The Parametric Complexity of Lossy Counter Machines

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The reachability problem in lossy counter machines is the best-known ACKERMANN-complete problem and has been used to establish most of the ACKERMANN-hardness statements in the literature. This hides however a complexity gap when the number of counters is fixed. We close this gap and prove $F_d$-completeness for machines with $d$ counters, which provides the first known uncontrived problems complete for the fast-growing complexity classes at levels $3 < d < \omega$. We develop for this an approach through antichain factorisations of bad sequences and analysing the length of controlled antichains.

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1 Introduction

Mayr and Meyer exhibited in 1981 ‘the first uncontrived decidable problems which are not primitive-recursive,’ namely the finite containment and equality problems in Petri nets [32]. McAloon [34], Clote [8], and Howell and Yen [24] subsequently proved Ackermannian upper bounds for these two problems, essentially matching Mayr and Meyer’s lower bound.

Such an astronomical complexity could have been an isolated phenomenon with only a few examples related to the original problems, but uncontrived problems with a similar complexity actually occur in logic (e.g., relevance logic [48], data logics [11, 16], interval temporal logic [36], linear logic [29], metric temporal logic [28]), verification (e.g., counter machines [44, 23], fragments of the $\pi$-calculus [4], broadcast protocols [43], rewriting systems [21], register automata [11, 18]), and games (e.g., partial observation energy games [37], bisimulation games on pushdown automata [25]), and even higher complexities also occur naturally [30, 6, 38, 20, 19, 10]; see [39, Sec. 6] for an overview.

This abundance of results is largely thanks to a framework [42, 43, 40] that comprises:

- The definition of an ordinal-indexed hierarchy $(F_\alpha)_{\alpha}$ of fast-growing complexity classes, along with assorted notions of reductions and completeness suitable to work with such high complexities [39]. The previous decision problems are complete for $\text{ACKERMANN} = F_\omega$ under primitive-recursive reductions; $F_\omega$ is the lowest non primitive-recursive class in the hierarchy, where $\text{TOWER} = F_3$ corresponds to problems solvable in time bounded by a tower of exponentials and where each $F_k$ for a finite $k$ is primitive-recursive.

- The identification of master decision problems, which allow to establish completeness more easily than from first principles. For instance, reachability in lossy counter machines [48, 44, 46] plays a similar role for $\text{ACKERMANN}$ as, e.g., 3SAT for $\text{NP}$ or QBF for $\text{PSPACE}$, and has been used to derive most of the known $\text{ACKERMANN}$-hardness results [4, 11, 16, 36, 28, 23, 21, 18, 25].
Lower bound techniques for establishing the complexity of such master problems: this typically relies on implementing weak computers for Hardy functions and their inverses in the formalism at hand, allowing to build a large but bounded working space on which a Turing or a Minsky machine can then be simulated [48, 44, 46, 6, 38, 20, 19].

Upper bound techniques relying on combinatorial statements, called length function theorems, on the length of controlled bad sequences over well-quasi-orders, which are used to prove the termination of the decision procedures [34, 8, 50, 7, 15, 41, 3, 38].

From an algorithmic perspective, these results are negative and one could qualify such problems as merely 'not undecidable.' What we gain however are insights into the computational power of the models, allowing to compare them and to identify the main sources of complexity—e.g., in lossy counter machines, the key parameter is the number of counters. Furthermore, from a modelling perspective, a formalism with a tremendous computational power that nevertheless falls short of Turing completeness can be quite satisfactory.

Contributions. In this paper, we revisit the proof of the best-known result in this area, namely the ACKERMANN-completeness of reachability in lossy counter machines (LCMs). Those are simply multi-counter Minsky machines with a lossy semantics that allows the counters to decrease in an uncontrollable manner during executions; see Section 2.

The gap in the current state of knowledge appears when one fixes the key complexity parameter, i.e., the number \(d\) of counters. Indeed, the best known lower bound for LCM reachability is \(F_d\)-hardness when \(d \geq 3\) [40, Thm. 4.9], but the best known upper bound is \(F_{d+1}\) [15, 42, 3]. This complexity gap reveals a serious shortcoming of the framework advertised earlier in this introduction, and also impacts the complexity of many problems shown hard through a reduction from LCM reachability.

Our first main contribution in Proposition 10 is an \(F_d\) upper bound, which together with the lower bound from [40, Thm. 4.9] entails the following completeness result.

> Theorem 1. LCM Reachability is \(F_\omega\)-complete, and \(F_d\)-complete if the number \(d \geq 3\) of counters is fixed.

Note that this provides an uncontrived decision problem for every class \(F_k\) with \(3 \leq k \leq \omega\), whereas no natural \(F_k\)-complete problems were previously known for the intermediate primitive-recursive levels strictly between TOWER and ACKERMANN, i.e., for \(3 < k < \omega\).

As we recall in Section 3, reachability in lossy counter machines can be solved using the generic backward coverability algorithm for well-structured systems [1, 17]. As usual, we derive our complexity upper bound by bounding the length of the bad sequences that underlie the termination argument for this algorithm. The main obstacle here is that the length function theorems in [15, 42, 3]—i.e., the bounds on the length of controlled bad sequences over \(\mathbb{N}^d\)—are essentially optimal and only yield an \(F_{d+1}\) complexity upper bound.

We circumvent this using a new approach in Section 4. We restrict our attention to strongly controlled bad sequences rather than the more general amortised controlled ones (see Section 4.1), which in turn allows us to work on the antichain factorisations of bad sequences (see Section 4.2). This entails that, in order to bound the length of strongly controlled bad sequences, it suffices to bound the length of strongly controlled antichains. This is tackled in Section 5, where we prove a width function theorem on the length of controlled antichains over \(\mathbb{N}^d\); to the best of our knowledge, this is the first statement of this kind specific to antichains rather than bad sequences. We wrap up with the proof of Proposition 10 in Section 6.
The developments of Sections 4 and 5 form our second main contribution. They are of wide interest beyond lossy counter machines, as they can be applied whenever the termination of an algorithm relies on \( \mathbb{N}^d \) having finite (controlled) bad sequences or antichains.

## 2 Lossy Counter Machines

**Syntax.** A lossy counter machine (LCM) [33] is syntactically identical to a Minsky machine \( M = (Q, C, \delta) \), where the transitions in \( \delta \subseteq Q \times C \times \{=0?, ++, --\} \times Q \) operate on a finite set \( Q \) of control locations and a finite set \( C \) of counters through zero-tests \( c=0? \), increments \( c++ \) and decrements \( c-- \).

**Operational Semantics.** The semantics of an LCM differ from the usual, ‘reliable’ semantics of a counter machine in that the counter values can decrease in an uncontrolled manner at any point of the execution. Formally, a configuration \( q(v) \) associates a control location \( q \) in \( Q \) with a counter valuation \( v \) in \( \mathbb{N}^C \), i.e. counter values can never go negative. The set of configurations \( Q \times \mathbb{N}^C \) is ordered by the product ordering: \( q(v) \leq q'(v') \) if \( q = q' \) and \( v(c) \leq v'(c) \) for all \( c \in C \).

A transition of the form \( (q, c, \text{op}, q') \in \delta \) defines a set of reliable computation steps

\[
q(v) \rightarrow q'(v'),
\]

where \( v(c') = v'(c') \) for all \( c \neq c' \) in \( C \) and

- if \( \text{op} = =0? \), then \( v(c) = v'(c) = 0 \),
- if \( \text{op} = ++ \), then \( v(c) + 1 = v'(c) \), and
- if \( \text{op} = -- \), then \( v(c) = v'(c) + 1 \).

A lossy computation step is then defined by allowing counter values to decrease arbitrarily between reliable steps: \( q(v) \rightarrow \ell q'(v') \) if there exist \( w \leq v \) and \( w' \geq v' \) such that \( q(w) \rightarrow q'(w') \).

We write as usual \( \rightarrow_\ell \) for the transitive reflexive closure of \( \rightarrow_\ell \).

**Reachability.** The decision problem we tackle in this paper is the following.

- **Problem (LCM Reachability).**
  - **instance** A lossy counter machine \( (Q, C, \delta) \) and two configurations \( q_0(v_0) \) and \( q_f(v_f) \).
  - **question** Is \( q_f(v_f) \) reachable from \( q_0(v_0) \), i.e., does \( q_0(v_0) \rightarrow_\ell q_f(v_f) \)?

Note that, due to the lossy semantics, this is equivalent to the coverability problem, which asks instead whether there exists \( v \geq v_f \) such that \( q_0(v_0) \rightarrow_\ell q_f(v) \). Indeed, such a \( v \) exists if and only if \( q_0(v_0) \geq q_f(v_f) \) or \( q_0(v_0) \rightarrow_\ell q_f(v_f) \).

While many problems are undecidable in LCMs [33, 45], these systems are in fact well-structured in the sense of [1, 17], which means that their coverability problem is decidable, as further discussed in Section 3. The ACKERMANN-hardness of reachability was first shown by Schnoebelen [44] in 2002, while an ACKERMANN upper bound follows from the length function theorems for Dickson’s Lemma [34, 8, 15, 42, 3]. Note that LCM reachability is equivalent to reachability in counter machines with incrementing errors [11] and to coverability in reset counter machines [46, Sec. 6], and this also holds if we fix the number of counters.

- **Example 2 (Weak \( \log^* \)).** Figure 1a shows the pseudo-code for a program computing \( \log^* n \) in \( y \), i.e., the inverse of a tower of exponentials of height \( y \). Indeed, when started with \( x = y = 0 \), each time the program visits the location \( q_2 \), it has performed \( n := \log n \) by

\[ \log \log \ldots \log n \text{ (times } y \text{ times)} \]

1 Urechhart [48] showed independently in 1999 and using a similar approach the same result for the closely related model of alternating expansive vector addition systems.


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\[ q_0: \text{while } n \mod 2 = 0 \text{ do} \]
\[ n := n/2; \]
\[ \text{if } x = n \text{ then} \]
\[ q_2: \quad y := y+1; \quad x := 0; \]
\[ \text{else} \]
\[ q_1: \quad x := x+1; \]
\[ \text{done} \]

(a) An integer program.  

(b) A counter machine implementing Figure 1a.

Figure 1 The counter machine for Example 2.

looping over \( q_1 \). Figure 1b displays a counter machine with counters \( \mathbb{C} \equiv \{x, y, n, n'\} \), where \( n' \) is an auxiliary counter used to perform division by two. Note that the deterministic choice if \( x = n \) is replaced by a non-deterministic choice between going to \( q_1 \) or \( q_2 \), but \( q_2 \) checks that this choice was correct by decrementing \( x \) and \( n' \) in lockstep and checking that they are both equal to zero before returning to \( q_0 \). If started in a configuration \( q_0(0, 0, n, 0) \) and using reliable semantics, this machine reaches \( q_0(0, y, 0, 0) \) exactly for \( y = \log^* n \). With lossy semantics, it might also reach smaller values of \( y \).

3 Well Structured Systems

Well-structured transition systems (WSTS) [1, 17] form a family of computational models where the (usually infinite) set of configurations is equipped with a well-quasi-ordering (see Section 3.1) that is ‘compatible’ with the computation steps (see Section 3.2). The existence of this well-quasi-ordering allows for the decidability of some important behavioural properties like termination (from a given initial configuration) or coverability, see Section 3.3.

3.1 Well-Quasi-Orders

A \textit{quasi-order} (qo) is a pair \((X, \leq)\) where \( \leq \subseteq X \times X \) is transitive and reflexive; we write \( x < y \) for the associated strict ordering, when \( x \leq y \) and \( y \not\leq x \) for incomparable elements, when \( x \not\leq y \) and \( y \not\leq x \), and \( x \equiv y \) for equivalent elements, when \( x \leq y \) and \( y \leq x \). The \textit{upward-closure} \( \uparrow Y \) of some \( Y \subseteq X \) is defined as \( \uparrow Y \equiv \{x \in X \mid \exists y \in Y. x \geq y\} \); we write \( \uparrow x \) instead of \( \uparrow \{x\} \) for singletons and say that a set \( U \subseteq X \) is \textit{upwards-closed} when \( U = \uparrow U \). We call a finite or infinite sequence \( x_0, x_1, x_2, \ldots \) over \( X \) \textit{bad} if for all indices \( i < j \), \( x_i \not\leq x_j \); if \( x_i \perp x_j \) for all \( i < j \), then \( x_0, x_1, x_2, \ldots \) is an \textit{antichain}.

A \textit{well-quasi-order} (wqo) [22, 27] is a qo \((X, \leq)\) where bad sequences are finite. Equivalently, \((X, \leq)\) is a wqo if and only if it is both \textit{well-founded}, i.e., there does not exist any infinite decreasing sequences \( x_0 > x_1 > x_2 > \cdots \) of elements in \( X \), and has the \textit{finite antichain condition}, i.e., there are no infinite antichains. Still equivalently, \((X, \leq)\) is a wqo if and only if it has the \textit{ascending chain condition}: any increasing sequence \( U_0 \subseteq U_1 \subseteq U_2 \subseteq \cdots \) of upwards-closed subsets of \( X \) eventually stabilises, i.e., \( \bigcup_{i \in \mathbb{N}} U_i = U_k = U_{k+1} = U_{k+2} = \cdots \) for some \( k \). Still equivalently, \((X, \leq)\) is a wqo if and only if it has the \textit{finite basis property}: any non-empty subset contains at least one, and at most finitely many minimal elements (up to equivalence); thus if \( U \subseteq X \) is upwards-closed, then \( \min U \) is finite and \( U = \uparrow (\min U) \).

For a basic example, consider any finite set \( Q \) along with the equality relation, which is a \( \text{wqo} \{(Q, =)\} \) by the pigeonhole principle. Any well-order is a \( \text{wqo} \), thus the set of natural numbers and any of its initial segments \([k] \equiv \{0, \ldots, k-1\}\) along with their natural ordering.
are also wqos. More examples can be constructed using algebraic operations: for instance, if \((X_0, \leq_{X_0})\) and \((X_1, \leq_{X_1})\) are wqos, then so are:

- their disjoint sum \((X_0 \cup X_1, \leq)\) where \(X_0 \cup X_1 \equiv \{(x, 0) \mid x \in X_0\} \cup \{(x, 1) \mid x \in X_1\}\) and \((x, i) \leq (y, j)\) if \(i = j\) and \(x \leq_{X_i} y\);

- their Cartesian product \((X_0 \times X_1, \leq)\) where \((x_0, x_1) \leq (y_0, y_1)\) if \(x_i \leq_{X_i} y_i\) for all \(0 \leq i \leq 1\); in the case of \((\mathbb{N}^d, \leq)\), this result is also known as Dickson’s Lemma [12].

Note that the set of configurations \((Q \times \mathbb{N}^d, \leq)\) of an LCM is a wqo for the product ordering.

### 3.2 Compatibility

An ordered transition system \(S = (S, \rightarrow, \leq)\) combines a set \(S\) of configurations with a transition relation \(\rightarrow \subseteq S \times S\) and a quasi-ordering \(\leq\) of its configurations. An ordered transition system \(S = (S, \rightarrow, \leq)\) is well-structured if \((S, \leq)\) is a wqo and

\[
\forall s_1, s_2, t_1 \in S, (s_1 \rightarrow s_2 \text{ and } s_1 \leq t_1) \implies \exists t_2 \in S, (t_1 \rightarrow t_2 \text{ and } s_2 \leq t_2) .
\]

(1)

This property is also called compatibility (of the ordering with the transitions). Formally, it just means that \(\leq\) is a simulation relation for \((S, \rightarrow)\), in precisely the classical sense of [35]. The point of (1) is to ensure that a larger configuration can do at least as much as a smaller configuration. For instance, lossy steps in a LCM are visibly compatible with \(\leq\) according to (1), and thus the transition system \((Q \times \mathbb{N}^d, \rightarrow, \leq)\) defined by the lossy operational semantics of a LCM is a WSTS.

### 3.3 Coverability

We focus here on the coverability problem: given a WSTS \((S, \rightarrow, \leq)\) and two configurations \(s, t \in S\), does \(s\) cover \(t\), i.e., does there exist \(t' \geq t\) such that \(s \rightarrow^* t'\)? The decidability of this problem uses a set-saturation method first introduced by Arnold and Latteux [5] for reset Petri nets, but the algorithm was independently rediscovered by Abdulla et al. [2] for lossy channel systems and its generic formulation was popularised in the surveys [1, 17].

**Backward Coverability.** The algorithm computes \(\text{Pre}_\leq^*(\uparrow t) \equiv \{s' \in S \mid \exists t' \geq t, s' \rightarrow^* t'\}\), i.e., the set of configurations that cover \(t\); it only remains to check whether \(s \in \text{Pre}_\leq^*(\uparrow t)\) in order to answer the coverability instance. More precisely, for a set of configurations \(U \subseteq S\), let us define its (existential) predecessor set as \(\text{Pre}_\leq(U) \equiv \{s \in S \mid \exists s' \in U, s \rightarrow s'\}\). The algorithm computes the limit of the sequence \(U_0 \subseteq U_1 \subseteq \cdots\) defined by

\[
U_0 \equiv \uparrow t , \quad U_{n+1} \equiv U_n \cup \text{Pre}_\leq(U_n) .
\]

(2)

Note that for all \(n, U_n = \{s' \in S \mid \exists t' \geq t, s' \rightarrow^{\leq n} t'\}\) is the set of configurations that cover \(t\) in at most \(n\) steps, and that we can stop this computation as soon as \(U_{n+1} \subseteq U_n\).

There is no reason for the chain defined by (2) to stabilise in general ordered transition systems, but it does in the case of a WSTS. Indeed, \(\text{Pre}_\leq(U)\) is upwards-closed whenever \(U \subseteq S\) is upwards-closed, thus the sequence defined by (2) stabilises to \(\bigcup_{i \in \mathbb{N}} U_i = \text{Pre}_\leq^*(\uparrow t)\) after a finite amount of time thanks to the ascending chain condition. Moreover, the finite basis property ensures that all the sets \(U_i\) can be finitely represented using their minimal elements, and the union or inclusion of two upwards-closed sets can be computed on this representation. The last ingredients are two effectiveness assumptions:

- \((S, \leq)\) should be effective, meaning that \(S\) is recursive and the ordering \(\leq\) is decidable,
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there exists an algorithm returning the set of minimal predecessors \( \min \text{Pre}_\exists(\uparrow s') \) of any given configuration \( s' \); this is known as the effective pred-basis assumption.

These two assumptions hold in LCMs: \( (Q \times \mathbb{N}^n, \leq) \) is certainly effective, and the minimal predecessors of a configuration \( q' (v') \) can be computed by

\[
\min \text{Pre}_\exists(\uparrow q' (v')) = \min\{ q(\text{pre}_{c \oplus p}(v')) \mid (q, c, p, q') \in \delta \text{ and } v'(c) = 0 \text{ if } p = 0 \}\tag{3}
\]

where \( \text{pre}_{c \oplus p}(v') \) is a vector in \( \mathbb{N}^n \) defined by \( \text{pre}_{c \oplus p}(v')(c') \equiv v'(c') \) for all \( c' \neq c \) in \( C \) and

\[
\text{pre}_{c \oplus p}(v')(c) \equiv 0, \quad \text{pre}_{c \ominus p}(v')(c) \equiv \max\{0, v'(c) - 1\}, \quad \text{pre}_{c \ominus p}(v')(c) \equiv v'(c) + 1. \tag{4}
\]

Let us consider again the lossy counter machine of Example 2. Observe that the backward coverability algorithm with target configuration \( q_0(0, y, 0, 0) \) terminates with \( q_0(0, 0, n, 0) \in \min \text{Pre}_\exists(\uparrow q_0(0, y, 0, 0)) \) for \( n = \text{tower}(y) \), and thus requires a non-elementary number of steps.

**Coverability Pseudo-Witnesses.** Let us reformulate the termination argument of the backward coverability algorithm in terms of bad sequences. We can extract a sequence of elements \( t_0, t_1, \ldots \) from the ascending sequence \( U_0 \subseteq U_1 \subseteq \cdots \) defined by (2) before saturation: \( t_0 \equiv t \) and \( t_{i+1} \in U_{i+1} \setminus U_i \) for all \( i \). Note that if \( i < j \), then \( t_j \in U_j \setminus U_i \) and therefore \( t_i \not\subseteq t_j \); the sequence \( t_0, t_1, \ldots \) is bad and therefore finite. In fact, we can even pick \( t_{i+1} \) at each step among the minimal elements of \( \text{Pre}_\exists(\uparrow t_i) \); we call such a bad sequence \( t_0, t_1, \ldots, t_n \) with

\[
t_0 \equiv t, \quad t_{i+1} \in \min \text{Pre}_\exists(\uparrow t_i) \setminus U_i. \tag{5}
\]

a pseudo-witness of the coverability of \( t \). The maximal length of pseudo-witnesses is therefore equal to the number of steps of the backward coverability algorithm, and this is what we will bound in the upcoming Sections 4 and 5.

## 4 Controlled Bad Sequences and Antichains

As we have just discussed, the running time of the backward coverability algorithm is essentially bounded by the length of the bad sequences constructed by its termination argument. Though bad sequences over a wqo are finite, we cannot bound their lengths in general; e.g., \( (0, n + 1), (0, n), \ldots, (0, 0) \) and \( (1, 0), (0, n), (0, n - 1), \ldots, (0, 1), (0, 0) \) are bad sequences over \( \mathbb{N}^2 \) of length \( n + 2 \) for all \( n \). Nevertheless, a bad sequence produced by an algorithm like the backward coverability algorithm of Section 3.3 is not arbitrary, because its elements are determined by the algorithm’s input and the complexity of its operations. We capture this intuition formally through controlled sequences.

### 4.1 Controlling Sequences

**Norms.** Given a wqo \( (X, \leq_X) \), we posit a norm function \( \cdot \big|_X : X \to \mathbb{N} \); if \( x \leq_X x' \) implies \( |x|_X \leq |x'|_X \), we call this norm monotone. In order to be able to derive combinatorial statements, we require \( X_{\leq n} \equiv \{ x \in X \mid |x|_X \leq n \} \) to be finite for every \( n \); we call the resulting structure \( (X, \leq_X, \cdot \big|_X) \) a normed wqo (nwo).

We will use the following monotone norms on the wqos we defined in Section 3.1: over a finite \( Q \), all the elements have the same norm 0; over \( \mathbb{N} \) or \([d] \), \( n \) has norm \( |n|_n = \min |d| = n \); over disjoint sums \( X_0 \sqcup X_1, (x, i) \) uses the norm \( |x|_{X_i} \) of its underlying set; finally, over Cartesian products \( X \times Y, (x, y) \) uses the infinite norm \( \max(|x|_X, |y|_Y) \).
Controls. Let \( n_0 \in \mathbb{N} \) and let \( g: \mathbb{N} \rightarrow \mathbb{N} \) be a monotone and inflationary function, i.e., for all \( x \leq x' \), \( g(x) \leq g(x') \) and \( x \leq g(x) \). We say that a sequence \( x_0, x_1, x_2, \ldots \) of elements in \( X \) is amortised \((g, n_0)\)-controlled if
\[
\forall i. \ |x_i|_X \leq g^i(n_0) ,
\]
where \( g^i \) denotes the \( i \)th iterate of \( g \). We say that it is strongly \((g, n_0)\)-controlled if
\[
|x_0|_X \leq n_0 \quad \text{and} \quad \forall i. \ |x_{i+1}|_X \leq g(|x_i|_X) .
\]
By definition, a strongly controlled sequence is also amortised controlled: \( |x_0|_X \leq g^0(n_0) = n_0 \), which prompts the name of initial norm for \( n_0 \), and amortised steps cannot grow faster than \( g \) the control function. A useful property of strongly \((g, n_0)\)-controlled sequences \( x_0, x_1, \ldots \) is that, for all \( i \), the suffix \( x_i, x_{i+1}, \ldots \) is also \((g, |x_i|_X)\)-controlled; this might not hold in an amortised controlled sequence.

Previous works like \([15, 42, 3]\) focused on the more general amortised controlled sequences, but strong controlled ones are actually more relevant in practice. For instance, in the case of LCM coverability, the computation of minimal predecessors in (3–4) shows that the pseudo-LCM coverability, the computation of minimal predecessors in (3–4) shows that the pseudo-

\[
\text{Length, Norm, and Width Functions.} \quad \text{The point of controlled sequences is that their length can be bounded. Consider for this the tree obtained by sharing the common prefixes of all the strongly \((g, n_0)\)-controlled bad sequences over a normed wqo \((X, \leq_X, |.|_X)\).}
\]

= This tree is finitely branching by (7)—its branching degree is bounded by the cardinal of \( X_{\leq g^i(n_0)} \) for a node at depth \( i \), and
\=

it has no infinite branches since bad sequences over \((X, \leq_X)\) are finite.

By König’s Lemma, this tree of bad sequences is therefore finite, of some height \( L^{s}_{g,X}(n_0) \) representing the length of the maximal strongly \((g, n_0)\)-controlled bad sequence(s) over \( X \), and we also let \( N^{s}_{g,X}(n_0) \) bound the norms encountered along such sequences; note that \( N^{s}_{g,X}(n_0) \leq g^{L^{s}_{g,X}(n_0)}(n_0) \) since \( g \) is monotone inflationary. Similarly, there exists a maximal length, denoted by \( W^{s}_{g,X}(n_0) \) (resp. \( W^s_{g,X}(n_0) \)), for amortised (resp. strongly) \((g, n_0)\)-controlled antichains over \( X \). We call \( L^{s}_{g,X}, N^{s}_{g,X}, \) and \( W^{s}_{g,X} \) the strong length, strong norm, and strong width functions respectively, and \( W^{s}_{g,X} \) the width function. By definition, a strongly controlled antichain is amortised controlled, i.e.,
\[
W^{s}_{g,X}(n_0) \leq W^{s}_{g,X}(n_0) ,
\]
and the length of a bad sequence where all the elements are of norm at most \( N = N^{s}_{g,X}(n_0) \) is bounded by the cardinal of \( X_{\leq_N} \), i.e.,
\[
L^{s}_{g,X}(n_0) \leq |X_{\leq N^{s}_{g,X}(n_0)}| .
\]
Observe that \( L^{s}_{g,X}(|t|_X) \) bounds the number of steps required by the backward coverability algorithm for a WSTS over \((X, \leq_X, |.|_X)\) with target configuration \( t \) where \( s' \in \min \text{Pre}(|t'|) \) implies \( |s'|_X \leq g(|t'|_X) \). In the case of LCMs, we are therefore interested in \( L^{s}_{H, Q, N}(|v_f|) \).

4.2 Antichain Factorisations

Let \((X, \leq_X, |.|_X)\) be a wqo where \(|.|_X\) is monotone—i.e., \( x \leq x' \) implies \( |x|_X \leq |x'|_X \)—, and let \( x_0, x_1, \ldots, x_{r-1} \in X^* \) be a strongly \((g, n_0)\)-controlled bad sequence over \((X, \leq_X, |.|_X)\).
As seen in Section 4, by suitably controlling how large the elements can grow in antichains, we can derive upper bounds on the time and space required by the backward coverability.
algorithm of Section 3. We prove in this section a width function theorem, a combinatorial statement on the length of amortised controlled antichains over tuples of natural numbers, which will allow to derive a complexity upper bound for reachability in lossy counter machines.

The high complexities at play here require the use of ordinal-indexed subrecursive functions in order to denote non-elementary growths. We first recall the definitions of two families of such functions in Section 5.1. We then prove in Section 5.2 a bound on the width function \( W_{g, N} \) using the framework of [41, 42].

5.1 Subrecursive Hierarchies

We employ notations compatible with those of Schwichtenberg and Wainer [47, Chap. 4], and refer the interested reader to their monograph and [42] for proofs and additional material.

Fundamental Sequences and Predecessors. Consider an ordinal term \( \alpha \) in Cantor normal form \( \omega^{\alpha_1} + \cdots + \omega^{\alpha_p} \) where \( \alpha_1 \geq \cdots \geq \alpha_p \). Then \( \alpha = 0 \) if and only if \( p = 0 \), an ordinal \( \alpha \) of the form \( \alpha' + 1 \) (i.e. with \( p > 0 \) and \( \alpha_p = 0 \)) is called a successor ordinal, and otherwise if \( \alpha_p > 0 \) it is called a limit ordinal, and can be written as \( \gamma + \omega^\beta \) by setting \( \gamma = \omega^{\alpha_1} + \cdots + \omega^{\alpha_{p-1}} \) and \( \beta = \alpha_p \). We write ‘\( \lambda \)’ to denote a limit ordinal.

A fundamental sequence for a limit ordinal \( \lambda \) is a strictly increasing sequence \( (\lambda(x))_{x < \omega} \) of ordinal terms with supremum \( \lambda \). We use the standard assignment of fundamental sequences to limit ordinals below \( \varepsilon_0 \) in Cantor normal form, defined inductively by

\[
(\gamma + \omega^{\beta+1})(x) \equiv \gamma + \omega^\beta \cdot (x + 1), \quad (\gamma + \omega^\lambda)(x) \equiv \gamma + \omega^{\lambda(x)}.
\]

(11)

This particular assignment satisfies e.g. \( 0 < \lambda(x) < \lambda(y) \) for all \( x < y \). For instance, \( \omega(x) = x + 1, (\omega^{x^1 + \omega^2})(x) = \omega^{x^1 + \omega^2 + \omega(x+1)} \).

The predecessor \( P_x(\alpha) \) of an ordinal term \( 0 < \alpha < \varepsilon_0 \) at \( x \in \mathbb{N} \) is defined inductively by

\[
P_x(\alpha + 1) \equiv \alpha, \quad P_x(\lambda) \equiv P_x(\lambda(x)).
\]

(12)

In essence, the predecessor of an ordinal is obtained by repeatedly taking the \( x \)th element in the fundamental sequence of limit ordinals, until we finally reach a successor ordinal and may remove 1. For instance, \( P_x(\omega^2) = P_x(\omega \cdot (x+1)) = P_x(\omega \cdot x + x + 1) = \omega \cdot x + x \).

Hardy and Cichoń Hierarchies. Let \( h: \mathbb{N} \to \mathbb{N} \) be a function. The Hardy hierarchy \( (h^\alpha)_{\alpha \in \varepsilon_0} \) and the Cichoń hierarchy \( (h_\alpha)_{\alpha \in \varepsilon_0} \) relative to \( h \) are defined for all \( 0 < \alpha < \varepsilon_0 \) by

\[
h^0(x) \equiv x, \quad h^\alpha(x) \equiv h^{P_\alpha}(h(x)), \quad h_0(x) \equiv 0, \quad h_\alpha(x) \equiv 1 + h_{P_\alpha(\alpha)}(h(x)).
\]

(13)

Observe that \( h^k(x) = h^{P_\alpha(k)}(h(x)) = h^{k-1}(h(x)) \) for some finite \( k \) is the \( k \)th iterate of \( h \). This intuition carries over: \( h^\alpha \) is a ‘transfinite’ iteration of the function \( h \), using diagonalisation in the fundamental sequences to handle limit ordinals. A standard choice for the function \( h \) is the successor function, noted \( H(x) \equiv x + 1 \); in that case, we see that a first diagonalisation yields \( H^2(x) = H^2(x + 1) = 2x + 1 \). The next diagonalisation occurs at \( H^{\omega + 1}(x + 1) = H^\omega(2x + 1) = 4x + 3 \). Fast-forwarding a bit, we get for instance a function of exponential growth \( H^{\omega^2}(x) = 2^{x+1} + 1 \), and later a non-elementary function \( H^{\omega^2}(x) \) akin to a tower of exponentials of height \( x \), and an ‘Ackermannian’ non-primitive-recursive function \( H^{\omega^2} \).

Both \( h^\alpha \) and \( h_\alpha \) are monotone and inflationary whenever \( h \) is monotone inflationary. Hardy functions are well-suited for expressing large iterates of a control function, and therefore for bounding the norms of elements in a controlled sequence. Cichoń functions are well-suited
for expressing the length of controlled sequences: we can compute how many times we should iterate $h$ in order to compute $h^\alpha(x)$ using the corresponding Cichoń function [7]:

$$h^\alpha(x) = h^{h_\alpha(x)}(x).$$ (14)

### 5.2 Width Function for Dickson’s Lemma

The starting point for our analysis is a descent equation for amortised controlled antichains through residuals, similar to the equations proven in [15, 41] for bad sequences (see Lemma 4). The key idea introduced in [15] is then to over-approximate the residuals of $\mathbb{N}^d$ by working over polynomial nwqos, where disjoint sums are also allowed. Then, the notion of ‘over-approximation’ of residuals of polynomial nwqos is captured formally by showing the existence of a normed reflection into another polynomial nwqo. The final step lifts this analysis to ordinals, allowing to relate $W^2_{g,X}$ for a polynomial nwqo $X$ to functions in the Cichoń hierarchy.

**Strict Polynomial Normed wqos.** Let us write $X \cdot p$ for $X \sqcup \cdots \sqcup X$, $X^d$ for $X \times \cdots \times X$, and $0$ for the empty nwqo. We call a nwqo of the form $\mathbb{N}^{d_1} \sqcup \cdots \sqcup \mathbb{N}^{d_m}$ for some $m \geq 0$ and $d_1, \ldots, d_m \geq 1$ a strict polynomial nwqo. The set of configurations $Q \times \mathbb{N}^{d}$ of an LCM with $|Q| = q$ locations and $|C| = d \geq 1$ counters, along with its ordering and infinite norm, is isomorphic to the strict polynomial nwqo $\mathbb{N}^d \cdot q$.

**Residuals and a Descent Equation.** Let $(X, \leq_X, |.|_X)$ be a normed wqo and $x$ be an element of $X$. We write $X_{\perp x} \equiv \{ y \in X \mid x \perp y \}$ for the residual of $X$ in $x$. By the finite antichain condition, there cannot be infinite sequences of residuations $((X_{\perp x_0})_{\perp x_1})_{\perp x_2} \cdots_{\perp x}$, because $x_i \perp x_j$ for all $i < j$ and it would create an infinite antichain.

Consider now an amortised $(g, n_0)$-controlled antichain $x_0, x_1, x_2, \ldots$ over $X$. Assuming the sequence is not empty, then for all $i > 0, x_0 \perp x_i$, i.e. the suffix $x_1, x_2, \ldots$ is actually an antichain over $X_{\perp x_0}$. This suffix is now amortised $(g, g(n_0))$-controlled, and of length bounded by $W^2_{g,X_{\perp x_0}}(g(n_0))$. This yields the following descent equation when considering all the possible amortised $(g, n_0)$-controlled antichains.

**Lemma 4** (Descent Equation). Let $(X, \leq_X, |.|_X)$ be a nwqo, $n_0 \in \mathbb{N}$ and $g : \mathbb{N} \rightarrow \mathbb{N}$. Then $W^2_{g,X}(n_0) = \max_{x \in X_{\leq n_0}} 1 + W^2_{g,X_{\perp x}}(g(n_0)).$

**Proof.** Any amortised $(g, g(n_0))$-controlled antichain $x_1, x_2, \ldots$ over $X_{\perp x}$ can be prefixed with any $x$ such that $|x|_X \leq n_0$ to yield an amortised $(g, n_0)$-controlled antichain $x, x_1, x_2, \ldots$, thus $W^2_{g,X}(n_0) \geq \max_{x \in X_{\leq n_0}} 1 + W^2_{g,X_{\perp x}}(g(n_0)).$

Conversely, let us pick an amortised $(g, n_0)$-controlled antichain $x_0, x_1, x_2, \ldots$ of maximal length $W^2_{g,X}(n_0)$; such a maximal antichain exists as discussed in Section 4.1. Then $x_0 \in X_{\leq n_0}$ and for all $i > 0, x_i \in X_{\perp x_0}$ and $|x_i|_X \leq g^{-1}(g(n_0))$, thus $W^2_{g,X}(n_0) \leq \max_{x \in X_{\leq n_0}} 1 + W^2_{g,X_{\perp x}}(g(n_0)).$ ▶

**Reflecting Normed wqos.** The descent equation, though it offers a way of computing the width function, quickly leads to complex residual expressions. We are going to over-approximate these $X_{\perp x}$’s using nwqo reflections, so that the computation can be carried out without leaving the realm of strict polynomial nwqos, leading to an inductive over-approximation of $X_{\perp x}$ over the structure of the strict polynomial nwqo $X$. 
A \textit{nwqo reflection} \cite{41} is a mapping $r: X \rightarrow Y$ between two nwqos $(X, \leq_X, |\cdot|_X)$ and $(Y, \leq_Y, |\cdot|_Y)$ that satisfies the two following properties:

\begin{align*}
\forall x, x' \in X. & \quad r(x) \leq_Y r(x') \implies x \leq_X x', \quad (15) \\
\forall x \in X. & \quad |r(x)|_Y \leq |x|_X. \quad (16)
\end{align*}

In other words, a nwqo reflection is an order reflection that is not norm-increasing. This induces a quasi-ordering between nwqos, written $X \hookrightarrow Y$. Remark that reflections are compatible with disjoint sums and products \cite[Prop. 3.5]{41}:

\[ X_0 \hookrightarrow Y_0 \text{ and } X_1 \hookrightarrow Y_1 \implies X_0 \sqcup X_1 \hookrightarrow Y_0 \sqcup Y_1 \text{ and } X_0 \times X_1 \hookrightarrow Y_0 \times Y_1 . \quad (17) \]

Crucially, nwqo reflections preserve amortised controlled antichains. Indeed, let $r: X \hookrightarrow Y$, and consider a sequence $x_0, x_1, \ldots$ over $X$. Then by (15), $r(x_0), r(x_1), \ldots$ is an antichain when $x_0, x_1, \ldots$ is, and by (16), it is $(g, n)$-controlled when $x_0, x_1, \ldots$ is. Hence

\[ X \hookrightarrow Y \implies W^g_{\sigma}(n) \leq W^g_{\tau}(n) \text{ for all } g, n. \quad (18) \]

\textbf{Remark 5.} By contrast with amortised controlled antichains, nwqo reflections to not preserve strongly controlled antichains. Consider for instance the strongly $(H, 4)$-controlled antichain $(4, 3) (5, 2) (6, 1) (7, 0)$ over $\mathbb{N}^2$ where $H(x) \equiv x + 1$. Let \{e, o\} denote a set with two incomparable elements of norm zero, and consider the nwqo reflection $r: \mathbb{N}^2 \rightarrow (\mathbb{N}^2 \times \{e, o\})$ defined by $r(2n, m) \equiv (2n, m, e)$ and $r(2n + 1, m) \equiv (2n, m, o)$. The image of our antichain is the antichain $(4, 3, e) (4, 2, o) (6, 1, e) (6, 0, o)$, but it is not strongly $(H, 4)$-controlled because $|(4, 2, o)| + 1 = 5 < 6 = |(6, 1, e)|$.

\textbf{Inductive Reflection of Residuals.} We provide a strict polynomial wqo reflecting $X_{\perp x}$ by induction over the structure of the strict polynomial nwqo $X$. The key difference compared to the analysis of bad sequences in \cite{15, 42} occurs for $X = \mathbb{N}$: if $k \in \mathbb{N}$,

\[ \mathbb{N}_{\perp k} = \{ 0 \} . \quad (19) \]

Regarding disjoint sums $X_0 \sqcup X_1$, it is plain that

\[ (X_0 \sqcup X_1)_{\perp(i, j)} = (X_i)_{\perp x} \sqcup X_{1-i} . \quad (20) \]

Consider now $(\mathbb{N}^d)_{\perp u}$ where $d > 1$ and $u \in \mathbb{N}^d$. Observe that if $u \in \mathbb{N}^d$ is such that $u \perp v$, then there exists $1 \leq i \leq d$ such that $u(i) < v(i)$, as otherwise we would have $u \geq v$. Thus

\[ (\mathbb{N}^d)_{\perp u} \hookrightarrow \bigsqcup_{1 \leq i \leq d} \mathbb{N}^{d-1} \times [v(i)] \hookrightarrow \mathbb{N}^{d-1} \cdot \sum_{1 \leq i \leq d} v(i) . \quad (21) \]

\textbf{Ordinal Notations.} As it is more convenient to reason with ordinals than with polynomial nwqos, we use the following bijection between strict polynomial nwqos and $\omega^\omega$:

\[ w(0) \equiv 0 , \quad w(\mathbb{N}^d) \equiv \omega^{d-1} , \quad w(X_0 \sqcup X_1) = w(X_0) \oplus w(X_1) . \quad (22) \]

where `$\oplus$' denotes the natural sum (aka \textit{Hessenberg sum}) on ordinals: the natural sum $\alpha \oplus \beta$ of two ordinals with Cantor normal forms $\alpha = \sum_{i=1}^p \omega^{\gamma_i}$ and $\beta = \sum_{j=1}^m \omega^{\delta_j}$ is $\omega^{\gamma_1 + \cdots + \gamma_p + \delta_1 + \cdots + \delta_m}$ where the exponents $\gamma_1 \geq \cdots \geq \gamma_p + m$ are a reordering of $\alpha_1, \ldots, \alpha_p, \beta_1, \ldots, \beta_m$. Given a strict polynomial nwqo $X = \bigsqcup_{i=1}^m \mathbb{N}^{d_i}$, its associated ordinal is $w(X) = \bigoplus_{i=1}^m \omega^{d_i-1}$. In the case of an LCM with $d \equiv |C|$ counters and $q \equiv |Q|$ locations, $w(Q \times \mathbb{N}^q) = \omega^{d-1} \cdot q$. 

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For each \( n \in \mathbb{N} \), we define a relation \( \partial_n \) over ordinals in \( \omega^\omega \) that mirrors the inductive residuation and reflection operations on strict polynomial nwqos \( X \) over the ordinals \( w(X) \):

\[
\partial_n \alpha \triangleq \{ \gamma \oplus \partial_n \omega^d \mid \alpha = \gamma \oplus \omega^d \}, \quad \partial_n \omega^d \triangleq \begin{cases} 0 & \text{if } d = 0, \\ \omega^{d-1} \cdot n(d + 1) & \text{otherwise.} \end{cases}
\]

(23)

The intuition here is that \( w(Y) \in \partial_n w(X) \) implies \( X_{\perp x} \rightarrow Y \) for some \( x \in X_{\leq n} \) (see Claim 6.1). Observe that \( \alpha' \in \partial_n \alpha \) implies \( \alpha' < \alpha \), thus \( \bigcup_n \partial_n \) is a well-founded relation. This leads to the definition of an over-approximation of the width function \( W_{g,X} \):

\[
M_{g,\alpha}(n) \triangleq \max_{\alpha' \in \partial_n \alpha} \{ 1 + M_{g,\alpha'}(g(n)) \}.
\]

(24)

**Proposition 6.** Let \( (X, \leq X, \cdot, |X|) \) be a strict polynomial nwqo, \( n_0 \in \mathbb{N} \), and \( g : \mathbb{N} \rightarrow \mathbb{N} \). Then \( W_{g,X}^2(n_0) \leq M_{g,w(X)}(n_0) \).

**Proof.** The derivation relation \( \partial_n \) over ordinals was designed to satisfy the following.

**Claim 6.1.** Let \( X \) be a strict polynomial nwqo and \( x \in X_{\leq n} \) for some \( n \). Then there exists a strict polynomial nwqo \( Y \) such that \( w(Y) \in \partial_n w(X) \) and \( X_{\perp x} \rightarrow Y \).

**Proof.** Let \( X = N^{d_1} \sqcup \cdots \sqcup N^{d_m} \) for some \( d_1, \ldots, d_m > 0 \) and \( x \in X_{\leq n} \); note that the existence of \( x \) rules out the case of \( m = 0 \). We proceed by induction over \( m > 0 \).

\(<\)

The base case is \( m = 1 \), i.e., \( X = N^d \) for some \( d > 0 \).

\(<\)

- If \( d = 1 \), then \( X = N \) and \( w(X) = \omega^0 \). Thus \( X_{\perp x} = \varnothing \) by (19) and \( \partial_n w(X) = \{ 0 \} \) by (23), and indeed \( w(0) = 0 \).

- Otherwise if \( d > 0 \), then \( w(X) = \omega^{d-1} \) and \( x \) is a vector \( v \in (N^d)_{\leq n} \). Equation (21) shows that \( X_{\perp x} \rightarrow N^{d-1} \cdot \sum_{1 \leq i \leq d} n_i = N^{d-1} \cdot nd \), while \( \partial_n w(X) = \{ \omega^{d-2} \cdot nd \} \) by (23), and indeed \( w(N^{d-1} \cdot nd) = \omega^{d-2} \cdot nd \).

\(<\)

For the induction step with \( m > 1 \), i.e., when \( X = Y_0 \sqcup Y_1 \) and \( w(X) = w(Y_0) \oplus w(Y_1) \) for two non-empty strict polynomial nwqos \( Y_0 \) and \( Y_1 \), then \( x = (y, i) \) for some \( i \in \{ 0, 1 \} \) and \( y \in (Y_i)_{\leq n} \). Equation (20) shows that \( X_{\perp x} \rightarrow (Y_i)_{\perp y} \sqcup Y_{1-i} \). By induction hypothesis, there exists \( Z_i \) such that \( w(Z_i) \in \partial_n w(Y_i) \) and \( (Y_i)_{\perp y} \rightarrow Z_i \), and the latter implies \( X_{\perp x} \rightarrow Z_i \sqcup Y_{1-i} \) by (17). By (23), \( w(Z_i) \sqcup Y_{1-i} = w(Z_i) \oplus w(Y_{1-i}) \in \partial_n w(X) \) as desired.

Let us return to the proof of Proposition 6. Either \( X_{\leq n_0} \) is empty and then \( W_{g,X}^2(n_0) = 0 \leq M_{g,w(X)}(n_0) \), or there exists some \( x \in X_{\leq n_0} \) that maximises \( W_{g,X_{\perp x}}(g(n_0)) \) in the descent equation from Lemma 4, i.e., such that

\[
W_{g,X}^2(n_0) = 1 + W_{g,X_{\perp x}}^2(g(n_0)).
\]

By Claim 6.1 there exists \( Y \) such that \( w(Y) \in \partial_n w(X) \) and \( X_{\perp x} \rightarrow Y \). By (18),

\[
W_{g,X}^2(n_0) \leq 1 + W_{g,Y}^2(g(n_0)).
\]

By well-founded induction on \( w(Y) \in \partial_n w(X) \),

\[
W_{g,Y}(g(n_0)) \leq M_{g,w(Y)}(g(n_0)).
\]

Thus by definition of \( M_{g,w(X)} \) in (24),

\[
W_{g,X}^2(n_0) \leq 1 + M_{g,w(Y)}(g(n_0)) \leq M_{g,w(X)}(n_0).
\]
It remains to compare $M_{g,\alpha}$ with standard subrecursive functions like the Cichoń functions, which was already done in [42, Sec. 2.4.3] for a very similar function.

**Proposition 7.** Let $n$ be in $\mathbb{N}$, $g: \mathbb{N} \to \mathbb{N}$ a monotone inflationary function, and $0 < \alpha < \omega^d$. Then $M_{g,\alpha}(n) \leq 1 + M_{g,\omega^d}(n)$.

**Proof.** The function $M_{g,\alpha}(n)$ is very nearly the same as the one studied in [42, Sec. 2.4.3], except for the coefficient $n(d+1)$ instead of $nd$ in the definition of $\partial_n \omega^d$ in (23). This minor difference turns out to have no impact on several of the arguments used in [42, Sec. 2.4.3], which we summarise here.

Let us start by introducing a *structural ordering* for ordinals in $\omega^\omega$ in Cantor normal form [42, Eq. 2.69]:

$$\omega^{d_1} + \cdots + \omega^{d_m} \sqsupseteq \omega^{d'_1} + \cdots + \omega^{d'_n} \quad \text{if} \quad m \leq k \text{ and } \forall 1 \leq i \leq m, d_i \leq d'_i.$$  \hspace{1cm} (25)

Clearly, for all $\alpha, \gamma \in \omega^\omega$,

$$\alpha \sqsubseteq \alpha \oplus \gamma.$$  \hspace{1cm} (26)

The $M_{g,\alpha}$ functions are monotone with respect to the structural ordering: if $\alpha$ and $\alpha'$ are two ordinals in $\omega^\omega$, then [42, Lem. 2.31]

$$\alpha \sqsubseteq \alpha' \quad \text{implies} \quad M_{g,\alpha}(n) \leq M_{g,\alpha'}(n).$$  \hspace{1cm} (27)

This allows to prove that there exists a ’maximising strategy’ when choosing how to decompose $\alpha = \omega^{d_1} + \cdots + \omega^{d_m}$ as $\gamma \oplus \omega^{d_i}$ (where $\gamma = \sum_{j \neq i} \omega^{d_j}$), which is to pick $i = m$, i.e., a term with minimal exponent. Indeed, if $\alpha = \gamma + \omega^d$, then [42, Lem. 2.32]

$$M_{g,\alpha}(n) = 1 + M_{g,\gamma + \partial_n \omega^d}(g(n)).$$  \hspace{1cm} (28)

We can now prove Proposition 7 by adapting the arguments of [42, Cor. 2.33]. Let us write $\alpha$ in Cantor normal form as $\gamma + \omega^{d' + d}$ for some $\gamma < \alpha$ and $d' < d$. By (28), $M_{g,\alpha}(n) = 1 + M_{\gamma + \partial_n \omega^{d'}}(g(n))$. If $d' = 0$, then by (23),

$$\gamma + \partial_n \omega^0 = \gamma = P_{nd}(\gamma + 1) = P_{nd}(\alpha)$$

and the statement holds. Otherwise, by (23),

$$\gamma + \partial_n \omega^{d'} = \gamma + \omega^{d' - 1} \cdot n(d' + 1)$$

$$\sqsubseteq \gamma + \omega^{d' - 1} \cdot nd \quad \text{by (26) since } d' < d$$

$$\sqsubseteq \gamma + \omega^{d' - 1} \cdot nd + P_{nd}(\omega^{d' - 1}) \quad \text{by (26)}$$

$$= P_{nd}(\gamma + \omega^{d'}) \quad \text{by definition of predecessors in (11–12)}$$

$$= P_{nd}(\alpha),$$

and the statement holds by (27). ▶

This extra twist of using a predecessor function different from the standard one from (12) can be avoided by instead over-approximating the control function $g$.

**Theorem 8 (Width Function for Strict Polynomial nwqos).** Let $d > 0$, $(X, \leq, \cdot, |)$ be a strict polynomial nwqo with $w(X) < \omega^d$, $n_0 \in \mathbb{N}$, $g: \mathbb{N} \to \mathbb{N}$ monotone inflationary, and let $h: \mathbb{N} \to \mathbb{N}$ be a monotone function such that $h(x \cdot d) \geq g(x) \cdot d$ for all $x$. Then $W^2_{g,X}(n_0) \leq h(w(X)(n_0d))$.  


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Proof. By Proposition 6, it suffices to show that $M_{g,w(X)}(n) \leq h_w(X)(nd)$, which we do by induction over $\alpha \leq w(X)$. If $\alpha = 0$, then $d_\alpha = \emptyset$ thus $M_{g,\alpha}(n) = 0 \leq h_n(nd)$. Otherwise, by Proposition 7, $M_{g,\alpha}(n) \leq 1 + M_{g,\alpha'}(g(n))$. Because $P_{nd}(\alpha) < \alpha$, we can apply the induction hypothesis, yielding $M_{g,\alpha}(n) \leq 1 + h_{P_{nd}(\alpha)}(g(n)d) \leq 1 + h_{P_{nd}(\alpha)}(h(nd)) = h_n(nd)$, where the last inequality follows from $h(nd) \geq g(n)d$ and the monotonicity of $h_{P_{nd}(\alpha)}$. ▶

Setting $h(x) \equiv g(x)d$ always satisfies the conditions of the theorem. There are cases where setting $h \equiv g$ suffices: e.g., $g(x) \equiv 2x$, $g(x) \equiv x^2$, $g(x) \equiv 2^x$, and more generally whenever $g$ is super-homogeneous, i.e. satisfies $g(dx) \geq g(x)d$ for all $d$, $x \geq 1$. In the case of LCMs, where $w(Q \times \mathbb{N}) < \omega^\alpha$ if $d \geq |C| > 0$, a control function $g(x) \equiv x + 1 = H(x)$ fits, thus setting $h(x) \equiv x + d = H^d(x)$ satisfies $h(dx) = dx + d = (x + 1)d = g(x)d$.

By (8), Lemma 3, and (14), Theorem 8 also yields a bound on the strong norm function.

> Corollary 9 (Strong Norm Function for Strict Polynomial nwqos). Let $d$, $X$, $n_0$, $g$, and $h$ be as in Theorem 8. Then $N_{g,X}^{\beta}(n_0) \leq h_w(X)(n_0d)$.

6 Wrapping up

We have now all the ingredients needed to prove an $F_d$ upper bound on LCM Reachability. Let us first recall the definition of the fast-growing complexity classes from [39].

Fast-Growing Complexity Classes. The fast-growing complexity classes [39] form a strict ordinal-indexed hierarchy of complexity classes $(F_\alpha)_{\alpha < \omega}$ using the Hardy functions $(H^\alpha)_{\alpha < \omega}$ relative to $H(x) \equiv x + 1$ as a standard against which to measure high complexities. Let

$$F_\alpha \equiv \bigcup_{\beta < \omega^\alpha} \text{DTIME}(H^\beta(p(n))), \quad F_d \equiv \bigcup_{p \in F_\alpha} \text{TIME}(H^{\omega^\alpha}(p(n))). \quad (29)$$

Then $F_\alpha$ is the class of functions computed by deterministic Turing machines in time $O(H^\beta(n))$ for some $\beta < \omega^\alpha$; this captures for instance the class of Kalmar elementary functions as $F_{<3}$ and the class of primitive-recursive functions as $F_{<\omega}$ [31, 49]. The class $F_\alpha$ is the class of decision problems solved by deterministic Turing machines in time $O(H^{\omega^\alpha}(p(n)))$ for some function $p \in F_\alpha$. The intuition behind this quantification of $p$ is that, just like e.g. EXP = $\bigcup_{p \in \text{polyDTIME}(2^{p(n)})}$ quantities over polynomial functions to provide enough ‘wiggle room’ to account for polynomial reductions, $F_\alpha$ is closed under $F_\alpha$ reductions [39, Thms. 4.7 and 4.8]. For instance, TOWER $\equiv F_3$ defines the class of problems that can be solved in time bounded by a tower of exponentials of elementary height in the size of the input, $\bigcup_{k \in \mathbb{N}} F_k$.
is the class of primitive-recursive decision problems, and \( \text{ACKERMANN} \cong F_d \) is the class of problems that can be solved in time bounded by the Ackermann function applied to some primitive-recursive function of the input size; see Figure 3 for a depiction.

**Upper Bound.** Recall from Section 3.3 that a pseudo-witness for coverability of a configuration \( q_f(v_f) \) in a LCM with \( d \cong |C| > 0 \) counters and \( q \cong |Q| \) locations is a strongly \((H, |v_f|)\)-controlled bad sequence over \( Q \times \mathbb{N}^d \), which as discussed in Section 5.2 is a strict polynomial \( wqo \) with \( w(Q \times \mathbb{N}^d) = \omega^{d-1} \cdot q < \omega^d \), and that \( h \cong H^d \) fits the conditions of Theorem 8 and Corollary 9. As stated in Theorem 1, together with the lower bounds from [40], the following entails the \( F_d \)-completeness of LCM Reachability with a fixed number \( d \geq 3 \) of counters.

\[ \begin{align*}
\text{Proposition 10 (Upper Bound for LCM Reachability).} & \quad \text{LCM Reachability is in } F_d, \text{ and in } F_d \text{ if the number } d \geq 3 \text{ of counters is fixed.} \\
\text{Proof.} & \quad \text{Let } n_0 \cong |v_f| \text{ be the infinite norm of the target configuration, } d \cong |C| \geq 3 \text{ be the number of counters, and } q \cong |Q| \geq 1 \text{ the number of locations. By Corollary 9, the elements in a pseudo-witness of the coverability of } q_f(v_f) \text{ are of norm at most } N \cong N_{H,Q \times \mathbb{N}^d}(n_0) = h^{\omega^{d-1} \cdot q}(n_0) \text{ for } h(x) \cong H^d(x). \text{ Let } n \cong \max\{qd - 1, n_0\}. \text{ As shown in Lemma 11 in Appendix A, this means that} \\
N & \leq H^{\omega^{d-1} \cdot qd}(n_0) \leq H^{\omega^{d-1} \cdot qd}(n) \leq H^{\omega^d}(n) \tag{30} \\
\text{by monotonicity of the Hardy functions.} & \\
\text{Note that there are at most } q(N + 1)^d \text{ different configurations in } Q \times \mathbb{N}^d \text{ of norm bounded by } N, \text{ i.e., } |(Q \times \mathbb{N}^d)_{\leq N}| \leq q(N + 1)^d. \text{ By (9), this is also a bound on the strong length function } L_{H,Q \times \mathbb{N}^d}(n_0). \text{ Thus the number of steps in the backward coverability algorithm is bounded by } q(N + 1)^d, \text{ and each step can be carried in time } O(N), \text{ hence the algorithm works in deterministic time } O(q(N + 1)^{d+1} = O(f(N)) = O(f(H^{\omega^d}(n_0))) \text{ for some elementary function } f \in \mathcal{F}_{<3}. \text{ By [39, Cor. A.9], there exists an elementary inflationary function } p \in \mathcal{F}_{<3} \text{ such that } f(H^{\omega^d}(n)) \leq H^{\omega^d}(p(n)); \text{ the backward coverability algorithm therefore works in deterministic time } O(H^{\omega^d}(p(n))) \text{ for some } p \in \mathcal{F}_{<3}, \text{ which is an expression of the form (29).} \\
\text{Therefore, LCM Reachability is in } F_d \text{ when } d \text{ is fixed, and in } F_d \text{ otherwise because } p(n) \geq n \geq d - 1 \text{ and thus } H^{\omega^d}(p(n)) \leq H^{\omega^d}(p(n)). \quad \blacksquare 
\end{align*} \]

7 **Concluding Remarks**

We have shown the \( F_d \)-completeness of reachability in lossy counter machines with a fixed number \( d \geq 3 \) of counters. The key novelty is that we analyse the length of controlled antichains over \( \mathbb{N}^d \) rather than that of controlled bad sequences. A possible explanation why this leads to improved upper bounds is that the ordinal width of \( \mathbb{N}^d \), i.e., the ordinal rank of its antichains, is conjectured to be \( \omega^{d-1} \) [13], while its maximal order type, i.e., the ordinal rank of its bad sequences, is well-known to be \( \omega^d \) [9].

Our approach might be employed to tackle related parameterised complexity gaps, like the one between \( F_{m-2} \)-hardness [26] and \( F_{m-1+1} \) membership [41] of reachability in *lossy channel systems* with \( m \geq 4 \) channel symbols and a single channel. Those results rely however on the set of finite words over an alphabet of size \( m \) being a \( wqo \) for Higman’s scattered subword ordering [22], for which the ordinal width and maximal order type coincide at \( \omega^{m-1} \) [13, 9].
The Hardy functions enjoy some nice identities (which justify the superscript notation): for all $h$, $\alpha$, $\beta$, and $x$

$$h^\alpha \circ h^\beta(x) = h^{\alpha+\beta}(x), \quad (h^\alpha)^\beta(x) = h^{\alpha\cdot\beta}(x), \quad (31)$$

provided $\alpha + \beta$ (resp. $\alpha \cdot \beta$) satisfies specific conditions like being \textit{tree-structured} \cite{14,7}; this will the case in our applications of these identities. The condition on $\alpha \cdot \beta$ is necessary; for instance, (31) fails for $(H^2)^\omega(x) = (H^2)x(x+2) = 3x+2$ because $2 \cdot \omega = \omega$ and $H^\omega(x) = 2x+1$.

Lemma 11 circumvents this limitation in a special case. Let us first mention that the Hardy functions $(h^\alpha)_\alpha$ can also be defined by transfinite induction on $\alpha$ by

$$h^0(x) \triangleq x, \quad h^{\alpha+1}(x) \triangleq h^\alpha(h(x)), \quad h^\lambda(x) \triangleq h^{\lambda}(x). \quad (32)$$

\begin{lemma}
Let $\alpha \in \varepsilon_0$, $d \in \mathbb{N}$, $x \in \mathbb{N}$, and $h: \mathbb{N} \to \mathbb{N}$ be monotone inflationary. Then $(h^d)^\alpha(x) \leq h^{\alpha\cdot d}(x)$.
\end{lemma}

\textbf{Proof.} Let us first show by induction over $d \in \mathbb{N}$ that for all limit ordinals $\lambda < \varepsilon_0$,

$$x \leq y \quad \text{implies} \quad h^{\lambda d}(y) \leq h^{\lambda d}(y). \quad (33)$$

For the base case $d = 0$, $h^{\lambda d}(y) = y = h^{\lambda d}(y)$. For the induction step $d + 1$,

$$h^{\lambda d}(y) = h^{\lambda d}(h^{\lambda d}(y)) \quad \text{by (31)}$$

$$\leq h^{\lambda d}(h^{\lambda d}(y)) \quad \text{since} \quad h^{\lambda d}(y) \leq h^{\lambda d}(y) \quad \text{by monotonicity of} \quad h^{\lambda d}(y)$$

$$\leq h^{\lambda d}(h^{\lambda d}(y)) \quad \text{by ind. hyp. since} \quad x \leq y \leq h^{\lambda d}(y)$$

$$= h^{\lambda d+d}(y) \quad \text{by (31)}.$$

Let us now prove Lemma 11 by transfinite induction over $\alpha$. For the base case $\alpha = 0$,

$$(h^d)^0(x) = x = h^{0\cdot d}(x).$$

For the successor case $\alpha + 1$,

$$(h^d)^{\alpha+1}(x) = (h^d)^\alpha(h^d(x)) \quad \text{by (32)}$$

$$\leq h^{\alpha d}(h^d(x)) \quad \text{by ind. hyp. on} \quad \alpha < \alpha + 1$$

$$\leq h^{\alpha d+d}(x) \quad \text{by ind. hyp. on} \quad \alpha < \alpha + 1$$

$$= h^{(\alpha+1) d}(x).$$

For the limit case $\lambda$,

$$(h^d)^\lambda(x) = (h^d)^\lambda(x) \quad \text{by (32)}$$

$$\leq h^{\lambda d}(x) \quad \text{by ind. hyp. on} \quad \lambda(x) < \lambda$$

$$\leq h^{\lambda d}(x) \quad \text{by (33)}.$$

\section*{References}


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