# **CENTER FOR MASSIVE DATA ALGORITHMICS**

# **Locating Interesting Subsequences**

# Motivation

Finding *interesting* parts of sequences is a problem appearing repeatedly in data analysis. Locating G+C rich regions of DNA sequences or finding tandem repeats in chromosome data are such problems. In addition to Bioinformatics, similar and/or identical problems appear in Pattern Matching, Image Processing, and Data Mining.

## Problem definitions

Given a Sequence A[1,...,n] of Numbers

• Find a subsequence A[i, ..., j] maximizing  $\sum_{t=i}^{j} A[t]$ .

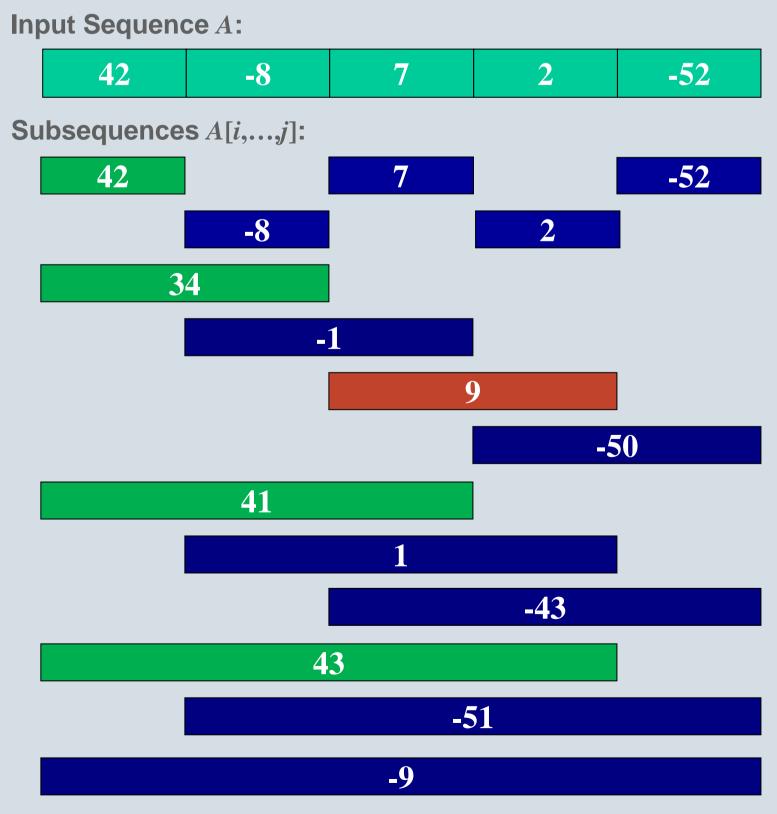
Given a Sequence *A*[1,...,*n*] of Numbers and Integer *k* 

- Find k subsequences maximizing  $\sum_{t=i}^{j} A[t]$ .
- Find a *k*'th largest subsequence.

Given a Sequence *A*[1,...,*n*] of Numbers, Integers *l*, *u* and *k* 

- Find k subsequences maximizing  $\sum_{t=i}^{j} A[t]$  among all subsequences of length at least *l* and at most *u*.
- Find a *k*'th largest subsequence among all subsequences of length at least *l* and at most *u*.

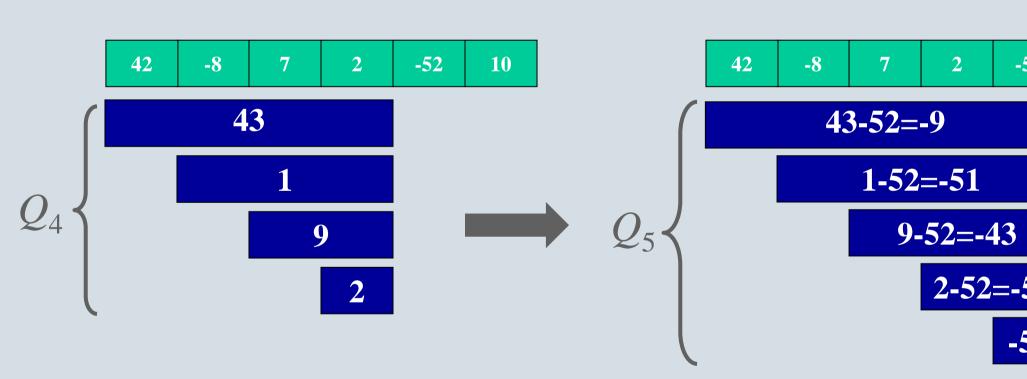
# Problem instance (k=5)

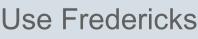


The k-1 largest are green, the k'th largest is red.

Insert all n(n+1)/2 sums in a heap ordered binary tree (values increase towards root). Find the *k* largest using Frederickson's heap selection algorithm.

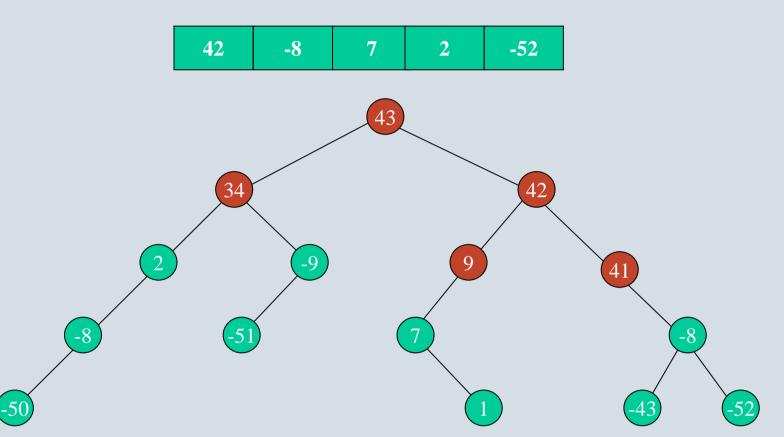








# Locating the *k* largest subsequences: Main ideas



Frederickson's algorithm finds the red nodes in O(k) time (no particular order)

Represent the  $O(n^2)$  sums implicitly in a heap ordered binary tree using O(n) space.

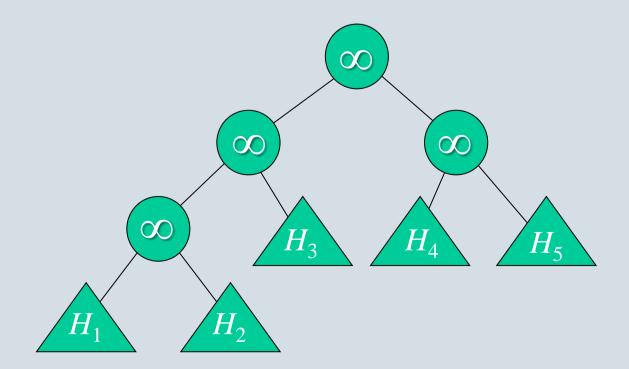
2

2-52=-50

-52

-52

A representation of  $Q_i$  can be build by adding the same value to all elements in  $Q_{i-1}$ and adding one new element. This allows for efficient construction of each set. Represent each set  $Q_i$  using a heap ordered binary tree  $H_i$  and join them.



Use Frederickson's algorithm and find the k+n-1 largest and discard the  $\infty$  values.

- An optimal algorithm finding the subsequence A[i,...,j]maximizing  $\sum_{t=i}^{j} A[t]$  in O(n) time is described in [1].
- In [2] we design an optimal O(n+k) time algorithm using O(k)space. This algorithm can be used to solve the 2-dimensional version of the problem in  $O(n^3+k)$  time using O(n+k) space and in general the d-dimensional problem in  $O(n^{2d-1}+k)$  time and  $O(n^{d-1}+k)$  space.
- In [3] we generalize this algorithm to find the subsequences inducing the *k* largest sums among all subsequences of length between l and u.
- In [3] we show an optimal  $O(n \log k/n)$  bound for selecting the subsequence inducing the *k*'th largest sum, by providing an algorithm with this running time and by proving a matching lower bound.
- We also generalize the selection algorithm to select the *k*'th largest subsequence among all subsequences of length between *l* and *u*, in  $O(n \log k/n)$  time. Remark that in this case  $k \leq (u-l+1)n$ .
- [1] Bentley, J.: *Programming Pearls: algorithm design techniques*. Commun. ACM 27(9)(1984) 865-873.
- [2] Brodal, G. S. and Jørgensen, A. G.: A linear time algorithm for the k maximal sums problem. Proc. 32<sup>nd</sup> International Symposium on Mathematical Foundations of Computer Science.
- [3] Brodal, G. S. and Jørgensen, A. G.: Sum selection in arrays. In submission.



### Results

## Bibliography