RMA+ and RMA++ using the RefPlus package

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Abstract

In this vignette, we introduce the ideas behind Extrapolation Strategy(RMA+) and Extrapolation Averaging (RMA++) methods, and give examples of using the functions in this package.

1 Introduction

The Extrapolation Strategy and Extrapolation Averaging are Affymetrix GeneChip microarray data pre-processing methods proposed by Goldstein (2006). These methods were independently developed by Chang, Harbron and South (2006), termed RMA+ and RMA++. Katz et al. (2006) also independently developed the RMA+ method, termed refRMA. This vignette will use the "RMA+" and "RMA++" nomenclature for these algorithms. RMA+ is an extension to the RMA algorithm by Irizarry et al. (2004), and RMA++ is a further extension based on the RMA+ method.

The RMA+ algorithm calculates the microarray intensities using a pre-stored RMA model trained on a reference microarray set (can be standard reference microarrays, microarrays from an independent study, or an incomplete set of microarrays in a study). RMA+ measurements of a microarray can be considered as an approximation to the RMA measurements of this microarray when the microarray is RMAed with the reference set microarrays in one batch.

RMA++ measurements of a microarray are the average of multiple RMA+ measurements of a microarray based on several reference sets. If the reference sets cover more information of the microarrays to be pre-processed than a single reference set does, the RMA++ measurements will provide a better approximation to the RMA measurements.

2 RMA+

RMA+ procedure:

1. Fit the RMA model on the reference set and store the normalizing quantiles and the estimated probe effects;

- 2. Background correct the probe intensities of the microarrays to be preprocessed;
- 3. Normalize the background-corrected probe intensities to the normalizing quantiles (reference quantiles);
- 4. Derive the probeset intensity using the estimated probe effects and normalized background-corrected probe intensity data.

Step 1 can be done using the **rma.para** function in the package. The normalizing quantiles and the estimated probe effects are returned. Step 2-4 can be done using the **rmaplus** function.

Both functions provide an option of skipping the background correction step. In this case, the microarrays can be background-corrected independently.

3 RMA++

RMA++procedure

- 1. Fit multiple RMA models on several reference sets and store the normalizing quantiles and the estimated probe effects of these reference sets;
- 2. Calculate the RMA+ measurements of the microarrays of interest for each reference set;
- 3. Average multiple RMA+ measurements of the microarray based on these reference sets.

4 Example

4.1 RMA+

The Dilution dataset in the affydata package consists of 4 microarray samples.

```
> ##Use Dilution in affydata package
> library(RefPlus)
> library(affydata)
```

```
Package LibPath Item
[1,] "affydata" "/home/biocbuild/bbs-3.6-bioc/R/library" "Dilution"
Title
[1,] "AffyBatch instance Dilution"
> data(Dilution)
```

```
> sampleNames(Dilution)
```

[1] "20A" "20B" "10A" "10B"

Firstly, we calculate the RMA measurements of the 4 microarays Ex0:

```
> ##Calculate RMA intensities using the rma function.
```

```
> Ex0<-exprs(rma(Dilution))</pre>
```

Background correcting Normalizing Calculating Expression

Secondly, we form a reference set using the first 3 samples and derive the reference quantiles and the reference probe effects:

```
> ##Background correct, estimate the probe effects, and calculate the
> ##RMA intensities using rma.para function.
> Para<-rma.para(Dilution[,1:3],bg=TRUE,exp=TRUE)
> Ex1 <- Para[[3]]</pre>
```

Then, we calculate the RMA+ measurements of all microarrays Ex2. Figure 1 compares the RMA measurements and the RMA+ measurements of these 4 microarrays.

```
> ##Calculate the RMA+ intensity using rmaplus function.
> Ex2 <- rmaplus(Dilution, rmapara=Para, bg = TRUE)</pre>
```

Use rmapara.

4.2 RMA++

Now, we form another reference set using the 2-4 samples and calculate a new set of RMA+ measurements Ex3.

```
> Para2 <- rma.para(Dilution[,2:4],bg=TRUE,exp=TRUE)
> Ex3 <- rmaplus(Dilution, rmapara=Para2, bg = TRUE)</pre>
```

Use rmapara.

We can then obtain a set of RMA++ measurements by averaging these two sets of RMA+ measurements Ex4. Figure 2 compares the RMA measurements and the RMA++ measurements of these 4 microarrays.

> Ex4 <- (Ex2+Ex3)/2





Figure 1: RMA (Ex0) vs. RMA+ (Ex2).





Figure 2: RMA (Ex0) vs. RMA++ (Ex4).

The root mean squares differences (RMSD) between \mathtt{RMA} measurements and 2 RMA+ measurements, are

> sqrt(mean((Ex0-Ex2)^2))

[1] 0.2138337

> sqrt(mean((Ex0-Ex3)^2))

[1] 0.228387

and the RMSD between RMA measurements and RMA++ measurements is

> sqrt(mean((Ex0-Ex4)^2))

[1] 0.06549345

We can see that the RMA++ measurements can provide a better approximation to the RMA measurements, which is consistant with the comparison between figure 1 and figure 2.

References

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