

Aalto University
School of Science
Bachelor's Programme in Science and Technology

Detecting multilinear monomials with algebraic fingerprints

Bachelor's Thesis

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ABSTRACT OF
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1 Introduction

In recent years, there have been rapid advances in the algorithms for combinatorial problems. This has been greatly sparked by the development in algebraic techniques for solving the multilinear monomial detection problem, i.e., finding whether a multivariate polynomial contains a multilinear monomial, first introduced by Koutis in [Kou05], where the set packing problem is reduced to multilinear monomial detection.

Namely, the technique of algebraic fingerprinting, first introduced in [Kou08] and further developed in [Wil09], has found great success for many combinatorial problems. For example, with algebraic fingerprinting, the k -path problem that previously could be solved in $\mathcal{O}^*(4^k)$ time by Chen et al. in [Che+07], could be solved in $\mathcal{O}^*(2^{3k/2})$ time in [Kou08]. This result was quickly improved in [Wil09], where an $\mathcal{O}^*(2^k)$ algorithm was given.

Of course, this technique was further developed, and soon after in [Bjö14] Björklund et al. showed an algorithm that solved the Hamiltonian problem (Hamiltonicity), i.e., finding whether a given graph contains a simple path that visits every vertex, in $\mathcal{O}^*(1.657^n)$ time. Soon enough, for k -path, an $\mathcal{O}^*(1.66^k)$ algorithm was found [Bjö+17]. The fastest algorithms for Hamiltonicity before this ran in $\mathcal{O}^*(2^n)$ and were known since 1962 [HK62], [Bel62]. This was a significant improvement on a problem that had seen no progress in nearly fifty years.

1.1 Research goals and thesis structure

AM: This section can be shortened to one paragraph

The goals of this thesis are to find out how multilinear monomial detection is relevant in combinatorial problems, and how algebraic fingerprints can be utilized to design faster algorithms for problems that use multilinear monomial detection. Also, interesting ideas regarding algebraic fingerprints are explored for.

Multilinear monomial detection is a fundamental problem, since many important combinatorial problems can be reduced into it via a problem specific algebraization. Thus, faster algorithms and new ideas for multilinear monomial detection are important.

Multilinear monomial detection is essentially searching for solutions among non-solutions, both of which are encoded as monomials in a polynomial. The technique of algebraic fingerprinting is present in multilinear monomial detection. With algebraic fingerprints, unwanted cancelation of solution monomials due to the characteristic of a field can be prevented. Moreover, algebraic fingerprints can be used to cancel non-solution monomials by abusing the characteristic.

AM: What is the meaning of $\mathcal{O}^*(\cdot)$?

AM: Note use of \citeauthor{Björklund2014}, AM: Did Björklund really develop the technique further? or did he only show how to apply it to another problem? AM: n or k ?

AM: No need for this since you already have ToC

In the next subsections, the thesis discusses algebraization and reduction into multilinear monomial detection. The section 2 covers preliminaries. In section 3, the thesis discusses general multilinear monomial detection. In section 4, some problem specific instances of multilinear monomial detection are given, and clever utilizations of algebraic fingerprints are shown. Section 5 concludes the thesis.

1.2 Algebraization of combinatorial problems

A combinatorial problem asks whether a given finite set of objects satisfies some given constraints. For example, the k -path problem asks for, given a finite set of vertices and edges, a simple path of k vertices. The solutions and non-solutions (solution space) to combinatorial problems can be thought of as combinations of the given objects. The solution space for the k -path problem consists of combinations of k vertices and $k - 1$ edges. A non-solution combination would contain duplicate vertices, or edges that contain vertices outside the combination.

AM: This sounds like a very simple problem... But I guess the catch is that we want an algorithm with complexity depending only on k ?

Algebraization is reducing a given problem into an algebraic form, i.e. AM: that is, a question regarding some algebraic property of some algebraic entity. In an algebraization of a combinatorial problem, the algebraic entity can be constructed from algebraic elements defined from the set of objects given as an input. The motivation behind the construction is some algebraic property that, when satisfied, gives a solution to the problem.

In [Val92], it was observed that multivariate polynomials in certain algebras have natural combinatorial interpretations. Utilizing this idea, [Kou05] managed to reduce a combinatorial problem into an algebraic form, that is, multilinear monomial detection. First, we introduce multiple variables that correspond to elements from the set of objects given as input. Then, we construct an arithmetic circuit representing a multivariate polynomial, such that it encodes all solutions and non-solutions as multivariate monomials, with multilinear monomials corresponding to solutions. Thus, the task of finding a satisfying combination to the combinatorial problem has been reduced to finding a multilinear monomial from the multivariate polynomial. It follows that a decision problem is answered by the existence of a multilinear monomial, and a counting problem by the number of multilinear monomials.

Appropriate definitions for the variables are problem-specific. In the following section, this thesis gives a reduction into multilinear monomial detection, shown in [KW15], for the k -3D matching problem. Another simple example can be found, for the set packing problem, in [Kou05].

AM:
Algebraization

AM:
Probably one
can drop the
vertices and
think only
about edges?

AM:
Somewhere
we need to
reference the
reader to
Sect. 2 for
non-familiar
terms like
multivariate,
algebras, etc.

AM: I would merge the two sections and use the k -3D matching as an example to the explanation above

1.3 Reducing k -3D matching into multilinear monomial detection

The k -3D matching problem is defined as follows:

k -3D MATCHING

Input: Three disjoint sets A , B and C , and a set of triples $T \subset A \times B \times C$.

Question: Is there a subset $M \subseteq T$, such that $|M| = k$ and $\forall m \in M$: None of the elements in m appear in $M \setminus \{m\}$ AM: $M \setminus \{m\}$

We begin by defining new variables corresponding to the elements in A , B and C , labeled as a_i , b_j and c_k , respectively, where $i \in [|A|]$, $j \in [|B|]$ and $k \in [|C|]$.

For every triple $t \in T$, we define a multilinear monomial x that is a product of the elements in t . We introduce a set X that satisfies the following:

$$\forall x \in X : x = abc : (a, b, c) \in T.$$

AM: Probably meant $\forall x \in X, x = abc \dots$?

Next, we define multivariate polynomials P_1 and P_k as follows:

$$P_1 = \sum_X, P_k = P_1^k.$$

Following this construction, we observe that P_k , when expanded into a sum of multivariate monomials, contains a multilinear term if and only if the original k -3D matching instance can be answered in the positive. Furthermore, every multilinear monomial in the expanded P_k corresponds to a solution to the problem, and the solutions can be directly found from the variables in the multilinear monomial. Thus, a successful reduction into multilinear monomial detection has been given for the k -3D mapping.

An example instance of k -3D matching with this exact algebraization can be found in [KW15]. [show the example?]

2 Preliminaries

It is necessary to recall basic algebraic concepts before further discussing multilinear monomial detections and algebraic fingerprinting. This section gives definitions for a group, ring and field, and some useful concepts regarding them. The second subsection goes through other notation and terminology used throughout the thesis.

AM: Use \setminus instead of \backslash
AM: $T = A \times B \times C$ not allowed?
AM: Either convert \forall into text or what follows into maths
AM: Def. of $|\cdot|$?
AM: Over what object are we doing multiplication?
AM: Use $\lfloor \dots \rfloor$ for display math
AM: Meaning of \sum_X ?

2.1 Groups, rings and fields

A group \mathbf{G} is a tuple $(G, +)$, where G is a set of elements, $+: G \times G \rightarrow G$ is a binary operation closed under the elements in G , $+$ is associative, every element $g \in G$ has an inverse $g^{-1} \in G$, and G contains an identity element e such that $g + e = g$, $g + g^{-1} = e$ and $e = e^{-1}$. Moreover, \mathbf{G} is called *Abelian* if $+$ is also commutative.

AM: There is also $\mathbf{\mathbb{Z}}$
AM: See TeX source
AM: I think $e = e^{-1}$ follows from the other requirements (but that is not so important)

A ring \mathbf{R} is a tuple $(R, +, \cdot)$, where $R = (G, +)$ is an Abelian group, $\cdot : G \times G \rightarrow G$ is a binary operation closed under G . We call the binary operations $+$ and \cdot addition and multiplication, respectively. Note, that from here on we use R as the set of elements defined for \mathbf{R} . In general, a bold typeface \mathbf{X} represents a group, ring or field and X its set of elements. R must contain a multiplicative identity $\mathbf{1} \in R$ such that $\forall a \in R: a \cdot \mathbf{1} = a$. We notate the additive identity e required for the group as $\mathbf{0}$ from here on. Observe, that for any $R \neq \{\mathbf{0}\}$, $\mathbf{1} \neq \mathbf{0}$. Left and right distributive laws hold for rings, i.e.,

$$\forall a, b, c \in R: a \cdot (b + c) = (a \cdot b) + (a \cdot c) \wedge (b + c) \cdot a = (b \cdot a) + (c \cdot a).$$

$u \in R$ is called *unit* if it holds that $\exists v \in R: u \cdot v = v \cdot u = \mathbf{1}$, i.e., it has a multiplicative inverse $v \in R$.

A field $\mathbf{F} = (F, +, \cdot)$ is defined with the following conditions:

- $(F, +)$ is an Abelian group
- $(F \setminus \{\mathbf{0}\}, \cdot)$ is an Abelian group
- Left and right distributive laws hold for \mathbf{F}

Equivalently, a ring is a field if every non-zero element is unit, $\mathbf{1} \neq \mathbf{0}$, and multiplication is commutative. The *characteristic* of a field \mathbf{F} is defined as follows:

$$\text{char}(\mathbf{F}) = \begin{cases} \min\{n \in \mathbb{N} : n \cdot \mathbf{1} = \mathbf{0}\} \\ 0 \end{cases} \quad \text{if such } n \text{ does not exist} \quad (1)$$

Note, that a field \mathbf{F} with characteristic 2 satisfies the following:

$$\forall u \in F: u + u = u \cdot (\mathbf{1} + \mathbf{1}) = u \cdot \mathbf{0} = \mathbf{0}$$

AM: characteristic

TODO: group algebra, polynomial ring, linear dependency

2.2 Notation and other terminology

TODO: create a table or like, list terms: multilinearity, multivariety, sum of monomials form & generating form (arithmetic circuit) of polynomial, degree of multivariate monomial,

\mathcal{O} , Θ , FPT, \mathcal{O}^* , determinism & non-determinism [or use 'Monte Carlo' :)], Schwartz-Zippel lemma

AM: Standard \mathcal{O} and Θ notation should be known

3 General multilinear monomial detection

The detection of multilinear monomials is a fundamental problem, since many important problems can be reduced to it [TODO: quick examples (just refs?)]. Therefore, any progress in general multilinear monomial detection directly implies faster algorithms for all problems, that are reduced to and solved with general multilinear monomial detection.

The general, parameterized multilinear monomial problem is defined as follows:

k -MULTILINEAR MONOMIAL DETECTION

Input: A commutative arithmetic circuit A over a set of variables X representing a polynomial $P(X)$.

Question: Does the polynomial $P(X)$ extended as a sum of monomials contain a multilinear monomial of degree k ?

A naive expansion of A into $P(X)$ and evaluation of $P(X)$ is not optimal: an N -degree polynomial will have $2^{\Theta(N)}$ possible monomials, and most problems, that can use this algebraization, have been solved with faster algorithms (TODO: quick example?). This motivates the detection of multilinear monomials without fully expanding the polynomial into a sum of monomials.

In the next subsection, this thesis quickly overviews dynamic programming and color coding for k -path in the context of k -multilinear monomial detection. Although color coding is outperformed by the purely algebraic technique of fingerprinting, color coding is given as valuable background information, since some ideas carry on to the algebraic methods. Moreover, it appears that most parameterized problems that are solved with color coding can be reduced to multilinear monomial detection, and thus have faster algorithms [KW15; Fom+17].

In a later subsection, the thesis covers color coding with matrices as a purely algebraic method, which leads into the technique of algebraic fingerprinting. Then, a general framework for algebraic fingerprinting is discussed. In the last subsection under this section, the thesis discusses some limits and cons in general multilinear monomial detection.

AM: These problems should "live" in their own subsection, then you can just call them "problems from Sect. X.Y" AM: This gives an upper bound for solving the problem, but it is generally not optimal AM: I am having a hard time figuring out the overall structure AM: \cite can take multiple parameters, see source AM: Use \label and \cref

3.1 Non-algebraic methods

Before discussing color coding, the thesis briefly overviews dynamic programming in the context of multilinear monomial detection, since dynamic programming is used in color coding. Furthermore, [the thesis introduces AM: we discuss](#) an algebraically interesting idea of setting squared variables to zero.

3.1.1 Dynamic programming for smart expansion of the polynomial

Multilinear monomial detection can be solved with dynamic programming. Dynamic programming, used for e.g. Hamiltonicity in [HK62], is a method where the problem is recursively broken down into smaller subproblems. [Fom+17] gives a dynamic programming algorithm for solving an instance of k -multilinear monomial detection in the problem context of [representative families for product families](#).

AM: ??

In multilinear monomial detection, only the multilinear terms are important. This implies that any non-linear term can be discarded as soon as they are formed in the arithmetic circuit, since the arithmetic circuit will never decrease the degree of a monomial. Therefore, a smart algorithm can be designed for the expansion that applies an additional rule: any squared variable can be instantly discarded.

With dynamic programming, this can be implemented by redesigning the arithmetic gates in the arithmetic circuit. In [practise AM: practice](#), the algorithm would employ special addition and multiplication gates that discard squared variables. Thus, the evaluation of the (sum of monomials form) polynomial is reduced to just evaluating the multilinear terms. For a multivariate polynomial with n variables, there are 2^n multilinear terms, which implies that the expansion can be ran in $\mathcal{O}^*(2^n)$ time [KW15].

AM: I am not quite sure what is going on here

The rule of discarding squared variables can be written in algebraic form: a squared variable $x^2 \in X$ is set to the additive identity (zero) of the field \mathbf{F} , where $X \subset F$. With this additional rule for multilinear monomial detection, the following is deduced: if there are no solutions to the original problem, the polynomial will identically evaluate to zero. AM: So algebraically we are computing over $\mathbf{F}[x]/(x^2)$?

Note, however, that the polynomial evaluating to zero does not [AM: necessarily](#) imply that no solutions exist. Indeed, we will come across a problem where multilinear monomials cancel each other, and thus evaluate to zero (see Section X.X).

3.1.2 Non-deterministic color coding for faster evaluation

TODO: go over random assignment with colors, which results a polynomial of smaller domain (less variables), thus making the evaluation faster

AM: I am not sure if this is not a too big of a detour

Color coding was introduced in [AYZ95] as a randomized method for subgraph problems. Among others, a non-deterministic algorithm for the k -path problem is given. First, the thesis explains the idea behind the algorithm in [AYZ95]. After this, the idea is translated into multilinear monomial detection for easier relevance.

(DIRECTED OR UNDIRECTED) k -PATH (DECISION AND SEARCH)

Input: A (directed or undirected) graph $G = (V, E)$.

Question: Does G contain a simple path on k vertices? If so, give such a path.

TODO: go quickly over the algorithm, translate it in terms of multilinear monomial detection, give focus on the random assignment in order to reduce domain size (adds non-determinism, though)

TODO: end with hints toward algebraic assignment (for example, a non-zero matrix squared can equal zero)

3.2 Color coding with algebra

TODO: go through idea of algebraic assignment and specifications for a suitable algebra, then lead to fingerprinting (fingerprints solve an issue of unwanted cancellation due to characteristic)

3.2.1 Algebraic fingerprinting to prevent unwanted cancellation

Algebraic fingerprints are introduced to solve the problem of unwanted cancellation. To prevent the multilinear monomials from cancelling each other, we augment the polynomial with new unique indeterminates, i.e. fingerprints, such that multilinear monomials in the expanded polynomial are unique. The squared variables, and thus non-solution monomials, will still vanish, but the multilinear terms will prevail.

However, introducing new indeterminates raises the degree of the multilinear monomials. Therefore it may be, that there are not enough matrices to assign such that there would be no duplicates in a multilinear monomial. As a result, higher dimension matrices are needed, which implies exponentially slower matrix multiplication (TODO: check the time complexity of matrix multiplication). Thus, introducing new indeterminates slows the algorithm. (TODO: slower than $\mathcal{O}^*(2^k)$?)

In [Kou08] instead of uniqueness, Koutis uses random assignment from $\{0, 1\}$. With this assignment, there is a possibility that a multilinear monomial will have an odd coefficient, thus surviving the cancellation due to characteristic.

Koutis uses random assignment from $\{0, 1\}$ in order to find an odd k -mld problem, which

can be solved by previous methods, since there is no cancellation (multilinear terms have odd coefficients).

AM: What is cancellation?

TODO: explain how fingerprints prevent cancellation, give the general algebraic framework that appears in [Wil09], (also polynomial identity testing)

[Kou08] uses commutative group algebras of \mathbb{Z}_2^k (to have squared variables equal zero) for ODD k -multilinear detection (multilinear monomials have odd coefficients). randomized assignment, no false positives

AM: ODD = ordered decision diagram?

Koutis reduces an algebraization into ODD k -multilinear detection by randomly assigning fingerprints from 0,1 in the hopes of getting odd number of solutions (k -multilinear terms), i.e., hoping that for a pair that cancels each other, one of them gets removed by 0-assignment so that cancelation is prevented

[Wil09] reduces multilinear detection (after assigning variables values from some algebra) to polynomial identity testing for a polynomial $P(A)$ that has fingerprints as variables. Fingerprints are then assigned values from a field that has more elements than $P(A)$ has roots, which means that $P(A)$ evaluates to non-zero with high probability (if there are solutions)

3.2.2 Finding the solution

Multilinear monomial detection has only been given as a detector for a solution, i.e., a decision algorithm. Talk about actually finding the solution.

[Kou08] gives an algorithm that solves the decision problem for k -path. k -path is found with $\mathcal{O}^*(n + \min(k^2, m))$ applications of the algorithm.

[Wil09] solves the decision problem for k -path. [Wil09] also gives an algorithm that finds a path when it is known that a k -path exists.

[Bjö+17] solves the decision problem for k -path starting at some vertex s . A k -path can be found by just applying the algorithm for every vertex.

3.3 Limits of general multilinear monomial detection

TODO: give some cons wrt. color coding (difficult derandomization, are we able to add weights for optimization problems?)

TODO: explain the limit in general multilinear detection with this algebraic framework (impossible to find better algebra than what is used for the current fastest k -mld algorithm), [KW09]

TODO: lead to problem specific implementations (we can use fingerprinting cleverly, when

we understand the underlying problem well)

4 Problem-specific applications of algebraic fingerprints

TODO: add subsections for problem specific implementations with different utilizations of fingerprints

4.1 Fingerprinting for cancellation of non-solutions

TODO: go over Björklund et al. for k-path or Hamiltonicity, managed to design fingerprints such that non-solutions cancel

5 Conclusion

TODO: conclude

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AM: No need for URL if you have DOIs, also no need for ISSNs
 AM: You should use math style in the title of [Wil09]
 AM: use dblp.org
 AM: It is more common to sort by lastname, then year

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