

## BIOS 7659 Journal Club:

A comparison of normalization methods for high density oligonucleotide array data based on variance and bias. (Bolstad et al., 2003)

Tim Vigers

9/22/2020

# Introduction

- ▶ The goal of normalization is to separate the interesting biological variation from the variation that is a result of sample preparation, array production and processing, etc.
- ▶ Affymetrix proposes scaling the arrays so that each one has the same mean expression summary measure.
  - ▶ This does not work well when there are non-linear relationships between arrays.

# Alternatives

- ▶ Other approaches, such as non-linear smooth curves or transforming data to standardize the distribution of intensities across arrays, rely on picking a “baseline” array.
- ▶ Bolstad et al. compare three different approaches, all of which combine data from every single array rather than relying on a baseline.

## Cyclic loess

- ▶ Basically an extension of the  $M$  vs.  $A$  plots discussed in class, but applied to pairwise combinations of Affymetrix arrays.
- ▶  $M$  is the difference in log expression values and  $A$  is the average (a Bland-Altman plot).

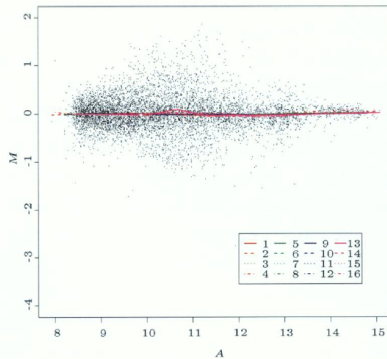
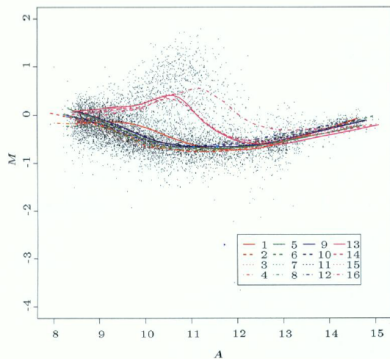


Figure 1: Dudoit et al., 2002

## Cyclic loess

1. Take two arrays  $i$  and  $j$ , each with probes  $k = 1, \dots, p$ .
2. Create an MA plot for these two arrays, and fit a loess curve through these data:

$$M_k = \log_2\left(\frac{x_{ki}}{x_{kj}}\right), A_k = \frac{\log_2(x_{ki}x_{kj})}{2}$$

3. Subtract the normalization curve fits  $M'_k = M_k - \hat{M}_k$  and obtain adjusted probe intensities:

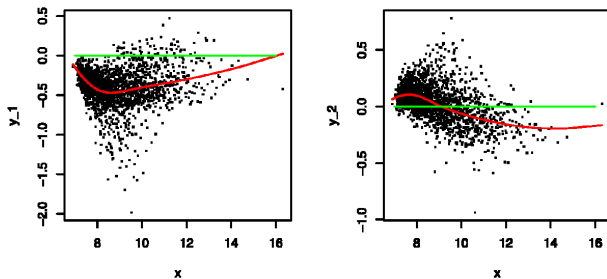
$$x'_{ki} = 2^{A_k + \frac{M'_k}{2}}, x'_{kj} = 2^{A_k - \frac{M'_k}{2}}$$

4. Take each of these adjustments (one for each pairwise comparison between arrays) and weight them equally across the set of arrays.

## Contrast method

- ▶ Very similar to the cyclic loess method, because it's another way of normalizing based on  $M$  vs.  $A$ :
- 1. Data is converted to the log scale and the basis is transformed (this is just a fancy linear algebra step).
- 2.  $n - 1$  normalizing curves are fit to the transformed basis as in cyclic loess.
- 3. Data is transformed again so that the normalizing curves lie on the horizontal, this time using a smooth function.
- 4. This normalized data is transformed back to the original basis and exponentiated.
- ▶ This is slightly faster than cyclic loess but fitting the curves can still be slow.

# Contrast method



**FIG. 2.** Contrast plots. Scatter plots of the 2 contrasts against the mean for 3 arrays, A, B, and C, prior to normalizing. The red curve is the fitted normalizing curve, and the green line is the reference line.

Figure 2: Åstrand, 2003

# The quantile method

- ▶ The goal of this method is to standardize the distribution of probe intensities across all arrays.
- ▶ The approach is an  $n$ -dimensional extension of the fact that, given a quantile-quantile plot where all of the points are on a straight diagonal line, you can be fairly sure that the two data vectors have the same distribution.
- ▶ We want to project our data onto the unit vector  $(\frac{1}{\sqrt{n}}, \dots, \frac{1}{\sqrt{n}})$ .
- ▶ Let  $\mathbf{q}_k = (q_{k1}, \dots, q_{kn})$  be the vector of  $k^{\text{th}}$  quantiles for  $k = 1, \dots, p$ . Then:

$$\text{proj}_{\mathbf{d}} \mathbf{q}_k = \left( \frac{1}{n} \sum_{j=1}^n q_{kj}, \dots, \frac{1}{n} \sum_{j=1}^n q_{kj} \right)$$



# Quantile normalization algorithm

1. Given  $n$  arrays with  $p$  probe intensity measurements, make the  $p \times n$  matrix  $X$ , where each column has all the data from a single array.
  2. Sort each column of  $X$  to produce  $X_{\text{sort}}$ . So, each row in  $X_{\text{sort}}$  is a quantile.
  3. Take the mean of each row, and replace every value in the row with the mean to produce  $X'_{\text{sort}}$
  4. Put each column of  $X'_{\text{sort}}$  back in the original ordering from  $X$  to produce  $X_{\text{normalized}}$
- This approach could theoretically be a problem for probes that have the same value across all arrays, but in practice this isn't an issue.

# Scaling method

- ▶ Based on the approach suggested by Affymetrix, but this paper uses a probe-level version.
- 1. Choose a baseline array  $x_{\text{base}}$ : Usually this is the median array, but doesn't necessarily have to be.
- 2. For each other array, calculate the mean trimmed intensity  $\tilde{x}_i$  and find

$$\beta_i = \frac{\tilde{x}_{\text{base}}}{\tilde{x}_i}$$

- 3. Normalized intensities are  $x'_i = \beta_i x_i$

## Non-linear method

- ▶ The scaling method is the same as fitting a straight line with intercept 0 between  $x_{\text{base}}$  and  $x_j$ .
- ▶ This can be extended to non-linear methods, usually a loess curve such that:

$$x'_j = \hat{f}_i(x_j)$$

where  $\hat{f}_i(\cdot)$  is the curve mapping from array  $i$  to baseline.

## References

1. Dudoit, S., Yang, Y. H., Callow, M. J., & Speed, T. P. (2002). STATISTICAL METHODS FOR IDENTIFYING DIFFERENTIALLY EXPRESSED GENES IN REPLICATED cDNA MICROARRAY EXPERIMENTS. *Statistica Sinica*, 12(1), 111–139. JSTOR.
2. Åstrand, M. (2003). Contrast Normalization of Oligonucleotide Arrays. *Journal of Computational Biology*, 10(1), 95–102.  
<https://doi.org/10.1089/106652703763255697>