

	operations	cost
$\tilde{v} = n$	$\frac{n}{2^0}$	
$\tilde{v} = \frac{n}{2}$	$\frac{n}{2^1}$	variable classification 2
$\tilde{v} = \frac{n}{4}$	$\frac{n}{2^2}$	assignment $2 + \log_2 n + 1$
\vdots		comparison $\log n + 1$
$\tilde{v} = \frac{n}{2^{k-1}}$		incrementation $\log n + 1$
\vdots		
$\tilde{v} = 1$	\rightarrow end	

$$2^{k-1} = n \Rightarrow k = \log_2 n + 1$$

$$\begin{aligned} \Rightarrow T(n) &= 2 + 2 + \log_2 n + \log_2 n + \log_2 n + 3 \\ &= 3 \log n + 7 \end{aligned}$$

$$T(n) = 3 \log n + 7 \leq 3 \log n + 7 \log n = 10 \log n$$

$$n_0 = 2, C = 10, g(n) = \log n$$

$$\Rightarrow T(n) = O(\log n)$$

$$T(n) = 3 \log n + 7 \geq 3 \log n$$

$$\Rightarrow T(n) = \Omega(\log n)$$

$$\Rightarrow T(n) = \Theta(\log n)$$

b. outer loop: $\tilde{j} = n, \frac{n}{2}, \frac{n}{4}, \dots, \frac{n}{2^{k-1}}, \dots, 1 \rightarrow \log n + 1 \text{ times}$

inner loop: $\tilde{v} = 0, 1, 2, \dots, n-1 \rightarrow n \text{ times}$

operations	cost
variable classification	$2 + \log n + 1$

assignment	$2 + \log n + 1$
comparison	$\log n + 1 + n(\log n + 1)$
incrementation	$2n(\log n + 1)$

$$\begin{aligned}\Rightarrow T(n) &= 3\log n + 7 + 3n(\log n + 1) \\ &= 3n\log n + 3n + 3\log n + 7 \leq 3n\log n + 3n\log n + 3n\log n + 7\log n \\ &\leq 16n\log n \\ \Rightarrow T(n) &= O(n\log n)\end{aligned}$$

$$T(n) \geq 3n\log n \Rightarrow T(n) = \Omega(n\log n)$$

$$\Rightarrow T(n) = \Theta(n\log n)$$

C. outer loop: $n = n, \frac{n}{2}, \frac{n}{4}, \dots \frac{n}{2^{k-1}} \dots 1 \rightarrow \log n + 1 \text{ times}$

inner loop: $i = 0, 1, 2, \dots, n-1 \rightarrow n \text{ times}$

operations	cost
variable declaration	$\log n + 1$
assignment	$2(\log n + 1)$
comparison	$\log n + 1 + n(\log n + 1)$
incrementation	$n(\log n + 1)$
function	$\frac{n}{2} \times n(\log n + 1)$

$$\begin{aligned}\Rightarrow T(n) &= 4\log n + 4 + \frac{1}{2}n^2(\log n + 1) + 2n(\log n + 1) \\ &= \frac{1}{2}n^2\log n + \frac{1}{2}n^2 + 2n\log n + 4\log n + 2n + 4\end{aligned}$$

$$\leq \frac{1}{2}n^2 \log n + \frac{1}{2}n^2 \log n + 2n \log n + 4n \\ = 13n^2 \log n$$

$$\Rightarrow T(n) = O(n^2 \log n)$$

$$T(n) \geq \frac{1}{2}n^2 \log n \Rightarrow T(n) = \Omega(n^2 \log n)$$

$$\Rightarrow T(n) = \Theta(n^2 \log n)$$

d. let C_1 = the runtime of `anotherMethod()`, then

$$\begin{aligned} T(n) &= 2T\left(\frac{n}{2}\right) + C_1 \\ &= 2(2T\left(\frac{n}{4}\right) + C_1) + C_1 = 4T\left(\frac{n}{4}\right) + 3C_1 \\ &= 2(2(2T\left(\frac{n}{8}\right) + C_1)) + C_1 = 8T\left(\frac{n}{8}\right) + 7C_1 \\ &\quad \vdots \\ &= 2^k T\left(\frac{n}{2^k}\right) + (2^k - 1) C_1 \end{aligned}$$

when $\frac{n}{2^k} = 1$, the loop will end, then

$$k = \log n$$

$$\begin{aligned} \Rightarrow T(n) &= 2^{\log n} T(1) + (2^{\log n} - 1) C_1 \\ &= n T(1) + (n - 1) C_1 \quad T(1) = 1 \end{aligned}$$

$$\Rightarrow T(n) = (C_1 + 1)n - C_1$$

$$\Rightarrow T(n) = O(n) = \Omega(n) = \Theta(n)$$

> a. growth rate of $O(n^2)$

① special case: if one of the lists is empty list,
then return (No intersections)

② use two while loop to find the intersections:
if there are n elements in list A and m elements

while (loopA : traverse listA) {

 while (loopB : traverse listB)

 if ($A == B$) {

 return;

}

}

$$\Rightarrow T(n) = n \cdot m \leq n^2 = O(n^2)$$

b. Since we assume the two linked lists don't have
duplicates, then we can use a TreeSet to hold
the elements of one of the lists, then see whether

Set A contains any element from list B

① Create an empty tree set:

Set<ListNode> set1 = new TreeSet<>();

② add the reference of each node in list A to set1

white (~) {

 set1.add(~); // $O(\log n)$

}

the add elements to TreeSet costs $O(fgn)$
then the while loop cost $O(n \log n)$.

- ③ compare the elements from set1 with listB,
if set1.contains(listB.current), then return.
thus part cost $O(n)$ time.

\Rightarrow the whole process costs $O(n \log n)$

c. If we use HashSet rather than TreeSet data structure, the growth rate will drop down to

$O(n)$ since HashSet.add(n) operation
only cost $O(1)$, better than the previous one.

while (loopA : listA) {
set1.add(listA.current); // $O(1)$
}

} $\Rightarrow T(n) = O(n)$

The steps are almost the same as previous one

3. let $f(n)$ represent method1, and $g(n)$ represent method2

$$\begin{aligned}
\text{then } f(4) &= g(3, 3) - (4-1) = g(3, 3) - 3 \\
&= 1 + g(2, 3) - 3 = 1 + 1 + g(1, 3) - 3 \\
&= 1 + 1 + 1 + g(0, 3) - 3 = g(0, 3) \\
g(0, 3) &= f(3) = g(2, 2) - (3-1) = g(2, 2) - 2 \\
&= 1 + g(1, 2) - 2 = 1 + 1 + g(0, 2) - 2 = g(0, 2) \\
g(0, 2) &= f(2) = g(1, 1) - (2-1) = g(1, 1) - 1 \\
&= 1 + g(0, 1) - 1 = g(0, 1) \\
g(0, 1) &= f(1) = g(0, 0) - (1-1) = g(0, 0) \\
g(0, 0) &= f(0) = 0 \\
\Rightarrow f(4) &= g(0, 3) = f(3) = g(0, 2) = f(2) = g(0, 1) \\
&= f(1) = g(0, 0) = f(0) = 0
\end{aligned}$$

\Rightarrow The output will be 0

$$\begin{aligned}
b. \text{ let } T_1(n) \text{ represent the running time function of } f(n) \\
\text{let } T_2(m, k) \text{ represent the running time function of } g(n) \\
f(n) &= g(n-1, n-1) - (n-1) \quad \text{let } C_1 = \text{time to calculate } n-1 \\
\Rightarrow T_1(n) &= T_2(m, k) + C_1, \quad m=n-1, \quad k=n-1 \\
\Rightarrow T_1(n) &= T_2(n-1, n-1) + C_1 \\
T_2(n-1, n-1) &= T_2(n-2, n-1) + C_2 = T_2(n-3, n-1) + 2C_2 \\
&= T_2(n-4, n-1) + 3C_2 \dots = T_2(n-k, n-1) + (k-1)C_2
\end{aligned}$$

when $n-k=0$, it ends $\Rightarrow k=n$

$$\Rightarrow T_2(n-1, n-1) = T_2(0, n-1) + (n-1) C_2 \\ = T_1(n-1) + (n-1) C_2$$

$$\Rightarrow T_1(n) = T_1(n-1) + (n-1) C_2 + C_1 \\ = T_1(n-2) + (n-2) C_2 + (n-1) C_2 + C_1 \\ = T_1(n-3) + (n-3) C_2 + (n-2) C_2 + (n-1) C_2 + C_1 \\ = T_1(0) + \sum_{i=0}^{n-1} i C_2 + C_1 \\ = \frac{n(n-1)}{2} C_2 + C_1 = \frac{1}{2} C_2 n^2 - \frac{1}{2} n C_2 + C_1$$

if $C_1 = C_2 = 1$, then

$$T_1(n) = \frac{1}{2} n^2 - \frac{1}{2} n + 1 = O(n^2)$$

From above, we get $T_2(n-1, n-1) = T_1(n-1) + (n-1) C_2$
use k, m replace $n-1$, then

$$T_2(m, k) = T_1(k) + k C_2 \\ = \frac{1}{2} k^2 - \frac{1}{2} k + 1 + k C_2 \\ = \frac{1}{2} k^2 + (C_2 - \frac{1}{2}) k + 1$$

$$\Rightarrow T_2(m, k) = O(n^2)$$

c. The growth rate is n^2 .