Algebraic Computation Models

February 20, 2012

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- Many areas of Computer Science use algebraic computations: computational algebra and geometry, numerical analysis, signal processing, robotics, et.c.
- Useful approximation to the asymptotic behavior of algebraic algorithms, as computers are allowed to use bigger precision day by day (progress in hardware).

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We can avoid such pitfalls by restricting the algorithm's ability to access individual bits.

Algebraic Straight-Line Programs

Something like the following:

Program. Comp	utes $e \times (x_1 +$	$e) + \pi \times x_2$ i	n ℝ.	
Input: x_1, x_2				
Output: y_4				
$y_1 = x_1 + e$				
$y_2 = \pi \times x_2$				
$y_3 = e \times y_1$				
$y_4 = y_3 + y_2$				

In the above straight-line program, x_1 and x_2 are given as inputs and the output y_4 is computed from previous y_i 's, which are the results of a binary operation in the field; π and e are built-in constants.

▶ Note: straight-line = no conditionals or loops.

Definition. An algebraic straight-line program of length T with input variables $x_1, x_2, \ldots, x_n \in \mathbb{F}$ and built-in constants $c_1, c_2, \ldots, c_n \in \mathbb{F}$ is a sequence of T statements of the form

$$y_i = z_{i_1} * z_{i_2},$$

where * is an operation in \mathbb{F} , and each of z_{i_1}, z_{i_2} is either an input variable, a built-in constant or y_j for j < i.

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The computation consists of executing these simple statements in order, finding values for y_1, y_2, \ldots, y_T . The *output* of the computation is the value of y_T .

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Lemma. The output of a straight-line program of length T with variables x_1, x_2, \ldots, x_n is a polynomial $p(x_1, x_2, \ldots, x_n)$ of degree at most 2^T .

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- FFT: $O(n \log n)$ [Cooley and Tukey].
- ▶ Matrix Multiplication: $O(n^3)$ with the naive method and this can be improved using techniques like Strassen's.
- Determinant: $O(n^3)$ using Gaussian elimination.

 $^{^1}z$ is a primitive $m {\rm th}$ root of unity $\Longleftrightarrow z^m = 1 \mbox{ and } z^k \neq 1, \, \forall k < m$

Algebraic Circuits

Again, an example $(x_1, x_2 \text{ inputs}, e, \pi \text{ constants})$:



The above circuit computes $e \times (x_1 + e) + \pi \times x_2$ in \mathbb{R} . Note the similarity with the straight-line program that we saw earlier.

Definition. An *algebraic circuit* consists of an acyclic graph. The leaves are called *input nodes*, are labeled x_1, x_2, \ldots, x_n and take values in a field \mathbb{F} . We also allow special input nodes labeled with arbitrary constants $c_1, c_2, \ldots, c_k \in \mathbb{F}$. Each internal node, called a *gate*, is labeled with one of the operations $+, \times$. We consider only circuits with a single output node and with the in-degree of each gate being 2.

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Lemma. Let $f : \mathbb{F}^n \to \mathbb{F}$ be some function. If f has an algebraic straight-line program of size S, then it has an algebraic circuit of size 3S. If it is computable by an algebraic circuit of size S then it is computable by an algebraic straight line program of length S.

Definition. Let \mathbb{F} be a field. We say that a family of polynomials $\{p_n\}_{n \in \mathbb{N}}$, where p_n takes n variables over \mathbb{F} , has *polynomially-bounded degree* if there is a constant c s.t. for every n the degree of p_n is at most cn^c .

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Definition. The class $AlgNP_{poly}$ is the class of polynomially bounded degree families $\{p_n\}$ that are definable as

$$p_n(x_1, x_2, \dots, x_n) = \sum_{e \in \{0,1\}^{m-n}} g_m(x_1, x_2, \dots, x_n, e_{n+1}, \dots, e_m),$$

where $g_m \in \mathbf{AlgP_{/poly}}$ and m is polynomial in n.

Definition. A function $f(x_1, x_2, \ldots, x_n)$ is a *projection* of a function $g(y_1, y_2, \ldots, y_n)$ if there is a mapping σ from $\{y_1, y_2, \ldots, y_n\}$ to $\{0, 1, x_1, x_2, \ldots, x_n\}$ s.t.

$$f(x_1, x_2, \ldots, x_n) = g(\sigma(y_1), \sigma(y_2), \ldots, \sigma(y_n)).$$

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Example. Let $f(x_1, x_2) = x_1 + x_2$; f is projection-reducible to $g(y_1, y_2, y_3) = y_1^2 y_3 + y_2$ since $f(x_1, x_2) = g(1, x_1, x_2)$.

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Completeness results based on the above:

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- permanent is AlgNP/poly-complete.

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Interesting fact: we need to show $AlgP_{/poly} \neq AlgNP_{/poly}$ before we can show $P \neq NP$.

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- Very powerful model (e.g. computes x^{2^n} in n steps).
- Could be even more powerful.
 - ► Allowing ≥ comparisons when branching would give it the ability to decide every language in P_{/poly}. (even undecidable!)
 - Allowing rounding as a basic operation would give it the ability of integer factorization in poly-time.

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Definition. For complex c, z define $p_c(z) = z^2 + c$. The *Mandelbrot set* is defined as

 $\mathcal{M} = \{ c \in \mathbb{C} \mid \text{the sequence } p_c(0), p_c(p_c(0)), \dots \text{ is bounded } \}.$



[http://upload.wikimedia.org/wikipedia/commons/2/21/Mandel_zoom_00_mandelbrot_set.jpg]

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Philosophical questions: Roger Penrose vs. Artificial Intelligence.

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