

## Design Paradigm: Divide and Conquer

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

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# Inversions

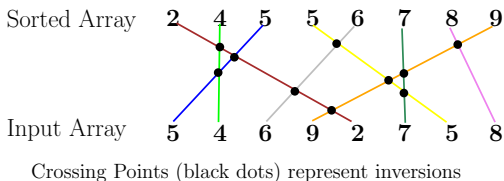
**Inversions** in an array  $A$  of numbers are the out-of-order pairs

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

$A =$

1	2	3	4	5	6	7	8
5	4	6	9	2	7	5	8

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



# Inversions

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**Inversions** in an array  $A$  of numbers are the out-of-order pairs

Pairs 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

▷ An array is sorted if there are zero inversions

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

## Applications

- Collaborative filtering
- Rank voting theory

# Recommendation Systems



Products, news, friends,  
websites,movies, courses



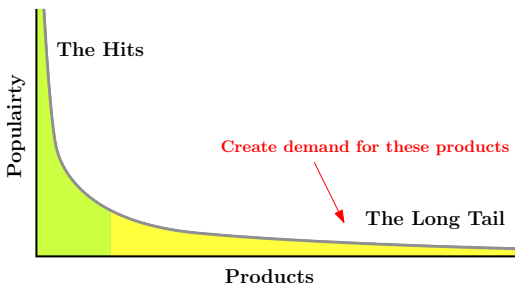
The Web, they say, is leaving the era of search and entering one of discovery. What's the difference? **Search** is what you do when you're looking for something. **Discovery** is when something wonderful that you didn't know existed, or didn't know how to ask for, finds you.

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)

# Recommendation Systems

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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			2	
$u_2$		1			2	1		2		1			3	
$u_3$		1	1	2			1					1	2	
			3	2			5		2		3	4		
	1			2						5				
$u_i$			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

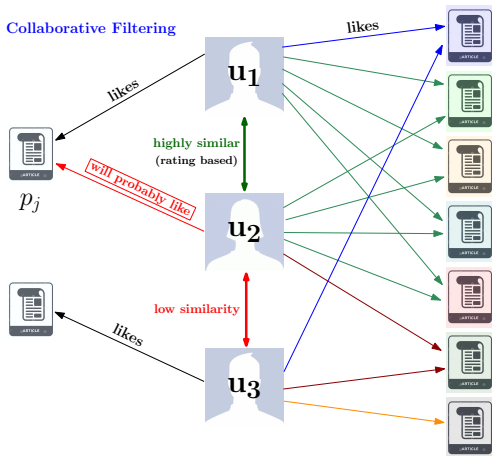
- 0 – 5 stars
- $\in [0, 1]$
- $\in \{0, 1\}$

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  $i$  rate  $j$  higher





# Inversions: Application in Collaborative Filtering

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

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  $u_y$

$$u_1 : \begin{array}{c} p_1 \quad p_2 \quad p_3 \quad p_4 \quad p_5 \quad p_6 \quad p_7 \\ \boxed{2} \quad \boxed{8} \quad \boxed{6} \quad \boxed{1} \quad \boxed{9} \quad \boxed{7} \quad \boxed{3} \end{array}$$
$$u_2 : \begin{array}{c} p_1 \quad p_2 \quad p_3 \quad p_4 \quad p_5 \quad p_6 \quad p_7 \\ \boxed{7} \quad \boxed{4} \quad \boxed{1} \quad \boxed{9} \quad \boxed{2} \quad \boxed{1} \quad \boxed{8} \end{array}$$
$$u'_2 : \begin{array}{c} p_4 \quad p_1 \quad p_7 \quad p_3 \quad p_6 \quad p_2 \quad p_5 \\ \boxed{9} \quad \boxed{7} \quad \boxed{8} \quad \boxed{1} \quad \boxed{1} \quad \boxed{4} \quad \boxed{2} \end{array}$$
$$u_3 : \begin{array}{c} p_1 \quad p_2 \quad p_3 \quad p_4 \quad p_5 \quad p_6 \quad p_7 \\ \boxed{2} \quad \boxed{7} \quad \boxed{6} \quad \boxed{1} \quad \boxed{9} \quad \boxed{8} \quad \boxed{2} \end{array}$$
$$u'_3 : \begin{array}{c} p_4 \quad p_1 \quad p_7 \quad p_3 \quad p_6 \quad p_2 \quad p_5 \\ \boxed{1} \quad \boxed{2} \quad \boxed{2} \quad \boxed{6} \quad \boxed{8} \quad \boxed{7} \quad \boxed{9} \end{array}$$

Is  $u_2$  closer to  $u_1$  or  $u_3$ ?

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

## Counting Inversions: Algorithm

Pairs 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  $A$

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**Algorithm** Counting Inversions - Brute force algorithm

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*count*  $\leftarrow$  0

**for**  $i = 1$  to  $n$  **do**

**for**  $j = i + 1$  to  $n$  **do**

**if**  $A[i] > A[j]$  **then**

*count*  $\leftarrow$  *count* + 1

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- 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
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Divide the list into two halves

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

Recursively count inversions in each half

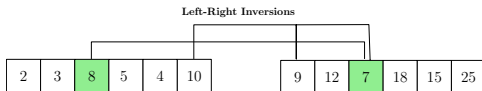
2	3	8	5	4	10
---	---	---	---	---	----

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

*Left-Left = 8-5, 8-4, 5-4*

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

Count inversions where  $a_i$  and  $a_j$  are in different halves



Return total inversions count

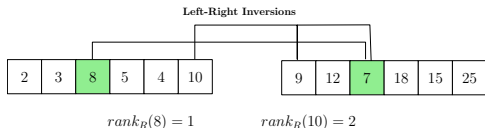
# Counting Inversions: Divide & Conquer

- 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

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## Algorithm Counting Inversions

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**function** COUNTINVERSIONS( $A$ )      ▷ returns both sorted  $A$  and number of inversion in  $A$

**if**  $|A| = 1$  **then**  
    **return**  $(A, 0)$

$L \leftarrow A[1, \dots, n/2]$

$R \leftarrow A[n/2 + 1, \dots, n]$

$(sortedL, LL_{inv}) \leftarrow \text{COUNTINVERSIONS}(L)$

$(sortedR, RR_{inv}) \leftarrow \text{COUNTINVERSIONS}(R)$

$LR_{inv} \leftarrow \text{SUM}(\text{FINDRANKS}(sortedL, sortedR))$       ▷  $n$  steps

**return**  $(\text{MERGE}(sortedL, sortedR), LL_{inv} + RR_{inv} + LR_{inv})$       ▷  $n$  steps

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## Counting Inversions: Recurrence Relation

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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  $O(n^2)$