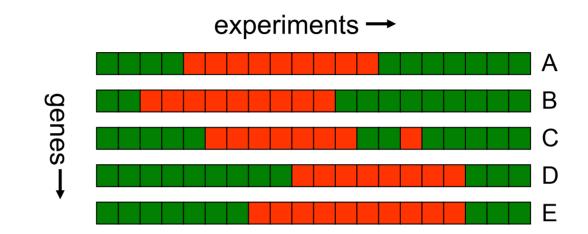
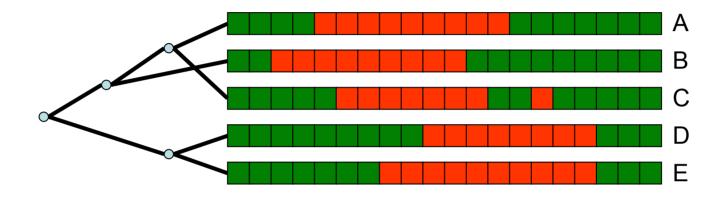
K-ary Clustering With Optimal Leaf Ordering for Gene Expression Data

Ziv Bar-Joseph, Erik Demaine, David Gifford, Angèle Hamel, Tommi Jaakkola, <u>Nathan Srebro</u>



Goal: Quickly and easily arrange the data for further inspection

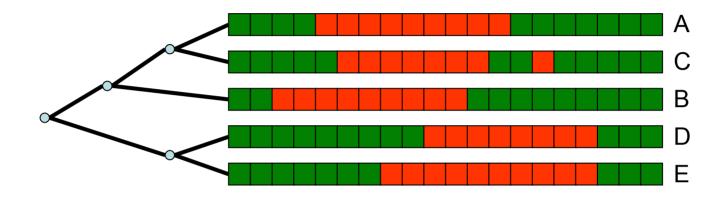
Hierarchical Clustering



• Greedily join nearest cluster pair [Eisen 1998]

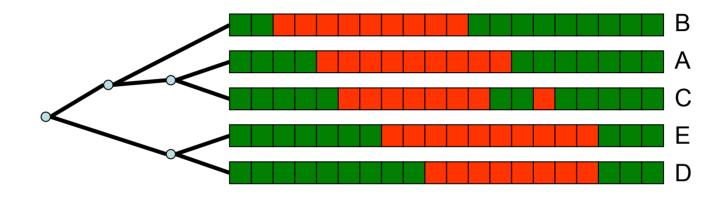
nearest: we use correlation coefficient (normalized dot product) can use other measures as well

Hierarchical Clustering

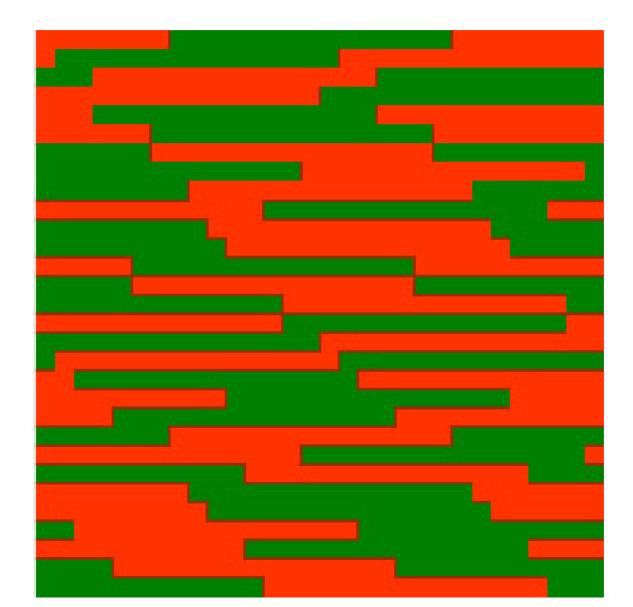


- Greedily join nearest cluster pair [Eisen 1998]
- Optimal ordering: minimize summed distance between consecutive genes
 - Criterion suggested by Eisen
 - n³ algorithm [Bar-Joseph et al 2001 + improvements]

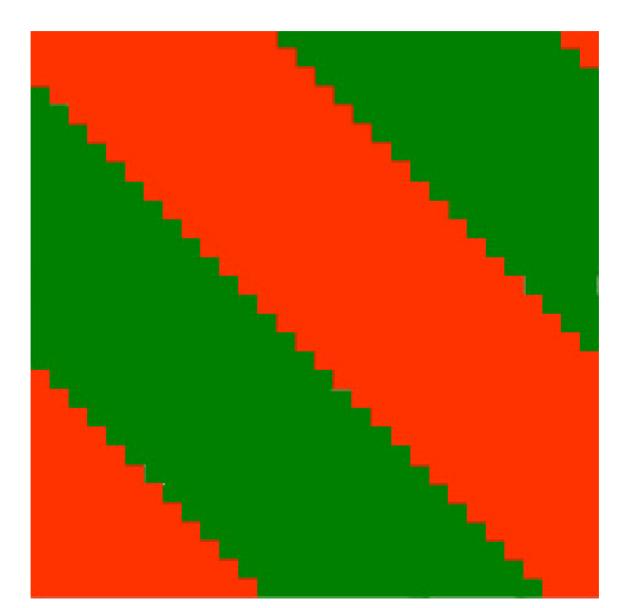
Hierarchical Clustering

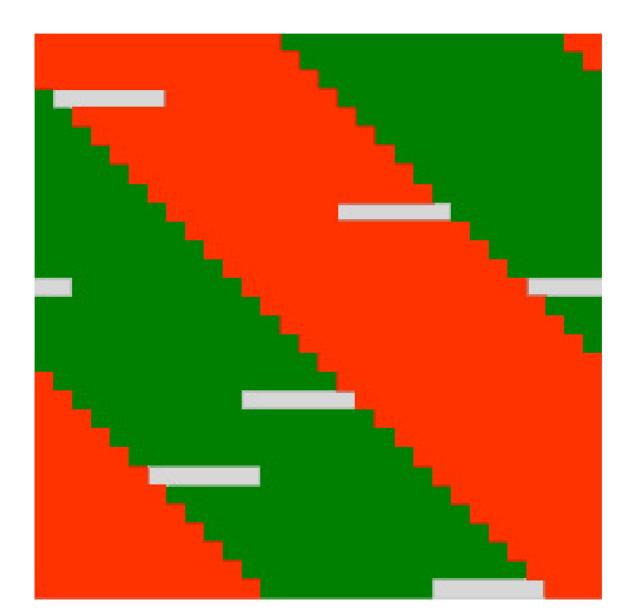


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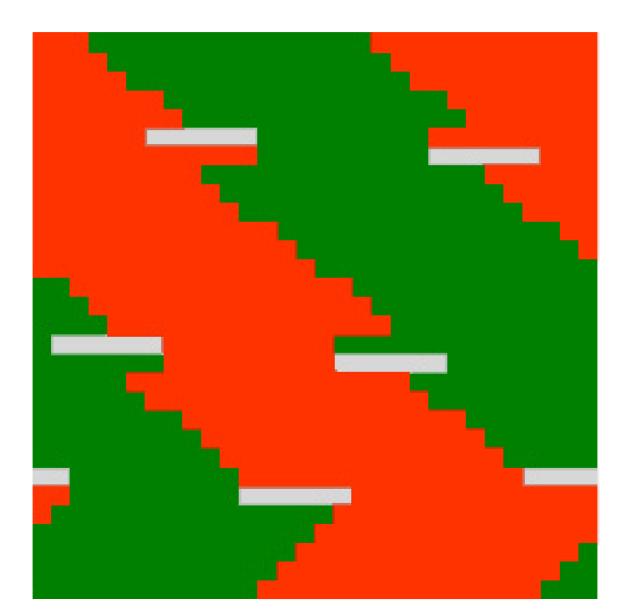


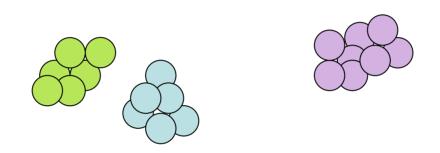
Optimally Ordered using Binary Clustering





Optimally Ordered using Binary Clustering





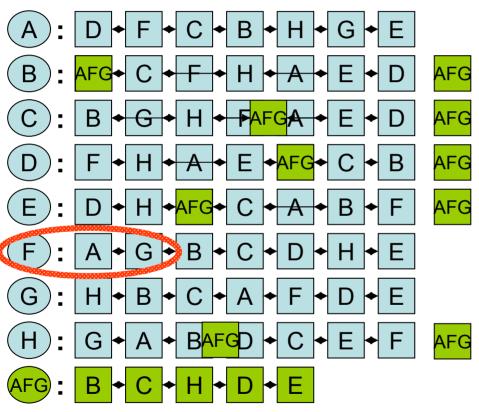
Main motivation for k-ary clustering: more information in k-wise distances leads to more robust clustering.

K-ary Hierarchical Clustering

- Greedily join tightest k clusters to form new cluster.
 - Tightest: maximal summed pairwise similarity
- Finding tightest k clusters: O(n^k)
- Running time: O(n^{k+1})
- Finding tightest k cluster is as hard as max-clique → W[1] hard, fixed parameter intractable: no poly(n)·f(k) algorithm

Sub-optimal Heuristic

k=3



- In order to find tight group of *k* clusters:
 - For each cluster, consider the cluster at its (k-1) nearest neighbors
 - Choose tightest of these neighborhoods

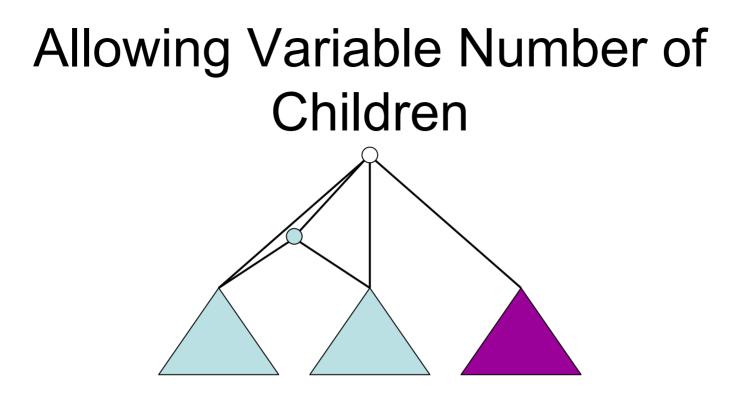
 For each cluster, maintain ordered linkedlist of neighbors

 $n^{2}\log n + n(nk^{2} + n \cdot \log n + n^{2} + n^{2}) = O(n^{3})$

Heuristic vs. Optimal

- Real data from 979 genes using k=3
- Heuristic 35 seconds vs Optimal 57 minutes (1.4 GHz Pentium III)
- Average node similarity: 0.6366 vs 0.6371

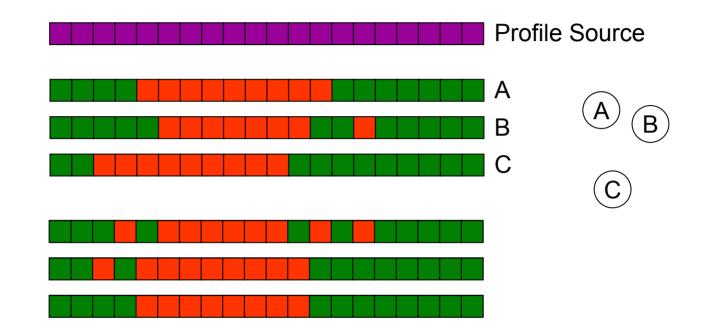
• For k=4, optimal is impractical



If the data clearly indicates a subset of less than *k* clusters that are significantly distinct from the rest, cluster them.

(Otherwise, cluster the tightest *k*-neighborhood)

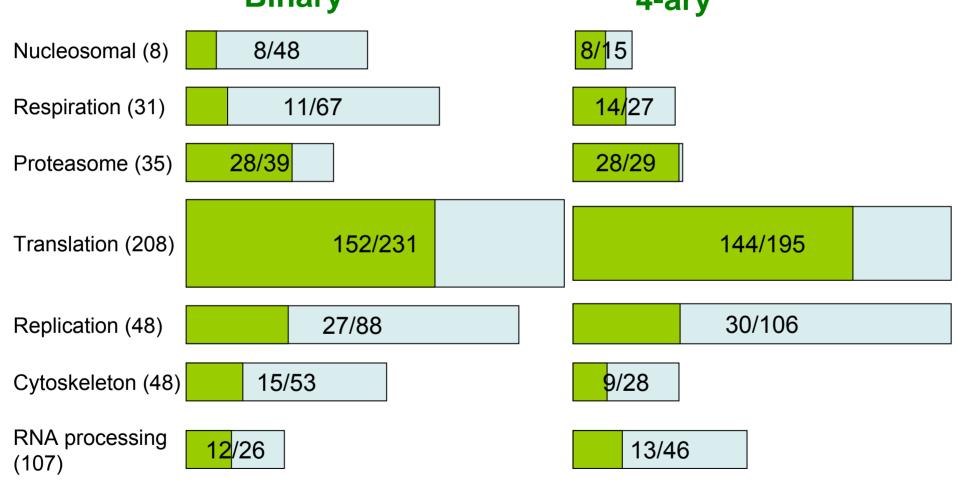
Permutation Test for Significance



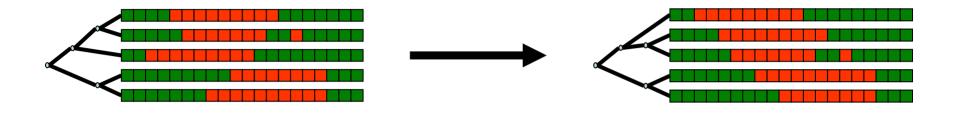
Might A,B and C all result from the same profile source, with the variations in distances being random variations?

Pr(2nd closest pair in bootstrap < 2nd closest pair in data)

Binary vs. 4-ary Hierarchical Clustering Binary 4-ary



Clusters of yeast condition response expression [Eisen et al 1998] corresponding to MIPS categories: genes from category/cluster size



Gene order consistent with hierarchical clustering tree that minimizes summed distance between consecutive genes (constrained TSP)

for binary trees [Bar-Joseph et al 01 + improvements]

a

V

v is least-common-ancestor of a,b

S(*a*,*b*)=Cost of opt ordering of descendents of *v*, such that *a*,*b* are at the two ends

for binary trees [Bar-Joseph et al 01 + improvements]

a

V

Va

Vb

d

••••

v is least-common-ancestor of a,b

S(*a*,*b*)=Cost of opt ordering of descendents of *v*, such that *a*,*b* are at the two ends

$$S(a,b) = \min_{c,d} S(a,c) + \delta(c,d) + S(d,b) \Rightarrow O(n^4)$$

$$T(a,d) = \min_c S(a,c) + \delta(c,d)$$

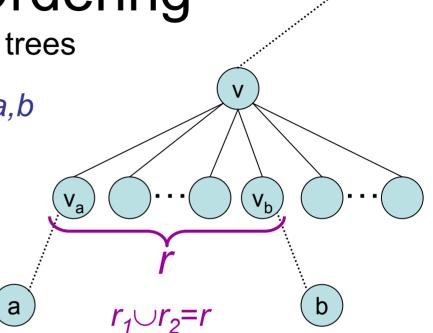
$$S(a,b) = \min_d T(a,d) + S(d,b)$$

for k-ary trees

v is least-common-ancestor of a,b

S(*a*,*b*,*r*)=Cost of opt ordering of **descendents of** *nodes in r*, such that *a*,*b* are at the two ends

r is a subset of children of *v*



 $S(a,b,r) = \min_{c \in r_1, d \in r_2} S(a,c,r_1) + \delta(c,d) + S(d,b,r_2)$

Dynamic Programming $r_2 = \{v_b\}$

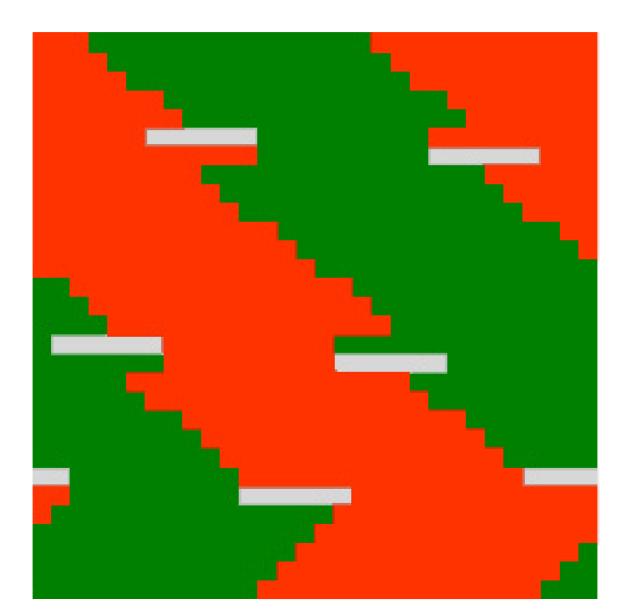
Divide and Conquer: no memoization equipartitions: $|r_1| = |r_2|$

> $O(n^3 4^k)$ $O(n^2 k)$ space

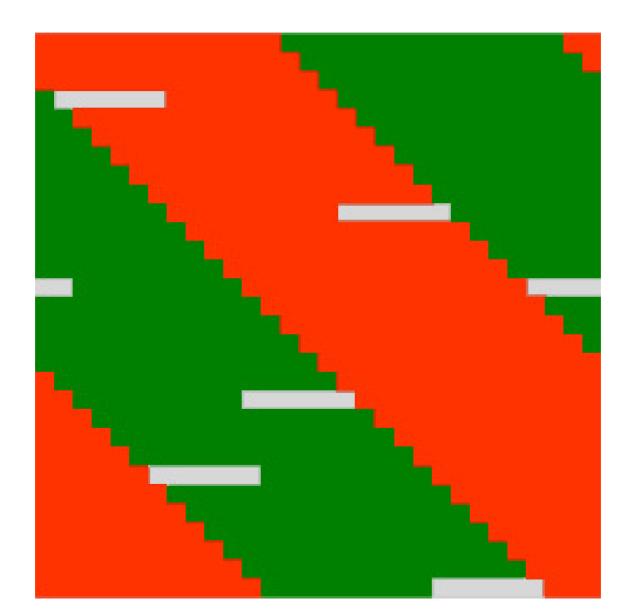
 $T(a,d,r_1) = \min_{\substack{c \in r_1}} S(a,c,r_1) + \delta(c,d)$ $S(a,b,r) = \min_{\substack{d \in r_2}} T(a,d,r_1) + S(d,b,r_2)$

O(n³2^k) O(n²2^k) space

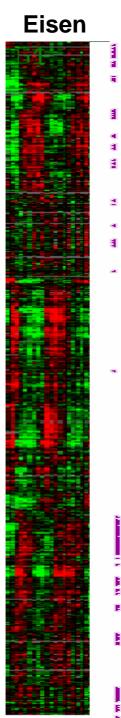
Optimally Ordered using Binary Clustering



Optimally Ordered using 4-ary Clustering



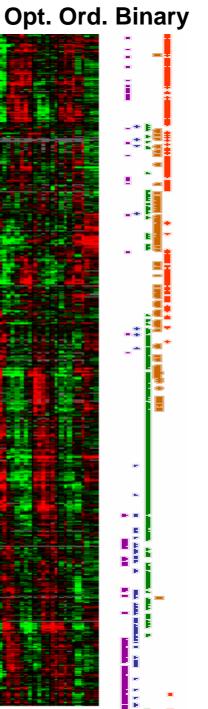


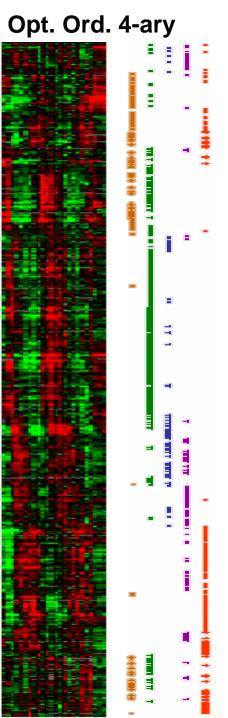




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Summary

We suggest using optimally ordered K-ary hierarchical clustering, instead of binary hierarchical clustering, for initial arrangement of gene expression data

- Simplicity and efficiency comparable to binary hierarchical clustering
- Clusters and order better reflect biology

http://psrg.lcs.mit.edu/~zivbj/