

Due Date: Thursday, 18 October 2007

1. This assignment uses data from the study of MLL that was described in our first lectures on dChip, and which comes from a paper by Armstrong et al, *Nature Genetics*, 2002;**30**:41–47. The data can be found at: [http://www.broad.mit.edu/cgi-bin/cancer/publications/pub\\_paper.cgi?mode=view&paper\\_id=63](http://www.broad.mit.edu/cgi-bin/cancer/publications/pub_paper.cgi?mode=view&paper_id=63)
  - (a) Begin by downloading the CEL files for the ALL and the MLL samples. In addition, download a copy of the file `scaling_factors_and_fig_key.txt`. How many CEL files are there? How many arrays are U95A and how many are U95Av2?
  - (b) Get a copy of the original publication. Briefly summarize the main results of the paper.
2. Prepare a “data list file” and a “sample info file” that can be used to load the U95A (but not the U95Av2) arrays into dChip. Your sample info file should include a column that describes whether the sample came from a patient with ALL or a patient with MLL. Analyze the data in dChip, producing a list of genes that are differentially expressed between ALL samples and MLL samples. How many genes are on the list? Use a permutation test to estimate the false discovery rate (FDR) for your list of genes.
3. We are now going to analyze the same data in R. Process the data using `just.rma` (or `justRMA` if you prefer the widget-based GUI for loading data) to quantify the files. What kind of object is produced, and what does it contain?
4. Perform two-sample t-tests for each gene.
  - (a) Which group of samples has a higher expression for a gene if the t-statistic is positive?
  - (b) Plot a histogram of the t-statistics.
  - (c) How many genes have a t-statistic greater than 4 in absolute value? How many have a t-statistic greater than 3.5?
  - (d) How many genes are significant using a Bonferroni correction and a significance level of 5%?
5. This problem is a continuation of the previous problem.
  - (a) Plot a histogram of the p-values.
  - (b) Fit a beta-uniform mixture (BUM) model to the set of p-values. Does the model appear to fit the data well?
  - (c) How many genes are called significant if you use the BUM model to bound the false discovery rate (FDR) at 5%?
  - (d) What p-value cutoff corresponds to FDR= 5%?
6. Perform a Wilcoxon test for each gene.
  - (a) If the rank-sum statistic for a gene is very large, which group of samples has a higher level of expression for that gene?
  - (b) Plot a histogram of the Wilcoxon statistics, and overlay a plot of the theoretical Wilcoxon distribution. Based on the plot, are there likely to be any differentially expressed genes? Are more genes overexpressed in ALL or in MLL samples?

- (c) Use the empirical Bayes method to compute posterior probabilities for differential expression as a function of the Wilcoxon rank-sum statistics. By trial and error, find the largest value of the prior probability of not being differentially expressed that ensures that the computed posterior probabilities are all positive.
  - (d) Using the prior probability from the previous part, determine how many genes have a posterior probability of being differentially expressed that is at least 90%.
7. Analyze the data using the tail-rank test. Since all the MLL samples have a similar genetic abnormality, we will assume they are more homogeneous and use them as the baseline group. Using a target specificity of 95% and a confidence level of 90%, how many genes are found to be significant by the tail rank test?
  8. At this point, we have analyzed the ALL and MLL samples multiple ways, and produced 4 different lists of differentially expressed genes (dChip, t-test, Wilcoxon test, tail-rank test). Construct a table that counts the overlap between each pair of analyses. (Note that you will have to export the results from dChip and read them into R in order to complete this problem.)