

(i) Each iteration of the outermost loop needs  $O(\frac{n}{B})$  I/Os to load each of the  $\frac{n}{B}$  blocks that make up the array  $Y$ . Thus, the total number of I/Os performed by the algorithm is  $O(\frac{n^2}{B})$ , as the outer loop runs for  $n$  iterations.

Alternatively:  
 $X$  is scanned once:  $O(\text{scan}(n)) = O(\frac{n}{B})$  I/Os  
 $Y$  is scanned  $n$  times:  $O(n \cdot \text{scan}(n)) = O(\frac{n^2}{B})$  I/Os  
 total:  $O(\frac{n^2}{B})$  I/Os

(ii) Idea of the cache-aware algorithm: instead of scanning over the entire array  $Y$  for each element of  $X$ , scan over tiles of  $X \times Y$  combinations which fit entirely in memory. More precisely scan over  $\frac{M}{2} - 2(B-1)$  items of  $X$  and  $\frac{M}{2} - 2(B-1)$  items of  $Y$  at the same time, which allows for using  $\frac{2\eta}{B}$  I/Os for analyzing  $\eta^2$   $X \times Y$  combinations, where  $\eta = \frac{M}{2} - 2(B-1)$

FindMin-CacheAware( $X, Y, M, B$ )

1.  $z \leftarrow +\infty$
2.  $\eta \leftarrow \lfloor \frac{M}{2} - 2(B-1) \rfloor$
3. for  $i \leftarrow 0$  to  $\lfloor \frac{n-1}{\eta} \rfloor$
4. for  $j \leftarrow 0$  to  $\lfloor \frac{n-1}{\eta} \rfloor$
5.  $z \leftarrow \min(z, \text{FindMin}(X[i \cdot \eta \dots (i+1) \cdot \eta], Y[j \cdot \eta \dots (j+1) \cdot \eta]))$
6. end for
7. end for
8. return  $z$

For each time line 5 is executed, we note that all elements needed to execute FindMin fit in memory: this means each such execution requires at most  $\frac{2\eta}{B}$  I/Os. The number of times line 5 is executed is  $(\lfloor \frac{n-1}{\eta} \rfloor)^2 \leq (\frac{n-1}{\eta} + 1)^2 = (\frac{n-1}{\eta})^2 + 2 \frac{n-1}{\eta} + 1 = O(\frac{n^2}{\eta^2}) = O(\frac{n^2}{M^2})$

Thus, the total number of I/Os scales as  $O(\frac{n^2}{M^2} \cdot \frac{M}{B}) = O(\frac{n^2}{MB})$  I/Os.

(iii) Idea of the cache-oblivious algorithm: we use a recursive approach, where we continuously split up the problem until it fits into memory.

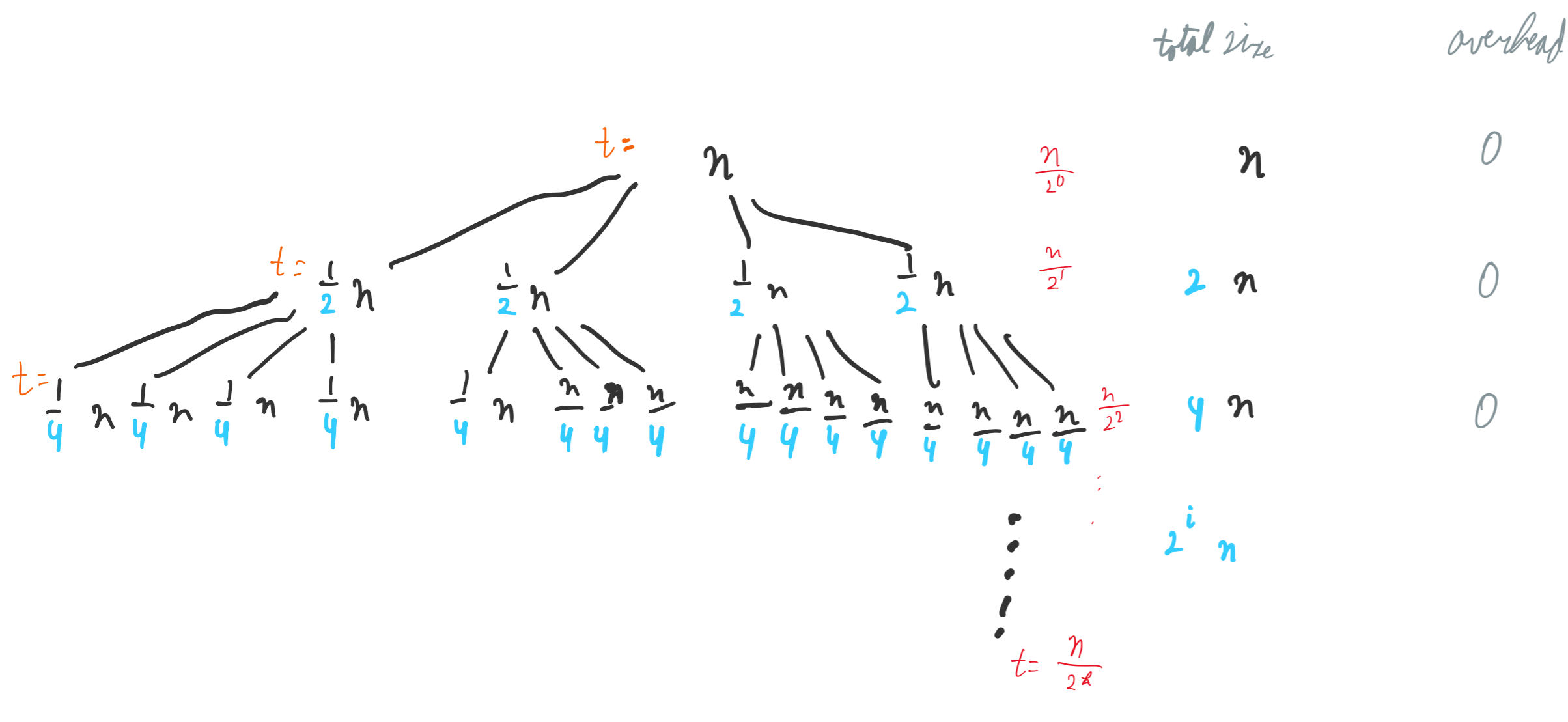
FindMin-CacheOblivious( $X, Y, i_1, i_2, j_1, j_2$ )

1. if  $i_1 = i_2$  or  $j_1 = j_2$  note: we do not know exactly what the problem fits in memory, hence, we need to split it until the problem size becomes 1.
2. return FindMin( $X[i_1 \dots i_2], Y[j_1 \dots j_2]$ )
3. else
4.  $i_{mid} \leftarrow \lfloor \frac{i_1 + i_2}{2} \rfloor$
5.  $j_{mid} \leftarrow \lfloor \frac{j_1 + j_2}{2} \rfloor$
6.  $z_1 \leftarrow \text{FindMin-CacheOblivious}(X, Y, i_1, i_{mid}, j_1, j_{mid})$
7.  $z_2 \leftarrow \text{FindMin-CacheOblivious}(X, Y, i_1, i_{mid}, j_{mid} + 1, j_2)$
8.  $z_3 \leftarrow \text{FindMin-CacheOblivious}(X, Y, i_{mid} + 1, i_2, j_1, j_{mid})$
9.  $z_4 \leftarrow \text{FindMin-CacheOblivious}(X, Y, i_{mid} + 1, i_2, j_{mid} + 1, j_2)$
10. return  $\min(z_1, z_2, z_3, z_4)$
11. end if

The number of I/Os  $T_{IO}(t)$  for running this algorithm with problem size  $t$ , is given by where  $t$  is the size of storing one subarray, we ignore loading issues (we read them off those)

$$T_{IO}(t) = \begin{cases} O(\frac{M}{B}) & \text{if the problem fits entirely into memory} \\ 4 T_{IO}(\frac{t}{2}) & \text{otherwise} \end{cases} = \begin{cases} O(\frac{M}{B}) & \text{if } 2(t+(B-1)) \leq M \\ 4 T_{IO}(\frac{t}{2}) & \text{otherwise} \end{cases}$$

Solving this recurrence for  $t = n$  gives us the following recursion tree



The total number of I/Os for the tree is the

$$T_{IO}(n) = \text{number of base cases} \cdot \text{number of I/Os per base case} + \text{total overhead}$$

$$= \# \text{ base cases} \cdot O(\frac{M}{B})$$

usually:  $\sum_{i=0}^{k-1} \text{overhead on level } i$   
 but here, this is zero, as overhead = 0 on all levels

We solve a simplified recurrence where the condition is replaced by  $2t = M$  at the base cases, we have

base case recurrence:  $2t = M$   
 $t = \frac{M}{2}$   
 using the total cache assumption to ignore blocks sticking out  
 at the base cases, we have that, for the height of the tree  
 $t = \frac{n}{2^k}$

$$= 4^k \cdot O(\frac{M}{B})$$

$$= (2^k)^2 \cdot O(\frac{M}{B})$$

$$= (\frac{2^n}{M})^2 \cdot O(\frac{M}{B})$$

$$= O(\frac{n^2}{M^2}) \cdot O(\frac{M}{B})$$

$$= O(\frac{n^2}{MB})$$