

A note on esApply

`exprSets` are complex objects. We will think of them as linked arrays: the `exprs` element of an `exprSet` is $G \times N$, where G is the number of genes on a chip and N is the number of tissues analyzed, and the `pData` element of the associated `phenoData` element is $N \times p$, where p is the number of phenotypic or demographic, etc., variables collected.

Abstractly, we are often interested in evaluating functions $f(y; x)$ where y is an N -vector of expression results for a specific gene and x is an N -dimensional structure, coordinated with y , that distinguishes elements of y for processing in the function f . A basic problem is to guarantee that the j th element of y is correctly associated with the j th component of x .

As an example, let's consider `eset` which is an `exprSet` supplied with Biobase. We will print a little report, then the first N -vector of gene expressions and some covariate data:

```
> print(eset)
```

```
Expression Set (exprSet) with
  500 genes
  26 samples
      phenoData object with 3 variables and 26 cases
varLabels
  cov1: Covariate 1; 2 levels
  cov2: Covariate 2; 2 levels
  cov3: Covariate 3; 3 levels
```

```
> print(exprs(eset)[1, ])
```

```
      A      B      C      D      E      F      G      H
192.7420 85.7533 176.7570 135.5750 64.4939 76.3569 160.5050 65.9631
      I      J      K      L      M      N      O      P
56.9039 135.6080 63.4432 78.2126 83.0943 89.3372 91.0615 95.9377
      Q      R      S      T      U      V      W      X
179.8450 152.4670 180.8340 85.4146 157.9890 146.8000 93.8829 103.8550
      Y      Z
64.4340 175.6150
```

```
> print(pData(eset)[1:2, 1:3])
```

```
      cov1 cov2 cov3
A      1      1      1
B      1      1      1
```

Now let's see how expressions and a covariate are related:

```
> print(rbind(exprs(eset[1, ]), cov1 = t(pData(eset))[1, ]))
      A      B      C      D      E      F      G      H
AFFX-MurIL2_at 192.742 85.7533 176.757 135.575 64.4939 76.3569 160.505 65.9631
cov1           1.000  1.0000   1.000   1.000   1.0000   1.0000   1.000   1.0000
      I      J      K      L      M      N      O      P
AFFX-MurIL2_at 56.9039 135.608 63.4432 78.2126 83.0943 89.3372 91.0615 95.9377
cov1           1.0000   1.000   1.0000   1.0000   1.0000   2.0000   2.0000   2.0000
      Q      R      S      T      U      V      W      X
AFFX-MurIL2_at 179.845 152.467 180.834 85.4146 157.989 146.8 93.8829 103.855
cov1           2.000   2.000   2.000   2.0000   2.000   2.0 2.0000   2.000
      Y      Z
AFFX-MurIL2_at 64.434 175.615
cov1           2.000   2.000
```

A function that evaluates the difference in median expression across strata defined using an abstract covariate *x* is

```
> medContr <- function(y, x) {
+   ys <- split(y, x)
+   median(ys[[1]]) - median(ys[[2]])
+ }
```

We can apply this to a small *exprSet* that gives back the data listed above:

```
> print(apply(exprs(eset[1, , drop = F]), 1, medContr, pData(eset)[["cov1"]]))
AFFX-MurIL2_at
-20.7607
```

That's a bit clumsy. This is where `esApply` comes in. We pay for some simplicity by following a strict protocol for the definition of the statistical function to be applied.

```
> medContr1 <- function(y) {
+   ys <- split(y, cov1)
+   median(ys[[1]]) - median(ys[[2]])
+ }
> print(esApply(eset, 1, medContr1)[1])
```

```
AFFX-MurIL2_at
-20.7607
```

The manual page on `esApply` has a number of additional examples that show how applicable functions can be constructed and used. The important thing to note is that the applicable functions *know* the names of the covariates in the `pData` dataframe.

This is achieved by having an environment populated with all the variables in the `phenoData` component of the *exprSet* put in as the environment of the function that will be applied. If that function already has an environment we retain that but in the second position. Thus, there is some potential for variable shadowing.