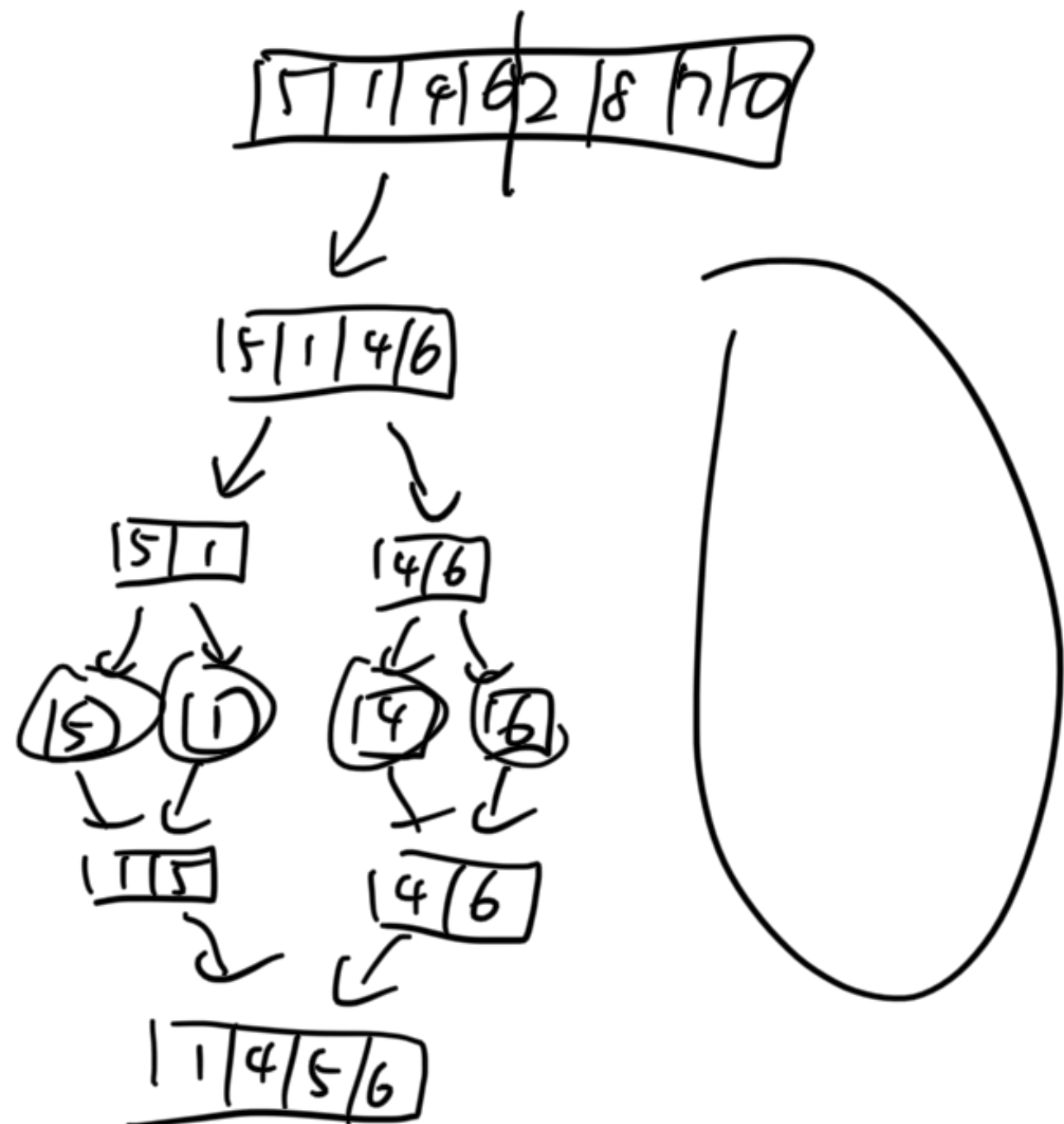
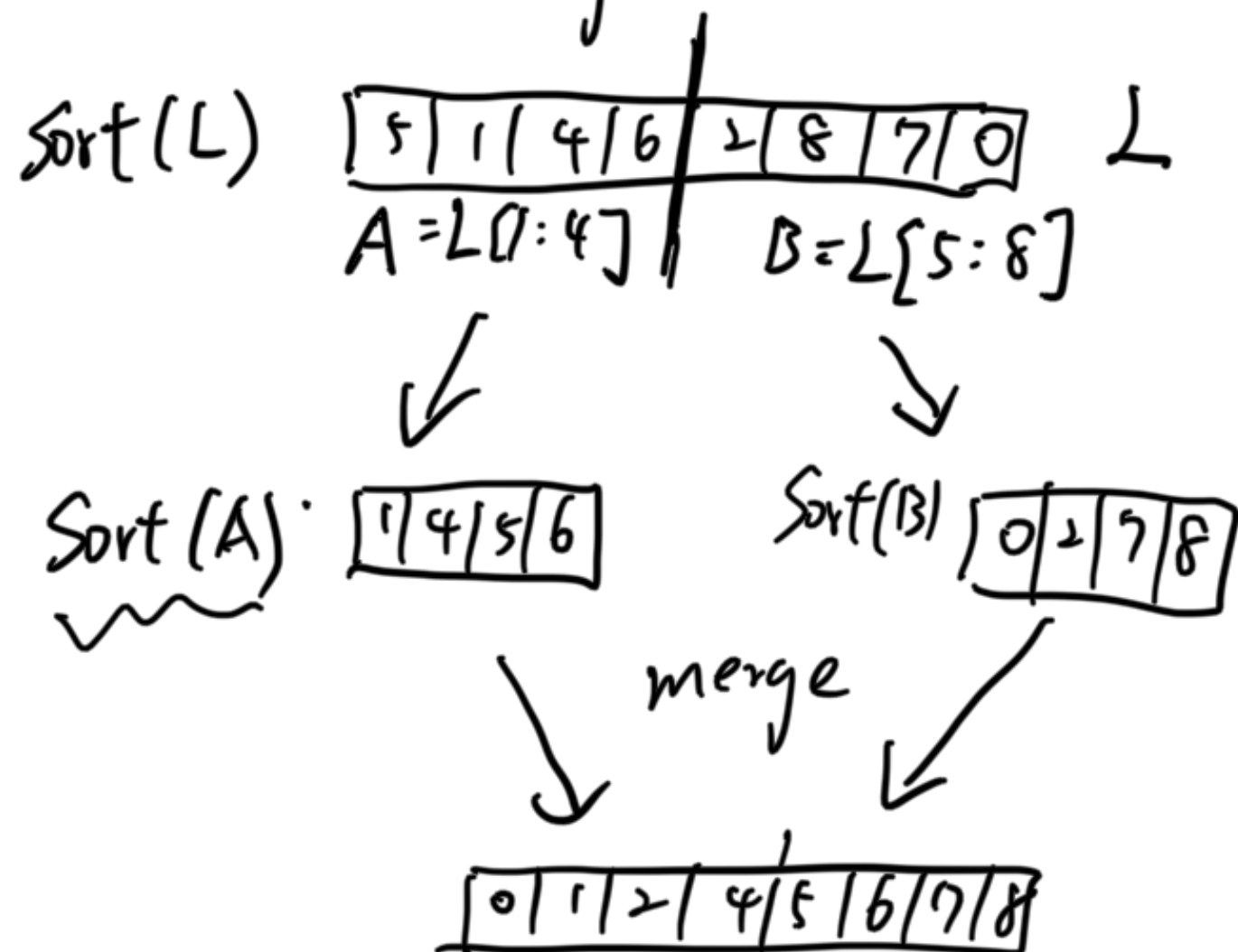


5.1 Divide-and-Conquer.

Merge Sort.



Merge Sort (L)

If L has ≤ 1 element
return L

Else:

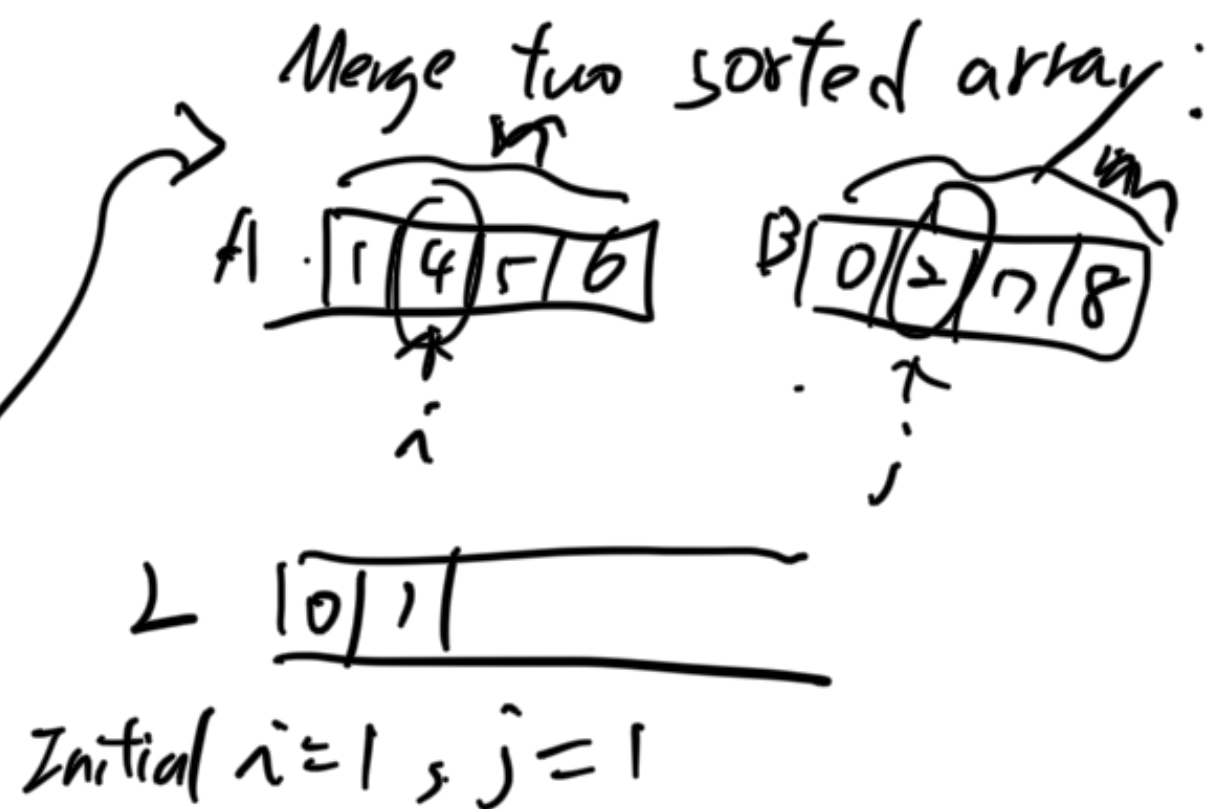
$O(1) \leftarrow A = L[1 : n/2], B = L[n/2 + 1 : n]$

$T(n/2) \leftarrow A \leftarrow \text{Merge Sort}(A)$

$T(n/2) \leftarrow B \leftarrow \text{Merge Sort}(B)$

$O(n) \leftarrow L \leftarrow \text{Merge}(A, B)$

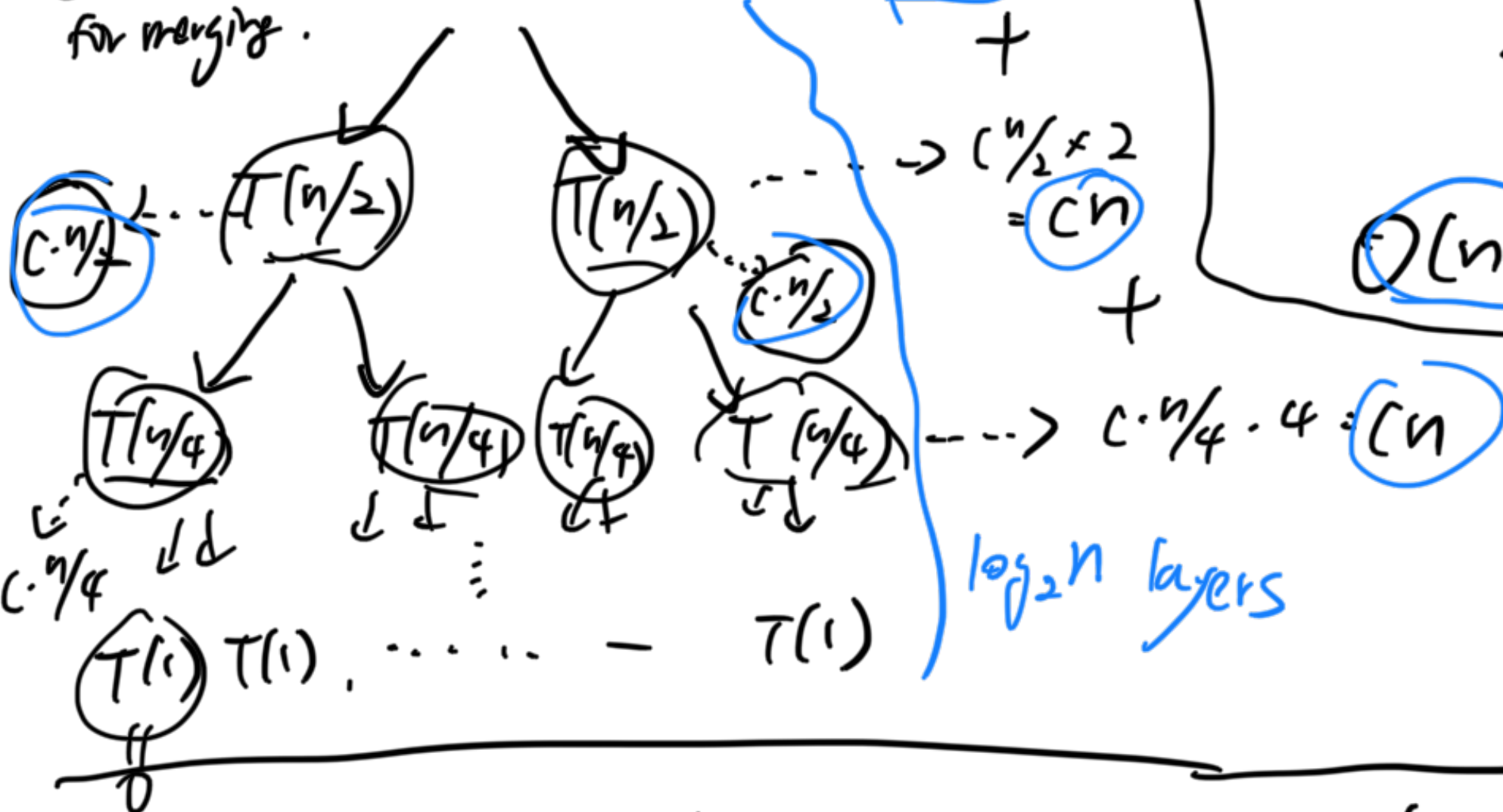
return L



$T(n)$: time for solving problem of size n .

Assume
 $c \cdot n$ time
for merging.

$$T(n) \rightarrow c \cdot n$$



while ($i \leq n$ and $j \leq n$)

if $A[i] \leq B[j]$

add $A[i]$ to L

$i++$.

else add $B[j]$ to L

$j++$.

$O(n)$ time for merging.

$$\text{total time} : \underbrace{\log_2 n}_{\text{layers}} \times \underbrace{c \cdot n}_{\substack{\downarrow \\ \text{total cost} \\ \text{for each layer}}} = c \cdot n \log_2 n = O(n \log n).$$

Induction to prove $T(n) = O(n \log_2 n)$. \nearrow const.

$$T(n) \leq 2 \cdot T(n/2) + c \cdot n, \quad T(1) = 0.$$

$= O(n \log n)$.

by induction: Base case $n=2$ $T(2) = 2 \cdot T(1) + c \cdot 2 = 2c = 2c \cdot \log_2 2$

Induction Hypothesis: Assume $T(m) \leq C m \cdot \log_2 m$ $\forall m < n$

Induction Step: prove this for n

$$\begin{aligned} T(n) &\leq 2 \cdot T(n/2) + Cn \leq 2 \cdot (C \cdot \frac{n}{2} \log_2 \frac{n}{2}) + Cn \\ &= Cn \underbrace{(\log_2 \frac{n}{2})}_{\text{e.g., } n-1} + Cn = Cn \log_2 n - Cn + Cn \\ &= Cn \log_2 n \quad \checkmark \end{aligned}$$

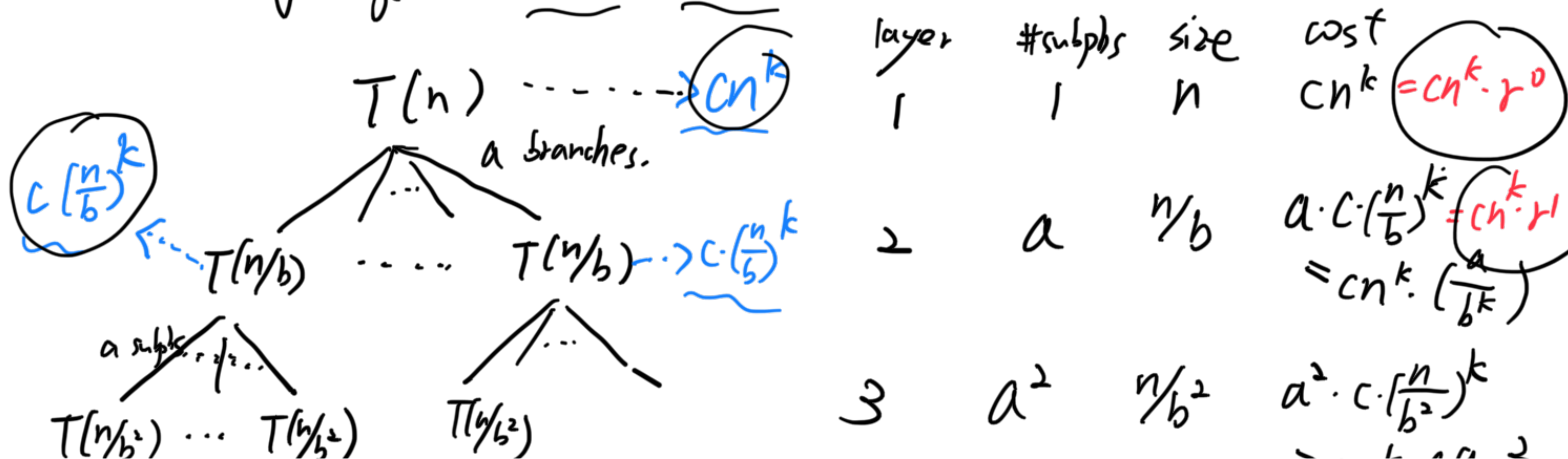
General Case:

subpb: a subpb size: n/b cost for merging Cn^k .

$$T(n) \leq a \cdot T(n/b) + Cn^k$$

$$r = \frac{a}{b^k}$$

e.g. merge sort $a=2$ $b=2$ $k=1$.



⋮
l

$$\begin{aligned}
 &= (n^k \cdot (\frac{a}{b^k})^l) \\
 &= cn^k \cdot r^2 \\
 &cn^k \cdot r^{l-1} +
 \end{aligned}$$

$\approx \log_b n$ layers.

$$\text{total cost} = \underline{cn^k (1 + r + r^2 + \dots + r^{\log_b n - 1})}$$

Case I: $r = 1$, $a/b^k = 1$ ← merge sort.

$$\text{total cost} = cn^k (\underbrace{1 + 1 + 1 + \dots + 1}_{\log_b n}) = C \cdot \log_b n \cdot n^k = O(n^k \log n).$$

Case II: $r > 1$, $a/b^k > 1$

$$\begin{aligned}
 \text{total cost} &= cn^k (1 + r + r^2 + \dots + r^{\log_b n - 1}) \\
 &= cn^k \left(\frac{r^{\log_b n} - 1}{r - 1} \right) \leq cn^k \frac{r^{\log_b n}}{r - 1}
 \end{aligned}$$

$$= O(n^k \cdot r^{\log_b n})$$

$$= O(\cancel{n^k} \cdot \frac{n^{\log_b a}}{\cancel{n^k}})$$

$$\begin{aligned}
 r^{\log_b n} &= \left(\frac{a}{b^k} \right)^{\log_b n} \\
 &= \frac{a^{\log_b n}}{r^{1/k \log_b n}} \rightarrow n^{\log_b a}
 \end{aligned}$$

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

$$(b^{\log_b n})^k = n^k$$

Case II: $r < 1$, $a/b^k < 1$

$$\begin{aligned} \text{total cost} &= cn^k (1 + r + r^2 + \dots + r^{\log_b n - 1}) \\ &= cn^k \left(\frac{1 - r^{\log_b n}}{1 - r} \right) \leq cn^k \frac{1}{1 - r} = O(n^k). \end{aligned}$$

Master Theorem: If $T(n) \leq \underline{a} \cdot T(n/b) + \underline{cn^k}$ then

$$T(n) = \begin{cases} O(n^k \log n) & \text{if } a/b^k = 1 \\ O(n^{\log_b a}) & \text{if } a/b^k > 1 \\ O(n^k) & \text{if } a/b^k < 1. \end{cases}$$