

COMP2054-ADE

Recurrence Relations

Recurrence Relations

- A recurrence relation is a recursively-defined function
 - But, generally, applied to the case when the function is some measure of resources ...
 - and we might only want the big-Oh family properties of the solution
- Suppose that the runtime of a program is $T(n)$, then a recurrence relation will express $T(n)$ in terms of its values at other (smaller) values of n .
 - (By “runtime” is usually meant something ‘abstract’ such as counting of operations. We do not care about difficulties of true timing in nanoseconds.)

Example: Merge Sort

- Suppose the runtime of merge-sort of an array of n integers is $T(n)$. Then

$$T(n) = 2 T(n/2) + b + a n$$

- “ $2 T(n/2)$ ” is due to having to sort the two sub-arrays each of size $n/2$
- “ b ” is the cost of doing the split
- “ $a n$ ” is the cost of doing the merge (and any copying to/from the workspace, etc.)

Example: Merge Sort

- Suppose the runtime of merge-sort of an array of n integers is $T(n)$.
 - Gave the recursive case
 - We also need a base-case.
 - We can take

$$T(1) = 1$$

- As just need to check the array length is 1.
- If make it some other number then we could just rescale the results for $T(n)$ to match. This convention is just for simplicity and convenience.

Example: Merge Sort

- Suppose the runtime of merge-sort of an array of n integers is $T(n)$.
 - Note that we simplified: if n is odd then we ought to have
 - " $T(n/2) + T(n/2+1) + \dots$ "
 - E.g. at $n=9$ $T(9) = T(4) + T(5) + \dots$
 - However, (generally), ignore such details, as they (generally) make no difference to the final statements of big-Oh behaviour.

Example: Merge Sort

- How would we solve

$$T(n) = 2 T(n/2) + b + a n$$

- We will do some special cases:

Example 1:

- How would we solve

$$T(n) = 2 T(n/2) \text{ with } T(1)=1$$

(If watching offline, pause and try)

Example 1:

- How would we solve
$$T(n) = 2 T(n/2) \text{ with } T(1)=1$$
- Given $T(1)=1$, what else can we evaluate?
 - $T(2)$ or $T(1/2)$
 - but want to solve for larger n , not smaller fractions, hence:
 - $T(2) = 2 T(1) = 2$, and then can get
 - $T(4) = 2 T(4/2) = 2 T(2) = 4$
 - $T(8) = 2 T(8/2) = 2 T(4) = 8$
 - Etc.
- It seems a good guess $T(2^k) = 2^k$
- But how do we prove it in general?
 - Induction!

Example 1 (cont):

- How would we solve
$$T(n) = 2 T(n/2) \quad \text{with} \quad T(1)=1$$
- Claim: for all k . $T(2^k) = 2^k$
- Proof by induction:
 - Base case: true at $k=0$. Recall: $2^0 = 1$
 - Step case. Suppose true at k (hypothesis), then need to show is true at $k+1$:
$$\begin{aligned} T(2^{k+1}) &= 2 T(2^{k+1}/2) \quad (\text{using the recurrence}) \\ &= 2 T(2^k) \quad (\text{simplification}) \\ &= 2 \cdot 2^k \quad (\text{using the hypothesis claim}) \\ &= 2^{k+1} \quad \text{QED. As matches the claim at } k+1 \end{aligned}$$

Make sure you understand the structure of this proof!

Example 1 (cont):

- How would we solve
$$T(n) = 2 T(n/2) \quad \text{with} \quad T(1)=1$$
- Showed $T(2^k) = 2^k$ for all k in \mathbb{N} , that is,
$$T(n) = n \quad \text{for all } n \text{ in } \{1,2,4,8,16,\dots\}$$
- What about other values of n ? E.g. what is $T(3)$?
- Depends what one wants!
 - Usually (in this module) just want the growth rate
 - So we can just be imprecise with, $T(n)=n$ for all n , and so then is $\Theta(n)$
 - Might need to refine the recurrence relation, use ceiling and floors to get integers.
 - Messy! But would be the same scaling answer, so do not (usually) bother

Example 2:

- How would we solve

$$T(n) = 2 T(n/2) + b \quad \text{with } T(1)=1$$

(Pause and try)

Example 2:

- How would we solve

$$T(n) = 2 T(n/2) + b \quad \text{with } T(1)=1$$

- We know $T(1)=1$, hence

- $T(2) = 2 T(1) + b = 2 + b$

- $T(4) = 2 T(4/2) + b = 2 (2 + b) + b = 4 + (2+1)b$

- $T(8) = 2 (4 + (2+1)b) + b = 8 + (4+2+1) b$

- It seems a good guess

$$T(2^k) = 2^k + (2^{(k-1)} + \dots + 1)b$$

$$= 2^k + (2^k - 1) b$$

So $T(n) = n + (n-1) b = (1+b)n - b$ for n in $\{1, 2, 4, 8, \dots\}$

Still $\Theta(n)$

Example 2: (cont)

- How would we solve
$$T(n) = 2 T(n/2) + b \quad \text{with } T(1)=1$$
- Claim: $T(2^k) = 2^k + (2^k - 1) b$
- Proof by induction:
 - Base case: $k=0, T(1) = 1 + (1-1)*b = 1$
 - Step case: assume true at k
 - $T(2^{k+1}) = 2 T(2^k) + b$
$$\begin{aligned} &= 2 (2^k + (2^k - 1) b) + b \\ &= 2^{k+1} + (2^{k+1} - 2 + 1) b \\ &= 2^{k+1} + (2^{k+1} - 1) b \end{aligned}$$

QED.

Example 3:

- How would we solve

$$T(n) = 2 T(n/2) + a n \quad \text{with } T(1)=1$$

(Pause and try, even if you saw the answer before!)

Example 3:

- How would we solve

$$T(n) = 2 T(n/2) + a n \quad \text{with } T(1)=1$$

- We know $T(1)=1$, hence try with $n = 2^k$
 - $k=1$: $T(2) = 2 T(1) + 2 a = 2 + 2 a$
 - $k=2$: $T(4) = 2 T(4/2) + 4 a = 2 (2 + 2 a) + 4 a$
 $= 4 + 2 * 4 a$
 - $k=3$: $T(8) = 2 (4 + 8 a) + 8 a = 8 + 3 * 8 a$
 - $k=4$: $T(16) = 2 (8 + 3 * 8 a) + 16 a = 16 + 4 * 16 a$
- It seems a good guess
 $T(2^k) = 2^k + k 2^k a = 2^k (1 + k a)$

So $T(n) = n + \log_2(n) n a$ for $n = \{1,2,4,8...\}$

Now $\Theta(n \log n)$: what we expect of merge-sort !!

Example 3: (cont)

- How would we solve

$$T(n) = 2 T(n/2) + a n \quad \text{with } T(1)=1$$

- Claim: $T(2^k) = 2^k + k 2^k a = 2^k (1 + k a)$
- Base case: $k=0 \quad T(1) = 1 + 0 * 1 * a = 1$
- Step case: assume true at k
 - $$\begin{aligned} T(2^{k+1}) &= 2 T(2^k) + 2^{k+1} a \\ &= 2 (2^k + k 2^k a) + 2^{k+1} a \\ &= 2^{k+1} + k 2^{k+1} a + 2^{k+1} a \\ &= 2^{k+1} + (k+1) 2^{k+1} a \end{aligned}$$

QED

Example 4:

- How would we solve

$$T(n) = 4 T(n/2) \quad \text{with } T(1)=1$$

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- How would we solve

$$T(n) = 4 T(n/2) \quad \text{with } T(1)=1$$

- We know $T(1)=1$, hence

- $k=1$: $T(2) = 4 T(1) = 4 = 2 * 2$
- $k=2$: $T(4) = 4 T(4/2) = 4 * 4 = 16$
- $k=3$: $T(8) = 4 (16) = 64 = 8 * 8$
- $k=4$: $T(16) = 4 (8*8) = (2*8) * (2*8) = 16 * 16$

- It seems a good guess

$$T(2^k) = (2^k)^2$$

So $T(n) = n^2$ for n in $\{1,2,4,8...\}$

Hence $\Theta(n^2)$

Example 4:

- How would we solve
 $T(n) = 4 T(n/2)$ with $T(1)=1$
- Claim $T(2^k) = (2^k)^2$
- Proof by induction:
 - Base case: $k=1$ $T(1) = 1^2 = 1$
 - Step case: assume true at k .
 - $T(2^{k+1}) = 4 T(2^k) = 2 * 2 * 2^k * 2^k = (2^{k+1})^2$ QED.

Exercise (offline): Ensure you understand the details

Example 5:

- How would we solve

$$T(n) = 4 T(n/2) + d n \quad \text{with } T(1)=1$$

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- How would we solve

$$T(n) = 4 T(n/2) + d n \quad \text{with } T(1)=1$$

- We know $T(1)=1$, hence

- $k=1$: $T(2) = 4 T(1) + 2 d = 4 + 2 d = 2^2 + 2 * 1 * d$

- $k=2$: $T(4) = 4 T(4/2) + 4 d = 4 (4 + 2 d) + 4 d$
 $= 16 + 12 d = 4^2 + 4*3*d$

- $k=3$: $T(8) = 4 (16 + 12 d) + 8 d = 8^2 + 8*7 d$

- It seems a good guess that

$$T(n) = n^2 + n(n-1) d$$

Exercise: proof it by induction.

- So $T(n)$ is $\Theta(n^2)$. "Value of d does not matter"

Example 6:

- $T(n) = T(n/2) + d$ $T(1) = 1$
 - E.g. From binary search of a sorted array

Example 6:

- $T(n) = T(n/2) + d$ $T(1) = 1$
 - E.g. From binary search of a sorted array
 - $k=1$: $T(2) = T(1) + d = 1 + d$
 - $k=2$: $T(4) = T(2) + d = 1 + 2d$
 - $k=3$: $T(8) = T(4) + d = 1 + 3d$
 - $k=4$: $T(16) = T(8) + d = 1 + 4d$
 - Guess $T(2^k) = 1 + kd$
 - Hence: $T(n) = 1 + d \log_2(n)$
 - Exercise (offline): prove by induction.
 - That is, $T(n)$ is $\Theta(\log n)$, as expected for binary search

Simple sorting?

- Bubble sort etc. do not naturally generate recurrence relations as they are not naturally recursive.
 - But could be phrased that way
 - Bubble sort
 - $T(n) = T(n-1) + d n$
 - $d n$ for a pass of the outer loop
 - $T(n-1)$ for the remaining passes – which now only need to process $n-1$ numbers.

Example 7:

- $T(n) = T(n - 1) + d n$ $T(1) = 1$
 - (Bubble sort, etc.)
 - $T(2) = T(1) + 2 d = 1 + 2 d$
 - $T(3) = (1 + 2 d) + 3 d = 1 + (2 + 3) d$
 - $T(4) = (1 + (2+3)d) + 4d = 1 + (2+3+4)d$
- Guess $T(n) = 1 + (2 + \dots + n) d$
 $\quad = 1 + (n(n+1)/2 - 1) d$
- Exercise (offline): prove by induction
- Observe it is $\Theta(n^2)$ as expected.

Solving Recurrence

- General pattern

1. Starting from the base case, use the recurrence to work out many cases, by directly substituting and working upwards in values of n
2. Inspect the results, look for a pattern and make a hypothesis for the general results
3. Attempt to prove the hypothesis – typically using some form of induction

Often then extract the large n behavior using big-Oh family

Can be long, tedious, and error-prone, but many cases are covered by a general rule with the name of “Master theorem” (next lecture)

Expectations

- Be able to extract recurrence relations from algorithms
 - Typically, used for recursive algorithms and especially “divide and conquer”
- Be able to explicitly solve (fairly simple) cases
 - Apply the recursion formula for sequence of (small) n
 - Guess pattern
 - Prove using induction
 - (You should generate multiple examples yourself and practice at solving them, and doing the induction proofs)