Current Innovations in Microarray Analysis

A look at two-sided clustering and context-specific Bayesian clustering

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Overview

- To date, biologists have used (one-sided) clustering to analyze their data
- While clustering is informative, there is much more in the data that we can learn
- New techniques, such as two-sided clustering and context-specific Bayesian clustering attempt to make mine more information out of these microarrays

Clustering

- Cluster software released in 1998 by Michael Eisen, et al
- Implements a standard statistical algorithm to determine what genes show the most similar expression pattern over all data

How Cluster Works¹

- Compute a similarity score between every pair of genes
- Search for the highest score this represents the most similar pair of genes
- Combine these genes into one node, and compute a similarity score between this node and every other gene/node
- Repeat recursively until there is one node, which is the whole tree

Benefits of Clustering

Cluster has become the industry standard for analyzing a biologist's expression data. Why?

- One of the first programs able to process huge quantities of data in microarrays
- So easy to use, even a biologist can use it
- Acceptable results for a first-pass analysis

Drawbacks of Clustering

But there is a lot more information in the data, information that Cluster does not extract

- Can not determine correlation between subsets of genes and experiments
- The model is not flexible; it can not incorporate any prior knowledge we might have about the genes and their functions (ie promoter regions, clinical data)

Innovations in the Works

Two-sided Clustering

- Forms subsets of genes and experiments
- Context-specific Bayesian Clustering
 - Flexible, intuitive model for gene regulation

Two-sided Clustering⁴

- More representational of biology in that genes that have common function will act together for the duration of the time that they carry out that function; two-sided clustering lets you see this subset
- Algorithm very similar to that of one-sided clustering
- Allows data to be clustered into subsets of genes and experiments

Bayesian Statistics 2,3,4

- Attempts to look at data and model gene relations based on causal relationships
- Cluster can only model data based on amount of transcription

Bayesian Statistics?

- Relies on a Bayesian Network
- For examples of basic concepts, see citation (2), pages 1-3
- BN are very good for describing processes that are locally dependent on each other

Advantages over Clustering

- Much richer than clustering, since Bayesian methods, given the same data set as a clustering program, can discovering "causal relationships, interactions between genes other than positive correlation, and finer intra-cluster structure"²
- Can incorporate all kinds of information, not just mRNA output levels, and it makes biological sense

Citations

- Eisen, M., P. Spellman, P. Brown, and D. Botstein (1998). Cluster analysis and display of genomewide expression patterns. PNAS 95, 14863-14868.
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- Segal, E., and B. Taskar, A. Gasch, N. Friedman, and D. Koller. Rich Probabilistic Models for Gene Expression. Bioinformatics vol. 1 no. 1, 1-9.