

# CSCI 2100 Tutorial 6

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# Outline

- Counting sort again – a linked list version
- Dynamic array vs linked list
- Dynamic array: space and update tradeoff

# Multi-set Sorting Problem (Review)

- Problem input:
  - An array containing  $n$  **key-value pairs**, where each key is an integer from  $[1, U]$ .  
E.g.: (93, 1155123456)
- Goal:
  - An array storing all pairs in **nondescending** order of **key**.

# Multi-set Sorting Problem

- Input:  
 $\{\{9, v_1\}, \{7, v_2\}, \{2, v_3\}, \{6, v_4\}, \{2, v_5\}, \{7, v_6\}, \{1, v_7\}, \{2, v_8\}\}$
- Initially we will have the following array

Input Array

$k_1$	$v_1$	$k_2$	$v_2$	$k_3$	$v_3$	$k_4$	$v_4$	$k_5$	$v_5$	$k_6$	$v_6$	$k_7$	$v_7$	$k_8$	$v_8$
9	$v_1$	7	$v_2$	2	$v_3$	6	$v_4$	2	$v_5$	7	$v_6$	1	$v_7$	2	$v_8$

- Rearrange the elements so that their **keys are sorted**:

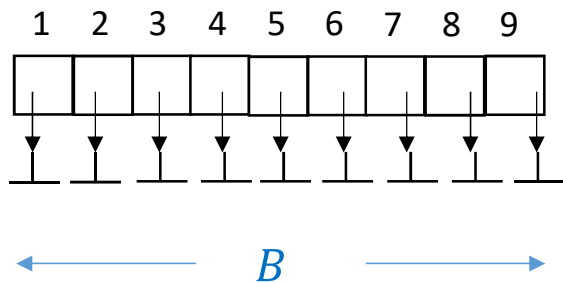
Sorted Array

1	$v_7$	2	$v_3$	2	$v_5$	2	$v_8$	6	$v_4$	7	$v_2$	7	$v_6$	9	$v_1$
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# Multi-set Sorting Problem

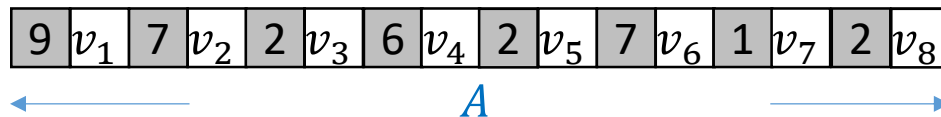
Today we will learn a simple variant of counting sort based on linked lists. The new algorithm also achieves the time complexity  $O(n + U)$ .

# Counting Sort (Linked List Ver.)

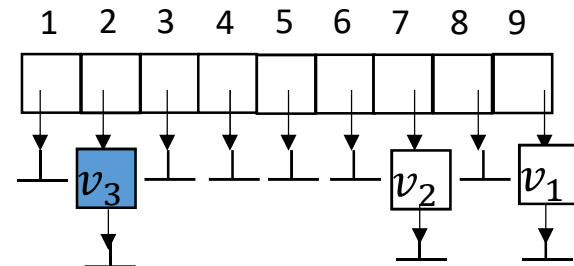
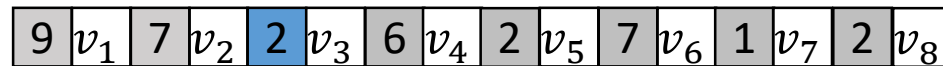
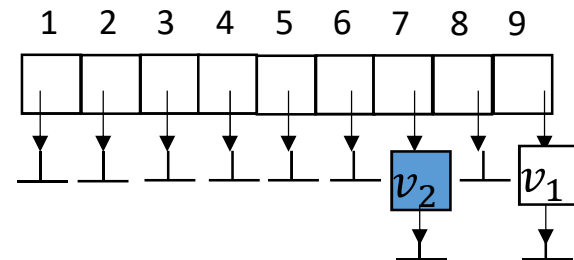
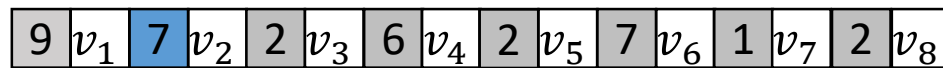
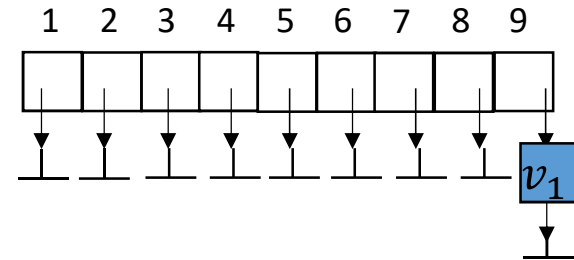
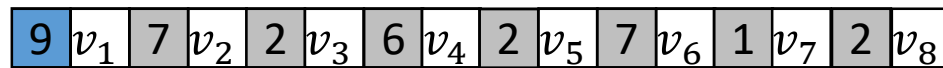


⊥: This means a null pointer

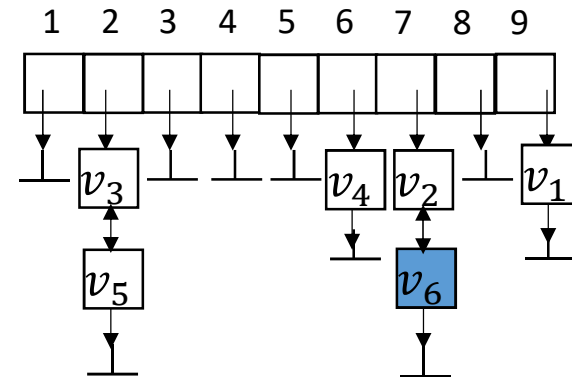
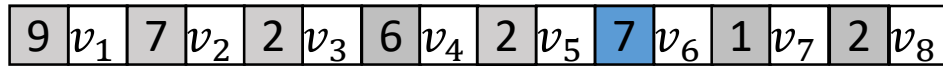
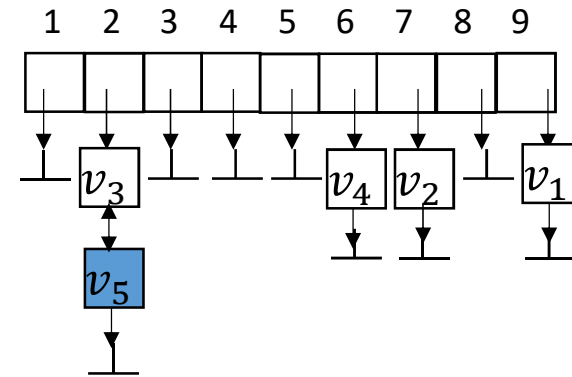
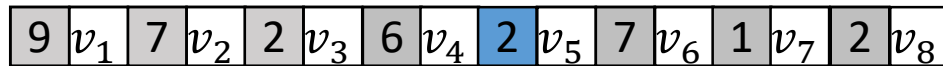
Compute  $B$



# Counting Sort (Linked List Ver.)

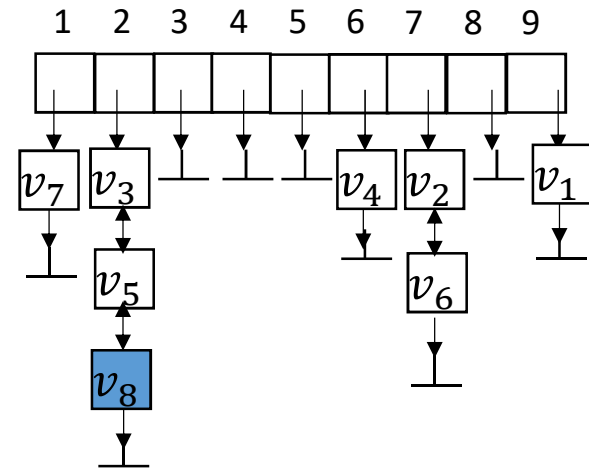
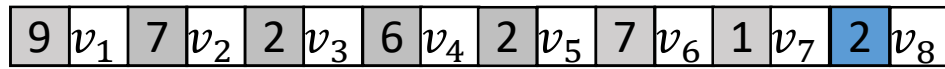


# Counting Sort (Linked List Ver.)

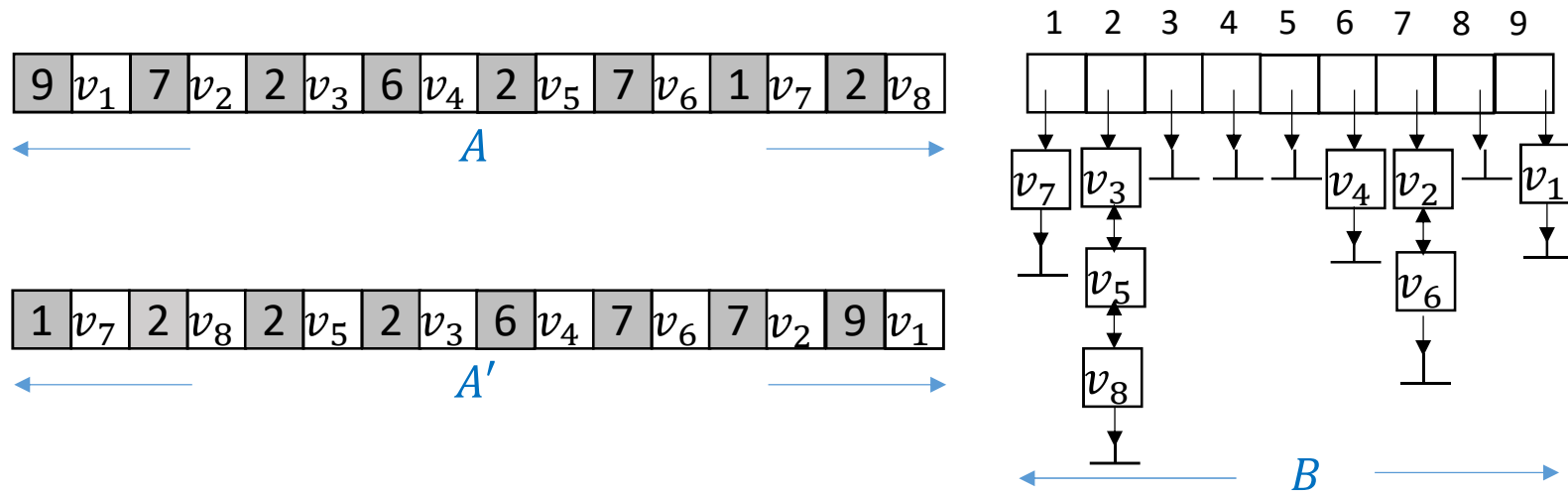




# Counting Sort (Linked List Ver.)



# Counting Sort (Linked List Ver.)



How do we produce the sorted array  $A'$ ?

Scan array  $B$ . For each cell pointing to a non-empty linked list, enumerate all the pairs therein.

Overall time complexity:  $O(n + U)$

# Dynamic Array vs Linked List

A linked list ensures  $O(1)$  insertion cost. A dynamic array guarantees  $O(1)$  insertion cost only after amortization.

However, a dynamic array provides constant-time access to any element, which a linked list cannot achieve.

# Dynamic Array vs Linked List

Question:

Design a data structure of  $O(n)$  space to store a set  $S$  of  $n$  integers to satisfy the following requirements:

- An integer can be inserted in  $O(1)$  time.
- We can enumerate all integers in  $O(n)$  time.

Answer: Linked list.

# Dynamic Array vs Linked List

Question:

Design a data structure of  $O(n)$  space to store a set  $S$  of  $n$  integers to satisfy the following requirements:

- An integer can be inserted in  $O(1)$  amortized time.
- We can enumerate all integers in  $O(n)$  time.
- For each  $i \in [1, n]$ , access  $i$ -th inserted integer in  $O(1)$  time.

Answer: Dynamic array

# Space-Update Tradeoff of the Dynamic Array

In the lecture, we expand the array from size  $n$  to  $2n$  when it is full.

What if we expand the array size to  $\lceil 1.5n \rceil$ ?

# Space-Update Tradeoff of the Dynamic Array

- Initially, size 2 (define  $s_1 = 2$ )
- 1<sup>st</sup> expansion: size from  $s_1$  to  $s_2 = \lceil 1.5s_1 \rceil = 3$ .
- 2<sup>nd</sup> expansion: from  $s_2$  to  $s_3 = \lceil 1.5s_2 \rceil = 5$ .
- ...
- $i$ -th expansion: from  $s_i$  to  $s_{i+1} = \lceil 1.5s_i \rceil$ .

We can prove:  $s_i \leq \left(\frac{8}{3}\right) 1.5^i - 2 = O(1.5^i)$  and  $s_i \geq 1.5^i$ .

# Space-Update Tradeoff of the Dynamic Array

- The total cost of  $n$  insertions is bounded by:

$$\left( \sum_{i=1}^n O(1) \right) + \sum_{i=1}^h O(1.5^{i+1}) = O(n + 1.5^{h+1})$$

where  $h$  is the number of expansions.

It must hold that  $n > s_h \geq 1.5^h$  (the  $h$ -th expansion happened because the array of size  $s_h$  was full).

Hence, the total cost is  $O(n)$ .



# Space-Update Tradeoff of the Dynamic Array

- Consider what happens in general. When the array is full, expand its size from  $n$  to  $\alpha n$ , for some constant  $1 < \alpha \leq 2$ .

# Space-Update Tradeoff of the Dynamic Array

- Initially, size 2 (define  $s_1 = 2$ )
- 1<sup>st</sup> expansion: size from  $s_1$  to  $s_2 = \lceil \alpha s_1 \rceil$ .
- 2<sup>nd</sup> expansion: from  $s_2$  to  $s_3 = \lceil \alpha s_2 \rceil$ .
- ...
- $i$ -th expansion: from  $s_i$  to  $s_{i+1} = \lceil \alpha s_i \rceil$ ..

We can prove:  $s_i = O\left(\frac{\alpha^i}{\alpha-1}\right)$  and  $s_i \geq \alpha^i$ .

# Space-Update Tradeoff of the Dynamic Array

The total cost of  $n$  insertions is bounded by:

$$\left( \sum_{i=1}^n O(1) \right) + \sum_{i=1}^h O\left(\frac{\alpha^{i+1}}{\alpha-1}\right) = O\left(n + \frac{\alpha^{h+2}}{(\alpha-1)^2}\right)$$

where  $h$  is the number of expansions.

It must hold that  $n > s_h \geq \alpha^h$  (the  $h$ -th expansion happened because the array of size  $s_h$  was full).

Hence, the total cost is  $O\left(n + \frac{\alpha^2}{(\alpha-1)^2} n\right)$ , namely, amortized cost =  $O\left(1 + \frac{\alpha^2}{(\alpha-1)^2}\right)$ .

# Space-Update Tradeoff of the Dynamic Array

$$\text{Amortized cost} = O\left(1 + \frac{\alpha^2}{(\alpha-1)^2}\right).$$

When  $\alpha$  decreases, the space consumption goes down, but the insertion cost goes up.