

4. ~~Fast~~ Fast exponentiation

Thm 4.1 Let G be a semigroup and assume we can multiply ^{two} el. of G in $\mathcal{O}(1)$.

Then, we can compute x^k for $x \in G$ and $k \geq 1$ in time $\mathcal{O}(\log k)$ (for large n) on an $\mathcal{O}(\log k)$ -bit RAM.

Pf If $2 \mid n$, then $x^k = (x^{k/2})^2$.

If $2 \nmid n$, then $x^k = (x^{(k-1)/2})^2 \cdot x$

$$\begin{aligned} & \text{computable} \\ & \text{in } \leq C \log_2 \frac{k}{2} \\ & = C(\log_2 k - 1) \end{aligned}$$

□

(See also pf 2 on following page.)

Thm 4.2 We can compute ~~the product~~ x^k for any binary integers $x \in \mathbb{Z}$ with $\leq n$ bits ~~and~~ ^{and $k \geq 1$} in time $\mathcal{O}(nk)$ on an $\mathcal{O}(\log(nk))$ -bit RAM.

Pf As for Thm 4.1, using that x^{nk} has $\mathcal{O}(nk)$ bits, so after computing $x^{\lfloor k/2 \rfloor}$ recursively in time $\leq C n \cdot \frac{k}{2}$, we can compute x^k in time $\mathcal{O}(nk)$. \rightarrow total time: geom. series.

Prnk The obvious method ~~of computing~~ $(x^k = x^{k-1} \cdot x)$ would ^(usually) have running time $\mathcal{O}(nk^2)$ ~~because~~ because the nr. of digits in step i is $\sim ni$ and there are k steps.

~~Q1~~
Q2 ^{of Thm 4.1} Write $k = \sum_{i=0}^m a_i z^i$, $a_i \in \{0, 1\}$

~~Compute~~

$$\Rightarrow x^k = \prod_{\substack{i: \\ a_i=1}} x^{2^i}.$$

compute $b_i := x^{2^i}$ for $i=0, \dots, m$ using the recurrence $b_{i+1} = b_i^2$.

then, compute the product $\prod_{\substack{i: \\ a_i=1}} b_i$. ~~starting with $i=0$~~ □

Principle similar for (\mathbb{Q}, \cdot) ~~(\mathbb{Q}, \cdot)~~ , if you allow nonreduced fractions $\frac{p}{q}$ (reducing the final result would take time $\mathcal{O}(nk \log(nk))$).

$(M_n(\mathbb{Q}), \cdot)$ ~~$(M_n(\mathbb{Q}), \cdot)$~~ , $(K[x], \cdot)$, ~~$(K[x], \cdot)$~~ , ...

Warning you can often do faster!

E.g. for the semigroup $(\mathbb{Q}, +)$: ~~you~~ you can compute $k \cdot x$ in $\mathcal{O}(n + \log k)$.

for 4.3 you can compute the n -th Fibonacci number F_n in time $\mathcal{O}(n)$ \uparrow $\mathcal{O}(n)$ digits
 you can compute $F_n \text{ mod } p$ in time $\mathcal{O}_p(\log n)$.

Pf
$$\begin{pmatrix} F_{n+1} \\ F_{n+2} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} F_n \\ F_{n+1} \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} F_n \\ F_{n+1} \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}^n}_{\substack{\mathcal{O}(n)\text{-bit matrix} \\ \text{computable} \\ \text{in } \mathcal{O}(n)}} \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

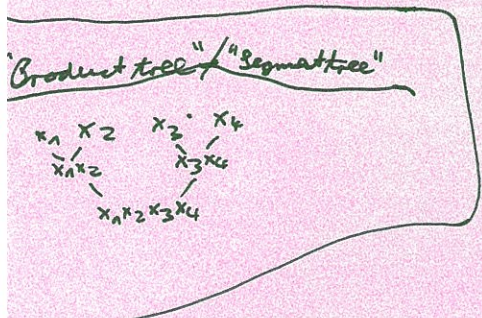
□

5. Multiplying more than two things

Thm 5.1 We can compute the prod. $x_1 \dots x_k$ for any bin. int. x_1, \dots, x_k with $\leq n$ bits in time $O(nk \log k)$ on an $O(\log(k))$ -bit RAM.

Pf ~~.....~~
w.l.o.g. $k = 2^a$.

$$x_1 \dots x_k = \underbrace{(x_1 \dots x_{2^{a-1}})}_{\substack{O(n \cdot 2^{a-1}) \text{ bits} \\ \text{time } \leq C n \cdot 2^{a-1} (a-1)}} \underbrace{(x_{2^{a-1}+1} \dots x_{2^a})}_{\dots}$$



$O(n \cdot 2^a)$ bits
time ~~.....~~ $\leq C n \cdot 2^a (a-1) + O(n \cdot 2^a)$
 $\leq C n \cdot 2^a a$ for large C .

□

Rule Obvious alg.: time $O(nk^2)$

Rule similar for (\mathbb{Q}, \cdot) , $(M_n(\mathbb{Q}), \cdot)$, $(\mathbb{Q}, +)$, ...

Rule There are better alg. for $(\mathbb{Z}, +)$, (\mathbb{Z}, max) , (F_2, \cdot) , ...
 \uparrow
 free group over two elements

~~.....~~
 Don't reduce intermediate results! (computing the gcd would take nonlinear time.)

~~Cor 5.2~~ Given integers x_1, \dots, x_k with $\leq n$ bits, you can compute the ~~(reduced)~~ numerator and denom. p, q of a continued fraction

$$\frac{p}{q} = x_1 + \frac{1}{x_2 + \frac{1}{\dots \frac{1}{x_k}}}$$

in ~~RAM~~ on an $O(\log(nk))$ -bit RAM.
 $O(nk \log(k))$

Pr By induction,

~~RAM~~

$$\begin{pmatrix} p \\ q \end{pmatrix} = \underbrace{\begin{pmatrix} x_1 & 1 \\ 1 & 0 \end{pmatrix} \dots \begin{pmatrix} x_k & 1 \\ 1 & 0 \end{pmatrix}}_{\text{compute this prod.}} \begin{pmatrix} 1 \\ 0 \end{pmatrix}.$$

in $O(nk \log k)$

□

Cor 5.3 Given integers a_0, \dots, a_n and x with $\leq m$ bits, you can compute $\sum_{i=0}^n a_i x^i$ (in binary) in time $O(mn \log n)$.

Pf By induction,

$$\begin{pmatrix} \sum_{i=0}^n a_i x^i \\ x^{n+1} \end{pmatrix} = \begin{pmatrix} 1 & a_n \\ 0 & x \end{pmatrix} \cdots \begin{pmatrix} 1 & a_0 \\ 0 & x \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix}.$$

□

What if x_1, \dots, x_k have very different numbers of bits?

Thm 5.4 Let x_1, \dots, x_k be integers with n_1, \dots, n_k bits.

We can compute $x_1 \dots x_k$ in time

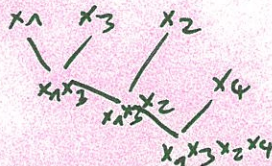
$$O\left(\sum_{i=1}^k n_i \left(\log \frac{n_1 + \dots + n_k}{n_i} + 1\right)\right) \text{ on an } O\left(\sum_{i=1}^k n_i\right) \text{ bit RAM.}$$

Pr ~~start with the list of numbers.~~

~~start with the list of numbers.~~ w.l.o.g. $x_1, \dots, x_n \neq 0, \Rightarrow n_i = \log |x_i| + O(1)$.

start with the list x_1, \dots, x_n .
In each step, replace the two integers ^{smallest $\log |x_i|$} from the list with the ~~smallest~~ by their product, until there's just one number.

\leadsto product tree



$$\text{Running time} \leq \sum_i \log |x_i| \cdot (\text{distance of } x_i \text{ from root})$$

$\leq (n_1 + \dots + n_k)$ times the average length of Huffman code over x_1, \dots, x_n if the probability of x_i is $p_i = \frac{n_i}{n_1 + \dots + n_k}$

~~Shannon entropy~~

$$= (n_1 + \dots + n_k) \cdot [\text{Shannon entropy} + O(1)]$$

$$\sum p_i \log \frac{1}{p_i}$$

□

A Theorem about Shannon codes

(Shannon: *Mathematical Theory of Communication*; or look up "Shannon-Fano coding" on Wikipedia)

Thm 5.5 Let x be an int, with n bits and let y_1, \dots, y_k be integers with m_1, \dots, m_k bits.

We can compute $x \bmod y_1, \dots, x \bmod y_k$ in time

$$O\left(n + \sum m_i \left(\log \frac{m_1 + \dots + m_k}{m_i} + 1\right)\right)$$

pf Consider the product tree for y_1, \dots, y_k constructed in the proof of Thm 5.4.

~~For~~ For each node (labeled t), compute $x \bmod t$, starting from the root. Note that if the parent node is labeled s , then $(x \bmod t) = \underbrace{(x \bmod s)}_{s|s} \bmod t$.

$$t|s, \text{ so}$$

$$s|s$$

□