1 Divide & Conquer Algorithms

1.1 $O(n \log n)$ Algorithm for Counting Inversions I

1.1.1 Divide and Conquer Paradigm

- Divide into smaller subproblems
- Conquer via recursive calls
- Combine solutions of subproblems into one for the original problem

The Problem

Input: Array, A, containing the numbers 1, 2, 3, ..., in the some order

Output: number of inversions = number of pairs (i, j) of array indices with i < j and A[i] > A[j].

If the array is not sorted, the number of inversions will be non-zero.

```
Example [4:00]: (1, 3, 5, 2, 4, 6)
Inversions: (3, 2), (5, 2), (5, 4)
```

Pictorially, the number of line crossings corresponds to inversions.

Motivation: Numerical similarity measure between two ranked lists.

Used in collaborative filtering.

In general, the maximum number of inversions in an array is $\binom{n}{2} = \frac{n(n-1)}{2}$.

High-level Approach

Brute-force: Setup a double for loop for i and j and check whether each pair is inverted, if so, add to count. This run in $O(n^2)$ time

Divide and Conquer: Classify inversions (i, j) with [i < j] of an array in one of three types:

```
    left: if i, j ≤ n/2
    right: if i, j > n/2
    split: if i ≤ n/2 < j</li>
```

The first two can be computed recursively, split needs separate subroutine.

Hign-Level Algorithm

```
\begin{array}{lll} Count(array\ A,\ length\ n) \\ & if\ n=1\ return\ 0 \\ & else \\ & x=Count(1st\ half\ of\ A,\ n/2) \\ & y=Count(2nd\ half\ of\ A,\ n/2) \\ & z=CountSplitInv(A,\ n) \\ & return\ x+y+z \end{array}
```

Goal: implement CountSplitInv in linear O(n) time => then Count will run in $O(n \log n)$ time.

1.2 $O(n \log n)$ Algorithm for Counting Inversions II

1.2.1 Piggybacking on Merge Sort

Key Idea #2: have recursive calls both count inversions and sort.

Motivation: Merge subroutine naturally uncovers split inversions.

output sorted A, B, C.

Merging uncovers number of inversions.

Pseudocode for Merge

```
D = ouput [length=n]

B = 1^{st} sorted array [n/2]

C = 2^{nd} sorted array [n/2]

i = 1
```

j = 1

Take B, C, traverse D using the k index, maintain pointers i, j to B and C respectively

for k = 1 to n

```
if \ A(i) < B(j) \ C(k) = A(i) \ i++ \ else \ [B(j) < A(i)] \ C(k) = B(j) \ j++ \ end \ (ignores \ end \ cases)
```

Suppose the input array A has no split inversions. What is the relationship between the sorted subarrays B and C?

All elements of B are less than all elements of C.

On an array with this property, MergeSort will copy everything from B to D before C even gets touched. So copying elements from array C has something to do with split inversions.

Ex.: [7:50]

General Claim: The split inversions involving an element, y, of the 2nd array C are percisely the numbers left in the 1st array B when y is copied to the output D.

Proof: let x be an element of the 1st array B

- if x copied to output D before y, then $x < y \Rightarrow$ no inversion involving x&y
- if y copied to output D before x, then $y < x \Rightarrow x &y$ are a split inversion

1.2.2 Merge and CountSplitInv

While merging the two sorted subarrays, keep running total of numbers of split inversions.

When element of 2nd array C gets copied to output D, increment total by number of elements remaining in 1st array.

1.2.3 Run time of subroutine

Linear time in merging, constant for count, sloppy:

$$O(n) + O(n) = O(n)$$

1.3 Strassen's Subcubic Matrix Multiplication Algorithm

$$X \times Y = Z$$

 $n \times n$ matrices, where

$$Z_{ij}i^{th}$$
 row of $\mathbf{Y} \times i^{th}$ column of $\mathbf{Y} = \sum_{k=1}^{n} X_{ik} \times Y_{kj}$

Note: input size = $O(n^n)$

Ex.: (n = 2)

$$\left[\begin{array}{cc} a & b \\ c & d \end{array}\right] \left[\begin{array}{cc} e & f \\ g & h \end{array}\right] = \left[\begin{array}{cc} ae + bg & af + bh \\ ce + dg & cf + dh \end{array}\right]$$

Running time from definition: $O(n^3)$ relative to matrix dimension, n.

1.3.1 Applying Divide and Conquer

Idea:

$$X = \left[\begin{array}{cc} A & B \\ C & D \end{array} \right]$$

$$Y = \left[\begin{array}{cc} E & F \\ G & H \end{array} \right]$$

where A through H are all $\frac{n}{2} \times \frac{n}{2}$ matrices.

then:

$$X \times Y = \left[\begin{array}{cc} AE + BG & AF + BH \\ CE + DG & CF + DH \end{array} \right]$$

1.3.2 Recursive Algorithm #1

Step 1: recursively compute the 8 necessary products

Step 2: do the necessary additions, $O(n^2)$ time

Fact: running time is $O(n^3)$ [fallows from master method]

1.3.3 Strassen's Algorithm (1969)

Step 1: recursively compute only 7 (cleverly chosen) products

Step 2: do the necessary (clever) additions + substractions (still $O(n^2)$ time)

Fact: better than cubic time! [see Master Method lecture for running time]

1.3.4 The Details:

The Seven Products:

$$P_1 = A(F - H)$$

$$P_2 = (A + B)H$$

$$P_3 = (C + D)E$$

$$P_4 = D(G - E)$$

$$P_5 = (A + D)(E + H)$$

$$P_6 = (B - D)(G + H)$$
$$P_7 = (A - C)(E + F)$$

 $\underline{\text{Cl}}\text{aim}$:

$$X \times Y = \begin{bmatrix} AE + BG & AF + BH \\ CE + DG & CF + DH \end{bmatrix}$$
$$= \begin{bmatrix} P_5 + P_4 - P_2 + P_6 & P_1 + P_2 \\ P_3 + P_4 & P_1 + P_5 - P_3 - P_7 \end{bmatrix}$$

Proof of one term:

$$P_5 + P_4 - P_2 + P_6 =$$

$$AE + \mathcal{A}H + \mathcal{D}E + \mathcal{D}H + \mathcal{D}G - \mathcal{D}E -$$

$$\mathcal{A}H - \mathcal{B}H + BG + \mathcal{B}H - \mathcal{D}G - \mathcal{D}H = AE + BG$$