

Gene Expression Platforms for Global Co-Expression Analyses A Comparison of spotted cDNA microarrays, Affymetrix microarrays, and SAGE

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cDNA Microarrays

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1. Abstract

We have conducted a comprehensive comparison of three major expression platforms: cDNA microarray, oligonal cottled microarray, oligonal cottled microarray, oligonal cottled microarray, oligonal cottled microarray and serial analysis of gene expressions (SAGE) using large sets of available data for Homo suppress. Several studies have compared two of the three platforms to evaluate the consistency of expression profiles compared two of the three platforms to evaluate the consistency of expression profiles are compared two of the three platforms are observed as recently published data set of 1202 cDNA microarray experiments (Stuart et al. 2003), 214 SAGE (Barrates form GOAF and internal sources, and 522 Affymetrs (GH-U133A) of the contraction of



Human gene expression data for three major expression platforms (see sidebar) were collected. We used a recently published data set of 1202 cDNA microarray experiments (Stuart et al., 2003), 214 SAGE libraries from the Caneer Genome Anatomy Project (CGAP) and internal sources, and 502 collections of the Caneer Genome Anatomy to the Caneer microarray experiments from the Gene Expression Omnibus (GEO). SAGE Expression Omnibus (GEO). SAGE tags were mapped to genes by the lowest sense tag predicted from Refseq or MGC sequences. Gene Ids from all three platforms were then mapped to Locus.Link and the intersection determined.



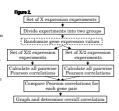
3. Methods

Gene Expression Analysis [sections 4-6]

Pearson correlations between genes were calculated using a modified version of the C clustering library (De Hoon et al., 2004). Correlations of correlations were calculated using the R statistical package (v. 1.8.1) and plotted with the R hexbin function.

rnal Consistency Analysis (section 4)

To evaluate the consistency of co-expression observed with each platform, we divide the experiments available and determine co-expressis for each subset independently. The results are then compared by results are then compared by calculating a correlation of the gene correlations. If the platform consistently finds co-expressed gene regardless of the exact experiments to 1. To determine whether the observed correlation is significant, we repeat the procedure with randomized gene expression values expecting a correlation close to 0.



Platform Comparison Analysis (section 5)

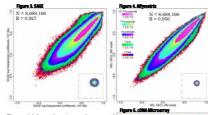
As with the internal consistency analysis, a correlation of gene correlations wa determined except for each of the three pairwise platform comparisons instead of between subsets of one platform. If the two platforms being compared report the same distance between each gene pair, the overall correlation between platforms should be near 1

Instead of considering the actual Peaseon correlation between each pease pair and to comparing between platforms, the correlation rath was considered. For example, it may be that for gene A. SAGE experiments identify its most similar gene (in terms of expression patterns to be gene B with a Peaseon correlation of 0.9. The CBNA control of the CB

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The Genro Ontology (GO) MySQL database dump (release 200402 of associb) was downloaded and a GO MySQL database was constructed. The most specific GO association of the Good o

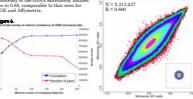
4. Internal Consistency Analysis



Figures 3-5: Internal consistency of expression datasets. Affymetrix shows the highest internal correlation of 0.96, then SAGE with correlation 0.92, and cDNA microarray with correlation 0.56. Inset: same plot for randomized data.

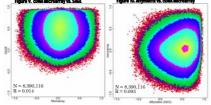
Figure 6: The low internal consistency observed in Fig. 5: is due to a large number of mossing values in the 2DVA microarray data. In the 2DVA microarray data experiments, and not all genes are present on all the arrays. Thus, for genes that are rarely present on the same array. Pourson overlapping data points. Increasing the wordsping data points Increasing the required number of overlapping data points increases the internal consistency.

Figure 7: When gene pairs sharing less than 100 data points are excluded, the internal consistency of the cDNA microarray dataset rises to 0.88, comparable to that seen for SAGE and Affymetrix.

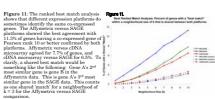


5. Platform Comparison Analysis

Figures 8-10: Cross platform comparisons. Despite the high levels of internal consistency observed above, surprisingly poor levels of consistency were observed between platforms. The distribution for each platform and the distribution for each platform correlations of r < 0.1. Affymetrix versus SAGE showed the best correlation of 0.094, then Affymetrix versus WAGE with 0.014. There are several possible explanations for this observation: One possibility is that one platform is correct one possibility is that one platform is correct explanation is that each platform identifies different to espersion patterns because the available data for each platform represents different to essenties and the platform in the control of the control o conditions. Yet another poss few genes are actually consis



6. Ranked Best Match Analysis





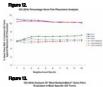


Figure 12: GO biological process domain knowledge was used to evaluate gene co-expression predictions for each platform. The proportion of gene pairs annotated at a specific common GO term for a given gene were enumerated and compared for a given gene were enumerated and compared against the maximum number of gene pairs that share GO terms for a given gene across each neighborhood distance. In general, the three neighborhood distance. In general, the three maximum of the compared to t

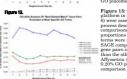
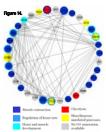


Figure 13. Gene pairs identified by more than one platform in the ranked best match analysis (Section pairs) in the pairs of the pairs

Composition related to enterple. Figure 11: Literallying groups or networks of coexpressed geners is the starting point for analyses such as functional annation of novel genes or identification of transcription factor between the control of the c



8. Conclusions

- Cocopyroused genes can be identified based on ingre-scale gene expression data instructed considerations are fairly high for excrypening pintures identified by Affymetrix (Be-9.08, SAGE (Be-9.29) and cDNA microarray (Be-0.50)

 DNA microarray data are more consistent when sufficient data is available (Be-0.88)

 Direct comparison of correlation values between platforms yields poor correlations (Be-0.1) platforms, particularly between Affymetrix and SAGE present plant destricted by different platforms, particularly between Affymetrix and SAGE present plant destricted by different platforms, particularly between Affymetrix and SAGE present plant of the control o
- piatiorms, particularly between Atlymetrix and SAGE.

 Gene pairs identified as occeptseed are more likely to share the same function

 Coexpressed gene pairs identified by more than one platform which also share functional
 annotations are most likely to be of biological interest, further analysis of these genes, using
 orthology and motif finding algorithms, can attempt to identify common transcription factor
 binding sites that may regulate the expression of these co-expression networks



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references | 1. Stuart et al. 2003. Science. 302(5643):249-255; 2. De Hoon et al. 2004. Bioinformatics. Feb 10 [epub ahead of print]; 3. Shannon et al. (2003). Genome Res 13:2498-2544.