Computational Algebra

-

Transcript

Fabio Gratl

April 30, 2015

Contents

1	Inte	ger Arit	thmetic 3
	1.1	Additi	on and Multiplication
		1.1.1	Algorithm 1 (Simple addition)
		1.1.2	Definition 2 (Bit-Operation)
		1.1.3	Definition 3 (Big O)
		1.1.4	Theorem 4 (Lower bound for addition) 4
		1.1.5	Algorithm 5 (Multiplication by "grid method") 5
		1.1.6	Theorem 6 (Runtime of Algorithm 5)
		1.1.7	Algorithm 7 (Karatsuba)
		1.1.8	Theorem 8 (Runtime of Algorithm 7) 6
		1.1.9	Definition 9 (Root of unity) 6
			Algorithm 10 (Fast Fourier transformation FFT)
			Theorem 11 (Runtime of Algorithm 10)
			Definition 12 (Good root of unity)
		1.1.13	Proposition 13 $(DFT_{\mu^{-1}})$
			Proposition 14 (Finding good roots of unity)
			Algorithm 15 (Polynomial multiplication using DFT) 9
			Theorem 16 (Runtime of Algorithm 15)
			Proposition 17 (Add and mul in $O(l)$)
			Proposition 18 (Sort of summary)
			Algorithm 19 (Multiplication using FFT) 10
			Theorem 20 (Runtime of Algorithm 19)
			Theorem 21 (Schönhage-Strassen 1971)
	1.2	Divisio	on with remainder, Euclidean algorithm
		1.2.1	Algorithm 1 (Division with remainder)
		1.2.2	Proposition 2 (Runtime of Algorithm 1)

2	Not	es		19
		1.3.5	Proposition 5:	18
		1.3.4	Definition 4	18
		1.3.3	Algorithm 3 (fast exponention)	17
		1.3.2	Algorithm 2 (Fermat Test)	17
		1.3.1	Theorem 1	16
	1.3	Prima	lity testing	16
		1.2.6	Algorithm 6 (Extended Euclidean Algorithm)	15
		1.2.5	Theorem 5 (Runtime of Algorithm 3)	15
			Theorem 4 (Correctness of Algorithm 3)	
		1.2.3	Algorithm 3 (Euclidean algorithm)	14

1 Integer Arithmetic

Topics:

- Addition and Multiplication
- GCD computation
- Primality testing
- Factorization

1.1 Addition and Multiplication

Agreement:

- $a, x \in \mathbb{N}$ represented as $x = \sum_{i=0}^{n-1} a_i \cdot B^i$ $B \in \mathbb{N}_{>1}$ fixed Base $(a_i \in \{0, B-1\})$
- if $x \neq 0$, assume $a_{n-1} \neq$ then define: length of x := l(x) = n = number of digits = $\lfloor \log_B(x) \rfloor + 1$ (mnemonic: $\log_B(B) + 1 = 2$)
- l(0) = 1 (Amount of memory required to store x = 0)
- l(x) := l(|x|)
- for $x \in \mathbb{Z}$ represent if as x = sgn(x) * |x|

1.1.1 Algorithm 1 (Simple addition)

input :
$$x = \sum_{i=0}^{n-1} a_i \cdot B^i$$
, $y = \sum_{i=0}^{n-1} b_i \cdot B^i$, $x, y \in \mathbb{N}$

output:
$$x + y = \sum_{i=0}^{n} c_i \cdot B^i$$

- (1) $\sigma = 0$
- (2) for i = 0, ..., (n-1):
- (3) set $c_i := a_i + b_i + \sigma_i$ and $\sigma := 0$
- (4) if $(c_i \geq B)$
- $(5) set c_i = c_i B$
- (6) $\operatorname{set} \sigma = 1$
- (7) set $c_n = \sigma$

If B = 2 then (3) - (6) can be realized by logic gates:



1.1.2 Definition 2 (Bit-Operation)

A bit operation is an operation that can be performed by a logic gate or by searching or writing a bit from / into memory.

1.1.3 Definition 3 (Big O)

Let M be a set (usually $M = \mathbb{N}$), $f, g: M \mapsto \mathbb{R}_{>0}$ we write $f \in O(g)$ if $\exists c \in \mathbb{R} : f(x) \leq c \cdot g(x) \forall x \in M$

1.1.4 Theorem 4 (Lower bound for addition)

Let $f: \mathbb{N} \to \mathbb{R}$, $n \mapsto$ maximal number of bit operations required by Algorithm 1 to add $x_y \in \mathbb{N}$ with $l(x), l(y) \leq n$

Let $g = id_{\mathbb{N}}$ Then $f \in O(g)$

We say Algorithm 1 requires O(n) bit operations for adding two numbers of length $\leq n$. ⇒ "linear complexity"

Set $M := \{ \text{Set of all algorithms for addition in } \mathbb{N} \}$

For $A \in M$ define $f_A : \mathbb{N} \to \mathbb{R}$ as above.

We would like to find $f_{odd} : \mathbb{N} \to \mathbb{R}$, $n \mapsto \inf\{f_A(n) | A \in M\}$

Since one needs to read x, y (and write the result) we can not do any better than linear complexity for addition.

Subtraction

let x, y as Algorithm 1, $x \ge y$

For
$$\bar{y} := \sum_{i=0}^{n-1} (B - 1 - b_i)B^i$$
 (digitwise / bitwise complement)

$$\Rightarrow x + \bar{y} = x - y + B^n - 1$$

 $\Rightarrow x - y = x + \bar{y} + 1 - B^n$ (initially set $\sigma = 1$)

Conclusion: Addition and Subtraction have cost O(n)

1.1.5 Algorithm 5 (Multiplication by "grid method")

input :
$$x = \sum_{i=0}^{n-1} a_i \cdot 2^i$$
, $y = \sum_{i=0}^{m-1} b_i \cdot 2^i$

output: $z = x \cdot y$

- (1) z := 0
- (2) for i = 0, ..., (n-1)

(3) if
$$(a_i \neq 0)$$
 set $z := z + \sum_{j=1}^{m-1} b_j 2^{i+j}$

1.1.6 Theorem 6 (Runtime of Algorithm 5)

Algorithm 5 requires O(n*m) bit operations.

As of the total input length n + m:

$$n \cdot m \le \frac{1}{2}(n+m)^2 \to O((n+m)^2)$$

⇒ Quadratic complexity

Karatsuba-multiplication:

Observation for polynomials:

$$a + bx, c + dx$$
 have $(a + bx)(c + dx) = ac + (ac + db - (a - b)(c - d))x + bdx^2$

The point: only used 3 multiplications instead of 4.

Specialize x = B "large" such that x = a + bB partition into two blocks. Then multiply the blocks by a recursive call.

1.1.7 Algorithm 7 (Karatsuba)

input : $x, y \in \mathbb{N}$

output: $z = x \cdot y$

- (1) Choose $k \in \mathbb{N}$ minimal such that $l(x), l(y) \leq 2^k$. Set $B = 2^{2^{k-1}}$
- (2) if (k = 0) return $x \cdot y$ (by bit-operation AND)
- (3) write $x = x_0 + x_1 B$, $y = y_0 + y_1 B$ with $l(x_i), l(y_i) \le 2^{k-1}$
- (4) compute $x_0 \cdot y_0$, $x_1 \cdot y_1$, $(x_0 x_1) \cdot (y_0 y_1)$ by a recursive call
- (5) return $z = x_0 y_0 + (x_0 y_0 + x_1 y_1 (x_0 x_1)(y_0 y_1))B + x_1 y_1 B^2$

1.1.8 Theorem 8 (Runtime of Algorithm 7)

For multiplying two numbers of length $\leq n$ Algorithm 7 requires $O(n^{\log_2 3}) \approx O(n^{1.59})$ bit operations.

Proof:

Set $\Theta(k) := \text{maximal numbers of bit operations for } l(x), l(y) \leq 2^k$

We have for
$$k > 0$$
: $\Theta(k) \le 3\Theta$ $(k-1)$ $+c$ 2^k with $(c \text{ some constant})$

Claim: $\Theta(k) \le 3^k + 2c(3^k - 2^k)$

Proof by Induction on k:

$$k = 0 : \Theta(k) = 1$$

$$k - 1 \to k = \Theta(k) = 3\Theta(k - 1) + c2^{k - 1}$$

$$\leq 3(3^{k - 1} + 2c(3^{k - r} - 2^{k - 1})) + c2^{k}$$

$$= 3^{k} + 2c(3^{k} - 2^{k})$$

So
$$\Theta(k) \le (2c+1)3^k$$

Now $l(x) \le n$ hence $2^{k-1} < n$ by minimality of k

So
$$k - 1 < \log_2 n$$

$$\Rightarrow \Theta(k) \le 3(2c+1)3^{\log_2(n)}$$

$$= 3(2c+1)2^{\log_2(3)\log_2(n)}$$

$$= 3(2c+1)n^{\log_2(3)} \quad \Box$$

One can modify the terminal condition of Karatsuba to switch to Grid-Multiplication, which is faster for small numbers.

Fast-Fourier Transform

Reminder: For a function $f: \mathbb{R} \to \mathbb{C}$ define:

$$\hat{f}: \mathbb{R} \to \mathbb{C}$$
 by $\hat{f}(\omega) = \int_{\mathbb{R}} f(t)e^{-i\omega t}dt$ (if it exists)

Think of ω as frequency.

Definition (Convolution)

Let
$$f, g : \mathbb{R} \to \mathbb{C}$$

 $(f * g)(x) = \int_{\mathbb{R}} f(t)g(x - t)dt$

Convolution is analogous to polynomial multiplication Formula: $(f * g) = \hat{f} \cdot \hat{g}$

For a function $M \mapsto C$ with $|M| < \infty$ we need the discrete Fourier transform (DFT)

1.1.9 Definition 9 (Root of unity)

Let R be a commutative ring with 1. An element $\mu \in R$ is called an n-th root of unity (= root of 1) if $\mu^n = 1$.

It is called primitive if $\mu^i \neq 1$ for (0 < i < n) i.e. $ord(\mu) = n$

let μ be a primitive n-th root of 1 (e.g. $e^{2\pi\frac{i}{n}}$ $\in \mathbb{C}$)

Then the map $DFT_{\mu}: \mathbb{R}^n \mapsto \mathbb{R}^n$

$$(\hat{a}_0, ..., \hat{a}_n) \mapsto (\hat{a}_0, ..., \hat{a}_n)$$
 with $\hat{a}_i = \sum_{j=0}^{n-1} \mu^{ij} a_j$

is called discrete Fourier transformation

For polynomials:

$$DFT_{\mu}: R[x] \mapsto R^n$$

 $f \mapsto (f(\mu^0), ..., f(\mu^{n-1}))$

Convolution rule: (from
$$f(\mu^i)g(\mu^i) = (f * g)(\mu^i)$$
)

$$DFT_{\mu}(f * g) = DFT_{\mu}(f) \cdot DFT_{\mu}(g)$$
 (component wise product)

Addition of two polynomials in R[x] of deg(n) require O(n) ring operations. Multiplication require $O(n^l)$.

With Karatsuba have $O(n^{\log_2(3)})$ ring operations.

Cost $DFT_{\mu}(f) \cdot DFT_{\mu}(g) : O(n)$ ring operations (with μ as 2n-th root of 1)

Want: Cheap way of doing DFT and back-transformation.

1.1.10 Algorithm 10 (Fast Fourier transformation FFT)

input : $f \in R[x]$, $\mu \in R$ primitive 2^k -th root of 1, such that $\mu^{2^{k-1}} = -1$ output: $DFT_{\mu}(f)$

(1) Write $f(x) = g(x^2) + xh(x^2)$ with $f, q, h \in R[x]$

(2) if
$$k = 1$$
 ($\Rightarrow \mu = 1$) return $DFT_{\mu}(f) = (g(1) + h(1), g(1) - h(1))$

(3) Recursive call: compute $DFT_{\mu^2}(g) = \hat{g}, DFT_{\mu^2}(h) = \hat{h} \in \mathbb{R}^{2^{k-1}}$

(4) return
$$DFT_{\mu}(f) = (\hat{f}_0, ..., \hat{f}_{2^k-1})$$
 with $\hat{f}_i = \hat{g}_i + \mu \hat{h}_i$ where $\hat{g}_i = \hat{g}_{i-2^{k-1}}$ for $i \ge 2^{k-1}$

Note: Components of \hat{g} and \hat{h} are:

$$\hat{g} = g(\mu^{2i}), \quad \hat{h}_i = h(\mu^{2i}) \text{ so}$$

 $\hat{f}_i := f(\mu^i) = \hat{g}_i(\mu^{2i}) + \mu h(\mu^{2i}) = \hat{g}_i + \mu \hat{h}_i$

Convention: $lg(x) = log_2(x)$

1.1.11 Theorem 11 (Runtime of Algorithm 10)

Let $n = 2^k$, $f \in R[x]$ with $deg(\psi) < n$ Then Algorithm 10 requires $O(n \cdot \lg(n))$ ring operations.

Better than $O(n^{1+\epsilon}), \quad \forall \epsilon > 0!$

Proof:

Set $\Theta(k) = \max$ number of ring operations required. By counting obtain for k > 1:

$$\Theta(k) \leq 2\Theta(k-1) + \underbrace{(\text{compute } \mu^{i}(i \leq 2^{k-1}))}_{2^{k-1}} + \underbrace{(\mu^{i}\hat{k}_{i})}_{2^{k-1}} + \underbrace{(\text{sums and differences})}_{2^{k}}$$

$$=2\Theta(k-1)+2^{k+1}$$

Claim:
$$\Theta(k) \le (2k-1)2^k$$

 $k = 1: f = a_0 + a_1 \cdot x \quad DFT_{\mu}(f) = (a_0 + a_1 \cdot a_0 - a_1) \Rightarrow \Theta(a) = 2$
 $k - 1 \to k: \Theta(k) \le 2 \cdot \Theta(k-1) + 2^{k+1} \le 2 \cdot (2k-3) \cdot 2^{k-1} + 2^{k+1} = (2k-1) \cdot 2^k$
since $k = \lg(n)$ obtain $O(k) \le (2 \cdot \lg(n) - 1) \cdot n \in O(n \cdot \lg(n))$

Back-transformation?

1.1.12 Definition 12 (Good root of unity)

A primitive n-th root of unity is called good (caveat: this is ad-hoc terminology) if: $\sum_{j=0}^{n-1} \mu^{ij} = 0 \quad \text{ for } (0 < i < n)$

example:

- (1) $\mu = e^{2\pi \frac{i}{n}}$ is a good primitive root of unity
- (2) $R = \mathbb{Z}/(8)$, $\mu = \bar{3} \Rightarrow \mu \cdot B$ is primitive 2^{nd} root of unity But $\bar{B}^0 + \bar{3}^1 = \bar{u} \neq \bar{0}$ so μ is not good.

1.1.13 Proposition **13** ($DFT_{u^{-1}}$)

let $\mu \in R$ be a good root of 1

$$(a) = (a_0, ..., a_{n-1}) \in R^n \Rightarrow DFT_{\mu}^{-1}(DFT_{\mu}(a)) = n \cdot (a)$$
 where $n = 1 + ... + 1 \in R$

$$DFT_{\mu}(a) = (\hat{a}) = (\hat{a}_0, ..., \hat{a}_{n-1})$$

with $\hat{a}_j = \sum_{k=0}^{n-1} \mu^{jk} a_k$

$$DFT_{\mu^{-1}}(\hat{a}) = (\hat{a}_0, ..., \hat{a}_1)$$

with
$$\hat{a}_i \sum_{i=0}^{n-1} \mu^{-ij} \sum_{k=0}^{n-i} \mu^{jk} a_k = \sum_{k=0}^{n-1} (a_k \cdot \sum_{i=0}^{n-1} \mu^{j(k-i)}) = a_i \cdot n$$

1.1.14 Proposition 14 (Finding good roots of unity)

let $\mu \in R, n \in \mathbb{N}$ Assume:

- a) R is an integral Domain and μ is a primitive or n-th root of 1 (Integral Domain: nonzero commutative ring in which the product of two nonzero elements is nonzero)
 - \Rightarrow Granted by FFT
- b) $n = 2^b$, $\mu^{\frac{n}{2}} = -1$, then $h > 0 \lor char(R) \neq 2$ $\to \mu$ is a good primitive n-th root of 1 ("root of unity")

Proof:

a) for
$$0 < i < n$$

$$\underbrace{(\mu^i - 1)}_{\neq 0} \underbrace{(\sum_{j=0}^{n-1} \mu^{ij})}_{=0} = \mu^{in} - 1 = 0$$

* Let
$$0 < i < n$$
, write $i = 2^{k-s} \cdot r$ with $r \text{ odd } \forall s > 0$

$$\sum_{j=0}^{2^k-1} \mu^{ij} = \sum_{l=0}^{2^{k-s}-1} \sum_{j=0}^{2^s-1} \mu^{i(l \cdot 2^s + j)}$$

$$\mu^{i \cdot 2^s} = 1$$

$$i \cdot 2^s = 2^{k-s} \sum_{j=0}^{2^s-1} \mu^{ij} = 2^{k-s} \sum_{j=0}^{2^{s-1}-1} (\mu^{ij} + \mu^{i(2^{s-1}+j)})$$
But $\mu^{i \cdot 2^{s-1}} = \mu^{2^{k-s} \cdot r \cdot 2^{s-1}} = \mu^{2^{k-1} \cdot r} = (-1)^r = -1$
So $\sum_{i=0}^{n-1} \mu^{ij} = 0$

b)
$$\mu^n = 1, n = 2^k \Rightarrow ord(\mu)|n \Rightarrow ord(\mu)$$
 is power of 2

1.1.15 Algorithm 15 (Polynomial multiplication using DFT)

input : $f,g \in R[x]$ with $deg(f) + deg(g) < 2^k =: n$ $\mu \in R$ as a good root of unity; Assume $2 \in R$ is invertible

output: $h = f \cdot g$

- (1) compute $\hat{f} = DFT_{\mu}(f)$, $\hat{g} = DFT_{\mu}(g)$ with $f, g \in \mathbb{R}^n$
- (2) compute $\hat{h} = \hat{f} \cdot \hat{g}$
- (3) compute $(h_0, ..., h_{n-1}) = DFT_{\mu^{-1}}\hat{h}$ (same as $DFT_{\mu}(\hat{h})$ but with different order) = Back-transformation $\cdot 2^k$ set $h = \frac{1}{2^k} \sum_{i=0}^{n-1} h_i x^i$

1.1.16 Theorem 16 (Runtime of Algorithm 15)

Algorithm 15 uses $O(n \cdot \log(n))$ ring operations for polynomials of deg < n**Proof:**

- Choose k minimal so that $deg(f) \cdot deg(g) < 2^k$ $\Rightarrow 2^{k-1} \le 2n \quad \Rightarrow k \le \log(n) + 2$
- $\bullet \ \ \underbrace{O(2k \cdot 2^k)}_{\text{Step 1}} + \underbrace{2^k}_{\text{Step 2}} + \underbrace{O(k \cdot 2^k) + 2^k}_{\text{Step 3}} \ \in \ O(2k \cdot 2^k) = O(n(g(n))) \qquad \Box$

Goal: Multiplication in $\mathbb N$ using DFT

Idea: find roots of 1 in a suitable $\mathbb{Z}/(m)$

Choose $m=2^l+1, \mu=\bar{2}\in R$

1.1.17 Proposition 17 (Add and mul in O(l))

Let $m = 2^{l} + 1$, $R = \mathbb{Z}/(m)$

Addition in R and multiplication by $\bar{2}^i \in R$ $(0 \le i < 2l)$ can be done in O(l) bit operations

Proof:

Let $\bar{x} \in R$ with $0 \le x \le 2^l$

- Addition: $x + \bar{y}$
 - (1) compute $x + y \in \mathbb{N}$: O(l)
 - (2) if $x + y > 2^l + 1$ subtract $2^l + 1$: O(l)
- Multiplication by $\bar{2}^i$ $(0 \le i < l)$
 - (1) Bit-shift i Bits to the left by relocating in memory:

$$\underbrace{O(\operatorname{length}(i))}_{\text{compute addr. of new first bit}} + \underbrace{l}_{\text{copying}} = O(\log(l)) + l \quad \in \quad O(l)$$

- Multiplication by $\bar{2}^i$ $(l \le i < 2l 1)$
 - (1) Multiplication by $\bar{2}^{i-l}$: O(l)
 - (2) take negative $\bar{2}^i \cdot \bar{x} = -\bar{2}^{i-l} \cdot \bar{x}$: O(l)

1.1.18 Proposition 18 (Sort of summary)

Let $k, r \in \mathbb{N}$, r > 0, $m = 2^{2^k \cdot r} + 1$, $R = \mathbb{Z}/(m)$, $\mu = \overline{2}^r \in R$ $\Rightarrow 2 \in R$ is invertible, μ is a good primitive 2^{k+1} -th root of $1 \Rightarrow \mu^{2^k} = 1$

Proof: \rightarrow from above

1.1.19 Algorithm 19 (Multiplication using FFT)

input : $x, y \in \mathbb{N}$

output: $Z = x \cdot y$

- (1) Choose $k \in \mathbb{N}$ minimal such that $l(x), l(y) \leq 2^{2k}$
- (2) if $k \leq 3$, compute $z = x \cdot y$ by Algorithm 5
- (3) set $B = 2^{2^k}$, $m = 2^{2^k \cdot 4} + 1$, $R = \mathbb{Z}/(m)$, $\mu = \overline{2}^4 \in R$ (\Rightarrow so μ is a good primitive 2^{k+1} -th root of 1)

- (4) write $x = \sum_{i=0}^{2^k 1} x_i \cdot B^i$, same for y with $(0 \le x_i, y_i < B)$ possible since $x, y < 2^{2^{2k}} = 2^{2^k \cdot 2^k} = B^{2^k}$
- (5) compute: $\hat{x} = DFT_{\mu}(\bar{x}_0, ..., \hat{x}_{2^k-1}, \underbrace{0, ..., 0}_{2^k \text{zeros}}) \in R^{2^{k+1}}$ same for y \rightarrow use FFT
- (6) compute: $\hat{z} = \hat{x} \cdot \hat{b} \in \mathbb{R}^{2^{k+1}}$ (component wise multiplication) Perform multiplication in R as follows: Multiply representatives (non negative and $\langle m \rangle$ by recursive call. Then reduce modulo m by "negative bit shift" (see proof of Proposition 17)
- (7) compute: $(\bar{z}_0, ..., \bar{z}_{2^{k+1}}) = \frac{1}{2^{k+1}} DFT_{\mu^{-1}}(\hat{z}) \in R$ with $0 \le z < m$

(8) set
$$z := \sum_{j=0}^{2^{k+1}-1} z_i \cdot B^i$$

1.1.20 Theorem 20 (Runtime of Algorithm 19)

Algorithm 19 correctly computes $t = x \cdot y$ and requires $O(n \cdot (\log n)^4)$ bit operations for $l(x), l(y) \le n$

Proof: Correctness

write
$$x(t) \sum_{i=0}^{2^k - i} x_i t^i \in \mathbb{Z}[t], \quad y(t), \quad \bar{x}(t) \in R[t], \bar{y}(t), \bar{z}(t)$$

write $x(t) \sum_{i=0}^{2^k - i} x_i t^i \in \mathbb{Z}[t], \quad y(t), \quad \bar{x}(t) \in R[t], \bar{y}(t), \bar{z}(t)$ by Proposition 18 and Proposition 13 we have $\bar{z}(t) = \bar{x}(t) \cdot \bar{y}(t)$ The l-th coefficient of $x(t) \cdot y(t)$ is $0 \le \sum_{i+j=l} x_i \cdot y_j < 2^k \cdot B^2 = 2^{k+2 \cdot 2^k} \le 2^{2^{k+2}} < m$

So
$$z(t) = x(t) \cdot y(t) \Rightarrow z = z(B) = x(B) \cdot y(B) = x \cdot y$$
 Cost:

Write $\Theta(k) := \max \text{ number of bit operations}$

Analyze Steps:

- (1) compute max $\{l(x), l(y)\}: O(l(n)) = O(k)$
- (2) O(1)
- (3) no bit operations
- (4) compute starting addresses of x_i, y_i in memory: $2 * 2^k$ increments of the address: $2 \cdot 2 \cdot 2^k = 2^{k+2}$ bit ops $\Rightarrow O(2^k)$
- (5) By Theorem 11 need $O(2 \cdot 2^{k+1} \cdot (k+1))$ operations in R which are additions and multiplications by powers of \bar{z} costing $O(2^{k+2})$ bit operations. Total for (5): $O(k \cdot 2^{2 \cdot k})$

- (6) 2^{k+1} multiplications of numbers < m, i.e. of length $\le 2^{k+2}$. So $k' \le \frac{k+3}{2}$ for k': the "new" k used in the next recursion level. For $\alpha \in R_{>0}$ define $\Theta(\alpha) := \Theta(\lfloor \alpha \rfloor)$ Total for (6): $2^{k+1}(\Theta(\frac{k+3}{2}) + \underbrace{O(2^{k+2})}_{\text{reduction mod m}})$
- (7) For $DFT_{\mu^{-1}}(\hat{z}): O(k \cdot 2^{2 \cdot k})as(5)$ Since \bar{z} is a n root of 1, multiplication by $\bar{2}^{-k-1}$ is multiplication by a positive power of $\bar{2}$, which costs $O(2^{k+2})$ Total for (7): $O(k \cdot 2^{2 \cdot k})$
- (8) For $j \leq 2^{k+1}$ have $\sum_{i=0}^{j-1} z_i \cdot B^i \leq (m-1) \sum_{i=0}^{j-1} B^i = (m-1) \frac{B^{j-1}}{B-1} < 2(m-1) \frac{B^j}{B} = (m-1) \frac{B^j}{B}$ $2^{1+2^{k+2}+(j-1)2^k}$ so the sum has length $(j+3)\cdot 2+1$ Adding $z_i \cdot B^j$ to this sum happens at $(j \cdot 2^k)$ -th bit and higher \Rightarrow cost is $O(2^k)$ Total for (8): $O(2^{2 \cdot k})$

Grad total: For
$$k \ge 4$$
: $\Theta(k) \le 2^{k+1} \cdot \Theta(\frac{k+3}{2}) + c \cdot k \cdot 2^{2 \cdot k}$ with c constant

Also for $k \in \mathbb{R}_{>4}$

Define
$$\Lambda(k) := \frac{\Theta(k)}{2^{2 \cdot k}} \Rightarrow \Lambda(k) \leq \frac{2^{k+1}\Theta(\frac{k+3}{2})}{2^{2 \cdot k}} + c \cdot k = 16 \cdot \Lambda(\frac{k+3}{2}) + c \cdot k$$

Define $\Omega(k) := \Lambda(k+3)$ So for $k \in \mathbb{R}_{\geq 1}$

Define
$$\Omega(k) := \Lambda(k+3)$$
 So for $k \in \mathbb{R}_{\geq} 1$

$$\Omega(k) \le 16 \cdot \Lambda(\frac{k}{2} + 3) + c \cdot (k+3) = \underbrace{16\Omega(\frac{\overline{k}}{2}) + c \cdot (k+3)}_{*}$$
Claim: For i $in\mathbb{N}$ with $2^{i-1} \le k-3$ have:

$$\Lambda(k) \le 16^{i} \Omega(\frac{k-3}{2^{i}}) + c \cdot (k+3)(1+8+\ldots+8^{i-1}) + 3 \cdot c \cdot (1+16+\ldots+16^{i-1})$$

Proof by induction:

$$i = 0\Lambda(k) = \Omega(k-3)$$

$$\begin{array}{l} i \to i+1: \Lambda(k) \leq 16^{i} \Omega(\frac{k-3}{2^{i}}) + c \cdot (k-3)(1+\ldots+8^{i-1}) + 3 \cdot c \cdot (1+\ldots+16^{i-1}) \leq 2^{i} \leq k-3 \quad * \\ \leq 16^{i} (16\Omega(\frac{k-3}{2^{i}+1})) + c(\frac{k-1}{2^{i}}+3) + c(k-3)\ldots = \text{claimed result} \\ \text{Take } u \in \mathbb{N} \text{ minimal with } 2^{u} > k-3 \Rightarrow \Omega(\frac{k-3}{2^{u}}) \leq \Omega(\lfloor \frac{k-3}{2^{u}} \rfloor) = \Omega(0) =: D \text{ (constant)} \end{array}$$

Note: u rouhly is recursion depth Have
$$2^{u-1} \le k-3 \Longrightarrow \Omega(\frac{2^u}{2^u}) \le \Omega(\lfloor \frac{2^u}{2^u} \rfloor) = \Omega(0) =: D$$
 (constant Note: u rouhly is recursion depth Have $2^{u-1} \le k-3 \Longrightarrow_{\text{claim}} \Lambda(k) \le 16^u \cdot D + c \cdot (k-3) \cdot \frac{8^u-1}{7} + 3c \cdot \frac{16^u-1}{15} \in O(16^u)$

Have
$$2^{u-1} \le k - 3 \Rightarrow u \le \lg(k - 3) + 1$$

$$\Rightarrow \Lambda(k) \in O(16^{\lg(k-3)}) = O((k-3)^4)$$

$$\Rightarrow \Theta(k) = 2^{2 \cdot k} \cdot \Lambda(k) \in O(2^{2k} \cdot (k-3)^4)$$

$$\Rightarrow \Theta(k) = 2^{2 \cdot k} \cdot \Lambda(k) \in O(2^{2k} \cdot (k-3)^4)$$
Have
$$2^{2(k-1)} < \underbrace{n}_{\max\{l(x) \cdot l(y)\}} \Rightarrow k \leq \frac{\lg(n)}{2} + 1$$

$$\max\{l(x)\cdot l(y)\}$$

So
$$\Theta(k) \in O(n \cdot (\lg(n))^4)$$

1.1.21 Theorem 21 (Schönhage-Strassen 1971)

Multiplication of integers of length $\leq n$ can be done in $O(n \cdot \lg(n) \cdot \lg(\lg(n)))$ bit operations. Schönhage-Strassen is used for integers of length > 100.000.

Asymptotically faster: Fürer's algorithm.

Comments on Bit complexity

- 1. Memory requirement may explode!
 - \Rightarrow No Problem as bit complexity is upper bound for memory requirements, since memory access is included in bit operations
 - $(\rightarrow \text{ only store what is calculated})$
- 2. Computation of addresses in memory take time
 - \Rightarrow length of addresses $\approx \lg(\text{memory space})$ computations of addresses $\approx \lg(\text{memory space})^2$
- 3. As memory requirement gets larger access times will get longer.
 - \Rightarrow transportation time for data $\geq \frac{\text{diameter of physical storage}}{2*\text{speed of light}}$

1.2 Division with remainder, Euclidean algorithm

1.2.1 Algorithm 1 (Division with remainder)

input :
$$b = \sum_{i=0}^{n-1} b_i 2^i$$
 $\sum_{i=0}^{n+m-1} a_i 2^i$ with $a_i, b_i, \in 0, 1, b_{n-1} = 1$

 $\text{output: } r,q \in \mathbb{N} \quad such that \\ a = q \cdot qb + r, \quad 0 \leq r < b$

- (1) $r = a_i$; q = 0
- (2) for i = m, m 1, ..., 0 do
- (3) if $r < 2^i \cdot b$ then set $r := r 2^i \cdot b$, $q = q + 2^i$

1.2.2 Proposition 2 (Runtime of Algorithm 1)

Algorithm 1 is correct and requires $O(n \cdot (m+1))$ bit operations.

Proof:

Always have $a = q \cdot b + r$

Claim:

before setp (3), have $0 \le 2^{i+1} \cdot b$

$$i = m;$$
 $0 \le r = a < 2^{m+n} = 2^{m+1}c \cdot 2^{n-1} \le 2^{m-1} \cdot b$ $i < m$ By step (3)

So after last passage through the loop $0 \le r < b$

Running Time: In step(3), have comparison and (possiby) subtraction. Only n bits involved $\Rightarrow O(n)$

Total: $O(b \cdot (m+1))$

Remarks:

(1) Division with remainder can be reduced to multiplication.

Precisely: given an algorithm for multiplication that requires M(n) bit operations, there exists an algorithm for division with remainder that requires O(M(n)) bit operations.

- (2) Practically relevant: Jebelean's algorithm (1997): $O(n^{\lg 3})$
- (3) Alternatively, may choose $r\mathbb{Z}$ such that $\lfloor \frac{-b}{2} \rfloor < r \leq \lfloor \frac{b}{2} \rfloor$
- (4) Algorithm 1 extends to \mathbb{Z} .
- (5) All Euclidean rings have division with remainder (by definition). (e.g., $R = K[x] \rightarrow \text{polynomial ring over field}$, $R = \mathbb{Z}[i] = \{a + bi | a, b \in \mathbb{Z}\} \subseteq \mathbb{C}, \quad i^2 = -1$)

1.2.3 Algorithm 3 (Euclidean algorithm)

input : $a, b \in \mathbb{N}$

output: gcd(a,b) "greatest common divisor"

- (1) set $r_0 := a$, $r_i := b$
- (2) for i = 1, 2, 3, ... perform steps (3) and (4)
- (3) if $r_i = 0$ then $gcd(a, b) = |r_{i-1}|$
- (4) Division with remainder: $r_{i-1} = q \cdot r_i + r_{i+1}$ $r_{i+1} \in \mathbb{Z}$ $|r_{i+1}| \leq \frac{1}{2}|r_i|$

Example:

$$a = 287, \quad b = 126$$

$$287 = 2 \cdot 126 + 35 \tag{1}$$

$$126 = 4 \cdot 35 - 14 \tag{2}$$

$$35 = (-2) \cdot (-14) + 7 \tag{3}$$

$$-14 = (-2) \cdot 7 + 0 \tag{4}$$

So:
$$7|(-14) \Longrightarrow 7|35$$

 $\Longrightarrow 7|126$
 $\Longrightarrow 7|287$

On the other hand take a common divisor d; d|287; d|126 $\Longrightarrow_{(1)} d|d \Longrightarrow_{(2)} d|14 \Longrightarrow_{(3)} d|7$

1.2.4 Theorem 4 (Correctness of Algorithm 3)

Algorithm 3 is correct.

Proof:

Since $r_{i-1} = q \cdot r_i + r_{i+1}$ every integer $x \in \mathbb{Z}$ satisfies the equivalence $x | r_{i-1}$ and $x|r_i \Leftrightarrow x|r_{i+1} \text{ and } x|r_i \text{ so } gcd(r_{i-1},r_i) = gcd(r_i,r_{i+1}) = gcd(a,b)$ when terminating have $gcd(a,b) = gcd(r_{i-1},0) = |r_{i-1}|$

1.2.5 Theorem 5 (Runtime of Algorithm 3)

Algorithm 3 requires $O(m \cdot n)$ bit operations for n = l(a), m = l(b)

Proof:

If a < b than the first passage yields $r_2 = a$, $r_1 = b$. Cost: O(n)May assume: $a \ge b$. Write $n_i = l(r_i)$

May assume: $a \ge b$. Write $n_i = \iota_{(i,i)}$ By Proposition 2 $\exists c$ constant such that the total time is $\le c \cdot \sum_{i=1}^k n_i \cdot (n_{i-1} - n_i + 1)$ $=:\sigma(n_0, ..., n_k)$

For i > 2: $n_i = n_{i-1} - 1$

Special Case: $n_i = n_{i-1} - 1$ for $i \ge 2$

 $\Rightarrow n_i = n_i - i + 1, \quad n_i = m, \quad k = m + 1$

Obtain $\sigma(n_0,...,n_k) = m \cdot (n-m+1) + \sum_{i=2}^{m+1} (m-i+1) \cdot 2 = m \cdot n - m^2 + m + m(m-1) = m * n.$

Claim: The special case is the worst (most expensive)!

From any sequence $n_1 > n_2 > ... > n_k$ get to the special case by iteratively inserting numbers in the gaps. Insert s with $n_{j-1} > s > n_j$.

 $\sigma(n_0, ..., n_{j-1}, s, n_j, ..., n_k) - \sigma(n_0, ..., n_k) = ... = s + (n_{j-1} - s) \cdot (s - n_j)$ $sp\sigma(n_0,...,n_k) \le \sigma(n,m,m-1,...,2,1,0) = n \cdot m$

Complexity is quadratic \rightarrow cheap

1.2.6 Algorithm 6 (Extended Euclidean Algorithm)

input : $a, b \in \mathbb{N}$

output: d = gcd(a, b) and $s, t \in \mathbb{Z}$ such that $d = s \cdot a + t \cdot b$

- (1) $r_0 := a, r_1 := b, s_0 := 1, t_0 := 0, s_1 := 0, t_1 = 1$
- (2) for i = 1, 2, ... perform steps (3) (5)
- if $r_i = 0$ set $d = |r_{i-1}|$ $s := sgn(r_{i-1}) \cdot s_{i-1}, t := sgn(r_{i-1}) \cdot t_{i-1}$ (3)
- division with remainder: $r_{i-1} = q_i \cdot r_i + r_{i+1}$, with $|r_{i+1}| \leq \frac{1}{2}|r_i|$ (4)
- (5) $set s_{i+1} := s_{i_1} - q_i \cdot s_i,$ $t_{i+1} := t_{i-1} - q_i \cdot t_i$

Justification: $r_i = s_i \cdot a + t_i \cdot b$ throughout

Application: $m \in \mathbb{N}$, $x \in \mathbb{N}$ such that m, x coprime (i.e. qcd(x, m) = 1

Algorithm 6 yields: $1 = s \cdot x + t \cdot m \Rightarrow s \cdot x \equiv 1 \mod m$. So obtain inverse of $\bar{x} \in \mathbb{Z}/(m)$

1.3 Primality testing

Let $\mathbb{P} \subseteq \mathbb{N}$ be the set of prime numbers.

Challenge: Given $n \in \mathbb{N}$ decide if $n \in \mathbb{P}$

Naive Method: Trivial division by $m \leq |\sqrt{n}|$.

Running time is exponential in l(n). Even when restricted to division by prime numbers,

need approximatily $\frac{\sqrt{n}}{|n|\sqrt{n}}$ trivial divisions (prime number theorem)

 \rightarrow hardly any better!

Reminder: (arithmetic modulo m)

 $G ext{ finite group } \Rightarrow \forall a \in G \quad a^{|G|} = 1$ Fermat's little theorem

For $G = (\mathbb{Z}/(p)^x \quad a^{p-1} \equiv 1 \mod p \in \mathbb{P} \quad \forall a \in \mathbb{Z} \quad \text{with pfa ???}$

Infact $(\mathbb{Z}/(p))^x \cong Z_{p-1}$ is cyclic

For $m = p_1^e, ...p_r^{e_r}$ with $p_i \in \mathbb{P}$, $e_i \in \mathbb{N}_{>0}$:

$$\mathbb{Z}_{(m)} \cong \mathbb{Z}_{(p_i^{e_i})}^{-} \oplus \ldots \oplus \mathbb{Z}_{(p_r^{e_r})}^{-} \Rightarrow \mathbb{Z}_{(m)}^{x} \cong \mathbb{Z}_{(p_i^{e_i})}^{x} \times \ldots \times \mathbb{Z}_{(p_r^{e_r})}^{x}$$

what is $\mathbb{Z}_{(p^e)}$ for $p \in \mathbb{P}$, $e \in \mathbb{N}_{>0}$?

1.3.1 Theorem 1

Let $p \in \mathbb{P}off \ e \in \mathbb{N}_{>0} \Rightarrow (\mathbb{Z}_{(p^e)})^x = Z_{(p-1)\cdot p^{e-1}}$ cyclic

Proof:

$$(\mathbb{Z}_{(p^e)})^x \cong \mathbb{Z}_{p-1} \Rightarrow \exists z \in \mathbb{Z} : order(z + p\mathbb{Z}) = p - 1$$

Set
$$a = \bar{z}^{p^{e-1}} \in (\mathbb{Z}_{(p^e)})^x =: G$$

 $a^{p-1} = \bar{z}^{(p-1) \cdot p^{e-1}} = \bar{z}^{|a|} = 1$

$$a^{p-1} = \bar{z}^{(p-1)\cdot p^{e-1}} = \bar{z}^{|a|} = 1$$

On the other hand, take $i \in \mathbb{Z}$ such that

$$a^{i} = 1 \Rightarrow z^{i \cdot p^{e-1}} \equiv 1 \mod p \Rightarrow (p-1)|(i-p^{e-1}) \Rightarrow (p-1)|i.$$

So ord(a) = p - 1.

Now consider $b = (p + 1) \in G$

Claim: $ord(b) = p^{e-1}$

Proof by induction on $k \in N_{>0}$ that $(p+1)^{p^{k-1}} \equiv p^k + 1 \mod p^{k+1}$

 $k \to k+1$: B; induction have $(p+1)^{p^{k-1}} = 1 + p^k + x \cdot p^{k+1}, \quad x \in \mathbb{Z}$ Compute:

$$(p+1)^{p^k} = ((1+p^k) + x \cdot p^{k+1})^p = \sum_{i=0}^p (p \text{ over } i)(i+p^k)^{p-i} \cdot x^i \cdot p^{i \cdot (k+1)}$$

$$\underbrace{\equiv}_{\text{Only 0-th summand}} (i+p^k) = \sum_{i=0}^p (p \text{ over } i) p^{i \cdot k} \underbrace{\equiv}_{p \text{ odd}} 1 + p^{k+1} \text{ mod } p^{k+2} \quad \checkmark$$

For
$$k = e : (p+1)^{p^{e-1}} \equiv | \mod p^e \Rightarrow b^{p^e} = 1 \Rightarrow ord(b)|p^{e-1}$$

But $(p+1)^{p^{e-2}} \equiv p^{e-1} + 1 \mod p^e \Rightarrow b^{p^{e-2}} \neq 1 \in G$

But
$$(p+1)^{p^{e-2}} \equiv p^{e-1} + 1 \mod p^e \Rightarrow b^{p^{e-2}} \neq 1 \in G$$

So $ord(b) = p^{e-1}$

Claim: $ord(a \cdot b) = (p-1)p^{e-1} \quad (\Rightarrow \text{Theorem})$ Let $(a \cdot b)^i = 1 \in G$ with $i \in \mathbb{Z}$ Then $1 = (a \cdot b)^{i \cdot (p-1)} = (a^{p-1})^i \cdot b^{i \cdot (p-1)} = b^{i \cdot (p-1)} \Rightarrow p^{e-1}|i \cdot i(p-1) \Rightarrow p^{e-1}|i$ Also $1 = (a \cdot b)^{p^{e-1} \cdot i} = a^{p^{e-1}} \Rightarrow (p-1)|p^{e-1} \cdot i \Rightarrow (p-1)|i \to (p-1) \cdot p^{e-1}|i$ Reminder: $(\mathbb{Z}/(2^e))^x \cong Z_2 \times Z_2^{e-2} \quad (e \ge 2)$

1.3.2 Algorithm 2 (Fermat Test)

input : $n \in \mathbb{N}_{>0}odd$

output: " $n \notin \mathbb{P}$ " or "probably $n \in \mathbb{P}$ "

- (1) Choose $a \in 2, ..., n-1$ randomly
- (2) Compute $a^{n-1} \mod n$
- (3) If $a^{n-1} \neq 1 \mod n$ then return " $n \notin \mathbb{P}$ " otherwise return "probably $n \in \mathbb{P}$ "

Not very satisfying. Is this fast

1.3.3 Algorithm 3 (fast exponention)

input : $a \in G$ G is a monoid, $e \in \mathbb{N}, e = \sum_{i=0}^{n-1} e_i 2^i, e_i \in \{0,1\}$

output: $a^e \in G$

- (1) Set b := a, y := 1
- (2) For i = 0, ..., n 1 perform (3) (4)
- (3) if $e_i = i$ set $y := y \cdot b$
- $(4) set b := b^2$
- (5) return y

this requires O(l(e)) operations in G

For $G = (\mathbb{Z}/(n)_i)$, each multiplication requires $O(l(n)^2)$ bit operations \Rightarrow Fermat test requires $O(l(n)^3)$ bit operations \rightarrow cubic complexity \rightarrow "fast"!

Example:

 $n=561=3\cdot 11\cdot 17$ For $a\in\mathbb{Z}$ with $gcd(a,n)\Rightarrow$ have $a^{n-1}=(a^2)^{280}\equiv 1\ mod\ 3$ $a^{n-1}\equiv 1\ mod\ n$ Fermat's test says "probably $n\in\mathbb{P}$ " in 57% of cases. $n=2207\cdot 6619\cdot 15443$: output "probably $n\in\mathbb{P}$ " in 99,93% of cases.

1.3.4 Definition 4

Let $n \in N_{>1}odd$, $a \in 1, ..., n-1$

- (a) n is pseudo-prime to base a if $a^{n-1} \equiv 1 \mod n$
- (b) otherwise a is called a witness of composition of n
- (c) If $n \notin \mathbb{P}$ but $a^{n-1} \equiv 1 \mod n$ $\forall a \text{ with } \gcd(n, a) = 1$ then n is called a Carmichael number

There are ∞ Carmichael numbers

1.3.5 Proposition 5:

Let $n \in N_{>1}$, $odd \notin \mathbb{P}$ not Carmichael $\Rightarrow |\{a \in \mathbb{Z} | 0 < a < n \text{ a witness of composite of } n\} > \frac{n-1}{2}$ **Proof:** Consider $\phi : (\mathbb{Z}/(n))^x =: G \to G, \quad \bar{a} \mapsto \bar{a}^{n-1}$ group homomorphism. By assumption, $|im(\phi| > 1 \Rightarrow |Ker(\phi)| \leq \frac{|a|}{2} < \frac{n-1}{2}$ $\Rightarrow |\{a \in \mathbb{Z} | 0 < a < n \text{ a witness of composite of } n\} > \frac{n-1}{2}$

polynomial rings measure not abs value but max power

2 Notes

- a|b a is divisible by b
- *ord*(*a*)
- char(A) the smallest positive n such that $\underbrace{1+\ldots+1}_{n\ summands}=0$ with 1 as the multiplicative identity element
- $\mathbb{Z}/(m)$ Ring modulo m
- $\lg(x) = \log_2(x)$
- Average number of bit operations for an increment:
 one operation for the last bit + 50% chance for one on the next bit + 25% on the
 following etc. ⇒ Geometrical row
 ⇒ on average two bit operations