

The *mgsa* package

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1 Introduction

Model-based Gene Set Analysis (MGSA, Bauer et al. [1]) is a Bayesian modeling approach for gene set enrichment. The package *mgsa* implements MGSA and tools to use MGSA together with the Gene Ontology [2].

2 Quick start

We start with a small simulated dataset which contains `example_go`, a random subset of yeast gene ontology annotations with 20 terms and `example_o`, a simulated set of observed positive genes. These genes could for example be the "hits" of some screen or a set of differentially expressed genes. In the simulation, the terms GO:0006109 and GO:0030663 were active, implying that genes annotated to these terms were more likely to be observed positives than other genes.

```
> library(mgsa)
> data("example")
> example_go
```

```
Object of class MgsaSets
10 sets over 158 unique items.
```

Set annotations:

	term	definition
GO:0046292	formaldehyde metabol...	The chemical reactio...
GO:0006109	regulation of carboh...	Any process that mod...
GO:0008113	peptide-methionine-(...	Catalysis of the rea...
GO:0016849	phosphorus-oxygen ly...	Catalysis of the cle...
GO:0046527	glucosyltransferase ...	Catalysis of the tra...
...	and 5 other sets.	

Item annotations:

	name
SFA1	Bifunctional enzyme ...
YJL068C	Non-essential intrac...
ADR1	Carbon source-respon...
CAT8	Zinc cluster transcr...
FYV10	Protein of unknown f...
...	and 153 other items.

```
> example_o
```

```
[1] "SFA1"      "ADR1"      "CAT8"      "FYV10"     "GCR1"      "GCR2"      "GID7"
[8] "HAP2"      "HAP3"      "HAP4"      "HAP5"      "PCL10"     "PCL6"      "PCL7"
[15] "PCL8"      "PFK26"     "PFK27"     "PH085"     "PIG1"      "PIG2"      "REG1"
[22] "SIP4"      "SNF1"      "SNF4"      "TYE7"      "UBC8"      "UBP14"     "VID28"
[29] "YLR345W"   "GSC2"      "CCT5"      "CPR6"      "CPR7"      "HSC82"     "PET100"
[36] "TIM9"      "COP1"      "GL03"      "RET2"      "RET3"      "SEC21"     "SEC26"
[43] "SEC27"
```

The method `mgsa` fits the MGSA model. It returns a `MgsaMcmcResults` object whose `print` method displays the most likely active terms. On this example, `mgsa` correctly reports largest posterior probabilities for the terms GO:0006109 and GO:0030663. The call to `set.seed()`, which sets the seed of the random number generator, simply ensures the example of this vignette to be reproducible. It is not required for `mgsa()` to work.

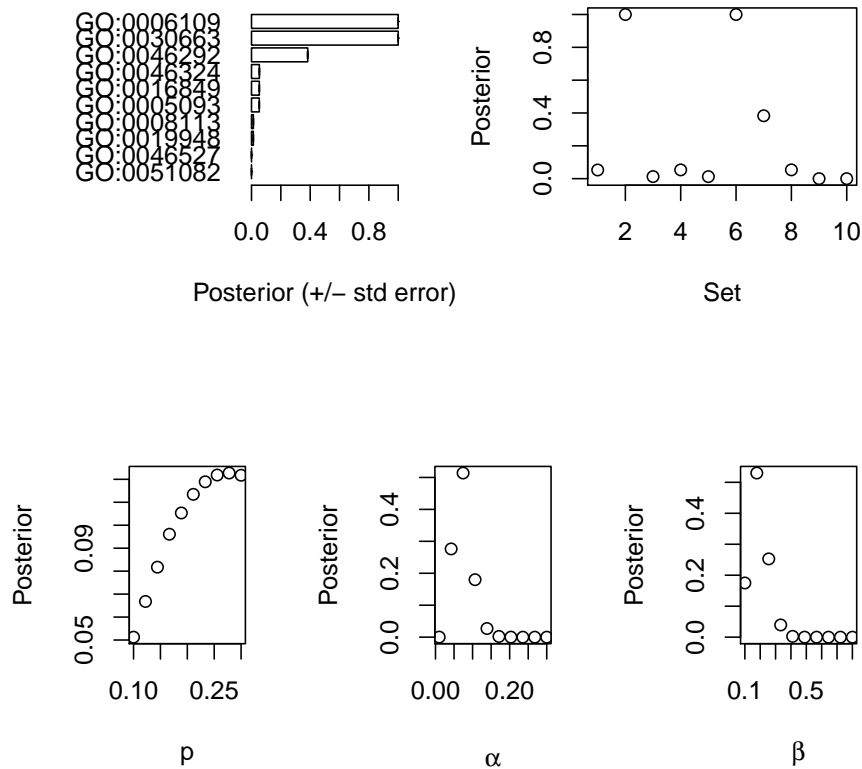
```
> set.seed(0)
> fit = mgsa(example_o, example_go)
> fit
```

```
Object of class MgsaMcmcResults
158 unique elements in population.
43 unique elements both in study set and in population.
'data.frame':      10 obs. of  4 variables:
 $ inPopulation: int   1 34 2 1 2 8 2 1 21 86
 $ inStudySet  : int   0 28 0 0 0 7 1 0 1 6
 $ estimate    : num  0.053 1 0.0128 0.0534 0.0124 ...
 $ std.error   : num  9.94e-04 5.50e-06 3.19e-04 6.18e-04 2.99e-04 ...
NULL
```

```
Posterior on set activity (decreasing order):
      inPopulation inStudySet estimate std.error
GO:0006109         34         28 0.9999814 5.500909e-06
GO:0030663          8          7 0.9999640 1.098180e-05
GO:0046292          2          1 0.3831750 1.474526e-03
GO:0046324          1          0 0.0537238 9.021826e-04
GO:0016849          1          0 0.0533820 6.177710e-04
GO:0005093          1          0 0.0529742 9.942573e-04
GO:0008113          2          0 0.0127592 3.187484e-04
GO:0019948          2          0 0.0124324 2.988773e-04
GO:0046527         21          1 0.0000000 0.000000e+00
GO:0051082         86          6 0.0000000 0.000000e+00
```

The method `plot` provides a graphical visualization of the fit.

```
> plot(fit)
```



3 Using the Gene Ontology

The Gene Ontology [2] (GO) provides structured annotations to genes. Genes with the same annotation constitute a gene set. MGSA can be run on these gene sets. GO annotation files for your organism of study can be downloaded at the GO web page: <http://www.geneontology.org>.

The function `readGAF` creates an `MgsaGoSets` object, a particular `MgsaSets`, from such a gene annotation file. Note that `readGAF` requires the package `GO.db` and `RSQLite` to be installed.

For illustration purposes, a simplified GO annotation file with only three yeast genes is provided:

```
> readGAF(
+   system.file(
+     "example_files/gene_association_head.sgd",
+     package="mgsa"
+   )
+ )
```

Object of class `MgsaGoSets`

111 sets over 3 unique items.

Set annotations:

	term	definition
G0:0000313	organellar ribosome	A ribosome contained...
G0:0000314	organellar small rib...	The smaller of the t...
G0:0000315	organellar large rib...	The larger of the tw...
G0:0003674	molecular_function	Elemental activities...
G0:0003735	structural constitue...	The action of a mole...
... and 106 other sets.		

Item annotations:

	symbol	name
S000004660	AAC1	Mitochondrial inner ...
S000007287	15S_RRNA	Ribosomal RNA of the...
S000007288	21S_RRNA	Mitochondrial 21S rR...

4 Using custom gene sets

MGSA is not restricted to Gene Ontology and can be applied to any gene sets. The method `mgsa` can directly be called on such gene sets provided as `list` as in the example below.

```
> mgsa( c("A", "B"), list(set1=LETTERS[1:3], set2=LETTERS[2:5]) )
```

```
Object of class MgsaMcmcResults
5 unique elements in population.
2 unique elements both in study set and in population.
'data.frame':      2 obs. of  4 variables:
 $ inPopulation: int  3 4
 $ inStudySet  : int  2 1
 $ estimate    : num  0.631 0.154
 $ std.error   : num  0.000654 0.000197
NULL
```

Posterior on set activity (decreasing order):

	inPopulation	inStudySet	estimate	std.error
set1	3	2	0.6312916	0.0006535761
set2	4	1	0.1542472	0.0001973723

Internally, the method `mgsa` indexes all elements of the sets before fitting the model. In case `mgsa` must be run on several observations with the same gene sets, computations can be speeded up by performing this indexing once for all. This can be achieved by building a `MgsaSets`.

```
> myset = MgsaSets( list(set1=LETTERS[1:3], set2=LETTERS[2:5]) )
> mgsa(c("A", "B"), myset)
```

```
Object of class MgsaMcmcResults
5 unique elements in population.
```

```

2 unique elements both in study set and in population.
'data.frame':      2 obs. of  4 variables:
 $ inPopulation: int  3 4
 $ inStudySet   : int  2 1
 $ estimate     : num  0.63 0.154
 $ std.error    : num  0.000594 0.000378
NULL

```

```

Posterior on set activity (decreasing order):
      inPopulation inStudySet estimate std.error
set1             3           2 0.6303232 0.0005937597
set2             4           1 0.1542558 0.0003784498

```

```
> mgsa(c("B", "C"), myset)
```

```

Object of class MgsaMcmcResults
5 unique elements in population.
2 unique elements both in study set and in population.
'data.frame':      2 obs. of  4 variables:
 $ inPopulation: int  3 4
 $ inStudySet   : int  2 2
 $ estimate     : num  0.527 0.289
 $ std.error    : num  0.000543 0.000175
NULL

```

```

Posterior on set activity (decreasing order):
      inPopulation inStudySet estimate std.error
set1             3           2 0.5272310 0.0005425706
set2             4           2 0.2889352 0.0001751067

```

References

- [1] S. Bauer, J. Gagneur and P. N. Robinson. GOing Bayesian: model-based gene set analysis of genome-scale data. *Nucleic acids research*, 2010.
- [2] The Gene Ontology Consortium. Gene Ontology: tool for the unification of biology. *Nature Genetics*, 25:25–29,2000.