

On the complexity of computing determinants

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Overview

1. Faster bit complexity without Strassen matrix multiplication
2. New speed-ups: the use of blocking
With Gilles Villard (middle)



Matrix determinant definition

$$\det(Y) = \det\left(\begin{bmatrix} y_{1,1} & \cdots & y_{1,n} \\ y_{2,1} & \cdots & y_{2,n} \\ \vdots & & \vdots \\ y_{n,1} & \cdots & y_{n,n} \end{bmatrix}\right) = \sum_{\sigma \in S_n} \left(\text{sign}(\sigma) \prod_{i=1}^n y_{i,\sigma(i)} \right),$$

where $y_{i,j}$ are from an *arbitrary commutative ring*, and S_n is the set of all permutations on $\{1, 2, \dots, n\}$.

Interesting rings: \mathbb{Z} , $\mathbb{K}[x_1, \dots, x_n]$, $\mathbb{K}[x]/(x^n)$

1. Bit complexity of linear algebra problems

Strassen's [1969] $O(n^{2.81})$ matrix multiplication algorithm

$$\begin{array}{l} m_1 \leftarrow (a_{1,2} - a_{2,2})(b_{2,1} - b_{2,2}) \\ m_2 \leftarrow (a_{1,1} + a_{2,2})(b_{1,1} + b_{2,2}) \\ m_3 \leftarrow (a_{1,1} - a_{2,1})(b_{1,1} + b_{1,2}) \\ m_4 \leftarrow (a_{1,1} + a_{1,2})b_{2,2} \\ m_5 \leftarrow a_{1,1}(b_{1,2} - b_{2,2}) \\ m_6 \leftarrow a_{2,2}(b_{2,1} - b_{1,1}) \\ m_7 \leftarrow (a_{2,1} + a_{2,2})b_{1,1} \end{array} \left| \begin{array}{l} a_{1,1}b_{1,1} + a_{1,2}b_{2,1} = m_1 + m_2 - m_4 + m_6 \\ a_{1,1}b_{1,2} + a_{1,2}b_{2,2} = m_4 + m_5 \\ a_{2,1}b_{1,1} + a_{2,2}b_{2,1} = m_6 + m_7 \\ a_{2,1}b_{1,2} + a_{2,2}b_{2,2} = m_2 - m_3 + m_5 - m_7 \end{array} \right.$$

Problems reducible to matrix multiplication:

linear system solving [Bunch and Hopcroft 1974],...

Coppersmith and Winograd [1990]: $O(n^{2.38})$

Life after Strassen: bit complexity

Linear system solving $x = A^{-1}b$ where $A \in \mathbb{Z}^{n \times n}$ and $b \in \mathbb{Z}^n$:

With Strassen [McClellan 1973]:

Step 1: For prime numbers p_1, \dots, p_k Do

Solve $Ax^{[j]} \equiv b \pmod{p_j}$ where $x^{[j]} \in \mathbb{Z}/(p_j)$

Step 2: Chinese remainder $x^{[1]}, \dots, x^{[k]}$ to $A\bar{x} \equiv b \pmod{p_1 \cdots p_k}$

Step 3: Recover denominators of x_i by continued fractions of $\frac{\bar{x}_i}{p_1 \cdots p_k}$.

Length of integers: $k = (n \max\{\log \|A\|, \log \|b\|\})^{1+o(1)}$

Bit complexity: $n^{3.38} \max\{\log \|A\|, \log \|b\|\}^{1+o(1)}$

With Hensel  lifting [Moenck and Carter 1979]:

Step 1: For $j = 0, 1, \dots, k$ and a prime p Do

Compute $\bar{x}^{[j]} = x^{[0]} + px^{[1]} + \dots + p^j x^{[j]} \equiv x \pmod{p^{j+1}}$

$$1.a. \quad b^{[j]} = \frac{b - A\bar{x}^{[j-1]}}{p^j} = \frac{b - (A\bar{x}^{[j-2]} + Ap^{j-1}x^{[j-1]})}{p^j}$$

$$1.b. \quad x^{[j]} \equiv A^{-1}b^{[j]} \pmod{p} \text{ reusing } A^{-1} \pmod{p}$$

Step 3: Recover denominators of x_i by continued fractions of $\frac{\bar{x}_i^{[k]}}{p^k}$.

With classical matrix arithmetic:

Bit complexity of 1.a: $n(n \max\{\log \|A\|, \|b\|\})^{1+o(1)} + n^2(\log \|A\|)^{1+o(1)}$

Total bit complexity: $(n^3 \max\{\log \|A\|, \log \|b\|\})^{1+o(1)}$

Bit complexity of the determinant

With Chinese remaindering: $O(n \log \|A\|)$ times matrix multiplication complexity.

Sign [Clarkson early 1990s]: $O(n^3)$ floating point operations with $O(n)$ precision.

Using denominators of linear system solutions [Abbot, Bronstein, Mulders 1999]: fast when first invariant factor is large.

Wiedemann's [1986] determinant algorithm

For $u, v \in \mathbb{K}^n$ and $A \in \mathbb{K}^{n \times n}$ consider the sequence of field elements

$$a_0 = u^T v, a_1 = u^T A v, a_2 = u^T A^2 v, a_3 = u^T A^3 v, \dots$$

The minimal polynomial of A linearly generates $\{a_i\}_{i=0,1,\dots}$.

By the Berlekamp/Massey [1967] algorithm we can compute in $n^{1+o(1)}$ arithmetic operations a minimal linear generator for $\{a_i\}_{i=0,1,\dots}$.

Wiedemann randomly perturbs A and chooses random u and v ; then

$$\det(\lambda I - A) = \text{minimal recurrence polynomial of } \{a_i\}_{i=0,1,\dots}.$$

Detail of algorithm

[exactly like my division-free determinant algorithm ISSAC 92]

For $i = 0, 1, \dots, 2n - 1$ Do Compute the $a_i = u^T A^i v$;

Done by baby steps/giant steps: let $r = \lceil \sqrt{2n} \rceil$ and $s = \lceil 2n/r \rceil$.

Substep 1. For $j = 1, 2, \dots, r - 1$ Do $v^{[j]} \leftarrow A^j v$;

Substep 2. $Z \leftarrow A^r$;
[$O(n^3)$ operations; integer length $(\sqrt{n} \log \|A\|)^{1+o(1)}$]

Substep 3. For $k = 1, 2, \dots, s$ Do $u^{[k]T} \leftarrow u^T Z^k$;
[$O(n^{2.5})$ operations; integer length $(n \log \|A\|)^{1+o(1)}$]

Substep 4. For $j = 0, 1, \dots, r - 1$ Do
 For $k = 0, 1, \dots, s$ Do $a_{kr+j} \leftarrow \langle u^{[k]}, v^{[j]} \rangle$.

Using fast rectangular matrix multiplication: $O(n^{3.064} \log \|A\|)$

Theorem 1

The determinant of an integer matrix can be computed in $O(n^{2.809} \log \|A\|)$ bit operations.

(Exponent 2.69 is a possibility but proofs need to be completed.)

Theorem 2

The determinant of a matrix over a commutative ring can be computed with $O(n^{2.69})$ ring additions, subtractions and multiplications.

Problem 1 (from my 3ECM 2000 talk)

Improve the bit complexity of algorithms for the determinant, resultant, linear system solution, over the integers.

2. Coppersmith's blocking

Use of the block vectors $\mathbf{x} \in \mathbb{K}^{n \times \beta}$ in place of u
 $\mathbf{y} \in \mathbb{K}^{n \times \beta}$ in place of v

$$\mathbf{a}_i = \mathbf{x}^{\text{Tr}} B^i \mathbf{y} \in \mathbb{K}^{\beta \times \beta}, \quad 0 \leq i < \frac{2n}{\beta} + 2.$$

Find a matrix polynomial $\mathbf{c}_0 + \mathbf{c}_1 \lambda + \cdots + \mathbf{c}_d \lambda^d \in \mathbb{K}^{\beta \times \beta}[\lambda]$,
 $d = \lceil n/\beta \rceil$, such that

$$\forall j \geq 0: \sum_{i=0}^d \mathbf{a}_{j+i} \mathbf{c}_i = \sum_{i=0}^d \mathbf{x}^{\text{Tr}} B^{i+j} \mathbf{y} \mathbf{c}_i = \mathbf{0} \in \mathbb{K}^{\beta \times \beta}$$

Probabilistic analysis

Theorem [K&V 2000]: *If B is nonsingular with distinct eigenvalues then we have for the **minimal** generating polynomial*

$$\det(\mathbf{c}_0 + \mathbf{c}_1\lambda + \cdots + \mathbf{c}_d\lambda^d) = \det(\lambda I - B)$$

for random \mathbf{x}, \mathbf{z} with probability

$$\geq 1 - \frac{2n - 1}{|\mathbb{K}|}.$$

Distinct eigenvalues can be obtained by preconditioning B à la [Wiedemann, 1986], for instance

$\tilde{B} \leftarrow V \cdot B \cdot W \cdot G$ where V is randomized butterfly network
 W is randomized butterfly network
 G is random diagonal

Proof idea for probabilistic analysis

$$(I - \lambda B)^{-1} = I + B\lambda + B^2\lambda^2 + \dots$$

$$\mathbf{x}^{\text{Tr}}(I - \lambda B)^{-1}\mathbf{y}(\mathbf{c}_d + \dots + \mathbf{c}_0\lambda^d) = R(\lambda) \in \mathbb{K}[\lambda]^{\beta \times \beta}$$

$$\mathbf{x}^{\text{Tr}}(I - \lambda B)^{-1}\mathbf{y} = R(\lambda)(\mathbf{c}_d + \dots + \mathbf{c}_0\lambda^d)^{-1}$$

Use theorems from multivariable control theory (irreducible realizations) to show that polynomial denominators are the same.

Show run-time estimates in Maple worksheet

```
> beta := n^sigma; # blocking factor
      beta := n^sigma
> s := n^tau; # number of giant steps
      s := n^tau
> r := simplify( (n/beta) / s); # number of baby steps
      r := n^(1-sigma-tau)
```

Standard matrix arithmetic, quadratic B/M, Chinese remainder integer arithmetic

Step 1.1: Compute $B^j y$, $j = 0, \dots, r$

```
> substep1 := simplify( r * beta * n^2 * r );
      substep1 := n^(4-sigma-2tau)
```

Step 1.2: Compute $Z = B^r$ by repeated squaring

```
> substep2 := simplify( n^3 * r );
      substep2 := n^(4-sigma-tau)
```

Step 1.3: Compute $x^{Tr} Z^k$, $k = 0, \dots, s$

```
> substep3 := simplify( s * beta * n^2 * n/beta );
      substep3 := n^(tau+3)
```

Step 1.4: Compute $(x^{Tr} Z^k) (B^j y)$

```
> substep4 := simplify( r * s * beta^2 * n * n/beta );
      substep4 := n^3
```

Step 2: Blocked Berlekamp/Massey for n moduli

```
> step2 := simplify( (n/beta)^2 * beta^3 * n );
```

$$\text{step2} := n^{(3+\sigma)}$$

Step 3: Determinant of generator matrix polynomial for n moduli

```
> step3 := simplify( beta^3 * n * n );
```

$$\text{step3} := n^{(3\sigma+2)}$$

Overall bit complexity

```
> eval([substep1, substep2, substep3, substep4, step2, step3],
```

```
> {sigma=1/3, tau=1/3});
```

$$[n^3, n^{(10/3)}, n^{(10/3)}, n^3, n^{(10/3)}, n^3]$$

“The asymptotically best algorithms frequently turn out to be worst on all problems for which they are used.”

— D. G. CANTOR and H. ZASSENHAUS (1981)

Fast matrix multiplication $O(n^\omega)$, linear B/M, linear integer arithmetic

Step 1.1: Compute $B^j y$, $j = 0, \dots, r$ by $B^{(2^i)}$ [$B^0 y$ — ... — $B^{(2^i-1)} y$]

```
> fastsubstep1 := simplify( n^omega * r );
```

$$\text{fastsubstep1} := n^{(\omega+1-\sigma-\tau)}$$

Step 1.2: Compute $Z = B^r$ by repeated squaring

```
> fastsubstep2 := simplify( n^omega * r );
```

$$\text{fastsubstep2} := n^{(\omega+1-\sigma-\tau)}$$

Step 1.3: Compute $x^{Tr} Z^k$, $k = 0, \dots, s$ as *fat* vectors times a *thin* matrix

```
> fastsubstep3 := simplify( s * (n/(beta*s))^2 * (beta*s)^omega * r,
```

```
> 'power', 'symbolic' );
```

$$\text{fastsubstep3} := n^{(-2\tau+3-3\sigma+(\sigma+\tau)\omega)}$$

Step 1.4: Compute $(x^{Tr} Z^k) (B^j y)$

```
> fastsubstep4 := simplify( r^2/s * (beta*s)^omega * n/beta, 'power',  
> 'symbolic' );
```

$$fastsubstep4 := n^{(3-3\sigma-3\tau+(\sigma+\tau)\omega)}$$

Step 2: Blocked Berlekamp/Massey for n moduli

```
> faststep2 := simplify( beta^2 * n * n);
```

$$faststep2 := n^{(2\sigma+2)}$$

Step 3: Determinant of generator matrix polynomial for n moduli

```
> faststep3 := simplify( beta^omega * n, 'power', 'symbolic' );
```

$$faststep3 := n^{(\sigma\omega+1)}$$

Overall bit complexity

```
> total := eval([fastsubstep1, fastsubstep2, fastsubstep3,  
> fastsubstep4, faststep2, faststep3]);
```

$$total := [\\ n^{(\omega+1-\sigma-\tau)}, n^{(\omega+1-\sigma-\tau)}, n^{(-2\tau+3-3\sigma+(\sigma+\tau)\omega)}, n^{(3-3\sigma-3\tau+(\sigma+\tau)\omega)}, n^{(2\sigma+2)}, n^{(\sigma\omega+1)} \\]$$

```
> expos:=simplify(map(x -> log[n](x), total), 'symbolic');
```

$$expos := [\omega + 1 - \sigma - \tau, \omega + 1 - \sigma - \tau, -2\tau + 3 - 3\sigma + \sigma\omega + \omega\tau, 3 - 3\sigma - 3\tau + \sigma\omega + \omega\tau, 2\sigma + 2, \sigma\omega + 1]$$

```
> numexpos:=eval(expos, omega=2.3755);
```

$$numexpos := [3.3755 - \sigma - \tau, 3.3755 - \sigma - \tau, .3755\tau + 3. - .6245\sigma, 3. - .6245\sigma - .6245\tau, 2\sigma + 2, 2.3755\sigma + 1]$$

```
> minexpo := solve({numexpos[2]=numexpos[3],
```

```
> numexpos[2]=numexpos[5]}, {sigma,tau});
```


$minexpo := \{\sigma = .4042922554, \tau = .1626232338\}$

> `eval(numexpos, minexpo);`

[2.808584511, 2.808584511, 2.808584510, 2.645961276, 2.808584511, 1.960396253]