

CSC3310 Algorithms

Divide and Conquer

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Divide and Conquer

Problem-solving strategy

Divide Break problem into smaller subproblems

- Similar structure to original problem

Conquer Solve the subproblems

- Generally recursively

Combine Combine subproblem solutions into a solution for the original problem

Example: Find Minimum

Name FINDMINIMUMVALUE

Description Given a sequence of n values, return a minimal value from the sequence

Input A sequence of $A = \langle a_0, a_1, \dots, a_{n-1} \rangle$

Output A value a_i such that a_i is in the sequence, and $a_i \leq a_j$ for all $0 \leq j < n$

Example: Find Minimum

Name FINDMINIMUMVALUE

Description Given a sequence of n values, return a minimal value from the sequence

Finding the minimum value with divide and conquer:

Divide split the array into 2 subarrays

Conquer Find the minimum value for each subarray

Combine The smallest value in the array will be the smaller of the two results

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Example: Find Minimum

Finding the minimum value with divide and conquer:

Divide split the array into 2 subarrays

Conquer Find the minimum value for each subarray

Combine The smallest value in the array will be the smaller of the two results

procedure FINDMIN(*a*)

n = *a.length*

if *n* = 1 **then**

return *a[0]*

else

left \leftarrow FINDMIN(*a[0 : n - 1]*)

right \leftarrow *a[n - 1]*

return *left* < *right*?*left* : *right*

end if

end procedure

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 : n – 1])
        right ← a[n – 1]
        return left < right?left : right
    end if
end procedure
```

Correctness

$n = 1$ The only element is the smallest

$n > 1$ Either the smallest element will be one of the first $n - 1$ elements, and FINDMIN will return it, or it will be the last element.

Thus the smaller of *left* and *right* is the smallest element.

Note that we assume FINDMIN is correct in order to prove it is correct. More precisely, we use induction on n .

Example: Find Minimum

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    if n = 1 then
        return a[0]
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        left ← FINDMIN(a[0 : n – 1])
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        return left < right?left : right
    end if
end procedure
```

Time Complexity

$$T(n) = \begin{cases} 3 & \text{if } n = 1 \\ 5 + T(n - 1) & \text{otherwise} \end{cases}$$

This is a *recurrence relation*.

Example: Find Minimum

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Time Complexity

$$T(n) = \begin{cases} 3 & \text{if } n = 1 \\ 5 + T(n - 1) & \text{otherwise} \end{cases}$$

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Substitution

$$T(n) = 5 + T(n - 1)$$

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    else
        left ← FINDMIN(a[0 : n - 1])
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        return left < right?left : right
    end if
end procedure
```

Substitution

$$T(n) = 5 + T(n - 1)$$

$$T(n - 1) = 5 + T(n - 2)$$

Example: Find Minimum

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procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 : n - 1])
        right ← a[n - 1]
        return left < right?left : right
    end if
end procedure
```

Substitution

$$T(n) = 5 + 5 + T(n - 2)$$

$$T(n - 1) = 5 + T(n - 2)$$

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 : n - 1])
        right ← a[n - 1]
        return left < right?left : right
    end if
end procedure
```

Substitution

$$T(n) = 5 + 5 + T(n - 2)$$

$$T(n - 2) = 5 + T(n - 3)$$

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 : n - 1])
        right ← a[n - 1]
        return left < right?left : right
    end if
end procedure
```

Substitution

$$T(n) = 5 + 5 + 5 + T(n - 3)$$

$$T(n - 2) = 5 + T(n - 3)$$

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 : n – 1])
        right ← a[n – 1]
        return left < right?left : right
    end if
end procedure
```

Substitution

$$T(n) = 5 + 5 + 5 + \cdots + 5 + T(n - (n - 1))$$

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 : n – 1])
        right ← a[n – 1]
        return left < right?left : right
    end if
end procedure
```

Substitution

$$T(n) = \overbrace{5 + 5 + 5 + \cdots + 5}^{n-1 \text{ times}} + T(n - (n - 1))$$

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
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    else
        left ← FINDMIN(a[0 : n – 1])
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end procedure
```

Substitution

$$T(n) = \overbrace{5 + 5 + 5 + \cdots + 5}^{n-1 \text{ times}} + T(n - (n - 1))$$
$$T(n - (n - 1)) = T(1) = 3$$

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end procedure
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Substitution

$$T(n) = \overbrace{5 + 5 + 5 + \cdots + 5}^{n-1 \text{ times}} + T(n - (n - 1))$$
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Substitution

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        left ← FINDMIN(a[0 : n – 1])
        right ← a[n – 1]
        return left < right?left : right
    end if
end procedure
```

Substitution

$$\begin{aligned} T(n) &= \overbrace{5 + 5 + 5 + \cdots + 5}^{n-1 \text{ times}} + 3 \\ &= (n - 1)5 + 3 \\ T(n) &= 5n - 2 \end{aligned}$$

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 :  $\frac{n}{2}$ ])
        right ← FINDMIN(a[ $\frac{n}{2}$  : n])
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Example: Find Minimum

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        left ← FINDMIN(a[0 :  $\frac{n}{2}$ ])
        right ← FINDMIN(a[ $\frac{n}{2}$  : n])
        return left < right?left : right
    end if
end procedure
```

Correctness

The smallest element must either be in the first half of the array or the second half. As *left* is the smallest element in the first half, and *right* is the smallest in the second, the smaller of the two must be the smallest overall (We can use *strong* induction to show this formally.)

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 :  $\frac{n}{2}$ ])
        right ← FINDMIN(a[ $\frac{n}{2}$  : n])
        return left < right?left : right
    end if
end procedure
```

Correctness

(We can use *strong induction* to show this formally.)

Hypothesis: FINDMIN returns a minimal element in an array of n elements

Base Case: If $n = 1$, the array contains a single element, which is the smallest.

Inductive Step: Assume FINDMIN returns a minimal element of any array whose length is less than n

By that assumption

$$left \leq a[i] \quad (\forall 0 \leq i < \lfloor \frac{n}{2} \rfloor)$$

$$right \leq a[j] \quad (\forall \lfloor \frac{n}{2} \rfloor \leq j < n)$$

$$\min(left, right) \leq left \leq a[i] \quad (\forall 0 \leq i < \lfloor \frac{n}{2} \rfloor)$$

$$\min(left, right) \leq right \leq a[j] \quad (\forall \lfloor \frac{n}{2} \rfloor \leq j < n)$$

$$\min(left, right) \leq a[k] \quad (\forall 0 \leq k < n)$$

\therefore The hypothesis holds for all positive integers n .

Example: Find Minimum

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    end if
end procedure
```

Time Complexity

$$T(n) = \begin{cases} 3 & \text{if } n = 1 \\ 5 + 2T\left(\frac{n}{2}\right) & \text{otherwise} \end{cases}$$

This is a *recurrence relation*.

Example: Find Minimum

```
procedure FINDMIN(a)
    n = a.length
    if n = 1 then
        return a[0]
    else
        left ← FINDMIN(a[0 : ⌊n/2⌋])
        right ← FINDMIN(a[⌈n/2⌉ : n])
        return left < right?left : right
    end if
end procedure
```

Time Complexity

$$T(n) = \begin{cases} 3 & \text{if } n = 1 \\ 5 + T(\lfloor \frac{n}{2} \rfloor) + T(\lceil \frac{n}{2} \rceil) & \text{otherwise} \end{cases}$$

This is a *recurrence relation*.

Example: Find Minimum

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procedure FINDMIN(a)
    n = a.length
    if n = 1 then
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        left ← FINDMIN(a[0 :  $\frac{n}{2}$ ])
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Substitution

$$T(n) = 5 + 2T\left(\frac{n}{2}\right)$$

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Time Complexity

$$T(n) = \begin{cases} 3 & \text{if } n = 1 \\ 5 + 2T\left(\frac{n}{2}\right) & \text{otherwise} \end{cases}$$

This is a *recurrence relation*.

Substitution

$$\begin{aligned} T(n) &= 5 + 2T\left(\frac{n}{2}\right) \\ &= 5 + 2\left(5 + 2T\left(\frac{n}{4}\right)\right) \\ &= 5 + 2\left(5 + 2\left(5 + 2T\left(\frac{n}{8}\right)\right)\right) \end{aligned}$$

Example: Find Minimum

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        return a[0]
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Substitution

Time Complexity

$$T(n) = \begin{cases} 3 & \text{if } n = 1 \\ 5 + 2T\left(\frac{n}{2}\right) & \text{otherwise} \end{cases}$$

This is a *recurrence relation*.

$$\begin{aligned} T(n) &= 5 + 2T\left(\frac{n}{2}\right) \\ &= 5 + 2\left(5 + 2T\left(\frac{n}{4}\right)\right) \\ &= 5 + 2\left(5 + 2\left(5 + 2T\left(\frac{n}{8}\right)\right)\right) \\ &= \dots \end{aligned}$$

We need a more systematic approach to solving recurrences

Divide-and-Conquer

Divide Break problem into smaller subproblems

Conquer Solve the subproblems

Combine Combine subproblem solutions into a solution for the original problem

If there are a subproblems of size $\frac{n}{b}$, we can represent the time complexity with the recurrence

$$T(n) = D(n) + aT\left(\frac{n}{b}\right) + C(n)$$

$D(n)$ is the time to divide the problem

$C(n)$ is the time to combine the solutions

Solving Recurrence Relations

Convert the recurrence into a “closed-form” solution

- Closed-form is *not* recursive

Example 1 Recursive $T(n) = \begin{cases} 0 & n = 1 \\ T\left(\frac{n}{2}\right) + 1 & \text{otherwise} \end{cases}$

Closed-form $T(n) \in \Theta(\log_2 n)$

Example 2 Recursive $T(n) = \begin{cases} 6 & n = 1 \\ 2T(n - 1) & \text{otherwise} \end{cases}$

Closed-form $T(n) = 6 \times 2^n \in \Theta(2^n)$

Methods for Solving Recurrences

- Repeated substitution
- Guess and check
- Recurrence Trees
- Master method
- Generating Functions

Methods for Solving Recurrences

- Repeated substitution
- **Guess and check**
- Recurrence Trees
- **Master method**
- Generating Functions

In this class, we will only discuss the *guess-and-check* and *master* methods

Guess and Check

Used for upper/lower asymptotic bounds

Approach:

- ① guess the closed form
- ② substitute the guess into the right-hand side
- ③ simplify the result

If the bound still holds, we know the guess was accurate

Upper Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big-O values to find an upper bound

$T(n) \in O(n^2)$ If so, then $T(n) \leq cn^2$, and $T(\frac{n}{2}) \leq c(\frac{n}{2})^2$. Then

$$\begin{aligned} T(n) &\leq 2(c(\frac{n}{2})^2) + 5 \\ &= \frac{c}{2}n^2 + 5 \end{aligned}$$

We then show that this is bounded by cn^2 (for some c, n_0)

$$\frac{c}{2}n^2 + 5 \leq cn^2$$

$$5 \leq \frac{c}{2}n^2$$

$$\frac{10}{c} \leq n^2 \text{ which is true for } c = 10, n_0 = 1$$

Upper Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big-O values to find an upper bound

$$T(n) \in O(n^2)$$

$T(n) \in O(n \log n)$ Show $T(n) \leq cn \log n$ assuming $T(\frac{n}{2}) \leq c\frac{n}{2} \log \frac{n}{2}$.

$$T(n) \leq 2 \left(c \frac{n}{2} \log \frac{n}{2} \right) + 5$$

$$= cn \log \frac{n}{2} + 5$$

$$= cn \log n - cn \log 2 + 5$$

$$cn \log n - cn \log 2 + 5 \stackrel{?}{\leq} cn \log n$$

$$5 \stackrel{?}{\leq} (c \log 2)n, \text{ true for } c = \frac{1}{\log 2}, n_0 = 5$$

Upper Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big-O values to find an upper bound

$$T(n) \in O(n^2)$$

$$T(n) \in O(n \log n)$$

$$T(n) \in O(n) \quad \text{Show } T(n) \leq cn \text{ assuming } T(\frac{n}{2}) \leq c\frac{n}{2}.$$

$$\begin{aligned} T(n) &\leq 2 \left(c \frac{n}{2} \right) + 5 \\ &= cn + 5, \text{ but} \\ &cn + 5 \not\leq cn \end{aligned}$$

That didn't work!

Upper Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big-O values to find an upper bound

$$T(n) \in O(n^2)$$

$$T(n) \in O(n \log n)$$

$$T(n) \in O(n - 5) \quad \text{We show } T(n) \leq c(n - 5) \text{ assuming } T\left(\frac{n}{2}\right) \leq c\left(\frac{n}{2} - 5\right)$$

$$T(n) \leq 2c\left(\frac{n}{2} - 5\right) + 5$$

$$= cn - 10c + 5$$

$$cn - 10c + 5 \stackrel{?}{\leq} c(n - 5) = cn - 5c$$

$$5 \stackrel{?}{\leq} 5c \text{ true for } c = 1, n_0 = 2$$

Upper Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big-O values to find an upper bound

$$T(n) \in O(n^2)$$

$$T(n) \in O(n \log n)$$

$T(n) \in O(n)$ We show $T(n) \leq c(n - 5)$ assuming $T(\frac{n}{2}) \leq c(\frac{n}{2} - 5)$

$$T(n) \leq 2c\left(\frac{n}{2} - 5\right) + 5$$

$$= cn - 10c + 5$$

$$cn - 10c + 5 \stackrel{?}{\leq} c(n - 5) = cn - 5c$$

$$5 \stackrel{?}{\leq} 5c \text{ true for } c = 1, n_0 = 2$$

Upper Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big-O values to find an upper bound

$$T(n) \in O(n^2)$$

$$T(n) \in O(n \log n)$$

$$T(n) \in O(n)$$

$T(n) \in O(\log n)$ If so, we can show $T(n) \leq c \log n$ assuming $T(\frac{n}{2}) \leq c \log \frac{n}{2}$

$$T(n) \leq 2 \left(c \log \frac{n}{2} \right) + 5$$

$$= 2c \log n - 2c \log 2 + 5$$

$$2c \log n - 2c \log 2 + 5 \stackrel{?}{\leq} c \log n$$

$$c \log n \stackrel{?}{\leq} 2c \log 2 - 5 \quad \text{Cannot be true as } n \rightarrow \infty$$

Upper Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big-O values to find an upper bound

$$T(n) \in O(n^2)$$

$$T(n) \in O(n \log n)$$

$$T(n) \in O(n)$$

$T(n) \notin O(\log n)$ If so, we can show $T(n) \leq c \log n$ assuming $T(\frac{n}{2}) \leq c \log \frac{n}{2}$

$$T(n) \leq 2 \left(c \log \frac{n}{2} \right) + 5$$

$$= 2c \log n - 2c \log 2 + 5$$

$$2c \log n - 2c \log 2 + 5 \stackrel{?}{\leq} c \log n$$

$$c \log n \stackrel{?}{\leq} 2c \log 2 - 5 \quad \text{Cannot be true as } n \rightarrow \infty$$

Upper Bound

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$$T(n) \in O(n^2)$$

$$T(n) \in O(n \log n)$$

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$T(n) \notin O(\log n)$ If so, we can show $T(n) \leq c \log n$ assuming $T(\frac{n}{2}) \leq c \log \frac{n}{2}$

$$T(n) \leq 2 \left(c \log \frac{n}{2} \right) + 5$$

$$= 2c \log n - 2c \log 2 + 5$$

$$2c \log n - 2c \log 2 + 5 \stackrel{?}{\leq} c \log n$$

$$c \log n \stackrel{?}{\leq} 2c \log 2 - 5 \quad \text{Cannot be true as } n \rightarrow \infty$$

$O(n)$ is the tightest upper bound we found

Lower Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big- Ω values to find an lower bound Since we know that $T(n) \in O(n)$, we can start at

$T(n) \in \Omega(n)$ We show $cn \leq T(n)$ assuming $c\frac{n}{2} \leq T(\frac{n}{2})$.

$$2(c\frac{n}{2}) + 5 \leq T(n)$$

$$cn + 5 \leq T(n)$$

$$cn \stackrel{?}{\leq} cn + 5 \quad \text{Always true!}$$

Lower Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big- Ω values to find an lower bound

$$T(n) \in \Omega(n)$$

$T(n) \in \Omega(n \log n)$ True if we show $cn \log n \leq T(n)$ assuming $c\frac{n}{2} \log \frac{n}{2} \leq T(\frac{n}{2})$. Then

$$2 \left(c\frac{n}{2} \log \frac{n}{2} \right) + 5 \leq T(n)$$

$$cn \log \frac{n}{2} + 5 \leq T(n)$$

$$cn \log n - cn \log 2 + 5 \leq T(n)$$

$$cn \log n \stackrel{?}{\leq} cn \log n - cn \log 2 + 5$$

$$cn \log 2 \stackrel{?}{\leq} 5 \text{ **false** for large } n$$

Lower Bound

$$T(n) = 2T(n/2) + 5$$

We will guess big- Ω values to find an lower bound

$$T(n) \in \Omega(n)$$

$$T(n) \notin \Omega(n \log n)$$

Guess and-Check Method

- Find upper and/or lower bounds
 - not Θ directly.
 - but if it is $O(n)$ and $\Omega(n)$...
- Guessing Strategy
 - Start with a high guess (e.g. $O(n^3)$) and tighten the bound until you cannot.
 - Be careful! A guess failing does not mean a *proportional* guess cannot succeed.

Master Theorem

Suppose $T(n)$ is a recurrence of the form

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

- If there is a constant $\epsilon > 0$ such that $f(n) \in O(n^{\log_b a - \epsilon})$, then $T(n) \in \Theta(n^{\log_b a})$
- If there is a constant $k \geq 0$ such that $f(n) \in \Theta(n^{\log_b a} \log^k n)$, then $T(n) \in \Theta(n^{\log_b a} \log^{k+1} n)$
- If there are constants $\epsilon > 0$ and $\delta < 1$ such that $f(n) \in \Omega(n^{\log_b a + \epsilon})$ and $af(n/b) \leq \delta f(n)$ for sufficiently large n , then $T(n) \in \Theta(f(n))$

Master Theorem

Suppose $T(n)$ is a recurrence of the form

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

- If there is a constant $\epsilon > 0$ such that $f(n) \in O(n^{\log_b a - \epsilon})$, then $T(n) \in \Theta(n^{\log_b a})$
- If there is a constant $k \geq 0$ such that $f(n) \in \Theta(n^{\log_b a} \log^k n)$, then $T(n) \in \Theta(n^{\log_b a} \log^{k+1} n)$
- If there are constants $\epsilon > 0$ and $\delta < 1$ such that $f(n) \in \Omega(n^{\log_b a + \epsilon})$ and $af(n/b) \leq \delta f(n)$ for sufficiently large n , then $T(n) \in \Theta(f(n))$

Intuition: $aT(\frac{n}{b}) \approx n^{\log_b a}$. The first case is when this dominates $f(n)$; the third is when $f(n)$ dominates it. When they are proportional, we get a logarithmic factor.

Master Theorem

Suppose $T(n)$ is a recurrence of the form

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

- If there is a constant $\epsilon > 0$ such that $f(n) \in O(n^{\log_b a - \epsilon})$, then $T(n) \in \Theta(n^{\log_b a})$
- If there is a constant $k \geq 0$ such that $f(n) \in \Theta(n^{\log_b a} \log^k n)$, then

$$T(n) \in \Theta(n^{\log_b a} \log^{k+1} n)$$

N.B. $k = 0$ is an important special case. If $f(n) \in \Theta(n^{\log_b a})$, then

$$T(n) \in \Theta(n^{\log_b a} \log n)$$

- If there are constants $\epsilon > 0$ and $\delta < 1$ such that $f(n) \in \Omega(n^{\log_b a + \epsilon})$ and $af(n/b) \leq \delta f(n)$ for sufficiently large n , then $T(n) \in \Theta(f(n))$

Intuition: $aT\left(\frac{n}{b}\right) \approx n^{\log_b a}$. The first case is when this dominates $f(n)$; the third is when $f(n)$ dominates it. When they are proportional, we get a logarithmic factor.

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Recall that divide-and-conquer algorithms tend to have the form

$$T(n) = D(n) + aT\left(\frac{n}{b}\right) + C(n)$$

Case 1

Let's solve

$$T(n) = 9T\left(\frac{n}{3}\right) + n$$

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Let's solve

$$T(n) = 9T\left(\frac{n}{3}\right) + n$$

Then,

$$a = 9$$

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Therefore, $T(n) \in \Theta(n^2)$

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Case 2

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Then,

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

- If there is a constant $k \geq 0$ such that $f(n) \in \Theta(n^{\log_b a} \log^k n)$, then $T(n) \in \Theta(n^{\log_b a} \log^{k+1} n)$

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Therefore, $T(n) \in \Theta(\log n)$

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Let's solve

$$T(n) = 3T\left(\frac{n}{4}\right) + n \log n$$

Then,

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Therefore, $T(n) \in \Theta(n \log n)$

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Example

$$T(n) = 2T\left(\frac{n}{2}\right) + 5$$

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Is this Case 1, Case 2, or Case 3?