

A note on esApply

ExpressionSets are complex objects. `exprs(ExpressionSet)` produces $G \times N$, where G is the number of genes on a chip and N is the number of tissues analyzed, and `pData(ExpressionSet)` produces $N \times p$, 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 `sample.ExpressionSet`, which is an *ExpressionSet* supplied with Biobase. We will print a little report, then the first N -vector of gene expressions and some covariate data:

```
> print(sample.ExpressionSet)
```

```
ExpressionSet (storageMode: lockedEnvironment)
assayData: 500 features, 26 samples
  element names: exprs, se.exprs
phenoData
  sampleNames: A, B, ..., Z (26 total)
  varLabels and varMetadata description:
    sex: Female/Male
    type: Case/Control
    score: Testing Score
featureData
  featureNames: AFFX-MurIL2_at, AFFX-MurIL10_at, ..., 31739_at (500 total)
  fvarLabels and fvarMetadata description: none
experimentData: use 'experimentData(object)'
Annotation: hgu95av2
```

```
> print(exprs(sample.ExpressionSet)[1, ])
```

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

```
> print(pData(sample.ExpressionSet)[1:2, 1:3])
```

```

      sex    type score
A Female Control  0.75
B   Male    Case   0.40

```

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

```

> print(rbind(exprs(sample.ExpressionSet[1, ]), sex <- t(pData(sample.ExpressionSet))
+           )))

```

```

      A          B          C          D          E
AFFX-MurIL2_at "192.742" "85.7533" "176.757" "135.575" "64.4939"
               "Female" "Male"    "Male"    "Male"    "Female"
      F          G          H          I          J
AFFX-MurIL2_at "76.3569" "160.505" "65.9631" "56.9039" "135.608"
               "Male"    "Male"    "Male"    "Female" "Male"
      K          L          M          N          O
AFFX-MurIL2_at "63.4432" "78.2126" "83.0943" "89.3372" "91.0615"
               "Male"    "Female" "Male"    "Male"    "Female"
      P          Q          R          S          T
AFFX-MurIL2_at "95.9377" "179.845" "152.467" "180.834" "85.4146"
               "Female" "Female" "Male"    "Male"    "Female"
      U          V          W          X          Y
AFFX-MurIL2_at "157.989" "146.8"   "93.8829" "103.855" "64.434"
               "Male"    "Female" "Male"    "Male"    "Female"
      Z
AFFX-MurIL2_at "175.615"
               "Female"

```

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 *ExpressionSet* that gives back the data listed above:

```

> print(apply(exprs(sample.ExpressionSet[1, ], drop = F),
+           1, medContr, pData(sample.ExpressionSet)[["sex"]]))

```

```

AFFX-MurIL2_at
-12.7935

```

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, sex)  
+   median(ys[[1]]) - median(ys[[2]])  
+ }  
> print(esApply(sample.ExpressionSet, 1, medContr1)[1])
```

```
AFFX-MurIL2_at  
-12.7935
```

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 `phenodata(ExpressionSet)` 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.

1 Session Information

The version number of R and packages loaded for generating the vignette were:

```
R version 2.7.0 (2008-04-22)  
x86_64-unknown-linux-gnu
```

```
locale:
```

```
LC_CTYPE=en_US;LC_NUMERIC=C;LC_TIME=en_US;LC_COLLATE=en_US;LC_MONETARY=C;LC_MESSAGES=en_US
```

```
attached base packages:
```

```
[1] tools      stats      graphics  grDevices  utils      datasets  
[7] methods    base
```

```
other attached packages:
```

```
[1] Biobase_2.0.1
```