6.2 Master Theorem

Note that the cases do not cover all possibilities.

Lemma 5

Let $a \ge 1, b > 1$ and $\epsilon > 0$ denote constants. Consider the recurrence

$$T(n) = aT\left(\frac{n}{b}\right) + f(n) \; .$$

Case 1. If $f(n) = O(n^{\log_b(a)-\epsilon})$ then $T(n) = O(n^{\log_b a})$.

Case 2. If $f(n) = \Theta(n^{\log_b(a)} \log^k n)$ then $T(n) = \Theta(n^{\log_b a} \log^{k+1} n)$, $k \ge 0$.

Case 3. If $f(n) = \Omega(n^{\log_b(a)+\epsilon})$ and for sufficiently large n $af(\frac{n}{b}) \le cf(n)$ for some constant c < 1 then $T(n) = \Theta(f(n))$.



We prove the Master Theorem for the case that n is of the form b^{ℓ} , and we assume that the non-recursive case occurs for problem size 1 and incurs cost 1.



The running time of a recursive algorithm can be visualized by a recursion tree:

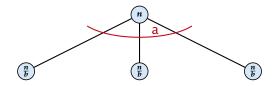


The running time of a recursive algorithm can be visualized by a recursion tree:

n

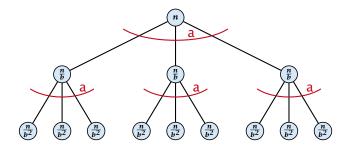


The running time of a recursive algorithm can be visualized by a recursion tree:





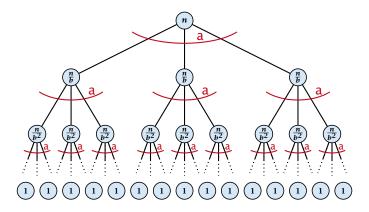
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6.2 Master Theorem

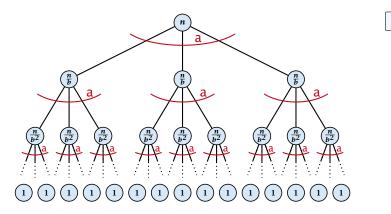
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6.2 Master Theorem

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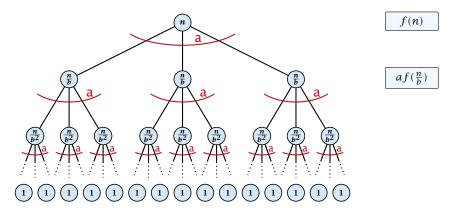


f(n)



6.2 Master Theorem

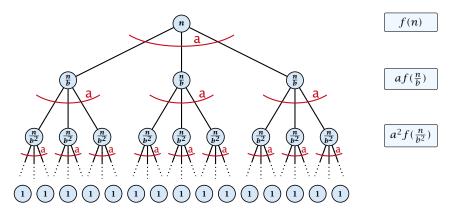
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6.2 Master Theorem

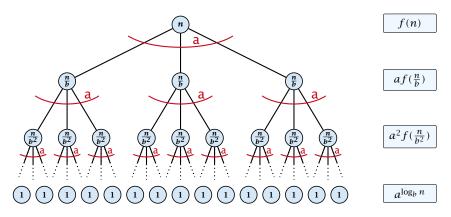
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6.2 Master Theorem

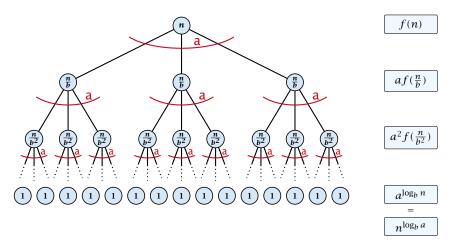
The running time of a recursive algorithm can be visualized by a recursion tree:





6.2 Master Theorem

The running time of a recursive algorithm can be visualized by a recursion tree:





6.2 Master Theorem

This gives

$$T(n) = n^{\log_b a} + \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right) \ .$$





 $T(n) - n^{\log_b a}$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
$$\leq c \sum_{i=0}^{\log_b n-1} a^i \left(\frac{n}{b^i}\right)^{\log_b a-\epsilon}$$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
$$\leq c \sum_{i=0}^{\log_b n-1} a^i \left(\frac{n}{b^i}\right)^{\log_b a-\epsilon}$$

 $b^{-i(\log_b a - \epsilon)} = b^{\epsilon i} (b^{\log_b a})^{-i} = b^{\epsilon i} a^{-i}$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
$$\leq c \sum_{i=0}^{\log_b n-1} a^i \left(\frac{n}{b^i}\right)^{\log_b a-\epsilon}$$
$$\underbrace{b^{-i(\log_b a-\epsilon)} = b^{\epsilon i}(b^{\log_b a})^{-i} = b^{\epsilon i}a^{-i}}_{b=0} = c n^{\log_b a-\epsilon} \sum_{i=0}^{\log_b n-1} (b^{\epsilon})^i$$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
$$\leq c \sum_{i=0}^{\log_b n-1} a^i \left(\frac{n}{b^i}\right)^{\log_b a-\epsilon}$$
$$\boxed{b^{-i(\log_b a-\epsilon)} = b^{\epsilon i}(b^{\log_b a})^{-i} = b^{\epsilon i}a^{-i}} = c n^{\log_b a-\epsilon} \sum_{i=0}^{\log_b n-1} (b^{\epsilon})^i$$
$$\boxed{\sum_{i=0}^k q^i = \frac{q^{k+1}-1}{q-1}}$$

q-1



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
$$\leq c \sum_{i=0}^{\log_b n-1} a^i \left(\frac{n}{b^i}\right)^{\log_b a-\epsilon}$$
$$\frac{b^{-i(\log_b a-\epsilon)} = b^{\epsilon i}(b^{\log_b a})^{-i} = b^{\epsilon i}a^{-i}}{\sum_{i=0}^{k-1} c n^{\log_b a-\epsilon}} = c n^{\log_b a-\epsilon} \sum_{i=0}^{\log_b n-1} (b^{\epsilon})^i$$
$$\frac{\sum_{i=0}^k q^i = \frac{q^{k+1}-1}{q-1}}{\sum_{i=0}^k c n^{\log_b a-\epsilon} (b^{\epsilon \log_b n} - 1)/(b^{\epsilon} - 1)}$$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
$$\leq c \sum_{i=0}^{\log_b n-1} a^i \left(\frac{n}{b^i}\right)^{\log_b a-\epsilon}$$
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$$\frac{\sum_{i=0}^k q^i = \frac{q^{k+1}-1}{q-1}}{\sum_{i=0}^k a^{-\epsilon}} = c n^{\log_b a-\epsilon} (b^{\epsilon \log_b n} - 1)/(b^{\epsilon} - 1)$$
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6.2 Master Theorem

$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
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$$\frac{\sum_{i=0}^{k} q^i = \frac{q^{k+1}-1}{q-1}}{\sum_{i=0}^{k-1} c n^{\log_b a-\epsilon} (b^{\epsilon}\log_b n-1)/(b^{\epsilon}-1)}$$
$$= c n^{\log_b a-\epsilon} (n^{\epsilon}-1)/(b^{\epsilon}-1)$$
$$= \frac{c}{b^{\epsilon}-1} n^{\log_b a} (n^{\epsilon}-1)/(n^{\epsilon})$$



$$T(n) - n^{\log_{b} a} = \sum_{i=0}^{\log_{b} n-1} a^{i} f\left(\frac{n}{b^{i}}\right)$$

$$\leq c \sum_{i=0}^{\log_{b} n-1} a^{i} \left(\frac{n}{b^{i}}\right)^{\log_{b} a-\epsilon}$$

$$\frac{b^{-i(\log_{b} a-\epsilon)} = b^{\epsilon i}(b^{\log_{b} a})^{-i} = b^{\epsilon i}a^{-i}}{\sum_{i=0}^{k-1} e^{i} c n^{\log_{b} a-\epsilon}} \sum_{i=0}^{\log_{b} n-1} (b^{\epsilon})^{i}$$

$$\frac{\sum_{i=0}^{k} q^{i} = \frac{q^{k+1}-1}{q-1}}{e^{-1}} = c n^{\log_{b} a-\epsilon} (b^{\epsilon \log_{b} n} - 1)/(b^{\epsilon} - 1)$$

$$= c n^{\log_{b} a-\epsilon} (n^{\epsilon} - 1)/(b^{\epsilon} - 1)$$

$$= \frac{c}{b^{\epsilon} - 1} n^{\log_{b} a} (n^{\epsilon} - 1)/(n^{\epsilon})$$

Hence,

$$T(n) \leq \left(\frac{c}{b^{\epsilon}-1}+1\right) n^{\log_b(a)}$$



$$T(n) - n^{\log_{b} a} = \sum_{i=0}^{\log_{b} n-1} a^{i} f\left(\frac{n}{b^{i}}\right)$$

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$$\frac{\sum_{i=0}^{k} q^{i} = \frac{q^{k+1}-1}{q-1}}{\sum_{i=0}^{k} a^{-\epsilon}} = c n^{\log_{b} a-\epsilon} (b^{\epsilon} \log_{b} n - 1)/(b^{\epsilon} - 1)$$

$$= c n^{\log_{b} a-\epsilon} (n^{\epsilon} - 1)/(b^{\epsilon} - 1)$$

$$= \frac{c}{b^{\epsilon} - 1} n^{\log_{b} a} (n^{\epsilon} - 1)/(n^{\epsilon})$$

Hence,

$$T(n) \leq \left(\frac{c}{b^{\epsilon}-1}+1\right) n^{\log_b(a)} \qquad \qquad \Rightarrow T(n) = \mathcal{O}(n^{\log_b a}).$$





 $T(n) - n^{\log_b a}$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$



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$$\leq c \sum_{i=0}^{\log_b n-1} a^i \left(\frac{n}{b^i}\right)^{\log_b a}$$
$$= c n^{\log_b a} \sum_{i=0}^{\log_b n-1} 1$$



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Hence,

 $T(n) = \mathcal{O}(n^{\log_b a} \log_b n)$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
$$\leq c \sum_{i=0}^{\log_b n-1} a^i \left(\frac{n}{b^i}\right)^{\log_b a}$$
$$= c n^{\log_b a} \sum_{i=0}^{\log_b n-1} 1$$
$$= c n^{\log_b a} \log_b n$$

Hence,

 $T(n) = \mathcal{O}(n^{\log_b a} \log_b n) \qquad \Rightarrow T(n) = \mathcal{O}(n^{\log_b a} \log n).$





 $T(n) - n^{\log_b a}$



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$$n=b^\ell \Rightarrow \ell = \log_b n$$



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$$\boxed{n = b^\ell \Rightarrow \ell = \log_b n} = c n^{\log_b a} \sum_{i=0}^{\ell-1} \left(\log_b \left(\frac{b^\ell}{b^i}\right)\right)^k$$



$$T(n) - n^{\log_b a} = \sum_{i=0}^{\log_b n-1} a^i f\left(\frac{n}{b^i}\right)$$
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$$= c n^{\log_b a} \sum_{i=0}^{\ell-1} (\ell - i)^k$$



$$T(n) - n^{\log_{b} a} = \sum_{i=0}^{\log_{b} n-1} a^{i} f\left(\frac{n}{b^{i}}\right)$$

$$\leq c \sum_{i=0}^{\log_{b} n-1} a^{i} \left(\frac{n}{b^{i}}\right)^{\log_{b} a} \cdot \left(\log_{b} \left(\frac{n}{b^{i}}\right)\right)^{k}$$

$$\overline{n = b^{\ell} \Rightarrow \ell = \log_{b} n} = c n^{\log_{b} a} \sum_{i=0}^{\ell-1} \left(\log_{b} \left(\frac{b^{\ell}}{b^{i}}\right)\right)^{k}$$

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$$T(n) - n^{\log_{b} a} = \sum_{i=0}^{\log_{b} n-1} a^{i} f\left(\frac{n}{b^{i}}\right)$$

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$$= c n^{\log_{b} a} \sum_{i=0}^{\ell-1} (\ell - i)^{k}$$

$$= c n^{\log_{b} a} \sum_{i=1}^{\ell} i^{k} \approx \frac{1}{k} \ell^{k+1}$$



$$T(n) - n^{\log_{b} a} = \sum_{i=0}^{\log_{b} n-1} a^{i} f\left(\frac{n}{b^{i}}\right)$$

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$$\approx \frac{c}{k} n^{\log_{b} a} \ell^{k+1} \qquad \Rightarrow T(n) = \mathcal{O}(n^{\log_{b} a} \log^{k+1} n)$$



Where did we use $f(n) \ge \Omega(n^{\log_b a + \epsilon})$?



6.2 Master Theorem

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From this we get $a^i f(n/b^i) \le c^i f(n)$, where we assume that $n/b^{i-1} \ge n_0$ is still sufficiently large.



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$$q < 1: \sum_{i=0}^{n} q^{i} = \frac{1-q^{n+1}}{1-q} \le \frac{1}{1-q}$$



From this we get $a^i f(n/b^i) \le c^i f(n)$, where we assume that $n/b^{i-1} \ge n_0$ is still sufficiently large.

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Hence,

 $T(n) \leq \mathcal{O}(f(n))$

Where did we use $f(n) \ge \Omega(n^{\log_b a + \epsilon})$?



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From this we get $a^i f(n/b^i) \le c^i f(n)$, where we assume that $n/b^{i-1} \ge n_0$ is still sufficiently large.

$$T(n) - n^{\log_{b} a} = \sum_{i=0}^{\log_{b} n-1} a^{i} f\left(\frac{n}{b^{i}}\right)$$
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$$\leq 1 : \sum_{i=0}^{n} q^{i} = \frac{1-q^{n+1}}{1-q} \leq \frac{1}{1-q} \leq \frac{1}{1-c} f(n) + \mathcal{O}(n^{\log_{b} a})$$

Hence,

q <

$$T(n) \leq \mathcal{O}(f(n))$$

$$\Rightarrow T(n) = \Theta(f(n)).$$

Where did we use $f(n) \ge \Omega(n^{\log_b a + \epsilon})$?



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Suppose we want to multiply two n-bit Integers, but our registers can only perform operations on integers of constant size.



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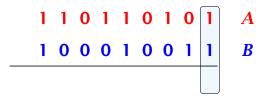
For this we first need to be able to add two integers **A** and **B**:

1 1 0 1 0 1 0 1 A 1 0 0 0 1 0 0 1 1 B



Suppose we want to multiply two n-bit Integers, but our registers can only perform operations on integers of constant size.

For this we first need to be able to add two integers **A** and **B**:



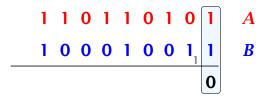


6.2 Master Theorem



Suppose we want to multiply two n-bit Integers, but our registers can only perform operations on integers of constant size.

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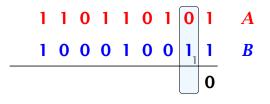




6.2 Master Theorem

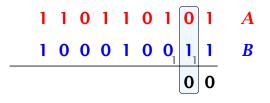


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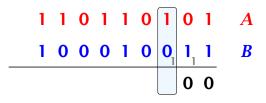
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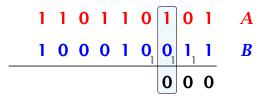




6.2 Master Theorem

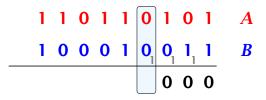
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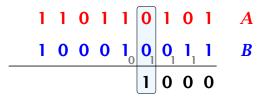


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6.2 Master Theorem

25. Oct. 2024 63/69

Suppose we want to multiply two n-bit Integers, but our registers can only perform operations on integers of constant size.

For this we first need to be able to add two integers **A** and **B**:

This gives that two *n*-bit integers can be added in time O(n).



Suppose that we want to multiply an *n*-bit integer A and an *m*-bit integer B ($m \le n$).

This is also nown as the "school method" for multiplying integers.
Note that the intermediate numbers that are generated can have at most m + n ≤ 2n bits.



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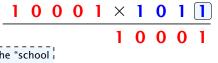


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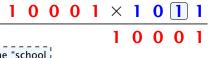
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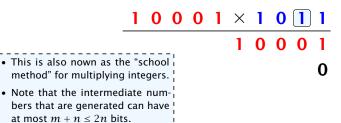
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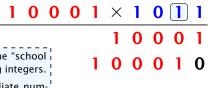


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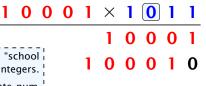






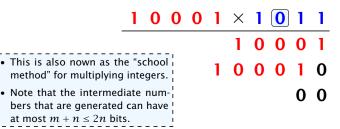
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	1	0	0	0	1	\times	1	0	1	1
						1	0	0	0	1
This is also nown as th method" for multiplying					1	0	0	0	1	0
Note that the intermed bers that are generated at most $m + n \le 2n$ bits	can			0	0	0	0	0	0	0



	1	0	0	0	1	\times	1	0	1	1
						1	0	0	0	1
• This is also nown as the method" for multiplying					1	0	0	0	1	0
 Note that the intermediate bers that are generated at most <i>m</i> + <i>n</i> ≤ 2<i>n</i> bits 	can			0	0	0	0	0	0	0



Suppose that we want to multiply an *n*-bit integer *A* and an *m*-bit integer *B* ($m \le n$).

	1	0	0	0	1	\times	1	0	1	1
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 This is also nown as th method" for multiplying 					1	0	0	0	1	0
• Note that the intermed bers that are generated				0	0	0	0	0	0	0
at most $m + n \le 2n$ bit								0	0	0



ı,

	1	0	0	0	1	\times	1	0	1	1
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at most $m + n \le 2n$ bits	5. 		1	0	0	0	1	0	0	0



	1	0	0	0	1	X	1	0	1	1
			_			1	0	0	0	1
• This is also nown as th method" for multiplying					1	0	0	0	1	0
• Note that the intermed bers that are generated				0	0	0	0	0	0	0
at most $m + n \le 2n$ bit			1	0	0	0	1	0	0	0



	1	0	0	0	1	×	1	0	1	1
						1	0	0	0	1
 This is also nown as the method" for multiplying 					1	0	0	0	1	0
Note that the intermedi bers that are generated	ate	num	- [0	0	0	0	0	0	0
at most $m + n \le 2n$ bits			1	0	0	0	1	0	0	0
-			1	0	1	1	1	0	1	1



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Time requirement:



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Time requirement:

• Computing intermediate results: O(nm).



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at most $m + n \le 2n$ bits			1	0	0	0	1	0	0	0
_			1	0	1	1	1	0	1	1

Time requirement:

- Computing intermediate results: O(nm).
- Adding *m* numbers of length $\leq 2n$: $\mathcal{O}((m+n)m) = \mathcal{O}(nm)$.



A recursive approach:

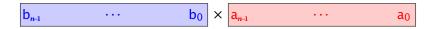


A recursive approach:





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$$\begin{array}{|c|c|c|c|c|c|} \hline B_1 & B_0 & \times & A_1 & A_0 \\ \hline \end{array}$$



A recursive approach:

Suppose that integers **A** and **B** are of length $n = 2^k$, for some k.

$$\begin{array}{|c|c|c|c|c|c|} B_1 & B_0 & \times & A_1 & A_0 \\ \hline \end{array}$$

Then it holds that

$$A = A_1 \cdot 2^{\frac{n}{2}} + A_0$$
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A recursive approach:

Suppose that integers **A** and **B** are of length $n = 2^k$, for some k.

Then it holds that

$$A = A_1 \cdot 2^{\frac{n}{2}} + A_0$$
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Hence,

$$A \cdot B = A_1 B_1 \cdot 2^n + (A_1 B_0 + A_0 B_1) \cdot 2^{\frac{n}{2}} + A_0 B_0$$



 Algorithm 3 mult(A, B)

 1: if |A| = |B| = 1 then

 2: return $a_0 \cdot b_0$

 3: split A into A_0 and A_1

 4: split B into B_0 and B_1

 5: $Z_2 \leftarrow mult(A_1, B_1)$

 6: $Z_1 \leftarrow mult(A_1, B_0) + mult(A_0, B_1)$

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We get the following recurrence:

$$T(n) = 4T\left(\frac{n}{2}\right) + \mathcal{O}(n)$$
.



Master Theorem: Recurrence: $T[n] = aT(\frac{n}{b}) + f(n)$.

- Case 1: $f(n) = O(n^{\log_b a \epsilon})$ $T(n) = O(n^{\log_b a})$
- Case 2: $f(n) = \Theta(n^{\log_b a} \log^k n)$ $T(n) = \Theta(n^{\log_b a} \log^{k+1} n)$
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In our case a = 4, b = 2, and $f(n) = \Theta(n)$. Hence, we are in Case 1, since $n = O(n^{2-\epsilon}) = O(n^{\log_b a - \epsilon})$.



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 \Rightarrow Not better then the "school method".



We can use the following identity to compute Z_1 :

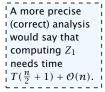
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would say that
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6.2 Master Theorem

We can use the following identity to compute Z_1 :

 $Z_1 = A_1 B_0 + A_0 B_1$

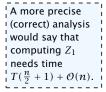




6.2 Master Theorem

We can use the following identity to compute Z_1 :

 $Z_1 = A_1 B_0 + A_0 B_1$ = (A_0 + A_1) \cdot (B_0 + B_1) - A_1 B_1 - A_0 B_0





6.2 Master Theorem

We can use the following identity to compute Z_1 :

$$Z_1 = A_1 B_0 + A_0 B_1 = Z_2 = Z_0$$

= (A_0 + A_1) \cdots (B_0 + B_1) - A_1 B_1 - A_0 B_0

A more precise (correct) analysis would say that computing Z_1 needs time $T(\frac{n}{2} + 1) + O(n)$.



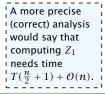
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Hence,	
Hence,	Algorithm 4 mult(A,B)
	1: if $ A = B = 1$ then
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A more precise	5: $Z_2 \leftarrow \operatorname{mult}(A_1, B_1)$
(correct) analysis	6: $Z_0 \leftarrow \operatorname{mult}(A_0, B_0)$
would say that computing Z_1	7: $Z_1 \leftarrow \text{mult}(A_0 + A_1, B_0 + B_1) - Z_2 - Z_0$
needs time $T(n + 1) + O(n)$	8: return $Z_2 \cdot 2^n + Z_1 \cdot 2^{\frac{n}{2}} + Z_0$
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Harald Räcke

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nence,	Algorithm 4 mult(A,B)	
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A more precise	5: $Z_2 \leftarrow \operatorname{mult}(A_1, B_1)$	
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needs time	8: return $Z_2 \cdot 2^n + Z_1 \cdot 2^{\frac{n}{2}} + Z_0$	
$T(\frac{n}{2}+1) + \mathcal{O}(n).$		

Harald Räcke

We can use the following identity to compute Z_1 :

$$Z_1 = A_1 B_0 + A_0 B_1 = Z_2 = Z_0$$

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6.2 Master Theorem

Harald Räcke

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A more precise	5: $Z_2 \leftarrow \operatorname{mult}(A_1, B_1)$	$T(\frac{n}{2})$
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$$T(n) = 3T\left(\frac{n}{2}\right) + \mathcal{O}(n)$$
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Master Theorem: Recurrence: $T[n] = aT(\frac{n}{b}) + f(n)$.

- Case 1: $f(n) = O(n^{\log_b a \epsilon})$ $T(n) = \Theta(n^{\log_b a})$
- Case 2: $f(n) = \Theta(n^{\log_b a} \log^k n)$ $T(n) = \Theta(n^{\log_b a} \log^{k+1} n)$
- Case 3: $f(n) = \Omega(n^{\log_b a + \epsilon})$ $T(n) = \Theta(f(n))$



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Again we are in Case 1. We get a running time of $\Theta(n^{\log_2 3}) \approx \Theta(n^{1.59})$.



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A huge improvement over the "school method".

