

About these slides

- ▶ Part of the Cryptanalysis course I taught at UCL in 2015 for the Master in Information Security
- ▶ Contain background computer algebra algorithms useful for both that course and this one
- ▶ The slides will not be covered during this course
- ▶ Best usage : know what is in them and consult when needed



Public Key Cryptanalysis

Algorithmic Number Theory Basics

Christophe Petit

University College London



Secure communications

- ▶ Alice wants to send a **private** message to Bob over a **public** channel
- ▶ Private key cryptography : Alice and Bob both have a key to some encryption box



- ▶ Public key cryptography : Alice uses a lock of which only Bob has the key



Public key vs Private key cryptography

- ▶ No preshared password needed with public key crypto
- ▶ Security *reduced* to “hard” number theory problems vs. “ad hoc” security for block ciphers, hash functions
- ▶ Mathematical problems have independent interest, so more scrutinized... for the best and the worst
- ▶ Typically ~ 1500 bits vs. ~ 160 bits



Module objectives

- ▶ Revise algorithmic number theory basics from IntroCrypto
- ▶ Revise Linear Algebra basics
- ▶ If time : learn root-finding algorithms
- ▶ Lab & tutorial : discover SAGE and connect theory to practice, play with some first attacks



Outline

Complexity measures

Algebra and number theory

First algorithmic number theory tools

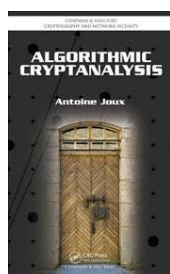
Linear algebra

Root-finding algorithms



Reference book

- ▶ Algorithmic Cryptanalysis, Chapters 1-3



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What do we mean by “hard” problem ?



- ▶ Is  hard ?
- ▶ Is adding two integers hard ?
- ▶ Is multiplying two integers hard ?
- ▶ Is factoring integers hard ? what about 15 ?
- ▶ Is inverting a matrix hard ? what if it has billions of rows and columns ?



Big Oh notation

- ▶ Let $f, g : \mathbb{N} \rightarrow \mathbb{R}$. We say $f = O(g)$ if there exist N and c such that for all $n > N$, we have $g(n) \geq cf(n)$.
- ▶ Examples :
 - ▶ $x = O(x^2)$
 - ▶ $10000000x = O(x^2)$
 - ▶ $x^n = O(e^x)$ for any n
 - ▶ $\log x = O(x)$



Measuring complexity (theory)

- ▶ Consider the multiplication problem : given two integers p and q , compute $n = pq$
- ▶ Hardness is function of $s := \log_2 p + \log_2 q$, the input size
- ▶ Trivial algorithm runs in time $O(\log_2 p \cdot \log_2 q) = O(s^2)$: multiply p by each bit of q , shift by appropriate powers of 2, and make additions with carries
- ▶ Best algorithms achieve $O(s \log s)$



Measuring complexity (theory)

- ▶ Consider exhaustive search on a key of n bits
- ▶ Hardness is function of n
- ▶ Complexity is $O(2^n)$: try every possible key
- ▶ Exponential complexity !



Measuring complexity (theory)

- ▶ Consider the factorization problem : given a positive composite integer n , find p and q such that $n = pq$
- ▶ Hardness is function of $\log_2 n$, that is the size of input
- ▶ The best algorithms today run in **subexponential** time

$$L_n(\alpha; c) = \exp(c(\log n)^\alpha (\log \log n)^{1-\alpha})$$

with $\alpha = 1/3$



P and NP

- ▶ A problem is in P if it can be solved in polynomial time (in other words, there is an integer n such that it can be solved in time $O(x^n)$ for an input of size x)
- ▶ Refinements to this : randomization, memory, etc.
- ▶ A problem is in NP if a solution can be checked in polynomial time
- ▶ P=NP? is worth a million dollars (and glory!)
- ▶ NP-complete problems are as hard as the hardest known NP problems such as 3-SAT, graph coloring, traveling salesman, etc
- ▶ Factorization, Dlog, are (probably) NOT NP-complete



In practice

- ▶ Hardness depends on your computer power, your time, your memory
- ▶ Hard for you might be easy for NSA
- ▶ Compare with exhaustive search : 2^{20} is certainly possible on a laptop, 2^{60} becomes very hard for most organizations
- ▶ See www.keylength.com for key sizes



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Group

- ▶ A group (G, \circ) is a set G with some binary operation $\circ : G \times G \rightarrow G$ such that
 - ▶ Neutral element : there exists $e \in G$ such that for all $x \in G$, we have $x \circ e = x = e \circ x$
 - ▶ Inverse : for all $x \in G$, there exists y such that $x \circ y = e = y \circ x$
 - ▶ Associativity : for all $x, y, z \in G$, we have $(x \circ y) \circ z = x \circ (y \circ z)$
- ▶ When \circ is implicit, we say G is a group
- ▶ A group is Abelian if for all x, y , we have $x \circ y = y \circ x$
- ▶ A group is finite if $|G|$ is finite



Group examples

- ▶ $(\mathbb{Z}, +)$ is a group with neutral element 0
- ▶ $(\mathbb{Q}, +)$ is a group with neutral element 0
- ▶ $(\mathbb{Q}, *)$ is not a group : 0 has no inverse
- ▶ $(\mathbb{Q}^*, *)$ is a group with neutral element 1
Here $\mathbb{Q}^* = \mathbb{Q} \setminus \{0\}$
- ▶ $(\mathbb{Z}_n, +)$ is a group for any positive integer n
Here $\mathbb{Z}_n = \mathbb{Z}/n\mathbb{Z}$ are integers modulo n
- ▶ $(\mathbb{Z}_p^*, *)$ is a group for any prime number p
Here $\mathbb{Z}_p^* = \mathbb{Z}_p \setminus \{0\}$
- ▶ ...



Rank of a group

- ▶ The rank of a group $(G, +)$ is the minimal number of elements needed to generate the whole group

$$\min\{k : \exists S = \{g_1, \dots, g_k\} \subset G \text{ s.t. } \forall g \in G, g = \sum_i g_{e_i} \text{ with } g_{e_i} \in S\}$$

- ▶ Example : $(\mathbb{Z} \times \mathbb{Z}, +)$ is a group of rank 2 with generating set $\{(1, 0), (0, 1)\}$
- ▶ A group of rank 1 is called a cyclic group



Lagrange theorem

- ▶ Let (G, \circ) a finite group
- ▶ For any integer k and any $g \in G$, we write g^k for $g \circ g \circ \dots \circ g$, k times
- ▶ Lagrange's theorem : for any $g \in G$, we have $g^{|G|} = e$ where e is the neutral element in the group
- ▶ Fermat's small theorem : for any prime p and any $g \neq 0 \pmod p$, we have $g^{p-1} = 1 \pmod p$



Field

- ▶ A field $(K, +, *)$ is a set K with two binary operations $+: K \times K \rightarrow K$ and $*: K \times K \rightarrow K$ such that
 - ▶ $(K, +)$ is an Abelian group
 - ▶ $(K^*, *)$ is a group, where $K^* = K \setminus \{e\}$, where e is the neutral element of K for $+$
- ▶ A field $(K, +, *)$ is finite if $|K|$ is finite



Field examples

- ▶ $(\mathbb{C}, +, *)$ is a field with neutral elements 0 and 1 for $+$ and $*$
- ▶ $(\mathbb{Q}, +, *)$ is a field with neutral elements 0 and 1 for $+$ and $*$
- ▶ $(\mathbb{Z}_p, +, *)$ is a finite field for any prime p
This field is often denoted \mathbb{F}_p



A more complicated example

- ▶ Let f be a polynomial of degree n with coefficients in \mathbb{F}_p , such that f has no factor of degree different than 0 or n .
- ▶ Consider $(K, +, *)$ where
 - ▶ $K =$ all polynomials over \mathbb{F}_p
 - ▶ $+$ and $*$ are addition and multiplication modulo the polynomial f
- ▶ Then $(K, +, *)$ is a finite field with p^n elements
- ▶ Example : let $f(x) = x^2 + x + 1 \in \mathbb{F}_2[x]$ then $\mathbb{F}_4 = \mathbb{F}_2[x]/(f(x)\mathbb{F}_2[x])$ is a finite field with 4 elements $\{0, 1, x, x + 1\}$



Vector space

- ▶ A vector space $(V, +, *)$ over some field K is a set $V \supset K$ with two operations $+: V \times V \rightarrow V$ and $*: K \times V \rightarrow V$ such that
 - ▶ $(V, +)$ is a group
 - ▶ For all $a, b \in K$ and all $v \in V$, we have $(a + b) * v = a * v + b * v$
 - ▶ For all $a \in K$ and $v, w \in V$, we have $a * (v + w) = a * v + a * w$
- ▶ The dimension of this vector space is the rank of $(V, +)$
- ▶ A basis of V is a set of $(\dim V)$ elements that generate V



Ring

- ▶ A ring $(R, +, *)$ is a set R with two operations $+ : R \times R \rightarrow R$ and $* : R \times R \rightarrow R$ such that
 - ▶ $(R, +)$ is an Abelian group
 - ▶ $(R, *)$ is associative and has a neutral element (but some elements may have no inverse)
 - ▶ Distributivity : for all $a, b, c \in R$, we have $(a + b) * c = a * c + b * c$



Ring examples

- ▶ Let K be a field and let $K[X]$ be the set of polynomials with coefficients in K . Then $(K[X], +, *)$ is a ring
- ▶ $\mathbb{Z}_n := \mathbb{Z}/n\mathbb{Z}$ (the integers modulo n) is a ring for any $n \in \mathbb{N}$. It is a field if and only if n is prime.
- ▶ Let K be a field. Let $f \in K[X]$ and let $\tilde{K} = K[X]/(f(X))$ be the set of polynomials over K "modulo $f(x)$ ". Then \tilde{K} is a ring. It is a field if and only if f is irreducible.



Prime numbers

- ▶ 2,3,5,7,11,... are prime numbers. 4,6,8,9,10,... are not
- ▶ Any integer n can be decomposed uniquely as a product of prime numbers
- ▶ There are infinitely many primes
- ▶ **Prime number theorem** : the number of primes up to some bound B is roughly equal to $B/\log B$



The RSA ring

- ▶ Let p, q be two primes and let $n = pq$
- ▶ Let $\mathbb{Z}_n := \mathbb{Z}/n\mathbb{Z}$ be the ring of integers modulo n
- ▶ Not a field : for any k , neither kp nor kq are invertible
- ▶ The map

$$\varphi : \mathbb{Z}_n \rightarrow \mathbb{Z}_p \times \mathbb{Z}_q : x \rightarrow (x \bmod p, x \bmod q)$$

is a ring isomorphism. Its inverse is given by

$$\begin{aligned} \varphi^{-1} : \mathbb{Z}_p \times \mathbb{Z}_q &\rightarrow \mathbb{Z}_n \\ (x_p, x_q) &\rightarrow x_p q (q^{-1} \bmod p) + x_q p (p^{-1} \bmod q) \end{aligned}$$



Chinese remainder theorem

- ▶ More generally if $n = \prod_{i=1}^N p_i^{e_i}$ then the map

$$\varphi : \mathbb{Z}_n \rightarrow \prod_{i=1}^N \mathbb{Z}_{p_i^{e_i}} : x \mapsto (x \bmod p_1^{e_1}, \dots, x \bmod p_N^{e_N})$$

is a ring isomorphism

- ▶ In other words given all residue values, there exists a unique value that corresponds to them modulo n



Euler's theorem

- ▶ Let $n = \prod_{i=1}^N p_i^{e_i}$ where the p_i are distinct primes
- ▶ Define the Euler totient function

$$\varphi(n) = \prod_{i=1}^N (p_i - 1)p_i^{e_i - 1}$$

- ▶ Then for all $x \in \mathbb{Z}_n^*$, we have

$$x^{\varphi(n)} = 1 \bmod n$$

- ▶ If $n = p$ a prime, then $\varphi(n) = p - 1$ and we recover Fermat's small theorem $x^{p-1} = 1 \bmod p$
- ▶ If $n = pq$ like in RSA, then $\varphi(n) = (p - 1)(q - 1)$



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Addition in \mathbb{F}_p

- ▶ Let p be a prime and let $K := \mathbb{F}_p = \mathbb{Z}/p\mathbb{Z}$
- ▶ Addition in K : given a and b , return $a + b \bmod p$
 - 1: $c \leftarrow a + b$
 - 2: **if** $c > p$ **then**
 - 3: $c \leftarrow c - p$
 - 4: **end if**
 - 5: **return** c
- ▶ Complexity $O(\log p)$ bit operations



Multiplication in \mathbb{F}_p

- ▶ Let p be a prime and let $K := \mathbb{F}_p = \mathbb{Z}/p\mathbb{Z}$
- ▶ Multiplication in K : given a and b , return $ab \bmod p$
 - 1: Let $b = \sum_{i=0}^n b_i 2^i$
 - 2: $a' \leftarrow a$; $c \leftarrow b_0 a$
 - 3: **for** $i=1$ **to** n **do**
 - 4: $a' \leftarrow 2a' \bmod p$
 - 5: $c \leftarrow c + b_i a' \bmod p$
 - 6: **end for**
 - 7: **return** c
- ▶ Complexity $O(n^2) = O(\log^2 p)$ bit operations
- ▶ Best algorithms achieve $O(\log p \log \log p)$



Modular exponentiation : Square-and-Multiply

- ▶ Let p be a prime and let $K := \mathbb{F}_p = \mathbb{Z}/p\mathbb{Z}$
- ▶ Exponentiation in K : given a and k , return $a^k \bmod p$
 - 1: Let $k = \sum_{i=0}^n k_i 2^i$
 - 2: $a' \leftarrow a$; $c \leftarrow a^{k_0}$
 - 3: **for** $i=1$ **to** n **do**
 - 4: $a' \leftarrow a'^2 \bmod p$
 - 5: $c \leftarrow c(a')^{k_i} \bmod p$
 - 6: **end for**
 - 7: **return** c
- ▶ Complexity $O(n) = O(\log p)$ multiplications



Remark on elementary operations

- ▶ The above algorithms can be generalized to a great extent to other fields, ring or group structures, with similar complexities



The discrete logarithm problem

- ▶ Let p be a prime and let $K := \mathbb{F}_p = \mathbb{Z}/p\mathbb{Z}$
- ▶ Exponentiation in K in $O(n) = O(\log p)$ multiplications
- ▶ What about the inverse operation ?
- ▶ **Discrete logarithm problem :**
Given g and $h = g^k \bmod p$, compute k
- ▶ Believed to be very hard : subexponential complexity $L_p(1/3, c)$
- ▶ More generally : given G , $g \in G$ and $h = g^k$, compute k
- ▶ Can be harder or easier depending on the group



Diffie-Hellman algorithm



- ▶ Designed by Diffie and Hellman in 1976
- ▶ Widely used today, e.g. in SSL
- ▶ Allows two parties to set up a common private key over a public channel
- ▶ Security requires hardness of discrete logarithm problem



Diffie-Hellman algorithm

- ▶ Public elements : G cyclic, $g \in G$ a generator
- ▶ Alice chooses random a and sends g^a to Bob
- ▶ Bob chooses random b and sends g^b to Alice
- ▶ Alice computes $(g^b)^a = g^{ab}$
- ▶ Bob computes $(g^a)^b = g^{ab}$



Diffie-Hellman security

- ▶ Solving discrete logarithm problem is sufficient to break Diffie-Hellman key exchange
- ▶ Solving discrete logarithm problem might not be necessary to break Diffie-Hellman key exchange
- ▶ Additional stuff is required for authentication, for example certificates



Primality testing

- ▶ Given an integer n , decide whether n is prime or not
- ▶ You can generate primes by picking random numbers smaller than B and checking whether they are prime : need about $\log B$ trials by the prime number theorem
- ▶ There are deterministic algorithms for primality testing (see AKS test)
- ▶ In practice, we use probabilistic algorithms (having a small probability to return prime for composite numbers) that are much faster



Fermat test

- ▶ Observation : if n is prime then $a^n = a \pmod n$ for all a (Fermat's small theorem)
- ▶ Idea : choose random a and check whether $a^n = a \pmod n$. If not then n is composite.
- ▶ Bad news : some numbers (Carmichael numbers) are composite and satisfy this equation for all $0 < a < n!$



Miller-Rabin test

- ▶ Observation : if n is prime, then the only x such that $x^2 = 1 \pmod n$ are $\pm 1 \pmod n$ whereas if n is composite, there are more of them
- ▶ Idea : write $n - 1 = 2^k q$, pick random a and compute $a_0 = a^q \pmod n$, then $a_i = a_{i-1}^2 \pmod n$, etc, up to $a_k = a^{n-1} \pmod n$
 - ▶ If n is prime : the sequence (a_0, a_1, \dots, a_k) will be $(*, *, \dots, *, -1, 1, \dots, 1)$ where $* \neq \pm 1$
 - ▶ If n is composite then it will be $(*, *, \dots, *, *, 1, \dots, 1)$ for at least 3/4 of the values a
- ▶ Complexity $O(-\log \epsilon)$ modular exponentiations, where ϵ is error probability



RSA algorithm



- ▶ Designed by Rivest-Shamir-Adleman in 1977
- ▶ One of the most widely used algorithms today, for both signatures and public key encryption
- ▶ Security requires hardness of integer factorization



RSA encryption algorithm

- ▶ Let p, q two distinct odd primes, and let $n = pq$
- ▶ Let e with no common divisor with $\varphi(n) = (p-1)(q-1)$
- ▶ **Public key** is (n, e) and **private key** is (p, q)
- ▶ Given private key, can also compute $d := e^{-1} \pmod{\varphi(n)}$
- ▶ Encryption of m is $c = m^e \pmod n$
- ▶ Decryption of c is $m' = c^d \pmod n$
- ▶ Correctness follows from

$$m' = (m^e)^d = m^{ed \pmod{\varphi(n)}} = mm^{(ed-1) \pmod{\varphi(n)}} = m \pmod n$$

by Euler's theorem



RSA security

- ▶ Solving the factorization problem is sufficient and necessary to reconstruct the private key
- ▶ Solving the factorization problem *might not be necessary* for other goals, such as decrypting without the private key
- ▶ In fact, “textbook RSA” insecure wrt some goals : for example given an encryption of m , can compute an encryption of $m^2 \bmod n$



RSA weak key generator attack

- ▶ Suppose Alice uses private key (p, q_a) and Bob uses private key (p, q_b) . Is it safe?
- ▶ Everybody sees $n_a := pq_a$ and $n_b := pq_b$
- ▶ Alice can compute $q_b = n_b/p$
- ▶ Bob can compute $q_a = n_a/p$
- ▶ **Anyone** can compute $\gcd(n_a, n_b) = p$ and then q_a and q_b
- ▶ Attack demonstrated in practice
Lenstra et al. *Ron was wrong, Whit is right*
Show that 2/1000 RSA keys are insecure



Euclidean algorithm

- ▶ Goal : given integers a and b , find $d = \gcd(a, b)$
- ▶ $d|a, d|b$ imply $d|(a + kb)$ for any integer k
Require: $a \geq b$
Ensure: $\gcd(a, b)$
 - 1: **if** $b|a$ **then**
 - 2: **return** b
 - 3: **else**
 - 4: Compute q such that $0 < a - qb < b$
 - 5: **return** $\gcd(b, a - qb)$
 - 6: **end if**
- ▶ Complexity $O(|a|^2)$; best algorithms achieve $O(|a| \log |a|)$



Example

$$\begin{aligned}\gcd(36, 16) &= \gcd(16, 36 - 2 \cdot 32) \\ &= \gcd(16, 4) \\ &= 4\end{aligned}$$



Extended Euclidean algorithm

- ▶ Goal : compute r and s such that $ra + sb = \gcd(a, b)$
Require: $a \geq b$
Ensure: $d = \gcd(a, b)$ and r, s , such that $ar + bs = d$
 - 1: **if** $b|a$ **then**
 - 2: **return** $a, 0, 1$
 - 3: **else**
 - 4: Compute q such that $0 < a - qb < b$
 - 5: $d, r, s \leftarrow \gcd(b, a - qb)$
 - 6: **return** $d, s, r - qs$
 - 7: **end if**
- ▶ Indeed if $rb + s(a - qb) = d$ then $sa + (r - qb)b = d$
- ▶ Complexity $O(|a|^2)$; best algorithms achieve $O(|a| \log |a|)$



Example

$$\begin{aligned}\gcd(36, 16) &= \gcd(16, 36 - 2 \cdot 32) \\ &= \gcd(16, 4) \\ &= 4\end{aligned}$$

$$\begin{aligned}4 &= 0 \cdot 16 + 1 \cdot 4 \\ &= 1 \cdot 36 + (0 - 2 \cdot 1)16\end{aligned}$$



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Main goals in linear algebra

- ▶ Multiply two square matrices
- ▶ Inverse a square invertible matrix
- ▶ Solve linear systems of equations



Complexity of linear algebra

- ▶ All these tasks have roughly the same complexity
- ▶ For an $n \times n$ matrix, complexity $O(n^\omega)$ multiplications where
 - ▶ Lower bound $\omega \geq 2$
 - ▶ Gauss elimination $\omega \leq 3$
 - ▶ Strassen $\omega \leq \log_2 7 \approx 2.8074$
 - ▶ In 2015 we know $\omega \leq 2.3728639$ (but not practical)
 - ▶ Conjecture : for any $\epsilon > 0$, we could have $\omega = 2 + \epsilon$
 - ▶ ω may be smaller for specific matrices



Scalar product

- ▶ Given two vectors $a = (a_1, \dots, a_n)$ and $b = (b_1, \dots, b_n)$, compute their scalar product $c = (a, b) = \sum_{i=1}^n a_i b_i$
- ▶ Complexity : n multiplications



Matrix multiplication

- ▶ Given two $n \times n$ matrices A and B compute $C = AB$
- ▶ See A and B as row and column matrices respectively

$$A = \begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix} \quad B = (b_1 \dots b_n)$$

- ▶ n^2 scalar products (a_i, b_j) , so n^3 multiplications in total



Strassen algorithm

- ▶ Idea : trade some multiplications for additions
- ▶ Compute a product of $2n \times 2n$ matrices using 7 (instead of 8) products of $n \times n$ matrices
- ▶ To compute MM' where $M = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ and $M' = \begin{pmatrix} a' & b' \\ c' & d' \end{pmatrix}$, compute

$$P_1 = (a + c)(a' + b'), \quad P_2 = (b + d)(c' + d'), \quad P_3 = (b + c)(c' - b')$$

$$P_4 = c(a' + c'), \quad P_5 = b(b' + d'), \quad P_6 = (c - d)c', \quad P_7 = (a - b)b'$$

$$M \cdot M' = \begin{pmatrix} P_1 + P_3 - P_4 - P_7 & P_5 + P_7 \\ P_4 - P_6 & P_2 - P_3 - P_5 + P_6 \end{pmatrix}$$

- ▶ Complexity :

$$T(2n) = 7 \cdot T(n) + O(n^2) \implies T(n) = n^{\log_2 7} = n^{2.807}$$



Best asymptotic algorithms

- ▶ Coppersmith-Winograd $\omega < 2.375477$
- ▶ Between 2010 and 2014 : ω decreased to 2.3728639
- ▶ Those fast asymptotic algorithms are not used in practice because of large constants involved
- ▶ Conjecture : $\omega = 2 + \epsilon$



From inversion to multiplication

$$D := \begin{pmatrix} I_n & A & 0 \\ 0 & I_n & B \\ 0 & 0 & I_n \end{pmatrix} \Rightarrow D^{-1} = \begin{pmatrix} I_n & -A & AB \\ 0 & I_n & -B \\ 0 & 0 & I_n \end{pmatrix}$$

- ▶ If inversion takes $O(n^\omega)$ then so does multiplication



From multiplication to inversion

- ▶ Given $M = \begin{pmatrix} A & B \\ C & D \end{pmatrix}$ to invert, let $N := \begin{pmatrix} I & 0 \\ -D^{-1}C & I \end{pmatrix}$
- ▶ We have $MN = \begin{pmatrix} S & B \\ 0 & D \end{pmatrix}$ for $S := A - BD^{-1}C$
- ▶ We have $(MN)^{-1} = \begin{pmatrix} S^{-1} & -S^{-1}BD^{-1} \\ 0 & D^{-1} \end{pmatrix}$
- ▶ Compute D^{-1} then $-D^{-1}C$
- ▶ Compute S then S^{-1} then $-S^{-1}BD^{-1}$
- ▶ Compute $M^{-1} = N(MN)^{-1}$
- ▶ Cost :

$$T_{inv}(2n) = 2T_{inv}(n) + 8T_{mul}(n) + O(n^2)$$

- ▶ If multiplication takes $O(n^\omega)$ then so does inversion



Gaussian elimination

- ▶ Observation : if $My = x$ then for any invertible N , we have $NMy = Nx$
- ▶ In particular, this is true when N is a matrix which
 - ▶ Swaps two rows of M
 - ▶ Multiplies one row by an invertible constant
 - ▶ Adds a multiple of one row of M to another row of M
- ▶ Gaussian elimination repeats these operations until the resulting matrix is upper triangular



Gaussian elimination

- ▶ Algorithm when M is invertible
 - 1: **for** each column i , from $i = 1$ **to** n **do**
 - 2: Find a nonzero element in this column
 - 3: Swap the row of this element with row i
 - 4: **for** each row j below row i **do**
 - 5: Let $c := -M_{j,i}/M_{i,i}$
 - 6: Add c times row i to row j to erase the value in (j, i)
 - 7: **end for**
 - 8: **end for**
- ▶ Adapt step 2 otherwise
- ▶ Cost is $O(n^3)$ multiplications



Resolution from Gaussian form

- ▶ Algorithm when M is invertible
 - 1: **for** each column i from n to 1 **do**
 - 2: Recover value of unknown i , using equation i and all values of previously computed unknowns $j > i$
 - 3: **end for**
- ▶ Adapt to determine the affine space of solutions $v + \ker M$ otherwise
- ▶ Cost is $O(n^2)$ multiplications
- ▶ Can be used to invert M in $O(n^3)$ multiplications



Hermite normal form

- ▶ If the matrices are defined over a ring (not a field) then not all elements are invertible
- ▶ Elimination in each column will be done with a kind of GCD algorithm :
 - 1: **for** each column i , with $i \in \{0, \dots, n\}$ **do**
 - 2: **while** some element below (i, i) is non zero **do**
 - 3: Find the smallest nonzero element in this column
 - 4: Swap the row of this element with row i
 - 5: For each row j below i , remove as many times row i as needed to have element (j, i) between 0 and element (i, i)
 - 6: **end while**
 - 7: **end for**



Hermite normal form

- ▶ A matrix is in Hermite normal form if it is upper triangular, has positive elements on the diagonal, and moreover all non-diagonal elements are non-negative and smaller than the diagonal elements in their column
- ▶ Last condition ensured by completing previous algorithm with
 - 1: **for** each column i , with i from 1 to n **do**
 - 2: For each row j above i , remove as many times row i as needed to have element (j, i) larger than 0 and smaller than element (i, i)
 - 3: **end for**



Sparse linear algebra

- ▶ A matrix is **sparse** if each row contains a small number of nonzero elements
- ▶ Very useful in index calculus algorithms (see topic 2) and many other contexts
- ▶ Can store larger size matrices by storing only $(i, j, M_{i,j})$ for nonzero elements $M_{i,j}$
- ▶ Gaussian elimination will kill the sparsity quickly
- ▶ Two approaches for sparse matrices :
 - ▶ Structured Gaussian elimination
 - ▶ Algorithms based on matrix-vector multiplications



Structured Gaussian elimination

- ▶ Consider the linear system $My = x$
- ▶ For the matrices M occurring in index calculus :
 - ▶ Each row contain few elements
 - ▶ The first columns contain much more elements than the last ones
- ▶ Structured Gaussian elimination involves several tricks such as removing variables that only appear once or twice
- ▶ Used as preprocessing to reduce the size in practice
- ▶ Heuristic



Lanczos algorithm

- ▶ If M is invertible, $My = x \Leftrightarrow M^t My = M^t x$ hence we can assume M is symmetric defining a scalar product $(x, y)_M := xMy^t$
- ▶ Lanczos is iterative : over the real/complex numbers, the algorithm can be stopped before the end with a reasonable approximation of the solution
- ▶ First compute a basis $\{v_i\}$ of orthogonal vectors with respect to the scalar product $(*, *)_M$ (see topic 3), then compute $x = \sum_{i=1}^n (x, v_i)_M v_i$
- ▶ First part involves $O(n)$ matrix-vector multiplications, each one at $O(n)$ cost if each row contains $O(1)$ elements



Wiedemann algorithm

- ▶ Reconstruct the **minimal polynomial** of M : smallest degree polynomial f such that $f(M) = 0$
- ▶ If $f(\alpha) = \sum_{i=0}^d f_i \alpha^i$, then $l = -\frac{1}{f_0} \sum_{i=1}^d f_i M^i$ then

$$x = -\frac{1}{f_0} \sum_{i=1}^d f_i M^i x = M \left(-\frac{1}{f_0} \sum_{i=1}^d f_i M^{i-1} x \right)$$

- ▶ We deduce y such that $My = x$
- ▶ The algorithm requires $O(n)$ matrix-vector products
- ▶ Recent discrete log records use Block Wiedemann <http://caramel.loria.fr/p180.txt>



Outline

Complexity measures

Algebra and number theory

First algorithmic number theory tools

Linear algebra

Root-finding algorithms



Root-finding algorithms

- ▶ Problem : given a polynomial $f \in \mathbb{F}_q[x]$, find all $x \in \mathbb{F}_q$ such that $f(x) = 0$
- ▶ Note that $f(\alpha) = 0 \Leftrightarrow (x - \alpha) | f(x)$
- ▶ First compute the **square-free split** part of f : the unique monic polynomial $\tilde{f} | f$ such that \tilde{f} has only linear factors, all distinct
- ▶ Then successively split \tilde{f} into smaller polynomials



Square-free split part

- ▶ We have $\alpha^q = \alpha$ for all $\alpha \in \mathbb{F}_q$ so

$$x^q - x = \prod_{\alpha \in \mathbb{F}_q} (x - \alpha)$$

- ▶ Therefore

$$\tilde{f}(x) = \gcd(x^q - x, f(x))$$

contains only factors of degree 1, no factor twice

- ▶ Compute $x^q \bmod f(x)$ with a square-and-multiply algorithm, subtract x , and compute gcd



Breaking out \tilde{f}

- ▶ If q odd we have

$$x^q - x = x(x^{\frac{q-1}{2}} + 1)(x^{\frac{q-1}{2}} - 1)$$

- ▶ Computing

$$\gcd(\tilde{f}, x^{\frac{q-1}{2}} \pm 1)$$

likely to break \tilde{f} into two parts

- ▶ Also notice that $x^q - x = (x - a)^q - (x - a)$ for all $a \in \mathbb{F}_q$
- ▶ Pick a random a and compute

$$\gcd(\tilde{f}, (x - a)^{\frac{q-1}{2}} \pm 1)$$



Remarks

- ▶ Several other algorithms
- ▶ Polynomial time in $\deg f$ and $\log q$
- ▶ Can be adapted when q is even
- ▶ Can be generalized to find other factors of f , not just degree 1 factors

