

Design Paradigm: Divide and Conquer

- Finding Rank - Merge Sort
- Counting Inversions
- Karatsuba Algorithm for Integers Multiplication
- Finding Closest Pair in Plane

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Inversions

An **inversion** in an array A of numbers is **an out-of-order pair**

A pair of indices (i, j) such that $i < j$ and $A[i] > A[j]$

$A =$	1	2	3	4	5	6	7	8
	51	43	64	91	22	75	51	86

Inversions: $\{(1, 2), (1, 5), (2, 5), (3, 5), (3, 7), (4, 5), (4, 6), (4, 7), (4, 8), (6, 7)\}$

Inversions

An **inversion** in an array A of numbers, is an out-of-order pair

A pair of indices (i, j) such that $i < j$ and $A[i] > A[j]$

Number of inversions is a measure of (dis)sorted-ness of array

- ▷ What is the maximum possible number of inversions in an array?
- ▷ What is the minimum possible number of inversions in an array?
- ▷ When is the $\min(A)$ and $\max(A)$ involved in an inversion?

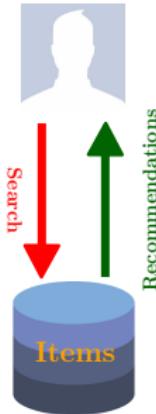
An array is sorted if there are zero inversions

Recall which sorting algorithm is better when A has few inversions?

Applications:

- Collaborative filtering
- Ranked choice voting

Recommendation Systems



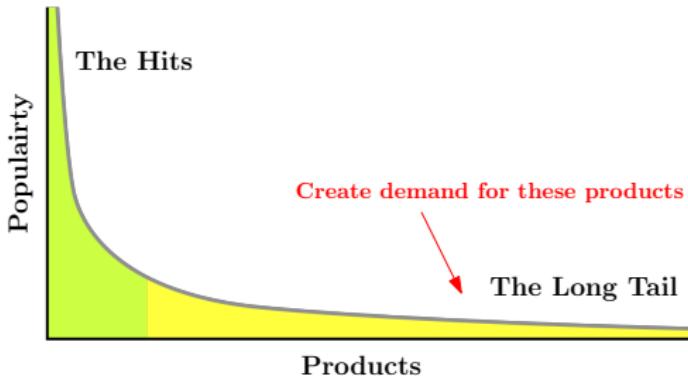
The Web, they say, is leaving the era of search and entering one of discovery. What's the difference? **Search** is what you do when you're looking for something. **Discovery** is when something wonderful that you didn't know existed, or didn't know how to ask for, finds you.

J. O'Brien, Nov 20, 2006 The race to create a 'smart' Google

Recommendation Systems

Retailers cannot shelve everything

- ▷ Online retailers and digital content providers have millions of products



Near zero-cost dissemination of information about products

Necessitates information filtering (customization and recommendation)

Filtering can be

- Hand-Curated: ▷ Chef's specials, editor's picks, favorites
- Simple aggregates: ▷ Top 10, Trending, Recent uploads
- Customized to individual users: ▷ **Recommendation Systems**

Recommendation Systems: Problem Formulation

- n users - $\{u_1, \dots, u_n\}$ and m items - $\{p_1, \dots, p_m\}$
- Utility Matrix U : $n \times m$ matrix row/column for each user/item
- $U(i, j)$: rating of user i for item j

	p_1	p_2	p_3	p_j				p_m			
u_1	1		2	1	4		2	3	2	5	
u_2		1				2	1		2		1
u_3	1	1	2			1				1	2
			3	2		5		2		3	4
u_i	1			2					5		
	3	2	1	4	5	?	1	3	1	2	1
	4								4		
	5			1					5		
	1		4				1	3	5	1	2
u_n		3		1	1	2	1		4		5

$U(i, j)$ could be

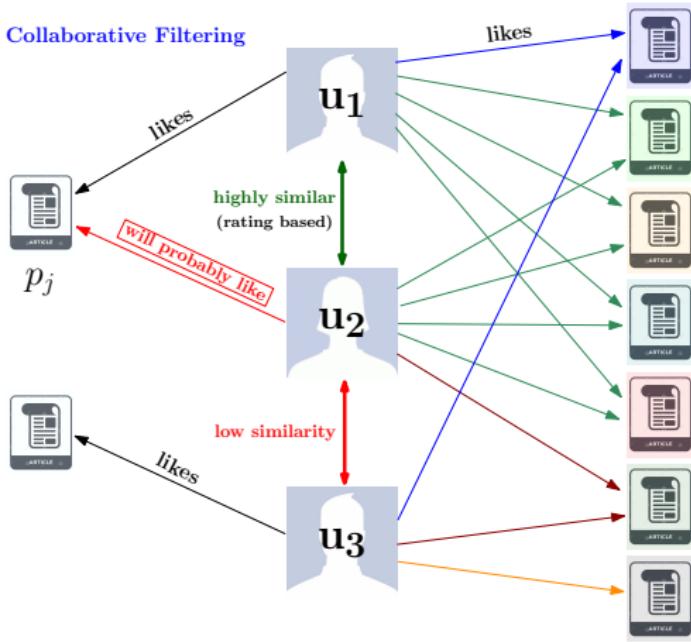
- 0 – 5 stars
- $\in [0, 1]$
- $\in \{0, 1\}$

If prediction for $U(i, j)$ is high, then recommend product j to user i

User-User Collaborative Filtering

Collaboratively filter (personalize) ratings using only the rating matrix U

User i will 'like' item j , if other users similar to i highly rate item j



Inversions: Application in Collaborative Filtering

A pair of indices (i, j) such that $i < j$ and $A[i] > A[j]$

Number of inversions is a measure of distance/similarity between two users

- Sort row of u_x by ratings of u_y
- Inversions in u_x row is distance between u_x and u_y

p_1	p_2	p_3	p_4	p_5	p_6	p_7
2	8	6	1	9	7	3

p_1	p_2	p_3	p_4	p_5	p_6	p_7
7	4	1	9	2	1	8

p_4	p_1	p_7	p_3	p_6	p_2	p_5
9	7	8	1	1	4	2

p_1	p_2	p_3	p_4	p_5	p_6	p_7
2	7	6	1	9	8	2

p_4	p_1	p_7	p_3	p_6	p_2	p_5
1	2	2	6	8	7	9

Is u_2 closer to u_1 or u_3 ?

Does u'_2 have more inversions or u'_3 ?

Counting Inversions: Algorithm

A pair of indices (i, j) such that $i < j$ and $A[i] > A[j]$

Input: An array A of n numbers

Output: Number of inversions in A

Algorithm Counting Inversions - Brute force algorithm

```
count ← 0
for i = 1 to n do
    for j = i + 1 to n do
        if A[i] > A[j] then
            count ← count + 1
```

- Correct by definition
- $\binom{n}{2}$ index pairs, number of comparisons is $O(n^2)$

Can we do better?

Counting Inversions: Divide & Conquer

2	3	8	5	4	10	9	12	7	18	15	25
---	---	---	---	---	----	---	----	---	----	----	----

1. Divide the list into two halves

2	3	8	5	4	10
---	---	---	---	---	----

9	12	7	18	15	25
---	----	---	----	----	----

2. Recursively count inversions in each half

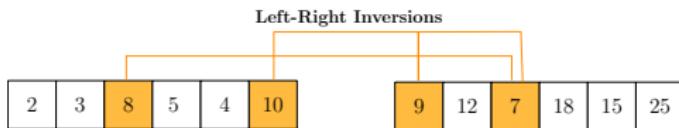
2	3	8	5	4	10
---	---	---	---	---	----

Left-Left : 8-5, 8-4, 5-4

9	12	7	18	15	25
---	----	---	----	----	----

Right-Right : 9-7, 12-7, 18-15

3. Count inversions where a_i and a_j are in different halves



4. Return total inversions count

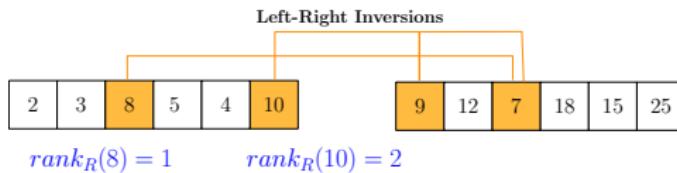
Counting Inversions: Divide & Conquer

- Divide the array into left and right halves
- Find left-left and right-right inversions recursively
- How to find left-right inversions?

How many L-R inversions a given element $x \in L$ is involved in?

Exactly the number of elements in R smaller than x , $\text{rank}_R(x)$

Finding L-R inversions is equivalent to finding ranks of all elements of L in R



L and R sorted \implies can find $\text{rank}_R(x) \forall x \in L$ (L-R inversions) in n steps

sorting L and R removes LL and RR inversions, LR inversions remain intact

Solution: First count LL and RR inversions, then sort L and R

Counting Inversions: Divide & Conquer

Algorithm Counting Inversions

```
function COUNTINVERSIONS( $A$ )           ▷ returns both sorted  $A$  and number of  
inversion in  $A$   
    if  $|A| = 1$  then  
        return ( $A, 0$ )  
  
     $L \leftarrow A[1, \dots, n/2]$   
     $R \leftarrow A[n/2 + 1, \dots, n]$   
    ( $sortedL, LL_{inv}$ )  $\leftarrow$  COUNTINVERSIONS( $L$ )  
    ( $sortedR, RR_{inv}$ )  $\leftarrow$  COUNTINVERSIONS( $R$ )  
     $LR_{inv} \leftarrow \text{SUM}(\text{FINDRANKS}(sortedL, sortedR))$            ▷  $n$  steps  
    return (MERGE( $sortedL, sortedR$ ),  $LL_{inv} + RR_{inv} + LR_{inv}$ )           ▷  $n$  steps
```

Counting Inversions: Recurrence Relation

The recurrence for runtime $T(n)$ on input size n is:

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

$$T(n) = 2n \log(n)$$

much better than $O(n^2)$