

Algebraic Complexity Classes

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1 Introduction

In this survey, I am going to try and describe the algebraic complexity framework originally proposed by Leslie Valiant [Val79, Val82], and the insights that have been obtained more recently. This entire article has an “as it appeals to me” flavour, but I hope this flavour will also be interesting to many readers. The article is not particularly in-depth, but it is an invitation to read [BCS97, Bür00a] and many recent papers on the topic, and to start attacking the open problems in the area.

Valiant started out with the mission of understanding the core essence of reductions and completeness, as witnessed in both recursive function theory and in computational complexity theory. He provided an algebraic framework in which to interpret the clustering of natural problems into completeness classes, even for problems of an algebraic rather than combinatorial nature. He had a remarkable hypothesis:

Linear algebra offers essentially the only fast technique for computing multivariate polynomials of moderate degree.

Clearly, then, we are going to talk about polynomials, not languages or functions.

2 Valiant’s original framework

Let \mathbb{F} be any field, and let $\mathbb{F}[x_1, \dots, x_n]$ be the ring of polynomials over indeterminates x_1, \dots, x_n with coefficients from \mathbb{F} . Consider a family (f) of polynomials $(f_n)_{n \geq 1}$, where each f_n is in $\mathbb{F}[x_1, \dots, x_{s(n)}]$ for some function $s : \mathbb{N} \rightarrow \mathbb{N}$. When should we say that (f) is tractable? Clearly, if there are too many variables to keep track of, there cannot be tractability. So we will henceforth demand that s is a polynomially bounded function ($\exists c, \forall n, s(n) \leq c + n^c$); then the n th polynomial f_n has at most n^c variables. But that is of course not enough.

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There are many ways in which we can set the bar for tractability. Here's a first attempt. Can (f) be computed by a *formula* of reasonable size? To elaborate further, a formula is an expression defined recursively:

1. for each $c \in \mathbb{F}$, " c " is a formula of size 0 computing the polynomial c ,
2. for each indeterminate x_i , " x_i " is a formula of size 0 computing the polynomial x_i , and
3. if F_1, F_2 are formulas computing polynomials f_1 and f_2 , then " $(F_1 + F_2)$ " and " $(F_1 \times F_2)$ " are formulas of size $\text{size}(F_1) + \text{size}(F_2) + 1$ each, computing the polynomials $f_1 + f_2$ and $f_1 \times f_2$ respectively.

Notice that $\text{size}(F)$ is just the number of ring/field operations used to construct F .

Instead of such a recursive definition, we could have a more intuitive picture: a formula is a rooted binary tree where internal nodes are labeled $+$ or \times and leaf nodes are labeled from the set $\mathbb{F} \cup X$, where X is the set of indeterminates. The size is just the number of non-leaf nodes.

Now, for tractability, we could require that there is a polynomially bounded function $t : \mathbb{N} \rightarrow \mathbb{N}$ and a family of formulas $(F_n)_{n \geq 1}$ such that each F_n computes f_n and has size at most $t(n)$. Let us use the notation VF to denote families of polynomials tractable in this sense. (VF: Valiant's Formulas — of course, Valiant didn't use this name! This class is also referred to as VP_e : Valiant's Polynomial-sized Expressions. Personally, I prefer VF.)

Here's a second attempt: Can (f) be computed by a *straight-line program* of reasonable size? As before, we will declare polynomial size to be reasonable. Straight-line programs are programs where instructions involve adding or multiplying previously computed polynomials, no divisions and no conditionals (no if-then-else). In the more intuitive picture, they correspond to directed *acyclic* graphs where each node is a source node (indegree 0) labeled from the set $\mathbb{F} \cup X$, or has indegree 2 and is labeled $+$ or \times . A designated sink node (outdegree 0) is the output node. Each node computes a polynomial in the obvious way, and the graph computes the polynomial at the output node. (The acyclicity constraint ensures that each node, or instruction, only uses previously computed polynomials) The size is the number of non-source nodes; again, this corresponds to the number of ring/field operations required. Such graphs are in fact exactly *algebraic circuits*, and we now look for polynomial size circuit families.

Clearly, this model generalises formulas. The catch is that it generalises it too much! To see why, consider the following circuit family: C_n has $n + 1$ nodes v_0, v_1, \dots, v_n , and the labeling is $v_0 = x_1$, $v_{i+1} = v_i \times v_i$ for $i \in [n]$. The family of polynomials (f_n) computed by (C_n) is $f_n = x_1^{2^n}$. Even for small integer values of x_1 , writing down the value of $f_n(x_1)$ is going to require exponentially many bits. How can we say that such a family (f_n) is tractable?

So we need to impose some additional restrictions. The obvious parameter to restrict is the degree of the polynomial. Say that the family (f_n) has moderate degree if for some polynomially bounded function $d : \mathbb{N} \rightarrow \mathbb{N}$, the degree of each polynomial f_n is at most $d(n)$. If $\text{degree}(f_n) = D$ is polynomially bounded, then on integer arguments with b -bit

representations, the value of f_n requires no more than $\text{poly}(n, b)$ bits. (In general, it needs no more than $\text{poly}(n, D, b)$ bits.) Henceforth, to qualify for the label tractable, a family (f_n) must have polynomially bounded degree.

(Why didn't we face this problem when considering VF ? Simply because a formula of size t cannot compute a polynomial of degree more than $t + 1$. Don't just believe me; check this by induction on formula size.)

Now we have our second possible definition of tractability: (f_n) is tractable if the sequence $\text{degree}(f_n)$ is polynomially bounded, and there is a polynomially bounded function $t : \mathbb{N} \rightarrow \mathbb{N}$ and a family of straight-line programs, or algebraic circuits, $(C_n)_{n \geq 1}$, such that each C_n computes f_n and has size at most $t(n)$. Let us use the notation VP to denote families of polynomials tractable in this sense. (VP: Valiant's analogue of the Boolean complexity class P. Valiant called these families p -computable [Val82].)

The well-studied polynomial family from linear algebra, the determinant of a matrix of indeterminates, is known to be tractable in this second sense. (To define the family (Det_n) , imagine an $n \times n$ matrix A_n with a new indeterminate x_{ij} at each position (i, j) , and let Det_n be the polynomial that represents the determinant of A_n . Thus $\text{Det}_1 = x_{11}$, $\text{Det}_2 = x_{11}x_{22} - x_{12}x_{21}$, and so on. Clearly, this family satisfies the mandatory conditions: Det_n has n^2 variables and is of degree n .) This is not surprising; we know that the determinant can be computed efficiently (in polynomial time) over instantiated matrices using, say, Gaussian elimination. But to compute the symbolic determinant via a straight-line program, Gaussian elimination is apparently not directly of use because we can't search for non-zero pivots and eliminate them (remember, no divisions and no conditionals). However, Strassen [Str73] gave a generic method of converting any straight-line program with divisions to a division-free straight-line program; the resulting program's size is polynomially bounded in the original size, the number of variables, and the degree. Thus we can conclude that there are polynomial-sized straight-line programs for the symbolic determinant. There are more direct algorithms as well; Samuelson, Berkowitz, Csanky, See [MV97] for an explicit description of circuits of size $O(n^4)$ (my favourite one – no surprise!).

Whether the determinant can be computed efficiently by formulas (is Det_n in VF ?) is still famously open. We know that it needs formula size at least $\Omega(n^3)$, see [Kal85]. But we do know that it can be computed by formulas of sub-exponential size $2^{O(\log^2 n)}$. This can be shown in many different ways, one of which we will look at a bit later, but the earliest demonstration of this follows from Csanky's algorithm [Csa76], which uses binary arithmetic operations and $O(\log^2 n)$ parallel time. Thus if we use quasi-polynomial ($2^{\log^c n}$ for some constant c) formula-size as the defining property for tractability (giving a class that we can call VQF), then again the family (Det_n) has long been known to be tractable. We could also use quasi-polynomial circuit size as the defining property for tractability, giving a class that we can call VQP. But VQP obviously contains VP and VQF, so (Det_n) is in VQP; nothing new there. (Note: in defining VQF and VQP, the quasi-polynomial limit on formula or circuit size is over and above the requirement that the degree and number of variables are polynomially bounded.)

Does VP include essentially all interesting and natural polynomial families? We do not

know. In fact, there is a large list of such polynomial families not known to be in VP. The most natural one is the permanent family (Perm_n) where Perm_n is the polynomial representing the permanent of an $n \times n$ matrix A_n of indeterminates. It is tantalisingly similar to the determinant; just the sign term is missing.

$$\text{Det}_n = \sum_{\sigma \in S_n} \text{sign}(\sigma) \prod_{i=1}^n x_{i\sigma(i)} \qquad \text{Perm}_n = \sum_{\sigma \in S_n} \prod_{i=1}^n x_{i\sigma(i)}$$

Yet, it does not seem to be tractable. How “intractable” is it?

Mirroring the definitions of the Boolean complexity classes P and NP, Valiant proposed a notion of p -definability in [Val79]. A polynomial family (f_n) is p -definable if it can be written as an exponential sum, over partial Boolean instantiations, of another tractable family. Formally, a family (f_n) over $s(n)$ variables and of degree $d(n)$ is p -definable if $s(n)$ and $d(n)$ are polynomially-bounded, as always, and further, there exist a polynomially-bounded function m , and a family of polynomials (g_n) in VF, such that g_n has $s(n) + m(n)$ variables denoted $\{x_1, \dots, x_{s(n)}, y_1, \dots, y_{m(n)}\}$, and

$$f_n(\tilde{x}) = \sum_{y_1=0}^1 \sum_{y_2=0}^1 \dots \sum_{y_{m(n)}=0}^1 g_n(\tilde{x}, \tilde{y}).$$

This looks like an algebraic analogue $\sum \cdot \text{VF}$ of the boolean class $\exists \cdot F$, where F is the class of languages decided by polynomial-size formulas. But it is well-known that $\exists \cdot F = \text{NP}$, so this should be algebraic NP. Later, Valiant redefined p -definability (no, that is not a circular definition!) as exponential sums of families in VP, rather than VF; that is, $\text{VNP} = \sum \cdot \text{VP}$. For clarity, let us agree to temporarily refer to these two definitions as VNF (or VNP_e) and VNP. However, Valiant [Val82] showed that these two classes are in fact the same, so just VNP will do. The proof involves showing that VP is contained in $\sum \cdot \text{VF}$. And it is of course easier to show upper bounds with the definition of VNP rather than VNF.

Now Valiant observed that not only (Det_n) , even (Perm_n) is p -definable. This should be similar to showing that the 0-1 permanent is in $\#P$, right? Almost. We are dealing with symbolic polynomials, so we do not have the liberty of looking at an input value and deciding what to do next. Still, the basic idea is the same. For a statement S , let $[S]$ denote the 0-1 valued Boolean predicate that takes value 1 exactly when S is true. Then

$$\begin{aligned} \text{Perm}_n = \sum_{\sigma \in S_n} \prod_{i=1}^n x_{i\sigma(i)} &= \sum_{Y \in \{0,1\}^{n \times n}} [\text{Y is a permutation matrix}] \cdot \prod_{i=1}^n \left(\sum_{j=1}^n Y_{ij} x_{ij} \right) \\ [\text{Y is a permutation matrix}] &= [\text{Y has at least one 1 in each row}] \times \\ &\quad [\text{Y has at most one 1 in each line (=row or column)}] \\ &= \left(\prod_{i=1}^n \sum_{j=1}^n Y_{ij} \right) \left(\prod_{i=k \text{ or } j=m} (1 - Y_{ij} Y_{km}) \right) \end{aligned}$$

Clearly, the polynomial family

$$g_n = \left(\prod_{i=1}^n \sum_{j=1}^n Y_{ij} \right) \left(\prod_{i=j \text{ or } j=m} (1 - Y_{ij} Y_{km}) \right) \prod_{i=1}^n \left(\sum_{j=1}^n Y_{ij} x_{ij} \right)$$

has formulas of size $O(n^3)$, and $\text{Perm}_n(\tilde{x}) = \sum_{Y \in \{0,1\}^{n \times n}} g_n(\tilde{x}, Y)$, so (Perm_n) is in VNP.

So we have some families in VP (even VF), and some in VNP but maybe not in VP. How do we compare families? For comparing languages, we have many-one reductions and Turing reductions – what is the algebraic analogue? Valiant proposed projections, a most restrictive kind of reduction when dealing with Boolean classes, but completely natural in the algebraic context. We say that $g \in \mathbb{F}[y_1, \dots, y_m]$ is a projection of $f \in \mathbb{F}[x_1, \dots, x_n]$ if substituting a value in $\mathbb{F} \cup Y$ for each variable in X and simplifying the polynomial resulting from f yields exactly g . (For instance, if $f = x_1 x_2 + x_3 x_4$, then the following are all projections of f : $y_1 + y_2$, $y_1 y_2 + 5$, $y_1 y_2 + y_2 y_3$. But $y_1^2 y_2$, $y_1 + y_2 + y_3$ are not, because a projection cannot increase the degree or number of monomials.) Further, we say that a family (g_n) is a p -projection of a family (f_n) if each g_n is a projection of some f_m for an m not too far from n . That is, there is a polynomially bounded function t , and each g_n is a projection of $f_{t(n)}$. If we allow t to be quasi-polynomially bounded, we obtain qp -projections.

Using these notions of reductions, we have the usual notions of hardness and completeness for algebraic classes. Here's what Valiant showed:

1. (Det_n) is hard for VF under p -projections (and is known to be in VP).
2. (Det_n) is complete for the class of quasi-polynomial size formulas VQF under qp -projections.
3. Over fields with characteristic other than 2, (Perm_n) is complete for VNP under p -projections. Over fields of characteristic 2, Perm_n equals Det_n and hence is in VP and VQF.
4. Polynomial families associated with a number of NP-complete languages are complete for VNP under p -projections.

The first two follow from a proof that a polynomial computed by a size s formula is a projection of Det_{s+2} . (It uses the combinatorial definition of determinant. as the signed weighted sum of cycle covers in an associated graph.) The hardness of (Perm_n) for VNP mirrors the hardness of the Boolean permanent for the counting class $\#P$. As in the case of the upper bound, additional care is needed to take into account non-access to an input instance and fully symbolic computations; in particular, the proof requires a multiplicative inverse of 2 and hence fails over fields of characteristic 2. See [Val79, BCS97, Bür00a] for various versions of these proofs.

3 The current status

We now know much more about the classes VF, VP, VQP, VNP defined above, and about other similarly defined classes. Let's review these results one by one.

Say that a family of polynomials (f_n) is a p -family if the number of variables in f_n and the degree of F are polynomially bounded functions of n . We only consider p -families.

Recall that VP consists of p -families with polynomial-sized circuits. Also note that algorithmically, circuit size roughly corresponds to number of processors needed in a parallel algorithm (associate one processor per gate), while circuit depth – the length of a longest path from the output node to an input node – corresponds to parallel time.

A clever construction due to Hyafil [Hya79] shows that any polynomial of degree D in N variables, computable by a circuit of size t , can be computed in parallel time $O(\log D \times \log(D^2t + N))$. This is a depth-reduction of the circuit, and generalises Csanky's result which was specifically tailored for the determinant. Further, this algorithm has parallel multiplicative depth only $O(\log D)$; this is worth noting since multiplication seems a more costly operation than addition or subtraction. Unfortunately, the resulting circuit, while shallow and depth-reduced, is rather large, roughly $t^{\log D}$. Applying this construction would take us from polynomial-size circuits to shallow quasi-polynomial size circuits. Soon after this, an improved construction was presented by Valiant, Skyum, Berkowitz and Rackoff [VSB83]; they achieved the same depth-reduction (and also $O(\log D)$ multiplicative depth) with size polynomial in tD . In particular, applying this construction to a circuit family (C_n) witnessing that a polynomial family (f_n) is in VP, we see that (f_n) is in $\text{VSAC}^1 \subseteq \text{VNC}^2$.

Wait, what exactly are these new classes? Again, we can think of them as analogues of Boolean classes. The Boolean circuit class NC^i has circuits of polynomial size and $O(\log^i n)$ depth. The class SAC^i is similarly defined, polynomial size, $O(\log^i n)$ \wedge -depth, and negations only at inputs. That is, if \vee nodes are allowed to have unbounded in-degree, but \wedge nodes must have in-degree 2, then these circuits have depth $O(\log^i n)$. (Hence the name SAC, for semi-unbounded alternation.) Clearly, $\text{NC}^i \subseteq \text{SAC}^i \subseteq \text{NC}^{i+1}$. Now define the classes VNC^i and VSAC^i as algebraic analogues of these, with \times and $+$ playing the roles of \wedge and \vee respectively. In the Boolean world, we know that $\text{NC}^1 \subseteq \text{SAC}^1 \subseteq \text{NC}^2 \subseteq \dots \subseteq \text{NC} \subseteq \text{P}$. In the algebraic world, however, $\text{VNC}^1 \subseteq \text{VSAC}^1 = \text{VNC}^2 = \dots = \text{VNC} = \text{VP}$.

An important consequence of the depth reduction result of [VSB83] is that the $(\text{Det}_n) \in \text{VQF}$ result generalises to all of VP; $\text{VP} \subseteq \text{VQF}$. Another important consequence is that at quasi-polynomial size, formulas are as powerful as circuits; VQF equals VQP. Such an equivalence is not known for p -families at polynomial size. (It holds at exponential size, because polynomials in any p -family have only exponentially many monomials. An explicit sum-of-monomials expression gives an exponential sized formula.)

Even before the results of [Hya79, VSB83], Spira [Spi71] and Brent [Bre74] had shown that depth-reduction is possible for VF. Any formula F can be rebalanced by identifying in it a suitably chosen node N and rewriting F as a linear form in N , say $AN + B$. If N is properly chosen, then the polynomials A and B are computed by small sub-formulas (size at most half of F) of F , and can be recursively rebalanced. The appropriate N is identified by using the tree separator lemma. This process yields a $O(\log \text{size}(F))$ depth formula. Thus

we conclude $VF = VNC^1$.

The depth reduction for VP from [VSBR83] proceeds similarly, but works on “proof-trees” or *parse trees*. Unfolding a circuit into a formula by systematically duplicating reused nodes may yield an exponential sized formula (recall the example X^{2^n} .) Let us nonetheless do so. Now, a minimal sub-formula that includes the output node, both children of an included \times node, and exactly one child of an included $+$ node, computes a potential monomial whose degree is the number of leaf nodes in the sub-formula. Call such a sub-formula a proof tree. For a circuit computing a p -family of polynomials, we can ignore proof trees of super-polynomial size. For each polynomial-sized proof tree, the balancing technique described above should work. The catch is, there can be too many proof trees (there can be exponentially many monomials), and each proof tree could require cutting at a different node. The clever twist is the following: in the formula depth-reduction, A can be computed recursively because it is the partial derivative of F with respect to N . If F is now a circuit rather than a formula, then F may not be linear in N , so computing the partial derivative will not help. But if N is chosen to have degree more than half the degree of F , then this is indeed the case. So the algorithm of [VSBR83] computes, for each pair of nodes N, N' , a new polynomial $F(N, N')$; these polynomials are recursively constructed, and whenever $2\text{degree}(N) > \text{degree}(N')$, $F(N, N')$ equals the partial derivative of N' with respect to N . Putting this together carefully gives the depth-required circuit. For details, see [VSBR83] itself. Also see [AJMV98a] and [Vol99] for *uniform* versions, where the task of describing the depth-reduced circuit given the original circuit is achieved using limited computational resources.

A couple of things slipped by almost unnoticed. We know what is meant by the degree of a polynomial, but what do we mean by $\text{degree}(N)$? This should be the degree of the polynomial computed at the node N , and indeed [VSBR83] use degree in this sense. But the uniform versions cannot do so, because computing the degree of a specified node in a given circuit is a completely non-trivial task! See the discussion about DegreeSLP in [ABKPM09, Kay10]. Fortunately, we can equally easily work with an upper bound on the degree of each node. And an upper bound $u(N)$ on the degree at each node N is easy to obtain: $u(N) = 1$ if N is a leaf, $u(N_1 + N_2) = \max\{u(N_1), u(N_2)\}$, $u(N_1 \times N_2) = u(N_1) + u(N_2)$. This upper bound is referred to as the complete formal degree of the circuit (as opposed to the degree of the polynomial it computes). However, just because the output node of C computes a polynomial of degree d , this does not imply that each node computes a polynomial of degree at most d . Higher degree monomials may get computed along the way, and get cancelled finally. Is it necessary, in terms of efficiency, to compute them? No! If C is of size s and computes a polynomial f of degree d , then we can construct a circuit C' of size $O(sd^2)$ computing the same polynomial and with each node computing a polynomial of degree at most d : just compute the homogeneous parts of f separately in the obvious way. Now C' will have complete formal degree $O(d^3s)$. (See [MP08] for details.) Thus we could have defined VP in terms of circuits of polynomial size and polynomially bounded complete formal degree as well.

There is a much simpler proof of the fact that VP is contained in VNC. This proof

yields a weaker upper bound of VSAC² rather than VSAC¹, but is still beautiful, and is still enough to conclude that $\text{VP} \subseteq \text{VQF}$. I first saw this proof in a survey talk by Pascal Koiran at Dagstuhl [Koi10], and I wish I had come up with it myself! Let (f_n) be in VP, as witnessed by a circuit family (C_n) with complete formal degree bounded by (d_n) . To depth-reduce C_n , partition the nodes into $1 + \lceil \log d_n \rceil$ parts; part k has nodes with formal degree in $[2^{k-1}, 2^k)$. Treating the polynomials from parts $i < k$ as variables, the nodes in part k form a *skew* circuit, where each \times node has at most one child that is not an input node. (Multiplying two nodes both in part k would create high degree, giving rise to a node in part $k + 1$.) Now, skew circuits can be depth-reduced to VSAC¹ rather easily, using a divide-and-conquer argument dating back to Savitch [Sav70]. Doing this separately for each part gives a VSAC² circuit.

We just introduced a new kind of circuit there: skew circuits. Are they as powerful as general circuits? We do not know! Let's define VP_{skew} ; p -families of polynomials computed by polynomial-sized skew circuits. It turns out this is a great class to study, because it *exactly* characterises the complexity of the determinant. Recall what we have already seen; (Det_n) is hard for $\text{VF} = \text{VNC}^1$ and is in VP. The upper bound proof from [MV97] actually gives a skew circuit of size $O(n^4)$, but skew circuit constructions were known much earlier: in [Ven92], Venkateswaran first defined Boolean skew circuits to capture nondeterministic circuits, and subsequently many authors independently extended that study to arithmetic rings, [Dam91, Tod92, Vin91, Val92]. And the lower bound proof from [Val79] shows that polynomials computed by skew circuits are p -projections of the determinant, though it is not stated this way. Valiant showed that a formula can be converted to a certain kind of graph that we nowadays call an algebraic branching program or ABP (more about this below), and that polynomials computed by ABPs are p -projections of (Det_n) . And we now know that ABPs are essentially skew circuits.

Time to define ABPs. These are directed acyclic graphs, with a designated source node s and a designated target sink node t (sometimes there may be multiple target nodes), and with edges labeled from $\mathbb{F} \cup X$ (similar to input nodes in a circuit). For any directed path ρ , the weight of ρ is the product of the labels of the edges on ρ . The polynomial p_v computed at a node v is the sum of the weights of all directed sv paths. The polynomial computed by the ABP is just p_t . Families computed by polynomial-size ABPs form the class VBP. (In some parts of the literature, edge labels are allowed to be linear forms in X . This does not significantly change the properties of ABPs as we discuss here. We'll stick to the convention that labels are in $\mathbb{F} \cup X$.)

So why are ABPs and skew circuits essentially the same? ABPs to skew circuits: clearly, $p_s = 1$, and for any other source node (in-degree 0) s' , $p_{s'} = 0$. Look at an edge $u \rightarrow v$ of the ABP with label ℓ . Then p_v has a contribution from $p_u \times \ell$. Summing this over all incoming edges at v gives a small circuit computing p_v from previously computed values, and this circuit is skew. For the reverse simulation, reverse this construction: (1) introduce a source node s , (2) for each input node u labeled ℓ , add an edge $s \rightarrow u$ labeled ℓ , (3) for each node $v = u + u'$, create edges $u \rightarrow v$ and $u' \rightarrow v$ labeled 1, and (4) for each node $v = u \times \ell$, create an edge $u \rightarrow v$ labeled ℓ .

So now we can add to the list of results at the end of Section 2: (Det_n) is complete for $\text{VBP} = \text{VP}_{skew}$ under p -projections.

In fact, we can add more. What makes the simulation from skew circuits to ABPs possible is the fact that at each \times gate, one argument is *easy*. Toda [Tod92] took this argument further – it is enough if one argument is independent of the rest of the circuit. That is, for each \times node $\alpha = \beta \times \gamma$, the entire sub-circuit rooted at either β or γ has no connection to the rest of the circuit except via this edge to α . (Equivalently, one of the edges into α is a bridge in the circuit.) Call such circuits *weakly skew* circuits. Toda showed that weakly skew circuits can be converted to skew circuits with linear size blow up. See also [MP08], where Malod and Portier made the size bounds in the conversions even more precise. So now we can say $\text{VBP} = \text{VP}_{skew} = \text{VP}_{ws}$, where the subscript *ws* stands for weakly skew.

Taking this idea further, Malod and Portier provide a brilliant characterization of the class VP. Say that a circuit is *disjoint* if at every node $\alpha = \beta \circ \gamma$, where \circ could be $+$ or \times , the sub-circuits rooted at β and γ are disjoint. This is just a fancy (convoluted?) way of saying that the circuit is a formula. But now relax this constraint a bit. Say that a circuit is *multiplicatively disjoint* or MD if at every \times node $\alpha = \beta \times \gamma$, the sub-circuits rooted at β and γ are disjoint. No restrictions apply to $+$ nodes. Like formulas, MD circuits of size s have complete formal degree bounded by s . But the MD restriction seems to allow more computation than formulas; for instance, weakly skew circuits are MD, and so MD circuits can compute (Det_n) in polynomial size. Malod and Portier showed that in fact polynomial-size MD circuits can compute everything in VP, but nothing more. That is, $\text{VP} = \text{VP}_{\text{MD}}$. While this fact can also be deduced once we have depth-reduction to VSAC^1 , Malod and Portier give a completely self-contained combinatorial proof which is very neat. Basically, imagine that each node in the VP circuit is labeled with its formal degree. Now make multiple copies of each node, inversely proportional to the formal degree. By carefully deciding which copies of its children to use to construct a copy of a node, multiplicative disjointness can be achieved with only polynomial blow-up in size.

A nice consequence of this characterisation of VP is a simpler proof of the fact that VP is contained in $\sum \cdot \text{VF}$. The key observation used is that a circuit is multiplicatively disjoint exactly when every proof tree is already a sub-graph of the circuit (even without any unfolding into a formula). See [MP08] for details.

Before we move on, we note another surprising relation between ABPs and formulas: VF equals the class of p -families computed by polynomial-size ABPs of constant width. What is this resource “width”? Recall that an ABP is a DAG with edges going “in the direction from s to t ”. Suppose we impose a layering constraint. The nodes of the DAG must be laid out at the vertices of a rectangular $w \times \ell$ grid, the node s must be at position $(S, 1)$ for some $S \in [w]$, the node t must be at position (T, ℓ) for some $T \in [w]$, and edges can only go across one layer, from (i, k) to $(j, k + 1)$ for some $i, j \in [w]$, $k \in [\ell - 1]$. Of course, any ABP can be converted to one of this form: just sub-divide edges when necessary and label the sub-division path so that its weight is the original edge’s label (use lots of 1s). Now we say that w is the width of the layered ABP and ℓ is the length. A bounded-width branching program family (B_n) is one where for some absolute constant c , each B_n has width at most c . Seems quite a squeeze – if

we view moving from s towards t as an incremental computation, then at each stage we can carry forward just c intermediate polynomials. We shouldn't be able to do much this way, right? Wrong! Ben-Or and Cleve [BOC92] showed, in a proof cleverly extending Barrington's famous characterisation [Bar89] of NC^1 by Boolean bounded-width branching programs, that every formula of depth D has an equivalent bounded-width branching program (that's quite a mouthful; let's agree to call it BWBP) of length 4^D and width just 3! Since we already know that formulas can be depth-reduced and VF equals VNC^1 , we see that VF is contained in a class that we can name VBWBP: polynomial-sized constant-width ABPs. The converse inclusion is easily seen to hold, again using a Savitch-style divide-and-conquer. Thus we have another characterisation of VF .

As a matter of curiosity, one may want to know: is the width-3 upper bound tight? Allender and Wang [AW11] recently settled this question affirmatively: they show that a very simple polynomial cannot be computed by any width-2 ABP. On the other hand, width-3 ABPs are universal, since every polynomial family has some formula family computing it. The question is one of efficiency: which families have polynomial-size width 3 ABPs?

OK, so we've had a plethora of class definitions, but just a handful of distinct classes: $\text{VF} = \text{VP}_e = \text{VNC}^1 = \text{VBWBP}$, $\text{VBP} = \text{VP}_{skew} = \text{VP}_{ws}$, $\text{VP} = \text{VP}_{\text{MD}}$, $\text{VQF} = \text{VQP}$, $\text{VNF} = \text{VNP}$.

As stated in [Bür00a], **Valiant's hypothesis** says that $\text{VNP} \not\subseteq \text{VP}$, and **Valiant's extended hypothesis** says that $\text{VNP} \not\subseteq \text{VQP}$. Over fields of characteristic not equal to 2, these imply: Perm_n is not a p -projection of Det_n , and Perm_n is not a qp -projection of Det_n , respectively.

Some miscellaneous results, in no specific order:

1. Let SymDet_n be the polynomial that represents the determinant of a symmetric $n \times n$ matrix of indeterminates B_n . (For instance, $\text{SymDet}_2 = x_{11}x_{22} - x_{12}^2$.) Clearly, (SymDet_n) is a p -projection of (Det_n) . The converse is also almost true. As shown by Grenet, Kaltofen, Koiran and Portier in [GKKP11], over any field of characteristic other than 2, Det_n is a projection of SymDet_{n^3} . Characteristic 2 is a problem: symmetric matrices correspond to undirected graphs, so each undirected cycle gives rise to two directed cycles, and so to get a projection we need division by 2. The best that we can currently say in characteristic 2 is that $(\text{Det}_n)^2$ is a projection of SymDet_{2n^3+2} ; this is also shown in [GKKP11].
2. VQP is also characterized by quasi-polynomial-size weakly skew circuits of polynomial degree. (From [VSB83] it follows that $\text{VQP} = \text{VQF}$; hence the above characterization. A direct proof is presented in [MP08].) Several natural polynomials are complete for this class under qp -reductions: the (Det_n) family, of course, but also, the trace of iterated matrix product and the trace of a matrix power. These families are all complete for VBP under p -reductions.
3. While we do not know the exact relationship between VQP and VNP , (they both contain VP), we do know that VQP does not equal either VP or VNP . Bürgisser ([Bür00a], Section 8.2) has shown that there is an explicit family of polynomials (f_n) in VQP that

is provably not in VNP, let alone in VP. This family is defined as follows: Consider numbers in base n . Let μ range over all such numbers with $m(n) = \lceil \log n \rceil$ digits. More precisely, let μ range over length- $m(n)$ sequences over the alphabet $\{0, 1, \dots, n-1\}$, and let $k_n(\mu)$ denote the value of this sequence, $k_n(\mu) = \sum_{j=1}^{m(n)} \mu_j n^{j-1}$. Define f_n as:

$$f_n(x_1, \dots, x_{m(n)}) = \sum_{\mu \in \{0, \dots, n-1\}^{m(n)}} 2^{2^{k_n(\mu)}} \prod_{j=1}^{m(n)} x_j^{\mu_j}$$

Exploiting the fact that the distinct double exponentials appear as coefficients in f_n , Bürgisser shows that f_n cannot be in VNP.

Furthermore, using $m(n) = \lceil \log^i n \rceil$ gives a family of polynomials f^i in VQP with size $O(n^{\log^i n})$ but provably not in size $O(n^{\log^{i-1} n})$, so within VQP there is a strict hierarchy.

4. From the qp -completeness of (Det_n) for VQP, and the p -completeness of (Perm_n) for VNP, it follows that $\text{VNP} \subseteq \text{VQP}$ if and only if (Perm_n) is a qp -projection of (Det_n) . This is a very long-standing open question. Originally the question of whether (Det_n) and (Perm_n) are p -equivalent was posed by Pólya [P13], who also showed that there is no way of expressing the permanent as the determinant by only changing the signs of selected entries (except for $n = 2$; flip the sign of a_{12} to get matrix B with $\text{Det}(B) = \text{Perm}(A)$). (I haven't myself seen Pólya's note, but have seen it referred to in various places.) Marcus and Minc [MM61] showed that there is no size-preserving transformation $(\text{Perm}_n \text{ to } \text{Det}_n)$, even if we relax the notion of projections to allow linear form substitutions for each variable. For many years, a linear lower bound was the best known ($\Omega(\sqrt{2}n)$ due to [vzG87, Cai90, Mes89]), until Mignon and Ressayre [MR04] showed that over the fields of characteristic 0 (eg real or complex numbers), even if linear form substitutions are allowed in projections, to express Perm_n as a projection of Det_m , we need $m \geq n^2/2$. The same lower bound was obtained for fields of characteristic other than 2 by Cai, Chen and Li [CCL10]. From Ryser's work [Rys63] it follows that Perm_n is a projection of Det_m for some $m < n^2 2^n$. More recently, Grenet showed [Gre12] via a very simple and neat construction that Perm_n is a projection of Det_m for $m = 2^n - 1$. This is the best known so far. Thus there is a huge gap between the lower and upper bounds on what is called the determinantal complexity of the permanent.
5. It is natural to believe that the complexity of a p -family (f_n) in this framework is closely related to the computational complexity of *evaluating* f_n for a given instantiation of its variables. In [Bür00b], Bürgisser gave this belief a firm footing. Consider a p -family (f_n) where f_n depends on n variables. Define its Boolean part $\text{BoolPart}(f)$ as a string function mapping $x \in \{0, 1\}^n$ to the binary encoding of $f_n(x)$. Note that we have considered only Boolean values. Even so, evaluation may seem difficult, because the circuits for (f_n) can involve arbitrary constants from the field. Bürgisser showed that assuming the generalised Riemann hypothesis GRH, over fields of characteristic zero, $\text{BoolPart}(\text{VP})$ has non-uniform multi-output NC^3 circuits. Furthermore, assuming

GRH, if Valiant’s hypothesis is false over such a field, then the entire polynomial hierarchy has (non-uniform) NC circuits.

6. An *extreme* depth-reduction result is given by the highly influential paper of Agrawal and Vinay [AV08]. To first see the context, note that any polynomial in n variables with degree d has an unbounded fan-in depth-2 circuit of size $2^{O(d+d\log \frac{n}{d})}$. (If $d \in \Omega(n)$, then $2^{O(d)}$ suffices, otherwise the second term in the exponent makes up.) This is because we can just explicitly compute all monomials of degree at most d , and add up the required ones with suitable weights. Now, can we find circuits substantially better than this, say even $2^{o(d+d\log \frac{n}{d})}$, if we allow depth to be increased a bit? Agrawal and Vinay showed that indeed this is possible, even with depth 4, provided there is some circuit (not necessarily depth-reduced) of that size to begin with. The idea is extremely simple. Perform the depth-reduction from [VSBR83] or [AJMV98b], and ensure with some additional care that degree provably drops at \times gates. (The price for this is small: a \times gate may have fanin upto 6, instead of 2.) Now, choose a horizontal cut in the depth-reduced circuit so that for the sub-circuit above it, and for the sub-circuits below it rooted at gates on the cut, the “brute-force” construction described above is small. Obviously there is a trade-off: if the cut is too high up, the lower sub-circuits can have large explicit forms, but if it is too low down, the upper sub-circuit can have large explicit forms. Cut in the right place, and everything works out!

This has significant implications for the quest for derandomizing algorithms for the well-studied problem ACIT (arithmetic circuit identity testing) — checking if a given circuit computes the identically-zero polynomial. But that is not directly connected with this survey. One question it raises here is: what kind of extreme depth-reduction can we achieve for VQP? Can we stay within quasi-polynomial size?

4 The syntactic multilinear world

Much of the study concerning VP and VNP involves the families (Det_n) and (Perm_n) . The polynomials in both families are *multilinear*. In principle, to compute a multilinear polynomial via a circuit, we need never compute intermediate polynomials that are not multilinear. Let us call such circuits, where the polynomial computed at each node is multilinear, *multilinear circuits*. However, often it is the case that allowing non-multilinear terms at intermediate stages, and eventually cancelling them out, allows more efficient computation (smaller circuits). This leads to the following quest: what kind of multilinear p -families have efficient multilinear formulas, or even multilinear circuits, where each intermediate polynomial is required to be multilinear? Even for the (Det_n) family, which we know is multilinear and in VP, we do not know of polynomial-size multilinear circuits. That being the case, can we prove lower bounds?

This question is trickier than it seems at first glance, because given a circuit, even checking whether it is multilinear is non-trivial. Fournier, Malod and Mengel [FMM12] recently observed that checking multilinearity of a given circuit is computationally equivalent to the

well-studied problem ACIT (arithmetic circuit identity testing) — checking if a given circuit computes the identically-zero polynomial.

So we may want a notion of certifiably multilinear circuits. One such notion is that of syntactic multilinearity, SM. A circuit is said to be syntactically multilinear if at every \times node $\alpha = \beta \times \gamma$, the sub-circuits rooted at nodes β and γ operate on disjoint sets of variables. Note that this is much more restrictive than multiplicative disjointness. But it certifies multilinearity, since no variable can ever get multiplied by itself. And syntactic multilinearity is easy to check computationally: it is violated if there is some node $\alpha = \beta \times \gamma$, some variable x , two input nodes I, I' labeled x , and paths from I to β and I' to γ .

If a family has efficient (polynomial-sized) SM circuits, then it has efficient multilinear circuits. The converse may not be true. But it is true if we look at formulas. Given a multilinear formula, identify an SM violation α, β, γ, x as above. Then we know by multilinearity of the polynomial $p(\alpha)$ that x does not appear in either $p(\beta)$ or $p(\gamma)$. In the appropriate sub-formula, set all instances of x to 0; the polynomials computed at and above α remain unchanged. Doing this systematically gives an SM formula of size no more than the original multilinear formula.

In the first major breakthrough, Ran Raz [Raz09] showed that for computation by SM formulas, and hence by multilinear formulas, both (Det_n) and (Perm_n) need size $n^{\Omega(\log n)}$. Clearly, this also means that they are not in SM-VNC^1 .

Since (Det_n) is in VP and even in VBP, SM-VF is strictly weaker than VBP. But this is hardly a fair comparison: we have restricted VF to be SM, but not VBP and VP. Can we say that SM-VF is strictly weaker than SM-VBP or SM-VP? We do not know whether (Det_n) is in multilinear VP, let alone SM-VP, so a different family is needed as a separating example. Such an example was provided soon thereafter, again by Ran Raz [Raz06]. He constructed an explicit polynomial family that is in SM-VP and even in SM-VSAC^1 , and showed that it needs SM-formula size $n^{\Omega(\log n)}$ and hence is not in SM-VNC^1 . Improved lower bounds for constant-depth circuits and subclasses of formulas were subsequently obtained by Raz, Shpilka and Yehudayoff (see for instance [RY09], [RSY08]).

Let's step back a bit. Why did we say “in SM-VP, and even in SM-VSAC^1 ”? Aren't VP and VSAC^1 the same? Well, we know that VP and VF can be depth-reduced. But can we assume that these depth-reduction techniques preserve syntactic multilinearity? Fortunately, they do; Raz and Yehudayoff [RY08] showed that the depth-reduction of [VSBR83] preserves SM, so indeed $\text{SM-VP} = \text{SM-VSAC}^1$. Similarly, in [JMR12] it is observed that the formula depth-reduction of [Bre74] also does preserve SM, so $\text{SM-VF} = \text{SM-VNC}^1$.

What about other relationships between the algebraic classes? We had considered ABPs — what certifies multilinearity there? It is easy to see that a read-once restriction, where on each path in the ABP each variable appears as a label at most once, does so. Let us therefore use read-once as the definition of syntactic multilinearity in ABPs. Then, as observed in [JMR12], the Savitch-style divide-and conquer argument preserves SM. So does the conversion from formulas to ABPs, [Val79]. But the conversion from formulas to width-3 ABPs, [BOC92], does not. In fact, Rao [Rao10] showed that even a significant generalisation of Ben-Or and Cleve's technique, using polynomially many registers instead of just 3, cannot

preserve syntactic multilinearity. Of course, there may be other ways of going from SM-VF to SM-VBWP, but it could equally well be that the classes are distinct.

To get back perspective, in the SM world what we have seen so far is:

$$\text{SM-VBWP} \subseteq \text{SM-VF} \subseteq \text{SM-VBP} \subseteq \text{SM-VP}$$

As mentioned earlier, Raz [Raz06] showed that the inclusion from SM-VF to SM-VP is proper. Very recently, this was improved by Dvir, Malod, Perifel, and Yehudayoff [DMPY12]. They showed that in fact the inclusion $\text{SM-VF} \subseteq \text{SM-VBP}$ is strict. Whether the first and the last inclusion are strict is still open.

The proof of [DMPY12] is a clever adaptation of the original technique from [Raz06]. Let us briefly examine this.

The central ingredient in Raz's proof is randomly partitioning the variables and analysing the rank of the resulting partial derivatives matrix. Consider a polynomial f on $2n$ variables $X = \{x_1, \dots, x_{2n}\}$, and consider a partition of X into equi-sized sets Y, Z . Consider a $2^n \times 2^n$ matrix $M_f^{Y,Z}$ where rows and columns are indexed by subsets of Y and Z (equivalently, multilinear monomials over Y and Z respectively). The entry (m_y, m_z) is the coefficient of the monomial $m_y \cdot m_z$ in f . Intuitively, if $M_f^{Y,Z}$ has high rank, then f should be hard. But high rank with respect to what partition? Raz showed that if multilinear f has small SM-formula size, then for at least one partition (Y, Z) of X , $M_f^{Y,Z}$ will have low rank. (The existence of the partition witnessing low rank is proved using the probabilistic method; choose a partition at random, and analyse the probability that the resulting matrix has rank exceeding some threshold.) He also constructed an explicit family g in SM-VSAC¹ and showed that for every partition (Y, Z) of X , $M_g^{Y,Z}$ has high rank; hence g is not in SM-VF.

The non-trivial adaptation done in [DMPY12] is to consider not all partitions, but a fairly small set of what they call arc-partitions. They showed that if f is in SM-VF, then for at least one arc-partition (Y, Z) of X , $M_f^{Y,Z}$ will have low rank. They consider an explicit family g in SM-VBP and show that for every arc-partition (Y, Z) of X , $M_g^{Y,Z}$ has high rank. Hence g is not in SM-VF. The low-rank proof is again probabilistic, but it has a very appealing combinatorial flavour. So does the very definition of an arc-partition.

5 More on completeness

Assume that completeness is defined with respect to p -projections. If a family (f_n) is complete for a class, then understanding (f_n) better allows us to understand the class better. If a natural family is complete for a class, then this is evidence that the class itself is natural.

Valiant started off with a proof that Perm is VNP-complete. He also showed that polynomial families associated with a number of NP-complete languages are complete for VNP under p -projections. So let us agree that VNP is a natural class.

What about VP? The family that naturally contrasts with Perm is Det, but Det is not yet known to be complete for VP (unless we allow qp -projections; that is not quite satisfactory). If this turns out to be the case, it will solve a major open problem, showing that polynomial-degree polynomial-size circuits are no more powerful than polynomial-size

branching programs VBP. VBP seems a natural enough class, and Det and many other families are complete for it.

So what problems are complete for VP? One can construct a canonical family complete for VP. By canonical, I mean something similar to saying that

$$\{\langle M, x, 1^t \rangle \mid M \text{ is an NDTM that accepts } x \text{ in } t \text{ or fewer steps}\}$$

is NP-complete. Undoubtedly true, but it doesn't give any new intuition about what NP is about. In the case of VP, the canonical family is not so trivial to construct (but not very difficult either).

The first description, with a very general completeness result, appears in [Bür00a] (see section 5.6, Cor 5.32(b)). Bürgisser shows that for every p -family h , the *relativized* classes VP^h and VNP^h have complete families with respect to p -projections. Since $VP^h = VP$ and $VNP^h = VNP$ whenever h itself is in VP, this gives families complete for VP and VNP as well. (In fact, it shows the existence of VNP-complete families, independent of Valiant's original proof.) These complete families compute homogeneous components separately, to keep the degree small, and then add up the required parts. They are constructed by first defining *generic polynomials*, and then defining the appropriate projection / substitution. The generic polynomials capture the canonical notion referred to above.

Later, a more direct construction tailored for VP (as opposed to VP^h and VNP^h for all h) was described by Ran Raz [Raz10], and also appears in [SY10]. Here the proof of hardness exploits the fact that we can perform depth-reduction on VP circuits. (This was not needed in Bürgisser's proof.) Roughly, here's how it goes: For each natural number N , consider a circuit C_N with nodes arranged in $2 \log N + 1$ layers numbered $0, 1, \dots, 2 \log N$. All even layers have exactly N nodes, and compute polynomials $g_{i,j}$ where i is the layer number, $j \in [N]$. Odd layers are used to build these polynomials. At layer 0, the polynomials are just distinct variables, $g_{0,j} = x_j$. At higher layers, we have an inductive definition: $g_{i+1,j} = \sum_{k,\ell \in [N]} g_{i,k} \cdot g_{i,\ell} \cdot y_{i,j,k,\ell}$, where the $y_{i,j,k,\ell}$ are new variables. Thus the nodes at the odd layers are the fanin-3 \times nodes, and nodes at even layers (other than the 0 layer) are $+$ nodes with large fanin. (We can reduce the fanins to 2 later; it won't change the polynomial computed.) The polynomial computed by this circuit at $g_{2 \log N, 1}$ is p_N . The total number of variables is $O(N^3 \log N)$, and the circuit is also of size $O(N^3 \log N)$. The degree of p_N is $2N - 1$. So (p_N) is in VP. Why is it VP-hard? Take any family (f_n) in VP. By the depth-reduction of [VSB83], it can be computed in VSAC¹. The VSAC¹ circuit D_n can be normalised to have alternating $+$ and \times nodes, with all \times nodes having fanin 2, and all leaves at the same depth. Choose N at least as large as $\min\{\text{size}(D_n), 2^{\text{depth}(D_n)}\}$, and also at least as large as the number of variables in C_n . Now, the computation of D_n can be embedded into C_N : Choose the right number of $+$ nodes at each even layer, and by carefully assigning 0,1 values to the y variables, ensure that they compute the required combinations of polynomials from the previous even layer.

The circuits described above are called *universal circuits* in [SY10], because every circuit is a projection of the universal circuit of appropriate size. And if we start with VP circuits, the projections are p -projections.

So now we know that VP has complete families under p -projections as well. But generic polynomials, universal circuits, and the polynomials they compute, are rather artificial. Are there other families that are defined independent of circuits and are VP-complete? Actually, we know very few. Recently, Stefan Mengel [Men11] made further progress here, considering polynomial families associated with constraint satisfaction problems CSPs. (This builds on earlier work by Briquel, Koiran, Meer [BK09, BKM11], though they did not explicitly look for VP-completeness.) Let's first review what CSPs are. Think of them as generalising CNF-SAT. In CNF-SAT, each clause forbids one assignment to the variables in it. (eg the clause $x_1 \vee \overline{x_3}$ forbids $x_1 = 0, x_3 = 1$.) In a CSP, variables can take values from a larger domain, not necessarily 0,1. Each constraint is like a clause; it has a set of variables, and it forbids certain combinations of assignments to these variables. (eg on domain $\{a, b, c\}$ a constraint on x_1, x_2 could say that $x_1 \neq x_2$. That is, assignments aa, bb, cc are forbidden, the other 6 assignments satisfy this constraint.) As in SAT, we look for assignments satisfying all constraints. If the domain has size 2, the CSP is Boolean. If each constraint involves 2 (or less) variables, the CSP is binary. As usual, consider not just a CSP but a family of CSPs (Φ_n) , where Φ_n has domain D_n . For tractability, we will require that the CSP is p -bounded; that is, the CSP has bounded arity (for some fixed constant c , each constraint in every Φ_n looks at no more than c variables), and it has polynomial sized domains (in Φ_n , the variables take values from a set D_n , where the size of D_n is p -bounded). Now associate with each such CSP (Φ_n) a polynomial family $(Q_n = Q(\Phi_n))$, where Q_n is on the variable set $\{X_d \mid d \in D_n\}$ and is defined as follows:

$$Q(\Phi_n) = \sum_{a: \text{var}(\Phi_n) \rightarrow D_n} [a \text{ satisfies } \Phi_n] \prod_{x \in \text{var}(\Phi_n)} X_{a(x)} = \sum_{a: \text{var}(\Phi_n) \rightarrow D_n} [a \text{ satisfies } \Phi_n] \prod_{d \in D_n} X_d^{|a^{-1}(d)|}$$

(Recall, $[S]$ is Boolean, 1 if and only if statement S is true.) Mengel has this wonderful characterization of the complexity of the family (Q_n) . The characterization involves associating with the CSP a graph G ; this graph has a vertex for each variable and an edge between two variables if they occur simultaneously in some constraint. Now the treewidth and pathwidth of the graph (these parameters describe roughly how tree-like or path-like the graph is, if we can consider blobs of vertices. The smaller the blobs, the better the similarity. See [Bod98] for definitions and an overview.) relate to the complexity. It also involves an assignment bound: a CSP is c -assignment-bounded if for each constraint φ and each variable x in the constraint, the number of distinct values possible for x in assignments satisfying φ is bounded by c , even though the domain may be much larger. This seems like a strong condition, but recall that Boolean CSPs are by definition 2-assignment-bounded.

Enough of definitions! Here's what Mengel shows:

1. For each p -bounded CSP (Φ_n) , $(Q(\Phi_n))$ is in VNP. Every family (f_n) in VNP is a p -projection of $(Q(\Phi_n))$ for some p -bounded (Φ_n) .
2. For each p -bounded CSP (Φ_n) where G_n has bounded treewidth, $(Q(\Phi_n))$ is in VP. Every family (f_n) in VP is a p -projection of $(Q(\Phi_n))$ for some p -bounded binary (Φ_n) where G is a tree (treewidth 1).

3. For each p -bounded CSP (Φ_n) where G_n has bounded pathwidth, $(Q(\Phi_n))$ is in VBP. Every family (f_n) in VBP is a p -projection of $(Q(\Phi_n))$ for some p -bounded binary (Φ_n) where G is a path (pathwidth 1).
4. For each p -bounded c -assignment-bounded CSP (Φ_n) where G_n has bounded treewidth, $(Q(\Phi_n))$ is in VF. Every family (f_n) in VF is a p -projection of $(Q(\Phi_n))$ for some p -bounded 2-assignment-bounded binary (Φ_n) where G has pathwidth at most 26.

The hardness proofs involve looking at the structure of parse trees for VP, witnessing paths for VBP.

Note that as stated, this falls slightly short of providing a single complete family for VP. However, applying the hardness reduction from universal circuits will yield a single CSP family that is VP-complete. To the best of my knowledge, this is the first instance of a VP-hardness result for a family defined (almost) independent of circuits.

All the above results require that the CSP has bounded arity. Unbounded arity seems to immediately give rise to intractability. If arity is unconstrained, can other types of restrictions still result in families in VP? Some progress in this direction is reported in [DM11].

6 Computing integers

The questions concerning algebraic complexity classes are closely connected to another very intriguing question. Let $N > 1$ be any natural number. Suppose we want to build up N from 1, using only $+$, $-$ and \times . The most naive way of doing this would be $N = 1 + 1 + \dots + 1$. But depending on N there can be many other ways. Which is the *most efficient* way? That is, which way uses the least number of $+$ or \times operations? To do anything non-trivial, we must use $+$ at least once, and the first time we use it we will generate 2. So let us not even count this mandatory $+$. How many more operations are needed?

We can state this as a question about circuits. Each way of building up N is an arithmetic circuit, or a straight-line program (SLP), that uses no constants other than 1 and 2. Let us denote by $\tau(N)$ the size of the smallest such circuit computing N . (This is the τ complexity of N). By definition, $\tau(1) = \tau(2) = 0$, and for all $N > 2$, $\tau(N) > 0$. Algorithms for computing N give upper bounds on $\tau(N)$. For instance, to compute $N = 2^k$, here's an SLP: $g_0 = 2$, $g_{i+1} = 2 \times g_i$ for $0 \leq i \leq k-2$. Clearly, g_i computes 2^{i+1} , so $\tau(2^k) \leq k-1$. But I'm sure you can already see better ways of doing this. From the circuit viewpoint, an explanation of why this is not the best is that the circuit corresponding to this SLP is skew. Surely we should be able to use non-skew gates and compute large numbers faster. Here's another SLP that computes big numbers fast: $f_0 = 2$, $f_{i+1} = f_i \times f_i$ for $0 \leq i \leq \ell-1$. Clearly, f_i computes 2^{2^i} , so $\tau(2^{2^\ell}) \leq \ell$, a much better bound than the earlier $2^\ell - 1$ at least for numbers of this form. Note that the way we used non-skewness, we produced a circuit with exponential formal degree (the degree at f_i is 2^i), but we're not worried about that for now. Now, using these compact circuits for 2^{2^ℓ} , we can build a better circuit for 2^k by just using the binary expansion of k : $k = \sum_{i=0}^t b_i 2^i$, where $t = \lfloor \log k \rfloor$ and $b_t = 1$. So $2^k = 2^{\sum_{i=0}^t b_i 2^i} = \prod_{i=0}^t 2^{b_i \times 2^i} = \prod_{i: b_i=1} 2^{2^i}$.

Compute all the double powers using t operations, and then multiply the required ones using at most t operations. Overall, $\tau(2^k) \leq 2t = 2\lfloor \log k \rfloor$.

We can use the same binary expansion idea to compute any N , not just a power of 2. Compute all powers of 2 upto $\log N$, and add the required ones. This shows that for all N , $\tau(N) \leq 2\lfloor \log N \rfloor - 1$.

So far we have not used any subtractions. But they can be very useful too. For instance, $\tau(2^{2^\ell} - 1) \leq \ell + 1$; compute 2^{2^ℓ} and subtract 1.

What about a lower bound? We can actually formalise the intuition that the exponential degree circuits we saw above for 2^{2^ℓ} produce the largest possible number in that size. Hence, for any N , $\tau(N) \geq \log \log N$.

In particular, $\tau(2^{2^\ell}) = \ell$. That sounds impressive – we know the exact value of τ for 2^{2^ℓ} . But essentially just for that; for all other numbers, we still seem to have a pretty large gap. If $N = 2^k$, then $\log \log N \leq \tau(N) \leq 2\lfloor \log k \rfloor = 2\lfloor \log \log N \rfloor$, so we know $\tau(N)$ within a factor of 2. But for general N , all we know is $\log \log N \leq \tau(N) \leq 2\lfloor \log N \rfloor - 1$. How can we reduce this gap? An obvious search for an efficient way where the last operation is $+$ or $-$ is to express N as $M \pm k$, compute M , compute $k = \pm(N - M)$, and combine, and to choose M that minimizes $\tau(M) + \tau(k) + 1$. (A similar approach can be used for factors of N and a \times as the last operation.) But in computing M and $\pm(N - M)$ (or N/M), the complexity may be *sub-additive* since we can reuse intermediate numbers from the program for M while computing $\pm(N - M)$ or N/M . (We are looking for circuits, not formulas.) It is identifying the extent of this reuse that is a challenge.

Similar to Shannon’s bound for functions and circuits (most functions require exponential sized circuits), Melo and Svaiter [dMS96] showed that most numbers N have $\tau(N)$ closer to the upper bound. They showed that for every $\epsilon > 0$, most N satisfy $\tau(N) \geq \frac{\log N}{(\log \log N)^{1+\epsilon}}$. Moreira [Mor97] improved this by showing that this holds even for $\epsilon = 0$. (He also showed that for all $\epsilon > 0$, there is an N_ϵ such that for all $N \geq N_\epsilon$, $\tau(N) \leq \frac{(1+\epsilon) \log N}{(\log \log N)}$.) And yet, showing such lower bounds for specific numbers seems quite hard – the classic “searching for hay in a haystack” paradox.

Let’s move over from individual numbers to sequences of numbers. Let $(a_n)_{n \geq 1}$ be some sequence of natural numbers. When can we say that the sequence is easy to compute? Each number in the sequence should be “easy” relative to its position in the sequence. That is, the sequence (b_n) , where $b_n = \tau(a_n)$, should not grow very fast. One possible definition is that b_n should be polynomially bounded in n . For instance, for $a_n = 2^{2^n}$, we know that $b_n = n$. Is that not moderate growth? Not really. Consider a function that maps a position n to not just the number $\tau(a_n) = b_n$ but to an SLP of size b_n computing a_n . For the sequence (2^{2^n}) , this function takes an input n represented in $\Theta(\log n)$ bits, and outputs a circuit of size n , that is, exponential in the size of the input. That’s not moderate growth!

OK, so let’s say that a sequence (a_n) is easy to compute if for some polynomial $p(\cdot)$, for each n , $\tau(a_n) \leq p(\log n)$, and otherwise it is hard to compute. We’ve set up this definition so that (2^{2^n}) is hard to compute, while the sequences (n) , (2^n) are easy to compute. Makes sense? Now let’s ask, what other sequences are easy? And what sequences are hard?

A sequence with famously open status is $(n!)$. The completely naive SLP that constructs

the first n numbers with $n-2$ increments and then multiplies them shows that $\tau(n!) \leq 2n-4$. But can this be improved significantly? Or is this sequence hard? The best we know is that $\tau(n!) \in O(\sqrt{n} \log^2 n)$; see [BCS97]. Here is the interesting connection to algebraic circuit complexity. Building on a sequence of constructions by Cheng [Che04] and Koiran [Koi05], Bürgisser [Bür09] showed that if $(n!)$ is hard to compute, then any algebraic circuit for the (Perm_n) family that uses only the constants $-1, 0, 1$ must be of superpolynomial size. If we can't even compute the numbers $n!$ easily, then we cannot compute the polynomials Perm_n efficiently, unless we allow the use of constants that cannot themselves be built up efficiently.

Analogous to the τ complexity of natural numbers, we can define the τ complexity of polynomial families. Let $\tau(f)$ denote the size of the smallest algebraic circuit using only the constants $-1, 0, 1$ – call such a circuit *constant-free* – and computing f . We say that the family (f_n) has polynomially bounded τ complexity if for some polynomial $p(n)$, and for each n , $\tau(f_n) \leq p(n)$. Bürgisser's result can now be stated as: if $\tau(\text{Perm}_n)$ is polynomially bounded, then $(n!)$ is easy to compute.

Let's examine this a bit closely. Why do we state the hypothesis as “ $\tau(\text{Perm}_n)$ is polynomial”? Is this not equivalent to saying (Perm_n) is in VP, and hence $\text{VNP} = \text{VP}$? Actually, it may not be equivalent. It is possible that (Perm_n) has polynomial-sized circuits but no polynomial-sized constant-free circuits. Conceivably, using other constants in intermediate computation and then cancelling them out could help. Recall that the proof of VNP-hardness of (Perm_n) uses constants other than $-1, 0, 1$; $1/2$ is needed. (As another example, recall how in showing that Det_n is a projection of SymDet_n , we needed the constant $1/2$, even though all coefficients in Det_n are $-1, 0, 1$.) So we can define a subclass of VP: families with constant-free circuits of polynomial size.

What can we say about such a subclass? As described above, Bürgisser has shown that if this subclass contains (Perm_n) , then $(n!)$ is easy to compute. Under the same hypothesis, he also shows that the sequences $[2^n e]$, $[(3/2)^n]$ and $[2^n \sqrt{2}]$ are easy to compute.

Malod [Mal03] observed that unlike in the case of VP, for constant-free circuits we may not be able to bound complete formal degree. For VP, if the polynomial computed by a circuit of size s had degree d , we could find an equivalent circuit with formal degree d , and another with complete formal degree $O(d^3 s)$, with only polynomial blow up in size. Not so if constants aren't freely available! Consider the polynomial family $f_n = 2^{2^n}(x_1 + \dots + x_n)$. With arbitrary constants, we have a circuit of size n . With only $-1, 0, 1$, we have a circuit of size $2n + 1$: build 2, build 2^{2^n} , build the linear form, multiply. But this circuit has exponential formal degree, and in fact, using only the constants $-1, 0, 1$, any circuit must have exponential formal degree to build up 2^{2^n} . So this polynomial is in VP, it has constant-free circuits of polynomial size, but it does not have constant-free polynomial-size circuits with polynomially-bounded complete formal degree.

This leads to a definition of a further subclass VP^0 , first defined in [Mal03]: polynomial families computed by constant-free circuits with polynomially bounded complete formal degree. Define VNP^0 analogous to VNP as $\sum \cdot \text{VP}^0$. Check back; our proof that (Perm_n) is in VNP also shows that (Perm_n) is in VNP^0 .

The hypothesis $(\text{Perm}_n) \in \text{VP}^0$ is stronger than saying that $\tau(\text{Perm}_n)$ is polynomially

bounded. What does it imply? Can it lead to more sequences being easier to compute? Firstly, note that $(\text{Perm}_n) \in \text{VP}^0$ does not immediately imply $\text{VP}^0 = \text{VNP}^0$. All we can say is the following, shown by Koiran [Koi05]: If (Perm_n) is in VP^0 , then for every family $(f_n) \in \text{VNP}^0$, there is some polynomially-bounded function $p(n)$ such that the family $(2^{p(n)}f_n)$ is in VP^0 . That is, a “shifted” version of f_n is in VP^0 . The precise shift can be described as follows – we know that f_n is a projection of $\text{Perm}_{q(n)}$ for some polynomially bounded $q(n)$, we assumed that $\text{Perm}_{q(n)}$ can be computed by a circuit C_n of size and formal degree bounded by a polynomial function of n , we take $p(n)$ to be the formal degree of C_n . Now C_n can be massaged to compute $2^{p(n)}f_n$ instead of $\text{Perm}_{q(n)}$.

This motivates another variant of easy-to-compute. Let’s say that a sequence (a_n) of natural numbers is ultimately easy to compute if at least some shifted version of it is easy to compute. That is, there is some other integer sequence A_n such that the sequence $a_n A_n$ is easy to compute. Note that if (a_n) is not ultimately easy, then for infinitely many n , all non-zero multiples of a_n have large τ complexity. Using this property, under the hypothesis that $n!$ is not even ultimately easy to compute, we can obtain a non-trivial derandomization of the Arithmetic-Circuit-Identity-Testing problem; see the last section of [ABKPM09]. Earlier, Koiran showed in [Koi05] that if $n!$ is not even ultimately easy to compute, then we have some separation: either $\text{VP}^0 \neq \text{VNP}^0$, or $\text{P} \neq \text{PSPACE}$. This is curious: we have a consequence involving Boolean classes as well. But it should not be so surprising. VP^0 and VNP^0 are computed by (sums of) constant-free poly-formal-degree algebraic circuits, and these are the arithmetic circuits that arise when consider counting classes like $\#\text{P}$ that count accepting paths of Turing machines. This does not mean that $\text{VNP}^0 = \#\text{P}$; the former is a collection of polynomial families whereas the latter is a collection of functions from strings to whole numbers. But the complexity of evaluating polynomial families in the former collection, at Boolean arguments, is closely related to what the latter collection refers to. Koiran’s proof actually shows the contrapositive: he first shows that if $\text{VP}^0 = \text{VNP}^0$ and $\text{P} = \text{PSPACE}$, then the sequence $\tau((2^\ell)!)$ is polynomially bounded in ℓ . So consider instead of each $n!$ the possibly larger factorial $(2^{\ell(n)})!$, where $2^{\ell(n)-1} < n \leq 2^{\ell(n)}$. Then the sequence $(b_n) = ((2^{\ell(n)})!)$ is easy to compute, and each b_n is a multiple of $n!$, so $(n!)$ is ultimately easy to compute.

Since Perm_n is not known to be complete for VNP^0 , what is? It turns out that for several other VNP -complete families, the hardness proofs use no constants other than $-1, 0, 1$ and the membership proofs use circuits with small formal degree; hence these families are complete for VNP^0 as well. As a concrete example, consider the Hamilton cycle polynomial family HC_n defined as follows: Let distinct variables $x_{i,j}$ label the edges of the complete directed graph D_n . Let C_n denote the set of all directed Hamiltonian cycles in D_n ; elements of C_n can be described by cyclic permutations $\sigma \in S_n$. Then

$$\text{HC}_n(x_{11}, \dots, x_{nn}) = \sum_{\sigma \in C_n} \prod x_{i, \sigma(i)}$$

This family is complete for VNP^0 ; ([Mal03]).

Returning to the question “What does $(\text{Perm}_n) \in \text{VP}^0$ imply?”; Koiran [Koi05] showed that it implies the sequence $\lfloor 2^n \ln 2 \rfloor$ is easy to compute. He also improved the earlier-

mentioned result in two ways, from “[$(VP^0 = VNP^0) \wedge (P = PSPACE)$] $\Rightarrow (n!)$ is ultimately easy to compute” to “[$(Perm_n \in VP^0) \wedge (P = PSPACE)$] $\Rightarrow (n!)$ is easy to compute”.

Under the stronger hypothesis that $VP^0 = VNP^0$, we can show more (again due to [Koi05]). If $VP^0 = VNP^0$, then the sequences $(\sum_{i=1}^{2^n} 2^{i^2-1})$, $\lfloor 2^{2^n} \ln 2 \rfloor$, $\lfloor 2^{2^n} \ln 3 \rfloor$, $\lfloor 2^{2^n} \pi \rfloor$, all have polynomially bounded complexity, something that is not yet known unconditionally.

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References

- [ABKPM09] Eric Allender, Peter Bürgisser, Johan Kjeldgaard-Pedersen, and Peter Bro Miltersen. On the complexity of numerical analysis. *SIAM J. Comput.*, 38(5):1987–2006, 2009.
- [AJMV98a] E. Allender, J. Jiao, M. Mahajan, and V. Vinay. Non-commutative arithmetic circuits: depth reduction and size lower bounds. *Theoretical Computer Science*, 209:47–86, 1998.
- [AJMV98b] Eric Allender, Jia Jiao, Meena Mahajan, and V. Vinay. Non-commutative arithmetic circuits: Depth reduction and size lower bounds. *Theor. Comput. Sci.*, 209(1-2):47–86, 1998.
- [AV08] Manindra Agrawal and V. Vinay. Arithmetic circuits: A chasm at depth four. In *FOCS*, pages 67–75, 2008. See also ECCO TR15-062, 2008.
- [AW11] Eric Allender and Fengming Wang. On the power of algebraic branching programs of width two. In *ICALP (1)*, pages 736–747, 2011.
- [Bar89] D.A. Barrington. Bounded-width polynomial size branching programs recognize exactly those languages in NC^1 . *Journal of Computer and System Sciences*, 38:150–164, 1989.

- [BCS97] P. Bürgisser, M. Clausen, and M.A. Shokrollahi. *Algebraic Complexity Theory*. Springer, 1997.
- [BK09] Irénée Briquel and Pascal Koiran. A dichotomy theorem for polynomial evaluation. In *MFCS*, pages 187–198, 2009.
- [BKM11] Irénée Briquel, Pascal Koiran, and Klaus Meer. On the expressive power of cnf formulas of bounded tree- and clique-width. *Discrete Applied Mathematics*, 159(1):1–14, 2011.
- [BOC92] M. Ben-Or and R. Cleve. Computing algebraic formulas using a constant number of registers. *SIAM Journal on Computing*, 21:54–58, 1992.
- [Bod98] Hans L. Bodlaender. A partial k-arboretum of graphs with bounded treewidth. *Theoretical Computer Science*, 209(12):1 – 45, 1998.
- [Bre74] R P Brent. The parallel evaluation of general arithmetic expressions. *Journal of the ACM*, 21:201–206, 1974.
- [Bür00a] P. Bürgisser. *Completeness and Reduction in Algebraic Complexity Theory*, volume 7 of *Algorithms and Computation in Mathematics*. Springer, 2000.
- [Bür00b] Peter Bürgisser. Cook’s versus Valiant’s hypothesis. *Theor. Comput. Sci.*, 235(1):71–88, 2000.
- [Bür09] Peter Bürgisser. On defining integers and proving arithmetic circuit lower bounds. *Computational Complexity*, 18(1):81–103, 2009.
- [Cai90] Jin-yi Cai. A note on the determinant and permanent problem. *Inf. Comput.*, 84(1):119–127, 1990.
- [CCL10] Jin-yi Cai, Xi Chen, and Dong Li. Quadratic lower bound for permanent vs. determinant in any characteristic. *Computational Complexity*, 19(1):37–56, 2010.
- [Che04] Qi Cheng. On the ultimate complexity of factorials. *Theor. Comput. Sci.*, 326(1-3):419–429, 2004.
- [Csa76] L. Csanky. Fast parallel inversion algorithm. *SIAM J of Computing*, 5:818–823, 1976.
- [Dam91] C. Damm. $\text{DET}=\text{L}^{(\#L)}$. Technical Report Informatik–Preprint 8, Fachbereich Informatik der Humboldt–Universität zu Berlin, 1991.
- [DM11] Arnaud Durand and Stefan Mengel. On polynomials defined by acyclic conjunctive queries and weighted counting problems. *CoRR*, abs/1110.4201, 2011.

- [DMPY12] Zeev Dvir, Guillaume Malod, Sylvain Perifel, and Amir Yehudayoff. Separating multilinear branching programs and formulas. In *STOC*, pages 615–624, 2012.
- [dMS96] W. de Melo and B.F. Svaiter. The cost of computing integers. *Proceedings of the American Mathematical Society*, 124(5):1377–1378, 1996.
- [FMM12] Hervé Fournier, Guillaume Malod, and Stefan Mengel. Monomials in arithmetic circuits: Complete problems in the counting hierarchy. In *STACS*, pages 362–373, 2012.
- [GKKP11] Bruno Grenet, Erich Kaltofen, Pascal Koiran, and Natacha Portier. Symmetric determinantal representation of weakly-skew circuits. In *STACS*, pages 543–554, 2011.
- [Gre12] Bruno Grenet. An upper bound for the permanent versus determinant problem. 2012.
- [Hya79] Laurent Hyafil. On the parallel evaluation of multivariate polynomials. *SIAM J. Comput.*, 8(2):120–123, 1979.
- [JMR12] M Jansen, M Mahajan, and B V Raghavendra Rao. Resource trade-offs in syntactic multilinear arithmetic circuits. *Computational Complexity*, page to appear, 2012. preliminary versions in MFCS 2008 and CSR 2009.
- [Kal85] K. Kalorkoti. A lower bound for the formula size of rational functions. *SIAM J. Comput.*, 14(3):678–687, 1985.
- [Kay10] Neeraj Kayal. Algorithms for arithmetic circuits. *Electronic Colloquium on Computational Complexity (ECCC)*, 17:73, 2010.
- [Koi05] Pascal Koiran. Valiant’s model and the cost of computing integers. *Computational Complexity*, 13(3-4):131–146, 2005.
- [Koi10] Pascal Koiran. Complexity of arithmetic circuits (a skewed perspective). In *Slides from Dagstuhl seminar 10481*. DROPS, 2010. <http://www.dagstuhl.de/Materials/Files/10/10481/10481.KoiranPascal.Slides.pdf>.
- [Mal03] G. Malod. *Polynomes et coefficients*. PhD thesis, Univ. Claude Bernard Lyon 1, 2003.
- [Men11] Stefan Mengel. Characterizing arithmetic circuit classes by constraint satisfaction problems - (extended abstract). In *ICALP (1)*, pages 700–711, 2011.
- [Mes89] Roy Meshulam. On two extremal matrix problems. *Linear Algebra and its Applications*, 114115:261 – 271, 1989. jce:title;Special Issue Dedicated to Alan J. Hoffmanj/ce:title;.

- [MM61] M. Marcus and H. Minc. On the relation between the determinant and the permanent. *Illinois Journal of Mathematics*, 5:376–381, 1961.
- [Mor97] Carlos Gustavo T. de A. Moreira. On asymptotic estimates for arithmetic cost functions. *Proceedings of the American Mathematical Society*, 125(2):pp. 347–353, 1997.
- [MP08] Guillaume Malod and Natacha Portier. Characterizing valiant’s algebraic complexity classes. *J. Complexity*, 24(1):16–38, 2008.
- [MR04] Thierry Mignon and Nicolas Ressayre. A quadratic bound for the determinant and permanent problem. In *International Mathematics Research Notices*, pages 2004–4241, 2004.
- [MV97] M. Mahajan and V. Vinay. Determinant: combinatorics, algorithms, complexity. *Chicago Journal of Theoretical Computer Science* <http://www.cs.uchicago.edu/publications/cjtcs>, 1997:5, dec 1997. Preliminary version in Proceedings of the *Eighth Annual ACM-SIAM Symposium on Discrete Algorithms SODA 1997*, 730–738.
- [P13] G. Pólya. Aufgabe 424. *Archiv der Mathematik und Physik (3)*, 20:271, 1913.
- [Rao10] B. V. Raghavendra Rao. *A Study of Width Bounded Arithmetic Circuits and the Complexity of Matroid Isomorphism*. PhD thesis, The Institute of Mathematical Sciences, Chennai, India., 2010. <http://www.imsc.res.in/xmlui/handle/123456789/177>.
- [Raz06] Ran Raz. Separation of multilinear circuit and formula size. *Theory of Computing*, 2(1):121–135, 2006. preliminary version in FOCS 2004.
- [Raz09] Ran Raz. Multi-linear formulas for permanent and determinant are of super-polynomial size. *J. ACM*, 56(2), 2009. preliminary version in STOC 2004.
- [Raz10] Ran Raz. Elusive functions and lower bounds for arithmetic circuits. *Theory of Computing*, 6(1):135–177, 2010.
- [RSY08] Ran Raz, Amir Shpilka, and Amir Yehudayoff. A lower bound for the size of syntactically multilinear arithmetic circuits. *SIAM J. Comput.*, 38(4):1624–1647, 2008.
- [RY08] Ran Raz and Amir Yehudayoff. Balancing syntactically multilinear arithmetic circuits. *Computational Complexity*, 17(4):515–535, 2008.
- [RY09] Ran Raz and Amir Yehudayoff. Lower bounds and separations for constant depth multilinear circuits. *Computational Complexity*, 18(2):171–207, 2009.

- [Rys63] H.J. Ryser. *Combinatorial mathematics*. Carus mathematical monographs. Mathematical Association of America, 1963.
- [Sav70] Walter J. Savitch. Relationships between nondeterministic and deterministic tape complexities. *J. Comput. Syst. Sci.*, 4(2):177–192, 1970.
- [Spi71] P.M. Spira. On time hardware complexity tradeoffs for boolean functions. In *Proceedings of 4th Hawaii International Symposium on System Sciences*, pages 525–527, 1971.
- [Str73] V. Strassen. Vermeidung von divisionen. *Journal of Reine U. Angew Math*, 264:182–202, 1973.
- [SY10] Amir Shpilka and Amir Yehudayoff. Arithmetic circuits: A survey of recent results and open questions. *Foundations and Trends in Theoretical Computer Science*, 5(3-4):207–388, 2010.
- [Tod92] S. Toda. Classes of arithmetic circuits capturing the complexity of computing the determinant. *IEICE Transactions on Information and Systems*, E75-D:116–124, 1992.
- [Val79] Leslie G. Valiant. Completeness classes in algebra. In *STOC*, pages 249–261, 1979.
- [Val82] L. G. Valiant. Reducibility by algebraic projections. In *Logic and Algorithmic: International Symposium in honour of Ernst Specker*, volume 30, pages 365–380. Monograph. de l’Enseign. Math., 1982.
- [Val92] L. G. Valiant. Why is boolean complexity theory difficult? In M. S. Paterson, editor, *Boolean Function Complexity*. Cambridge University Press, 1992. London Mathematical Society Lecture Notes Series 169.
- [Ven92] H. Venkateswaran. Circuit definitions of nondeterministic complexity classes. *SIAM J. on Computing*, 21:655–670, 1992.
- [Vin91] V Vinay. Counting auxiliary pushdown automata and semi-unbounded arithmetic circuits. In *Proceedings of 6th Structure in Complexity Theory Conference*, pages 270–284, 1991.
- [Vol99] H. Vollmer. *Introduction to Circuit Complexity: A Uniform Approach*. Springer-Verlag New York Inc., 1999.
- [VSB83] Leslie G. Valiant, Sven Skyum, S. Berkowitz, and Charles Rackoff. Fast parallel computation of polynomials using few processors. *SIAM J. Comput.*, 12(4):641–644, 1983.
- [vzG87] Joachim von zur Gathen. Permanent and determinant. *Linear Algebra and its Applications*, 96(0):87 – 100, 1987.