

Recurrence

Defⁿ - Recurrence eqⁿ recursively defines a sequence of $f(n)$ with different argument.
 - Behaviour of recursive algorithm is better represented using recurrence eqⁿ.

Case 1:
Generic Form
 If $f(n) \in O(n^c)$ where $c > \log_b \frac{a}{b}$
 (using big O notation) then:
 $T(n) \in O(n^{\log_b a})$
Example:
 $a = 8, b = 2, f(n) = 1000n^2$
 $f(n) \in O(n^c)$ where $c = 2$

Case 3:
Generic Form
 If it is True that
 $f(n) \in \omega(n^c)$ where $c > \log_b a$
 and it is also true that
 $a f(\frac{n}{b}) \leq k f(n)$ for some constant $k < 1$
 and sufficiently large n (often called the regularity condition) then:

General format
 $T(n) = a \cdot T(n/b) + f(n)$
 $a \geq 1$
 $b \geq 1$
 $f(n) = +ve$
Cases

1. The Substitution Method
Forward substitution
 - This method finds the solⁿ of the smallest problem using base condition. A solution of the bigger problem is obtained using the previously computed solⁿ of the smaller problem. This process is repeated until the solⁿ for problem n is achieved.

Next, we see if we satisfy case 1 condition
 $\log_b a = \log_2 8 = 3 > c$
 It follows from the 1st case of the master thm
 $T(n) \in O(n^{\log_b a}) = O(n^3)$
 (The exact solⁿ of the recurrence relⁿ is assuming)
 $T(n) = 1001n^3 - 1000n^2$ assuming $T(1) = 1$
Case 2:
Generic Form
 If it is true, for some constant $k \geq 0$
 $f(n) \in \Theta(n^c \log^k n)$ where $c = \log_b a$

$T(n) \in \Theta(f(n))$
Example:
 $T(n) = 2T(\frac{n}{2}) + n^2$
 $a = 2, b = 2, f(n) = n^2$ so
 $f(n) \in \omega(n^c)$ where $c = 2$
 Next, we see if we satisfy case 3 condition
 $\log_b a = \log_2 2 = 1$ and therefore yes
 $c > \log_b a$

1) $f(n) < n^{\log_b a}$
 $T(n) = \Theta(n^{\log_b a})$
 2) $f(n) = n^{\log_b a}$
 $T(n) = \Theta(n^{\log_b a} \log n)$
 3) $f(n) > n^{\log_b a}$
 $T(n) = \Theta(f(n))$

Backward Substitution
 - This method substitutes the value of n by $n-1$ recursively, to solve, smaller and smaller problem. It works exactly in reverse order of forward substitution

then-
 $T(n) \in \Theta(n^c \log^k n)$
Example:
 $T(n) = 2T(\frac{n}{2}) + 10n$
 $a = 2, b = 2, f(n) = 10n$
 $f(n) \in \Theta(n^c \log^k n)$ where $c = 1, k = 0$
 Next, we see if we satisfy case 2 condition
 $\log_b a = \log_2 2 = 1$ and therefore yes, $c = \log_b a$
 It follows from the 2nd case of the master theorem
 $T(n) = \Theta(n^{\log_b a} \log^{k+1} n) = \Theta(n^1 \log^1 n) = \Theta(n \log n)$
 Thus, the given recurrence relation $T(n)$ was in $\Theta(n \log n)$.

The regularity condition also holds
 $2(\frac{n}{2}) \leq kn^2$ choosing $k = 1/2$
 It follows from the 2nd case of the master thm
 $T(n) = \Theta(f(n)) = \Theta(n^2)$
 Thus the given recurrence relation $T(n)$ was in $\Theta(n^2)$ that complies with the $f(n)$ of the original formula
 (This result is confirmed by the exact solⁿ of the recurrence relation which is, assuming $T(1) = 1$)

2. Master Method
 - The master method concerns relations of the form:
 $T(n) = aT(\frac{n}{b}) + f(n)$ where $a \geq 1, b > 1$
 - In the application to the analysis of a recursive algorithm, the constants and function take on the following significance:
 • n is the size of the problem.
 • a is the no of subproblems in the recursion
 • n/b is the size of each subproblem
 • $f(n)$ is the cost of work done outside the recursive calls, which includes the cost of dividing the problem and the cost of merging the solutions to the subproblems

(This result is confirmed by the exact solⁿ of the recurrence relation) \rightarrow
 $T(n) = n + 10n \log_2 2$ assuming $T(1) = 1$