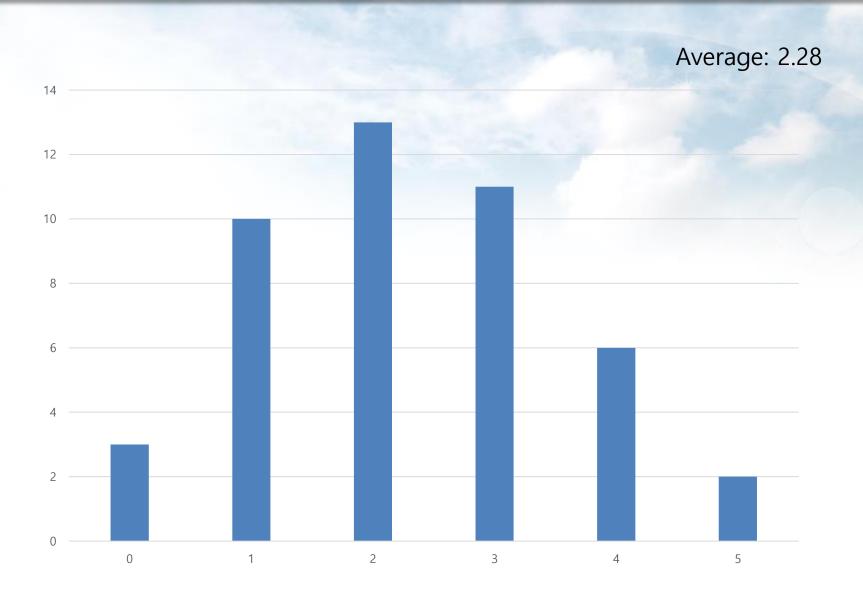


# **Level Test Result**



# **In This Lecture**

### Outline

- Order of Complexity
- 2. List

### Algorithm Analysis

- Execution time measurement
  - Measures actual execution times of two algorithms
  - Requires actual implementation
  - Should use identical hardware
- Complexity analysis
  - (Roughly) Analyze without actual implementation
  - Count the number of operations during algorithm
  - Time complexity
  - Space complexity

#### Measurement

An example code of measuring computing time

```
void main( void )
{
  clock_t start, finish;
  double duration;
  start = clock();
  // algorithm...
  // ...
  finish = clock();
  duration = (double)(finish - start) / CLOCKS_PER_SEC;
  printf("%f seconds.\(\frac{\psi}{\psi}\)n", duration);
}
```

#### **Complexity Analysis**

- How many algorithms can you imagine for solving a problem?
  - Many!
- Among them, what algorithm should we choose?
  - An efficient one!
  - So algorithm analysis is important!
- How to analyze algorithm (from an efficiency perspective)?
  - Complexity analysis!
    - Without actual implementation, roughly we can compare two algorithms
    - Independent for hardware or software environment

#### **Complexity Analysis**

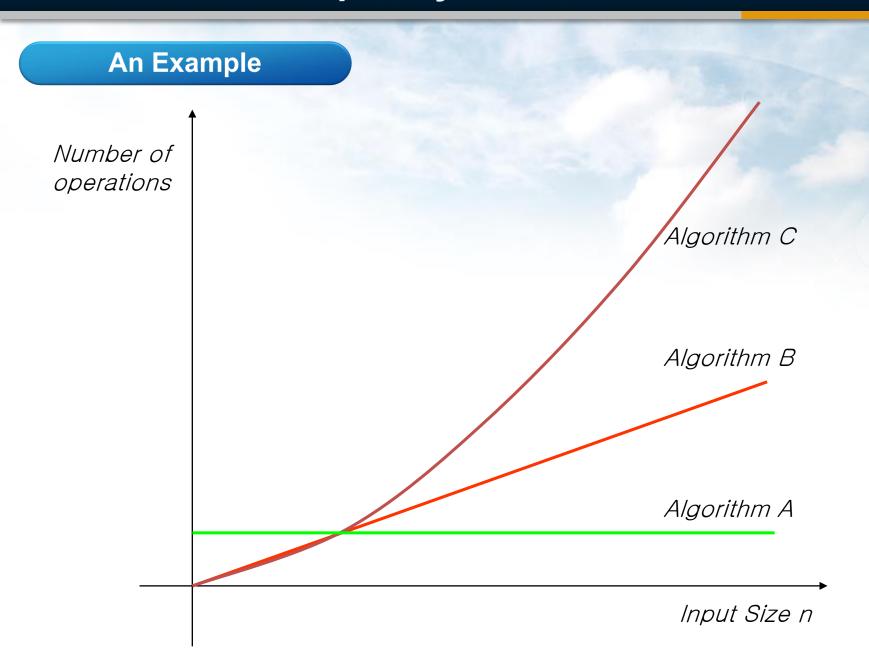
- Computing time complexity
  - Count the number of operations
    - Basic operations: comparison, assignment, arithmetic, etc.
  - Not measure actual execution time!
- Represented by a time complexity function -> T(n)
  - A function of n (input size)
  - Roughly estimate time for running algorithm

#### An Example

- Problem: sum n for n times
  - Let's count the number of operations
  - Let's not consider the for loop control operations

Algorithm A	Algorithm B	Algorithm C
sum ←n*n;	sum ← 0; for i ← 1 to n do sum ←sum + n;	<pre>sum ← 0; for i←1 to n do   for j←1 to n do   sum ←sum + 1;</pre>

	Algorithm A	Algorithm B	Algorithm C
Assignment	1	n + 1	n*n + 1
Addition		n	n*n
Multiplication	1		
Division			
Total	2	2n + 1	2n <sup>2</sup> + 1



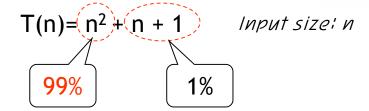
#### **Another Example**

 By analyzing the code, we can roughly calculate the time complexity for the given algorithm

```
\begin{array}{lll} \text{ArrayMax}(A,n) & & & & & \\ \text{tmp} \leftarrow A[0]; & & & & \\ \text{for } i \leftarrow 1 \text{ to n-1 do} & & & \\ & \text{if tmp} < A[i] \text{ then} & & \\ & & \text{tmp} \leftarrow A[i]; & & \\ & \text{return tmp}; & & & \\ & & & \\ & & & \text{total} = 2n \text{ (at most)} \end{array}
```

#### **Big O Notation**

- If n is large, the highest exponent part actually matters, ignoring other parts
  - E.g., n = 1000, T(n) = 1,001,001, first part accounts for about 99%

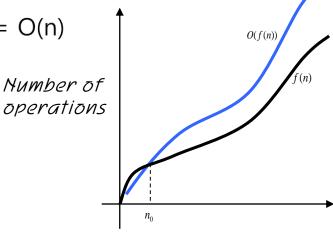


So, typically it is enough to consider the part that most affect

#### **Big O Notation**

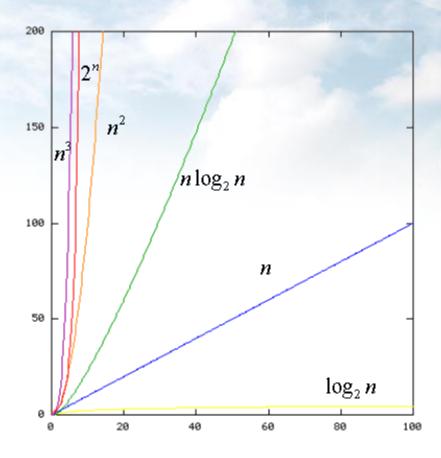
- Big O Definition: (Asymptotic Upper Bound)
  - For given f(n) and g(n),
     for all n≥n<sub>0</sub>, if there exist two constants c and n<sub>0</sub> satisfying |f(n)| ≤ c|g(n)|
     then f(n)=O(g(n))
- Big O represents the upper bound
  - E.g., if  $n \ge 5$ , 2n+1 < 10n -> 2n+1 = O(n)

if  $n_0 = 2$ , c = 2, for  $n \ge 2$ ,  $2n+1 \le 2n^2$  $\rightarrow O(n^2)$ 



#### **Big O Notation**

- O(1): constant
- O(logn): log
- O(n): linear
- O(nlogn) : log-linear
- O(n²) : quadratic
- O(n³) : cubic
- $O(2^n) : exponent$
- O(n!): factorial



### Comparisons

Complexity	n					
Complexity	1	2	4	8	16	32
1	1	1	1	1	1	1
logn	0	1	2	3	4	5
n	1	2	4	8	16	32
nlogn	0	2	8	24	64	160
n²	1	4	16	64	256	1024
n³	1	8	64	512	4096	32768
<b>2</b> <sup>n</sup>	2	4	16	256	65536	4294967296
n!	1	2	24	40326	20922789888000	26313 × 10 <sup>33</sup>

### Comparisons

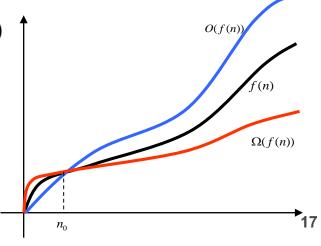
	А	В	С	D	E
	100n	$10nlog_2n$	$5n^2$	$n^3$	$2^n$
10	<b>10<sup>-3</sup></b> sec	<b>1.5*10<sup>-3</sup></b> sec	<b>5*10</b> <sup>-4</sup> sec	<b>10</b> -3 sec	<b>10<sup>-3</sup></b> , sec
100	<b>10<sup>-2</sup></b> sec	<b>0.03</b> sec	<b>5*10<sup>-2</sup></b> sec	<b>1</b> sec	<b>4*10<sup>14</sup></b> cent
1,000	<b>10</b> <sup>-1</sup> sec	<b>0.45</b> sec	<b>5</b> sec	<b>1.6</b> min	***
10,000	1 sec	<b>6.1</b> sec	<b>8.3</b> min	<b>11.57</b> d	***
100,000	<b>10</b> sec	<b>1.5</b> min	<b>13.8</b> hour	<b>31</b> .7 y	***

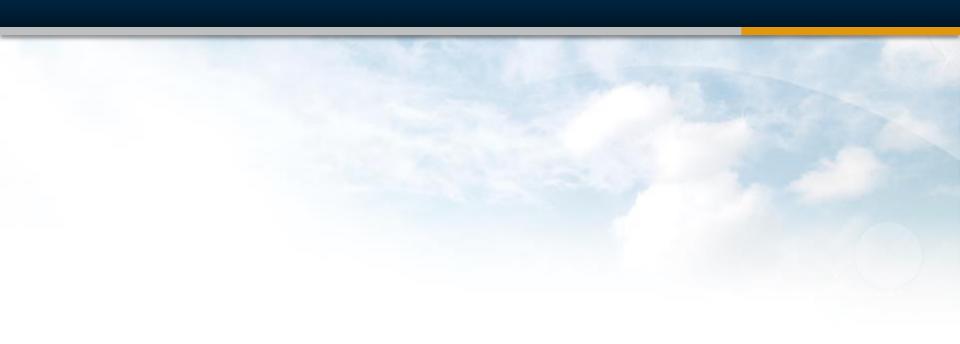
#### Big $\Omega$ Notation

- Big Omega Definition: (Asymptotic Lower Bound)
  - For given f(n) and g(n),
     for all n≥n<sub>0</sub>, if there exist two constants c and n<sub>0</sub> satisfying |f(n)| ≥ c|g(n)|
     then f(n)= Ω(g(n))
- Big Omega represents the lower bound
  - E.g., if  $n \ge 1$ ,  $2n+1 \ge 10n -> 2n+1 = \Omega(n)$

#### **Big** θ **Notation**

- Big Theta Definition: (Asymptotic Tight Bound)
  - For given f(n) and g(n),
     for all n≥n<sub>0</sub>, if there exist three constants c<sub>1</sub>, c<sub>2</sub>, and n<sub>0</sub> satisfying c<sub>1</sub>|g(n)| ≤ |f(n)| ≤ c<sub>2</sub>|g(n)|
     then f(n)= θ(g(n))
- Big Theta represents the lower and upper (tight) bound
  - f(n)=O(g(n)) and  $f(n)=\Omega(g(n)) \rightarrow f(n)=\theta(n)$
  - E.g., if  $n \ge 1$ ,  $n \le 2n+1 \le 3n -> 2n+1 = \theta(n)$

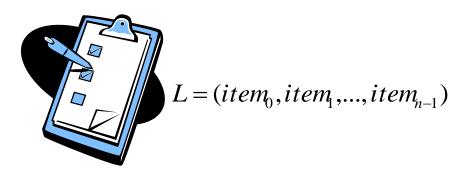




# List

#### **Definition**

- An abstract data type (ADT) that represents a countable number of ordered values, where the same value may occur more than once
- Examples
  - Days (Monday, Tuesday, …)
  - Alphabet (A, B, ...)
  - Card (Ace, 2, 3, ...)
  - Phone numbers



#### **ADT**

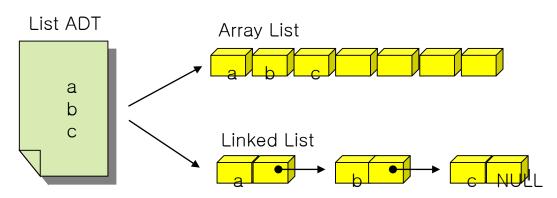
#### ·Object:

A sequence with n values

- ·Operations:
- add\_last(list, item)
- add\_first(list, item)
- add(list, pos, item)
- delete(list, pos)
- clear(list)
- replace(list, pos, item)
- is\_in\_list(list, item)
- get\_entry(list, pos)
- get\_length(list)
- is\_empty(list)
- is\_full(list)
- display(list)

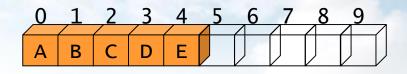
#### **List Implementation**

- Array List
  - Simple
  - Insertion and deletion may not be easy
  - Limited capacity
- Linked List
  - Difficult
  - Efficient in insertion and deletion
  - No limitation in capacity

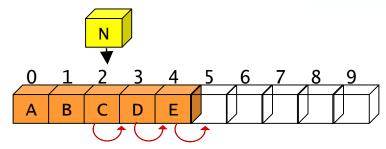


### **Array List**

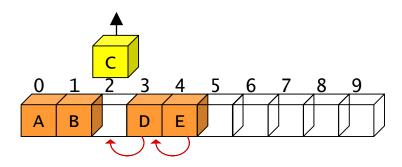
- Store data in one-dimensional array sequentially
  - L = (A, B, C, D, E)



Insertion

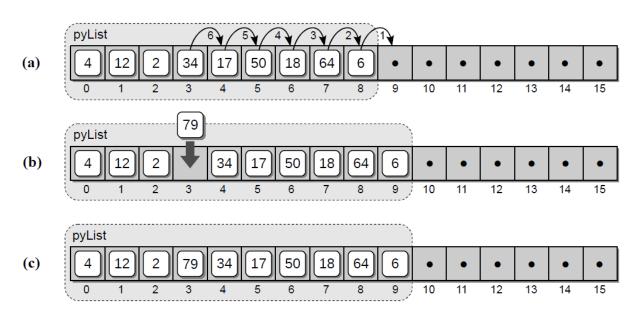


Deletion



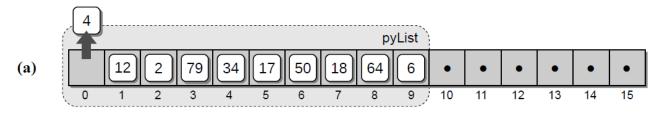
#### **Array List: Insertion**

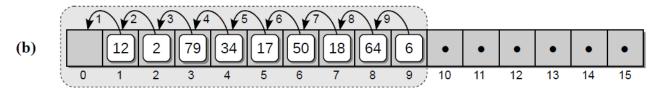
- Inserting an item into a list
  - (a) the array elements are shifted to the right one at a time, traversing from right to left
  - (b) the new value is then inserted into the array at the given position
  - (c) the result after inserting the item

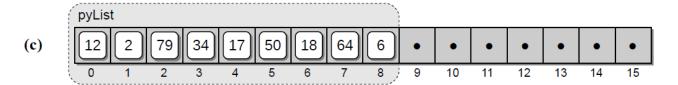


#### **Array List: Deletion**

- Removing an item from a list
  - (a) a copy of the item is saved
  - (b) the array elements are shifted to the left one at a time, traversing left to right
  - (c) the size of the list is decremented by one







#### **Linked Representation**

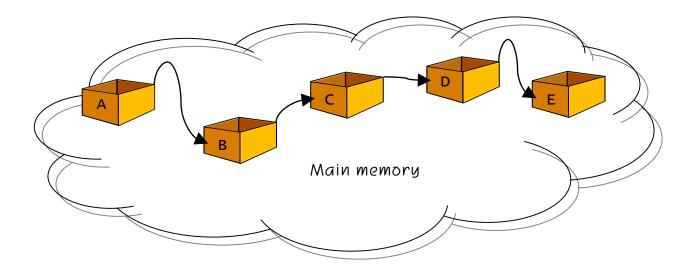
Linked Representation

Node: Data & Link

Data: data value

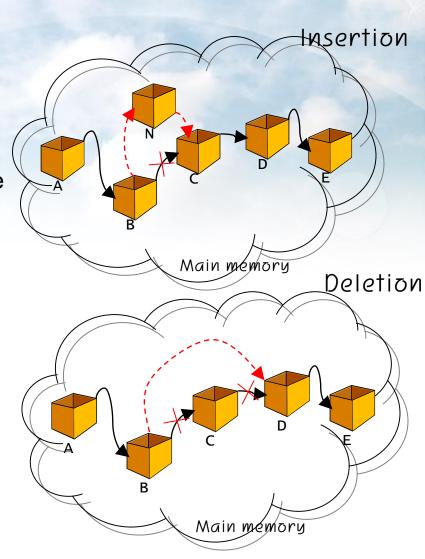
Link: next node

The sequence of link may not be identical to that in physical memory



#### **Linked Representation**

- Pros
  - Insertion/deletion are easy
  - Need not continuous memory space
  - No space limitation
- Cons
  - Difficult to implement
  - Possible errors



#### **Structure**

Node = (data, link)

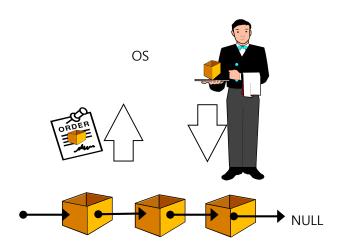
```
class ListNode :
    def __init__( self, data ) :
      self.data = data
      self.next = None
```

"Head" indicates the first node



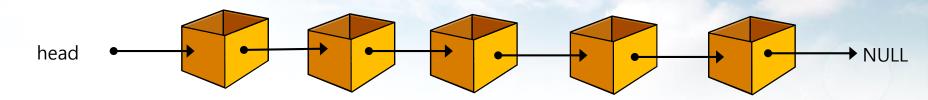
link

- Node creation
  - a = ListNode( 11 ); a.next = b
  - b = ListNode(52); b.next = c
  - c = ListNode( 18 )

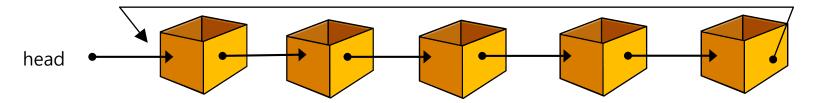


#### **Linked List Types**

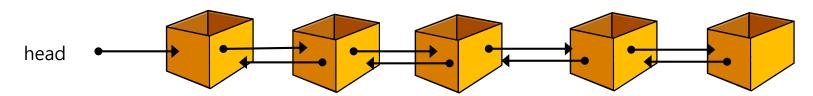
Singly linked list



Circular linked list

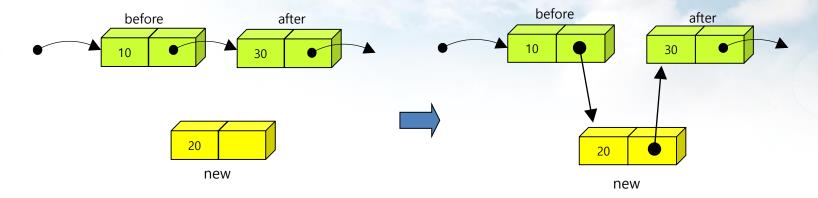


Doubly linked list



#### Insertion

Basic idea

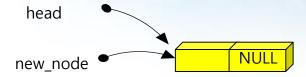


```
insert_node(L, before, new)

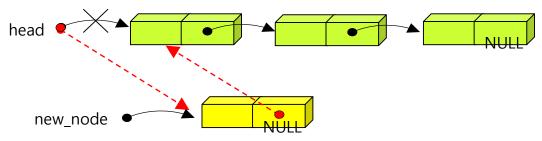
if L = NULL
then L←new
else new.link←before.link
before.link←new
```

#### Insertion

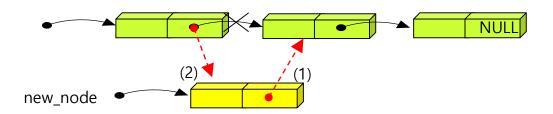
- 3 cases
  - Empty list



Add to first

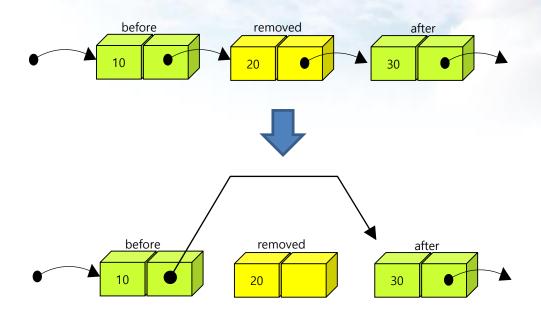


General case



#### **Deletion**

Basic idea



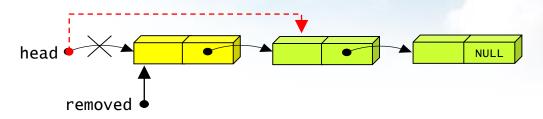
```
remove_node(L, before)

if L ≠ NULL

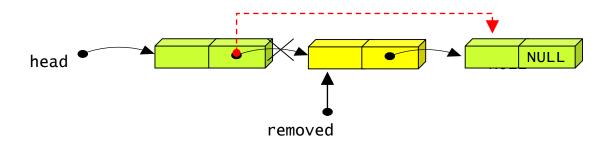
then before.link←removed.link
```

#### **Deletion**

- 2 cases
  - First node

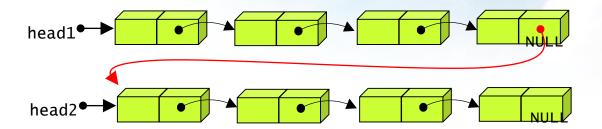


General case



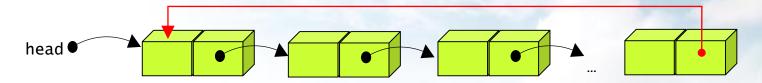
#### **Merge Two Lists**

An example of list operations: merging two lists

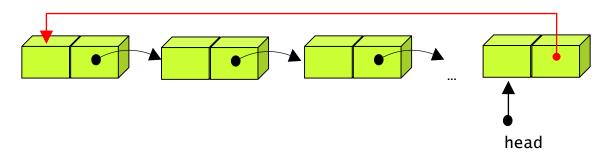


#### **Circular Linked List**

The last node points the first node

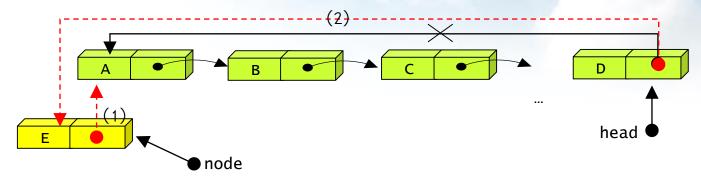


- We can traverse the list starting from any node
- Easier than single linked list in insertion / deletion
- If head points to the last node, "addFirst" and "addLast" can be easily implemented compared to single linked list

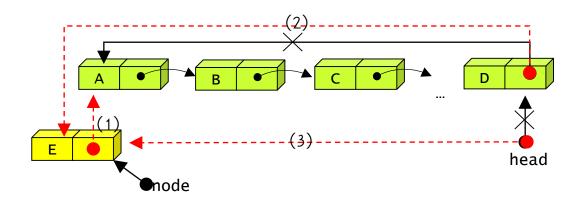


#### Insertion

- Two cases
  - Insertion in the first



Insertion in the last

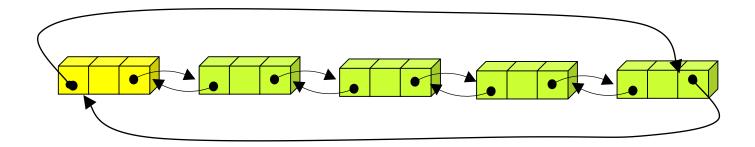


#### **Double Linked List**

- Double linked list
  - Node has two links for previous and next data
  - Link => bidirectional

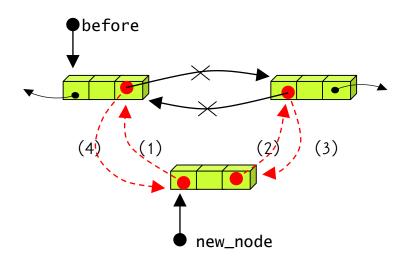


Practically, "double linked list + circular linked list" type is widely used



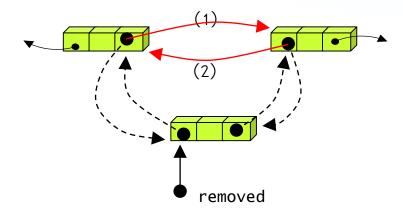
#### Insertion

- new\_node.llink = before; (1)
- new\_node.rlink = before.rlink; (2)
- before.rlink.llink = new\_node; (3)
- before.rlink = new\_node; (4)



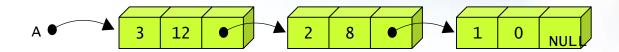
#### **Deletion**

- removed.llink.rlink = removed.rlink (1)
- removed.rlink.llink = removed.llink (2)



#### **Application: Polynomial**

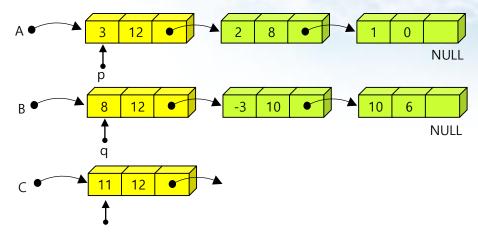
- A polynomial (in one variable) can be expressed as a list
  - $A=3x^{12}+2x^8+1$



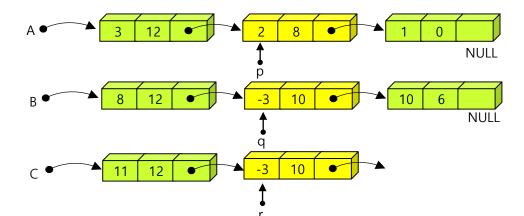
```
class _PolyTermNode( object ):
    def __init__( self, degree, coefficient ):
        self.degree = degree
        self.coefficient = coefficient
        self.next = None
```

### **Polynomial Addition**

① p.expon == q.expon

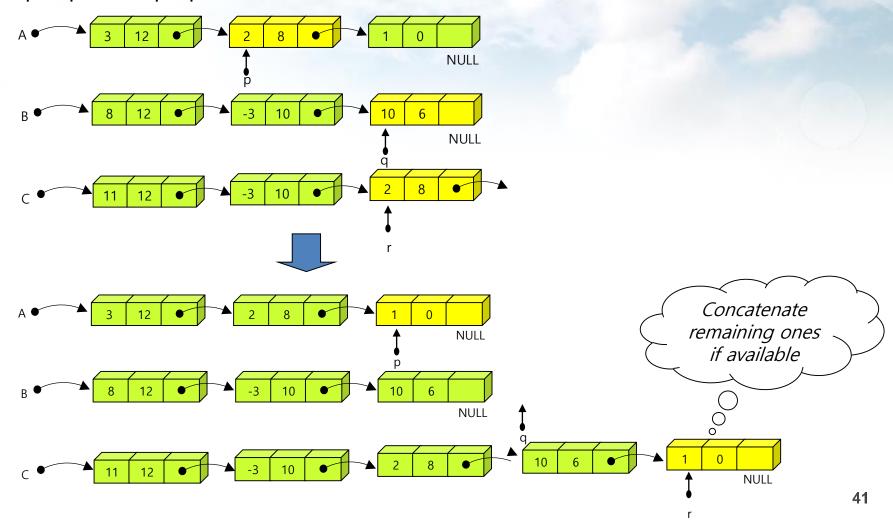


2 p.expon < q.expon



#### **Polynomial Addition**

3 p.expon > q.expon



### What You Need to Know

#### **Summary**

- Algorithm analysis
  - Time complexity
- List
  - Array list
  - Linked list
    - Linked representations
    - Singly linked list
    - Circular linked list
    - Double linked list
  - Application: Polynomials

