绪论

迭代与递归:分而治之

邓俊辉

deng@tsinghua.edu.cn

凡治众如治寡,分数是也

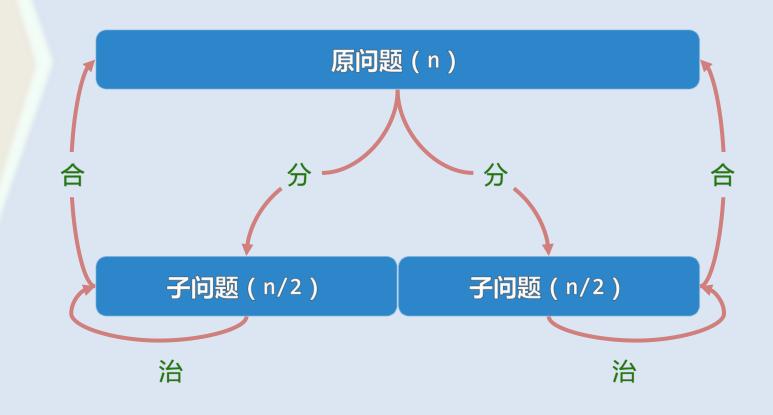
Divide-and-Conquer

- ❖ 为求解一个大规模的问题,可以...
- **❖ 将其划分为若干子问题**

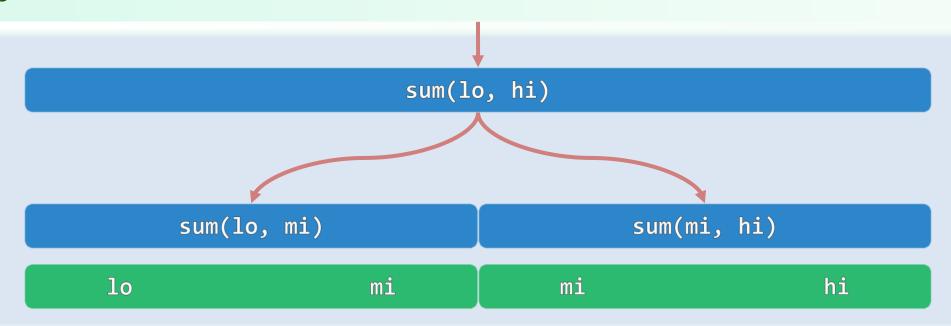
(通常两个,且规模大体相当)

- ❖ 分别求解子问题
- ❖ 由子问题的解

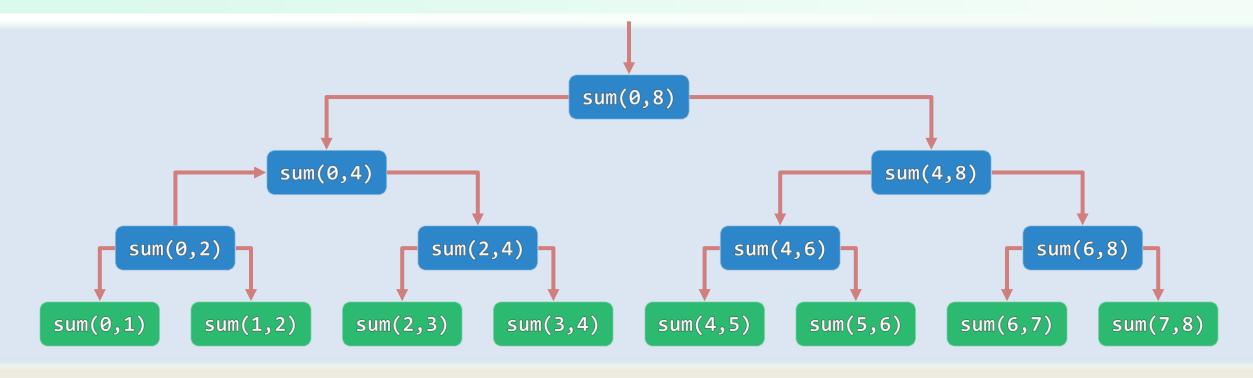
合并得到原问题的解



Binary Recursion



Binary Recursion: Trace



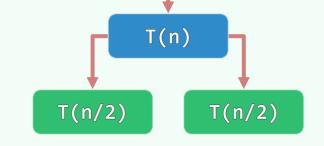
❖ T(n) = 各层递归实例所需时间之和 //递归跟踪

$$= \mathcal{O}(1) \times (2^0 + 2^1 + 2^2 + \dots + 2^{\log n})$$

$$=\mathcal{O}(1)\times(2^{1+\log n}-1)=\mathcal{O}(n)$$
 //更快捷地,作为几何级数...

Binary Recursion: Recurrence

- ❖ 从递推的角度看,为求解sum(A, lo, hi),需
 - 递归求解sum(A, lo, mi)和sum(A, mi+1, hi), 进而 //2*T(n/2)
 - **将子问题的解累加** //**0**(1)
- **᠅递推方程:** $T(n) = 2 \cdot T(n/2) + \mathcal{O}(1)$



$$T(1) = \mathcal{O}(1)$$
 //base: sum(A, k, k)

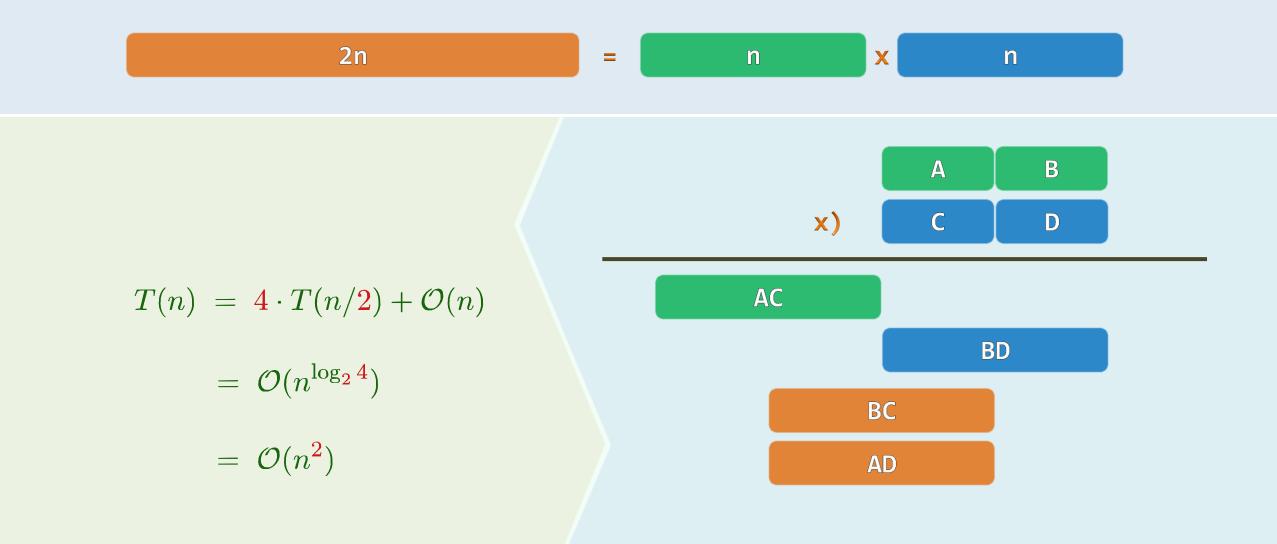
*求解:
$$T(n) = 4 \cdot T(n/4) + \mathcal{O}(3) = 8 \cdot T(n/8) + \mathcal{O}(7) = 16 \cdot T(n/16) + \mathcal{O}(15) = \dots$$

= $n \cdot T(1) + \mathcal{O}(n-1) = \mathcal{O}(2n-1) = O(n)$

Master Theorem

- $\ref{holimits}$ 分治策略对应的递推式,通常(尽管不总是)形如: $T(n) = a \cdot T(n/b) + \mathcal{O}(f(n))$
 - (原问题被分为a个规模均为n/b的子任务;任务的划分、解的合并耗时f(n))
- *若 $f(n) = \mathcal{O}(n^{\log_b a \epsilon})$,则 $T(n) = \Theta(n^{\log_b a})$
 - kd-search: $T(n) = 2 \cdot T(n/4) + \mathcal{O}(1) = \mathcal{O}(\sqrt{n})$
- *若 $f(n) = \Theta(n^{\log_b a} \cdot \log^k n)$,则 $T(n) = \Theta(n^{\log_b a} \cdot \log^{k+1} n)$
 - binary search: $T(n) = 1 \cdot T(n/2) + \mathcal{O}(1) = \mathcal{O}(\log n)$
 - mergesort: $T(n) = 2 \cdot T(n/2) + \mathcal{O}(n) = \mathcal{O}(n \cdot \log n)$
 - STL mergesort: $T(n) = 2 \cdot T(n/2) + \mathcal{O}(n \cdot \log n) = \mathcal{O}(n \cdot \log^2 n)$
- *若 $f(n) = \Omega(n^{\log_b a + \epsilon})$,则 $T(n) = \Theta(f(n))$
 - quickSelect (average case): $T(n) = 1 \cdot T(n/2) + O(n) = O(n)$

Multiplication: Naive + DAC



Multiplication: Optimal

$$T(n) = 3 \cdot T(n/2) + \mathcal{O}(n)$$

$$= \mathcal{O}(n^{\log_2 3})$$

$$\approx \mathcal{O}(n^{1.585})$$

$$B \cdot C + A \cdot D = A \cdot C + B \cdot D - (A - B) \cdot (C - D)$$

$$AD$$

$$(A-B)(C-D)$$