independent events which happen with probability \( a_1, a_2, \ldots \), resp. If now \( \sum_{k \in \mathbb{N}} a_k = \infty \), then with probability one infinitely many of the events \( A_k \) occur simultaneously.

**Proof.** See Feller \[\text{[III VIII.3]}\]. \[\square\]
Theorem 26.
The Law of the Logarithm for Linear Complexity of Multisequences

For all $M \in \mathbb{N}$ and for almost all (in the sense of Haar measure on $(\mathbb{F}_q^M)^\infty$) sequences $a \in (\mathbb{F}_q^M)^\infty$, we have

$$\limsup_{n \to \infty} \frac{d_a(n)}{\log n} = \frac{1}{(M + 1) \log q}$$

and

$$\liminf_{n \to \infty} \frac{d_a(n)}{\log n} = -\frac{1}{(M + 1) \log q}.$$

Proof. We fix some $\varepsilon > 0$ and apply the Borel–Cantelli Lemma 31(i) to the events

$A_n : \frac{d_a(n)}{\log n} > \frac{1 + \varepsilon}{(M + 1) \log q}$.

With $L := \left\lceil \frac{\log n}{(M+1) \log q} \right\rceil$, the probability for $A_k$ is

$$a_k = \sum_{d=L}^{\infty} \gamma(d, T, t) + \sum_{d=-L}^{-\infty} \gamma(d, T, t) \leq 2 \cdot C(M, T) \sum_{d=-L}^{\infty} q^{-|d|(M+1)} = 2C(M, T) \frac{q^{-L(M+1)}}{1 - q^{-(M+1)}}.$$ 

with accumulated probability

$$\sum_{n=1}^{\infty} a_n \leq \frac{2C(M, T)}{1 - q^{-(M+1)}} \sum_{n=1}^{\infty} q^{-(M+1)} \frac{\frac{\log k}{(M+1) \log q}}{1 - q^{-(M+1)}} = \frac{2C(M, T)}{1 - q^{-(M+1)}} \sum_{n=1}^{\infty} n^{-(1 + \varepsilon)} < \infty.$$

For the inner bounds, we need independent events:

Denote by $n_1, n_2, \ldots$ the timesteps, when $d = 0$. If this sequence is finite, $d \to -\infty$, since at least one battery no longer discharges. This event is of measure zero, requiring all discrepancies $\delta$ pertaining to that battery equal to zero from some $n_0$ on.

Assume now an infinite sequence of these timesteps. Let $L_k := \left\lceil \frac{\log k}{(M+1) \log q} \right\rceil$ and let $A_k$ be the event of $(M + 1) \cdot (L + 1)$ consecutive discrepancies, all zero, after $n_k$. The events $A_k$ are independent with probability $a_k = q^{-(M+1)(L+1)}$, since they belong to different, independent discrepancies. Now, within $(L + 1)(M + 1)$ symbols, we have at least $(L+1)^{M+1}$ columns and


thus \([ (L + 1)\frac{M+1}{M} \frac{M}{M+1} ] \geq L\) actions \(d_-\) (without intermediate discharges), and thus \(A_k\) leads to \(d \leq -L\).

With
\[
\sum_{n=1}^{\infty} a_n \geq \sum_{n=1}^{\infty} q^{-(M+1)(1+\frac{\log n}{(M+1)\log q})} = q^{-(M+1)} \sum_{n=1}^{\infty} n^{-1} = \infty
\]
and Lemma 31(ii), we get equality of the bounds. \(\square\)

**Corollary 27.** For all \(M \in \mathbb{N}\), for all \(\varepsilon > 0\), for almost all sequences from \((\mathbb{F}_q^M)^\infty\), it holds
\[
-\varepsilon < \liminf_{n \to \infty} \frac{d_a(n)}{n} \leq \limsup_{n \to \infty} \frac{d_a(n)}{n} < \varepsilon
\]

**Proof.** Almost always, we have
\[
\left| \frac{d_a(n)}{\log n} \right| \leq \frac{1}{(M+1)\log q} \iff \left| \frac{d_a(n)}{n} \right| \leq \frac{\log n}{n(M+1)\log q}
\]
by the last theorem, and with
\[
\frac{1}{(M+1)\log q} \lim_{n \to \infty} \frac{\log n}{n} = 0
\]
the statement follows. \(\square\)

**Theorem 28.** With measure one,
\[
\liminf_{n \to \infty} \frac{L_a(n)}{n} = \limsup_{n \to \infty} \frac{L_a(n)}{n} = \frac{M}{M+1}.
\]

**Proof.** From \(L_a(n) = d_a(n) + \left\lfloor \frac{nM}{M+1} \right\rfloor\) and the previous corollary, we have
\[
\frac{L_a(n)}{n} = \frac{d_a(n)}{n} + \frac{M}{M+1} + O\left(\frac{1}{n}\right)
\]
and thus
\[
\lim_{n \to \infty} \frac{L_a(n)}{n} = \lim_{n \to \infty} \frac{d_a(n)}{n} + \frac{M}{M+1} = \frac{M}{M+1}.
\]
\(\square\)

In other words, we obtain again the result of Niederreiter and Wang \(6\) that \(\frac{L_a(n)}{n} \to \frac{M}{M+1}\) with probability one, for all \(M \in \mathbb{N}\).
9. Finite Strings

Definition. For \( s \in S \), let the generation of state \( s \) be

\[
g(s) = \left\lceil \frac{\tilde{I}_1(s)}{M + 1} \right\rceil \cdot (M + 1).\]

Conjecture 29. For every state \( s \in S \) and every \( n \in \mathbb{N}_0 \)

\[
\mu_n(s) = \begin{cases} 
0, & n < g(s) \land (T, t) \not\equiv n \\
\mu_\infty(s) \cdot F(s), & n = g(s) \\
\mu_\infty(s), & n > g(s) \land (T, t) \equiv n
\end{cases}
\]

with

\[
F(s) = \prod_{m=M_1(s)}^{M} \frac{q^m}{q^m - 1},
\]

for

\[
M_1(s) = M + 1 - \# \{1 \leq m \leq M \mid b_m = \max\{b_1, \ldots, b_M, d\}\}.
\]

In the case of the empty product for \( M_1(s) = M + 1 \), \( F(s) = 1 \), and for \( M_1(s) = 1 \), \( F(s) = \mathcal{P}(M, q) \).

Conjecture 30. a) For \( g \in \mathbb{N}_0 \), let \( \#(g, M) \) be the number of states that are reachable in the \( g \)-th generation. Then

\[
\#(g, M) = \binom{g + M}{M}.
\]

b) \( \#\{s \in S \mid g(s) = g\} = \binom{g + M}{M} - \binom{g}{M} \).

Conclusion

We introduced the Battery–Discharge–Model BDM as a convenient container for all information about linear complexity deviations in \((\mathbb{F}_q^M)^\infty\).

We obtained a closed formula for measures and averages for the linear complexity deviation, numerically proven for the cases \( M = 1, \ldots, 8 \), and conjectured for any \( M \), which coincides with the results known before for \( M = 1, 2, 3 \), but gives a better account of the inner structure of these measures. In particular, the measure is a sum of \( M \) components of the form

\[
\Theta(q^{-|d|(M+1)^h}), h = 1, \ldots, M.
\]
References

[1] M. del P. Canales Chacón, M. Vielhaber, Structural and Computational Complexity of Isometries and their Shift Commutators, Electronic Colloq. on Computational Complexity, ECCC TR04–057, 2004.

[2] Z. Dai, X. Feng, Multi–Continued Fraction Algorithm and Generalized B–M Algorithm over $\mathbb{F}_2$, in [13].

[3] William Feller, An Introduction to Probability Theory and Its Application (3rd ed.), Wiley, 1968.

[4] F. G. Gustavson, Analysis of the Berlekamp-Massey linear feedback shift-register synthesis algorithm, IBM J. Res. Develop., 20, 204–212, 1976.

[5] H. Niederreiter, “The probabilistic theory of linear complexity”, in: Advances in Cryptology – EUROCRYPT ’88 (C.G. Günther, Ed.), LNCS 330, 191 – 209, Springer, Berlin, 1988.

[6] H. Niederreiter, L.–P. Wang, Proof of a Conjecture on the Joint Linear Complexity Profile of Multisequences, in: INDOCRYPT 2005 (T. Johansson, S. Maitra et al. Eds) , LNCS 3797, Springer 2005, 13–22.

[7] H. Niederreiter, L.–P. Wang, The Asymptotic Behavior of the Joint Linear Complexity Profile of Multisequences, Preprint.

[8] M. Rosenblatt, Random Processes, Springer, 1974.

[9] Victor Shoup, The Number Theory Library NTL, http://shoup.net/ntl

[10] M. Vielhaber, A Unified View on Sequence Complexity Measures as Isometries, in [13].

[11] L.–P. Wang, H. Niederreiter, Enumeration results on the joint linear complexity of multisequences, to appear in FFA.

[12] Handbook of Discrete and Combinatorial Mathematics (K.H. Rosen, ed.) CRC, Mouth of the Mouse, Fla., 2000.

[13] Proceedings SETA ’04, International Conference on Sequences and Their Applications, October 24 – 28, 2004, Seoul, Korea and LNCS 3486, Springer, 2005.