

GE-biplot Microarray data



"An approach to the ordination
of Gene Expression Data - the
GE-biplot"

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Outline

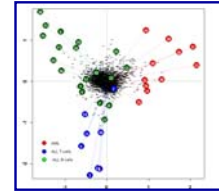
1. The Bi-plot

- GE-Biplot

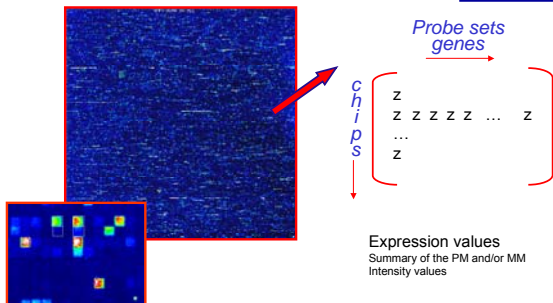


2. Applications

- Simulated Data
- Colon Data
- Leukemia Data

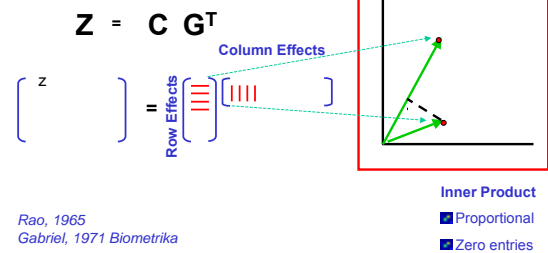


GeneChip Data



Biplot

Graphical Display of
a Rank 2 matrix



Rao, 1965
Gabriel, 1971 Biometrika

Biplot - Microarray Data

$$Z = C G^T$$

$$= \{C R^T\} \{G R^{-1}\}^T$$

Factorization

metric in which to represent the data.

Approximation of Z by a matrix of rank 2

SVD

Choice of Z

- Expression level transformations ? (logs)
- Row (chip) standardizing? (Scaling, Normalization)
- Column (gene) mean centering or standardization?
- Gene selection (filtering) ?

Approximate Biplot

$$Z_{(2)} = U_{(2)} \Lambda_{(2)} V_{(2)}^T$$

Eckhart and Young 1939
Good 1969

- cols of V = right singular vectors
= eigenvectors of Z'Z
- cols of U = left singular vectors
= eigenvectors of Z Z'

$$\Lambda = \text{diag}(\lambda_1, \lambda_2, \dots) \lambda_1 > \lambda_2 > \dots > 0$$

$$= \text{diag}(\sqrt{\lambda_1}, \sqrt{\lambda_2}, \dots)$$

- Minimizes $\sum_i \sum_j (z_{ij} - z_{(ij)})^2$
- Goodness of Fit is $\sum_{i=1}^n \lambda_i^2 / \sum_{i=1}^n \lambda_i^2$

$$Z = \underbrace{\begin{bmatrix} u_1 & u_2 \\ \vdots & \vdots \end{bmatrix}}_C \underbrace{\begin{bmatrix} \lambda_1 & \lambda_2 \\ \vdots & \vdots \end{bmatrix}^\alpha \begin{bmatrix} \lambda_1 & \lambda_2 \\ \vdots & \vdots \end{bmatrix}^{1-\alpha} \begin{bmatrix} v_1 & v_2 \\ \vdots & \vdots \end{bmatrix}}_{G^T}$$

GE-biplot

Factorization

$$Z = CG^T = (U\Lambda^\alpha)(\Lambda^{-1-\alpha}V^T)$$

• If $\alpha=0$, $C=U$ and $G^T=V^T$

Then if Z is mean corrected

and $C = \sqrt{N_c}U$ and $G = (1/\sqrt{N_c})\Lambda V^T$

$$GG^T = S_g$$

Matrix Z

- Log 2 expression level
- Chips - standardized
- Genes - mean corrected

Variant of the h-plot

Corsten and Gabriel (1976)

Display

- Chips - vectors with annotation.
- Genes - points or symbols

Focus: Similarity of up down regulation.

Goodness of Fit
variances

$$\sum_{i=1}^K \lambda_i^4 / \sum_{i=1}^K \lambda_i^2$$

Simulation Study

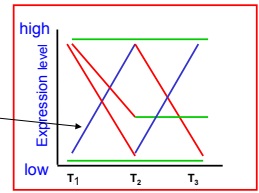
Multiplicative model

'True' expression values

low (1), medium (3), or high (9)

- 27 genes
- 3 treatment groups
- 3 replicates

Every combination

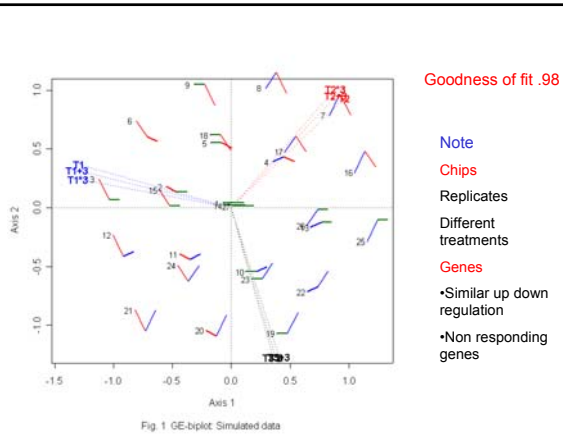


Gene

Treatment Groups

Matrix Z

1. Log 2 expression level
2. Chips - standardized
3. Genes - mean corrected



Goodness of fit .98

Note

Chips

Replicates

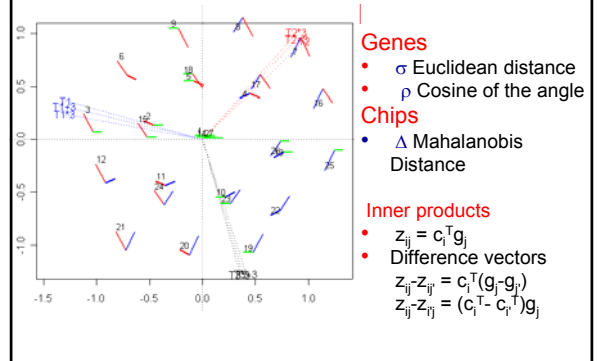
Different treatments

Genes

• Similar up down regulation

• Non responding genes

Some Interpretations



Genes

- σ Euclidean distance
- ρ Cosine of the angle

Chips

- Δ Mahalanobis Distance

Inner products

- $Z_{ij} = c_i^T g_j$
- Difference vectors $Z_{ij} - Z_{ij'} = c_i^T (g_j - g_{j'})$
- $Z_{ij} - Z_{ij'} = (c_i^T - c_{i'}^T) g_j$

Colon Data

Data

62 x 1988

- 22 matched normal and tumour colon tissue samples
- 18 unmatched tumour colon tissues

Pre-processing steps

- Filtering ; selection of the 2000 'highest minimal intensity' genes.

- considerable skewness

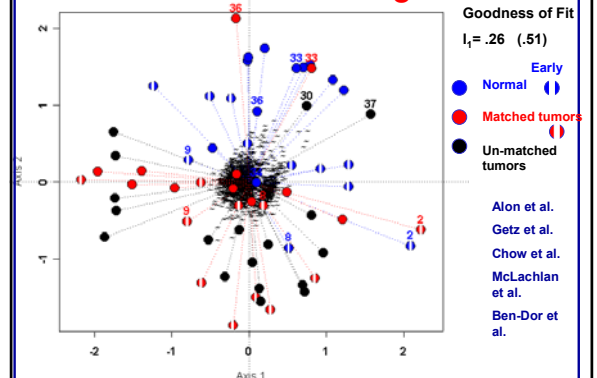
Problems

- mixture of matched and unmatched samples
- change of protocol after the first 11 samples of matched pair data
- contamination with muscle tissue (Alon et al., 1999).

<http://www-genome.wi.mit.edu>

Alon U., Barkai N., Notterman D.A., Gish K., Ybarra S., Mack D., Levine A.J., 1999. Broad patterns of gene expression revealed by clustering analysis of tumour and normal colon tissues probed by oligonucleotide arrays. *Proc Natl Acad Sci U S A*. 96(12) 6745-50.

Colon data -1988 genes



Goodness of Fit

$I_1 = .26$ (.51)

Early

Normal (●)

Matched tumors (●)

Un-matched tumors (●)

Alon et al.

Getz et al.

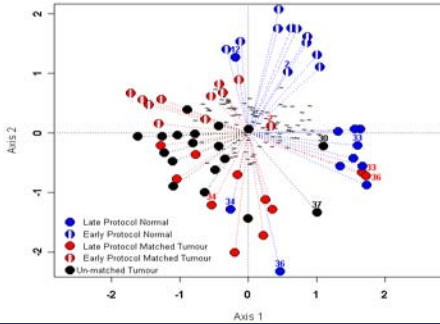
Chow et al.

McLachlan et al.

Ben-Dor et al.

Gene Selection MAD 100 genes

Goodness of Fit
 $I_1 = .41$ ($.79$)



Leukaemia Training Data Set

Data 38 x 7129

- 11 Acute Lymphoblastic Leukaemia (AML)
- 27 Acute Myeloid Leukaemia (ALL).
- 8 T-cells
- 19 B-cells

• Pre-processing steps as described in Dudoit et al (2002) + removal of controls

- Thresholding : floor 100 ceiling 16000
- Filtering : max/min ≤ 5 and max-min ≤ 100

• 3030 genes \log_2

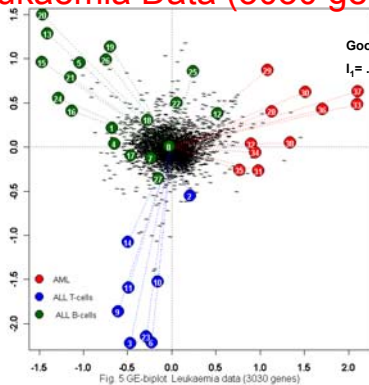
• Censored data

Golub, T.R., Slonim, D.K., Tamayo, P., Huard, C., Gaasenbeek, M., Mesirov, J.P., Coller, H., Loh, M.L., Downing, J.R., Caligiuri, M.A., Bloomfield, C.D., Lander, E.S., 1999. Molecular classification of cancer: Class discovery and class prediction by gene expression profiling. *Science*, 286, 531-537.



Leukaemia Data (3030 genes)

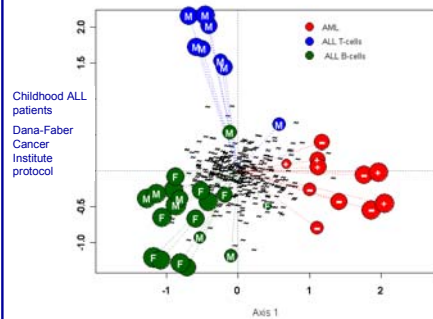
Goodness of Fit
 $I_1 = .26$, $I_2 = .6$



Golub et al.
Dudoit et al.
Ge et al.
Korenberg

10% Most varying genes

Goodness of Fit
 $I_2 = .74$



Childhood ALL patients
Dana-Faber
Cancer
Institute
protocol

Adult samples
Cancer and
Leukemia Group
B Leukemia
bank.

Gene Selection

Ge et al.
 $I_1 = .65$, $I_2 = .97$
Adjusted $p < .05$

Independent F-tests
3 groups
 $I_1 = .63$, $I_2 = .97$
 $p < .00001$

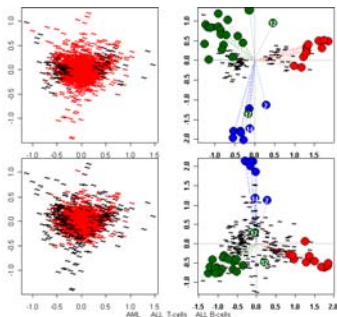


Fig 7. Top: Ge et al selection (52 genes). Bottom: Ge et al selection (245 genes). Left: Gene plot. Right: GE biplot.

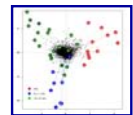
Summary

1. The biplot is a useful visualization tool for microarray data

Simultaneous plotting of the genes and chips on the same plot

2. Many types of biplots

- Factorization, Rank 2 approximation, Matrix
- GE-Biplot



CSDA -Computational Statistics & Data Analysis Journal

• Special issue on bioinformatics & biostatistics.

• Possibly an issue on the analysis of microarray data

