

# Generalized and Customizable Sets in R

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## Abstract

This document explains algorithms and basic operations of sets and some generalizations of sets (fuzzy sets, multisets, and fuzzy multisets) available in R through the `sets` package.

There is only rudimentary support in base R for *sets*. Typically, these are represented using atomic or recursive vectors (lists), and one can use operations such as `union()`, `intersect()`, `setdiff()`, `setequal()`, and `is.element()` to emulate set operations. However, there are several drawbacks: first of all, quite a few other operations such as the Cartesian product, the power set, the subset predicate, etc., are missing. Then, the current facilities do not make use of a class system, making extensions hard (if not impossible). Another consequence is that no distinction can be made between sequences (ordered collections of objects) and sets (unordered collections of objects), which is key for the definition of relations, where both concepts are combined. Also, there is no support for extensions such as fuzzy sets or multisets. Therefore, we decided to provide more formalized and extended support for sets, and, because they are needed for Cartesian products, also for tuples.

## 1 Tuples

The *tuple* functions in package `sets` represent basic infrastructure for handling tuples of general (R) objects. They are used, e.g., to correctly represent Cartesian products of sets, resulting in a set of tuples (see below). Although tuple objects should behave like “ordinary” vectors for the most common operations (see examples), some functions may yield unexpected results (e.g., `table()`) or simply not work (e.g., `plot()`) since tuple objects are in fact list objects internally. There are several constructors: `tuple()` for arbitrarily many objects, and `singleton()`, `pair()`, and `triple()` for tuples of lengths 1, 2 and 3, respectively. Note that tuple elements can be named.

```
> ## Do not quote strings
> sets_options("quote", FALSE)
> ## constructor
> tuple(1,2,3, TRUE)

(1, 2, 3, TRUE)

> triple(1,2,3)

(1, 2, 3)

> pair(Name = "David", Height = 185)

(Name = David, Height = 185)

> tuple_is_triple(triple(1,2,3))

[1] TRUE
```

```
> tuple_is_ntuple(tuple(1,2,3,4), 4)
```

```
[1] TRUE
```

```
> ## converter  
> as.tuple(1:3)
```

```
(1L, 2L, 3L)
```

```
> ## operations  
> c(tuple("a","b"), 1)
```

```
(a, b, 1)
```

```
> tuple(1,2,3) * tuple(2,3,4)
```

```
(2, 6, 12)
```

```
> rep(tuple(1,2,3), 2)
```

```
(1, 2, 3, 1, 2, 3)
```

The `Summary()` methods will also work if defined for the elements:

```
> sum(tuple(1, 2, 3))
```

```
[1] 6
```

```
> range(tuple(1, 2, 3))
```

```
[1] 1 3
```

In addition, there is a `tuple_outer()` function to apply functions to all combinations of tuple elements. Note that `tuple_outer()` will also work for regular vectors and thus can really be seen as an extension of `outer()`:

```
> tuple_outer(pair(1, 2), triple(1, 2, 3))
```

```
  1 2 3  
1 1 2 3  
2 2 4 6
```

```
> tuple_outer(1:5, 1:4, "^")
```

```
   1L 2L  3L  4L  
1L  1  1  1  1  
2L  2  4  8 16  
3L  3  9 27 81  
4L  4 16 64 256  
5L  5 25 125 625
```

## 2 Sets

The basic constructor for creating sets is the `set()` function accepting an arbitrary number of R objects as arguments (which can be named). In addition, there is a generic `as.set()` for converting suitable objects to sets.

```

> ## constructor
> s <- set(1, 2, 3)
> s

{1, 2, 3}

> ## named elements
> snamed <- set(one = 1, 2, three = 3)
> snamed

{one = 1, 2, three = 3}

> ## Indexing by label
> snamed[["one"]]

[1] 1

> ## subassignment using element-based indexing
> snamed[c(2,3)] <- c("a", "b")
> snamed

{a, three = b, one = 1}

> ## a more complex set
> set(c, "test", list(1, 2, 3))

{test, <<function>>, <<list(3)>>}

> ## set of sets
> set(set(), set(1))

{ {}, {1} }

> ## conversion functions
> s2 <- as.set(2:5)
> s2

{2L, 3L, 4L, 5L}

```

There are some basic predicate functions (and corresponding operators) defined for the (in)equality, (proper) sub-(super-)set, and element-of. Note that all the `set_is_foo()` functions are vectorized:

```

> set_is_empty(set())

[1] TRUE

> set_is_equal(set(1), set(1))

[1] TRUE

> set(1) == set(1)

[1] TRUE

> set(1) != set(2)

[1] TRUE

> set_is_subset(set(1), set(1, 2))

```

```

[1] TRUE
> set(1) <= set(1, 2)

[1] TRUE
> set(1, 2) >= set(1)

[1] TRUE
> set_is_proper_subset(set(1), set(1))

[1] FALSE
> set(1) < set(1)

[1] FALSE
> set(1, 2) > set(1)

[1] TRUE
> set_contains_element(set(1, 2, 3), 1)

[1] TRUE
> 1 %e% set(1, 2, 3)

[1] TRUE
> set_contains_element(set(1, 2, 3), 1:4)

[1] FALSE FALSE FALSE FALSE
> 1:4 %e% set(1, 2, 3)

[1] FALSE FALSE FALSE FALSE

```

Other than these predicate functions and operators, one can use: `c()` and `|` for the union, `-` for the difference (or complement), `&` for the intersection, `%D%` for the symmetric difference, `*` and `^n` for the ( $n$ -fold) Cartesian product (yielding a set of  $n$ -tuples), and `2^` for the power set. `set_union()`, `set_intersection()`, and `set_syndiff()` accept more than two arguments.<sup>1</sup> The `length` method for sets gives the cardinality. `set_combn()` returns the set of all subsets of specified length. `closure()` and `reduction()` compute the closure and reduction under union, intersection, and symmetric difference. Note that (currently) the `rep()` method for sets will just return its argument since set elements are unique.

```

> length(s)

[1] 3
> length(set())

[1] 0
> ## complement, union, intersection, symmetric difference:
> s - 1

```

---

<sup>1</sup>The  $n$ -ary symmetric difference of a collection of sets consists of all elements contained in an odd number of the sets in the collection.

```

{2, 3}

> s + set("a")

{a, 1, 2, 3}

> s / set("a")

{a, 1, 2, 3}

> s & s2

{}

> s %D% s2

{2L, 3L, 4L, 5L, 1, 2, 3}

> set(1,2,3) - set(1,2)

{3}

> set_intersection(set(1,2,3), set(2,3,4), set(3,4,5))

{3}

> set_union(set(1,2,3), set(2,3,4), set(3,4,5))

{1, 2, 3, 4, 5}

> set_symdiff(set(1,2,3), set(2,3,4), set(3,4,5))

{1, 3, 5}

> ## Cartesian product
> s * s2

{(1, 2L), (1, 3L), (1, 4L), (1, 5L), (2, 2L), (2, 3L), (2, 4L), (2,
  5L), (3, 2L), (3, 3L), (3, 4L), (3, 5L)}

> s * s

{(1, 1), (1, 2), (1, 3), (2, 1), (2, 2), (2, 3), (3, 1), (3, 2), (3,
  3)}

> s ^ 2 # same as above

{(1, 1), (1, 2), (1, 3), (2, 1), (2, 2), (2, 3), (3, 1), (3, 2), (3,
  3)}

> s ^ 3

{(1, 1, 1), (1, 1, 2), (1, 1, 3), (1, 2, 1), (1, 2, 2), (1, 2, 3), (1,
  3, 1), (1, 3, 2), (1, 3, 3), (2, 1, 1), (2, 1, 2), (2, 1, 3), (2, 2,
  1), (2, 2, 2), (2, 2, 3), (2, 3, 1), (2, 3, 2), (2, 3, 3), (3, 1, 1),
  (3, 1, 2), (3, 1, 3), (3, 2, 1), (3, 2, 2), (3, 2, 3), (3, 3, 1), (3,
  3, 2), (3, 3, 3)}

```

```

> ## power set
> 2 ^ s

{ {}, {1}, {2}, {3}, {1, 2}, {1, 3}, {2, 3}, {1, 2, 3} }

> ## subsets:
> set_combn(as.set(1:3),2)

{{1L, 2L}, {1L, 3L}, {2L, 3L}}

> ## closure and reduction (under union):
> cl <- closure(set(set(1), set(2), set(3)))
> print(cl)

{{1}, {2}, {3}, {1, 2}, {1, 3}, {2, 3}, {1, 2, 3}}

> reduction(cl)

{{1}, {2}, {3}}

```

The `Summary()` methods will also work if defined for the elements:

```

> sum(s)

[1] 6

> range(s)

[1] 1 3

```

Using `set_outer()`, it is possible to apply a function on all factorial combinations of the elements of two sets. If only one set is specified, the function is applied to all pairs of this set.

```

> set_outer(set(1, 2), set(1, 2, 3), "/")

  1  2      3
1 1 0.5 0.3333333
2 2 1.0 0.6666667

> X <- set_outer(set(1, 2), set(1, 2, 3), set)
> X[[2, 3]]

{2, 3}

> set_outer(2^set(1, 2, 3), set_is_subset)

      {}  {1}  {2}  {3} {1, 2} {1, 3} {2, 3} {1, 2, 3}
{}      TRUE TRUE TRUE TRUE  TRUE  TRUE  TRUE  TRUE
{1}     FALSE TRUE FALSE FALSE  TRUE  TRUE  FALSE  TRUE
{2}     FALSE FALSE TRUE  FALSE  TRUE  FALSE  TRUE  TRUE
{3}     FALSE FALSE FALSE  TRUE  FALSE  TRUE  TRUE  TRUE
{1, 2}  FALSE FALSE FALSE  FALSE  TRUE  FALSE  FALSE  TRUE
{1, 3}  FALSE FALSE FALSE  FALSE  FALSE  TRUE  FALSE  TRUE
{2, 3}  FALSE FALSE FALSE  FALSE  FALSE  FALSE  TRUE  TRUE
{1, 2, 3} FALSE FALSE FALSE  FALSE  FALSE  FALSE  FALSE  TRUE

```

Because set elements are unordered, it is not allowed to use positional subscripting. However, as shown before, it is possible to use the elements or their labels as index. Further, it is possible to iterate over *all* elements using `for()` and `lapply()/sapply()`:

```

> sapply(s, sqrt)

[1] 1.000000 1.414214 1.732051

> for (i in s) print(i)

[1] 1
[1] 2
[1] 3

```

### 3 Generalized Sets

There are several extensions of “ordinary” sets such as *fuzzy sets* and *multisets*. Both can be seen as special cases of *fuzzy multisets*. For all extensions, the approach is to define a generalized set  $X$  as a pair  $(D, f)$  where  $D$  is an ordinary set representing the domain, and  $f$  the characteristic function of  $X$ , mapping  $D$  to some image  $I$ . The subset of the domain for which  $f$  is non-zero is the support of  $X$ . If  $I = \{0, 1\}$ ,  $X$  represents an “ordinary” set. If  $I = \mathbb{N}$ ,  $X$  becomes a multiset whose elements  $e_i$  can appear multiple times.  $f(e_i)$  is then called the multiplicity of  $e_i$ . If  $I$  is the unit interval,  $X$  becomes a fuzzy set. In this context,  $f$  is typically called the membership function,  $f(e_i)$  the membership grade of  $e_i$ , and  $D$  the universe for  $X$ . If  $I$  is a multiset whose domain is the unit interval (0 excluded),  $X$  is a fuzzy multiset whose elements can each have several (possibly non-unique) membership grades. If for one element, the associated membership grades are all 1, we get a multiset. If there is at most one membership grade, we get a “simple” fuzzy set. If for the latter case the membership is 1, we fall back to an ordinary set.

Generalized sets are created using the `gset()` function. This can be done in four ways:

1. Specify the support only (this yields an ordinary set).
2. Specify support and memberships.
3. Specify universe and membership function.
4. Specify a set of elements along with their individual membership grades.

Note that by default, for efficiency reasons, `gset()` will not store elements with zero membership grades. The specification of an universe is thus only mandatory with membership functions. A default universe can be set using `sets_options()`.

```

> X <- c("A", "B", "C")
> ## ordinary set (X is converted to a set internally).
> gset(support = X)

{A, B, C}

> ## multiset
> multi <- 1:3
> gset(support = X, memberships = multi)

{A [1], B [2], C [3]}

> ## fuzzy set
> ms <- c(0.1, 0.3, 1)
> gset(support = X, memberships = ms)

{A [0.1], B [0.3], C [1]}

```

```

> ## fuzzy set using a membership function
> f <- function(x) switch(x, A = 0.1, B = 0.2, C = 1, 0)
> gset(universe = X, charfun = f)

{A [0.1], B [0.2], C [1]}

> ## fuzzy multiset
> ## Here, the membership argument expects a list of membership grades,
> ## either specified as vectors, or as multisets.
> ms2 <- list(c(0.1, 0.3, 0.4), c(1, 1),
+           gset(support = ms, memberships = multi))
> gset(support = X, memberships = ms2)

{A [{0.1, 0.3, 0.4}], B [{1 [2]}], C [{0.1 [1], 0.3 [2], 1 [3]}]}

```

`gset_support()`, `gset_memberships()`, `gset_height()` and `gset_core()` can be used to retrieve support, memberships, height (maximum membership degree), and the core (elements with membership 1), respectively, of a generalized set. `gset_charfun()` returns a (point-wise defined) characteristic function for a given set. Note that in general, this will be different from the characteristic function possibly used for the creation of the set.

As for ordinary sets, the usual operations such as union, intersection, and complement are available. Additionally, the sum and the difference of sets are defined, which add and subtract multiplicities:

```

> X <- gset(c("A", "B", "C"), 4:6)
> print(X)

{A [4], B [5], C [6]}

> Y <- gset(c("B", "C", "D"), 1:3)
> print(Y)

{B [1], C [2], D [3]}

> ## union vs. sum
> gset_union(X, Y)

{A [4], B [5], C [6], D [3]}

> gset_sum(X, Y)

{A [4], B [6], C [8], D [3]}

> ## intersection vs. difference
> gset_intersection(X, Y)

{B [1], C [2]}

> gset_difference(X, Y)

{A [4], B [4], C [4]}

> ## sum and difference for fuzzy sets
> X <- gset("a", 0.3)
> Y <- gset(c("a", "b"), c(0.3, 0.4))
> gset_sum(X, Y)

{a [0.6], b [0.4]}

```

```
> gset_sum(X, Y, set("a"))
```

```
{a [1], b [0.4]}
```

```
> gset_difference(Y, X)
```

```
{b [0.4]}
```

Note that + and - can be used instead, and that for fuzzy (multi-)sets, in general, complement and difference do not yield the same result (as for crisp sets):

```
> X - Y
```

```
{}
```

```
> gset_complement(X, Y)
```

```
{a [0.3], b [0.4]}
```

`gset_mean()` creates a new set by averaging corresponding memberships using the arithmetic, geometric or harmonic mean. Note that missing elements have 0 membership degree:

```
> x <- gset(1:3, 1:3/3)
```

```
> y <- gset(1:2, 1:2/2)
```

```
> gset_mean(x, y)
```

```
{1L [0.4166667], 2L [0.8333333], 3L [0.5]}
```

```
> gset_mean(x, y, "harmonic")
```

```
{1L [0.4], 2L [0.8]}
```

```
> gset_mean(set(1), set(1, 2))
```

```
{1 [1], 2 [0.5]}
```

The membership vector of a generalized set can be transformed using `gset_transform_memberships()`, using any *vectorized* function:

```
> x <- gset(1:10, 1:10/10)
```

```
> gset_transform_memberships(x, pmax, 0.5)
```

```
{1L [0.5], 2L [0.5], 3L [0.5], 4L [0.5], 5L [0.5], 6L [0.6], 7L [0.7],  
8L [0.8], 9L [0.9], 10L [1]}
```

Note the effect of applying transformations to (multi)sets:

```
> x = gset(1, 2)
```

```
> gset_transform_memberships(x, `*`, 0.5)
```

```
{1 [{0.5 [2]}]}
```

```
> rep(x, 0.5)
```

```
{1}
```

In addition, three convenience functions exist: `gset_concentrate()` and `gset_dilate()` apply the square and the square root function, and `gset_normalize()` normalizes the memberships to a specified maximum:

```

> gset_dilate(y)
{1L [0.7071068], 2L [1]}

> gset_concentrate(y)
{1L [0.25], 2L [1]}

> gset_normalize(y, 0.5)
{1L [0.25], 2L [0.5]}

```

For fuzzy (multi-)sets, the user can choose the logic underlying the operations using the `fuzzy_logic()` function. Fuzzy logics are represented as named lists with four components `N`, `T`, `S`, and `I` containing the corresponding functions for negation, conjunction (“t-norm”), disjunction (“t-conorm”), and implication. The fuzzy logic is selected by calling `fuzzy_logic()` with a character string specifying the fuzzy logic “family”, and optional parameters. Available families include: “Zadeh” (default), “drastic”, “product”, “Lukasiewicz”, “Fodor”, “Frank”, “Hamacher”, “Schweizer-Sklar”, “Yager”, “Dombi”, “Aczel-Alsina”, “Sugeno-Weber”, “Dubois-Prade”, and “Yu”. A call to `fuzzy_logic()` without arguments returns the currently set fuzzy logic.

```

> X <- gset(c("A", "B", "C"), c(0.3, 0.5, 0.8))
> print(X)

{A [0.3], B [0.5], C [0.8]}

> Y <- gset(c("B", "C", "D"), c(0.1, 0.3, 0.9))
> print(Y)

{B [0.1], C [0.3], D [0.9]}

> ## Zadeh logic (default)
> gset_intersection(X, Y)

{B [0.1], C [0.3]}

> gset_union(X, Y)

{A [0.3], B [0.5], C [0.8], D [0.9]}

> gset_complement(X, Y)

{B [0.1], C [0.2], D [0.9]}

> !X

{A [0.7], B [0.5], C [0.2]}

> ## switch logic
> fuzzy_logic("Fodor")
> ## Fodor logic
> gset_intersection(X, Y)

{C [0.3]}

> gset_union(X, Y)

{A [0.3], B [0.5], C [1], D [0.9]}

```

```
> gset_complement(X, Y)
```

```
{D [0.9]}
```

```
> !X
```

```
{A [0.7], B [0.5], C [0.2]}
```

The `cut()` method for generalized sets “filters” all elements with membership not less than a specified level—the result, thus, is a crisp (multi)set:

```
> cut(X, 0.5)
```

```
{B, C}
```

```
> cut(X)
```

```
{}
```

## 4 Characteristic Functions and their Visualization

The `sets` package provides several generators of characteristic functions to be used as templates for the creation of fuzzy sets, including the following shapes: gaussian curve (`fuzzy_normal()`), double gaussian curve (`fuzzy_two_normals()`), bell curve (`fuzzy_bell()`), sigmoid curve (`fuzzy_sigmoid()`), trapezoid (`fuzzy_trapezoid()`), and triangle (`fuzzy_triangular()`, `fuzzy_cone()`).

```
> N <- fuzzy_normal(mean = 0, sd = 1)
```

```
> N(-3:3)
```

```
[1] 0.01110900 0.13533528 0.60653066 1.00000000 0.60653066 0.13533528 0.01110900
```

```
> gset(charfun = N, universe = -3:3)
```

```
{-3L [0.01110900], -2L [0.1353353], -1L [0.6065307], 0L [1], 1L  
 [0.6065307], 2L [0.1353353], 3L [0.01110900]}
```

For convenience, we also provide wrappers that directly generate corresponding sets, given a specified (default) universe:

```
> fuzzy_normal_gset(universe = -3:3)
```

```
{-3L [0.01110900], -2L [0.1353353], -1L [0.6065307], 0L [1], 1L  
 [0.6065307], 2L [0.1353353], 3L [0.01110900]}
```

It is also possible to create function generators for characteristic functions from other functions (such as distribution functions):

```
> fuzzy_poisson <- make_charfun_generator(dpois)
```

```
> gset(charfun = fuzzy_poisson(10), universe = seq(0, 20, 2))
```

```
{0 [0.00036288], 2 [0.018144], 4 [0.1512], 6 [0.504], 8 [0.9], 10 [1],  
 12 [0.7575758], 14 [0.4162504], 16 [0.1734377], 18 [0.05667898], 20  
 [0.01491552]}
```

`make_fuzzy_tuple()` generates a sequence (tuple) of sets based on any of the generating functions (except `fuzzy_trapezoid()` and `fuzzy_triangular()`). The chosen generating function is called with different values (chosen along the universe) passed to the first argument, thus varying the position or the resulting graph:

```
> make_fuzzy_tuple(fuzzy_normal, 5)

(<<gset(201)>>, <<gset(201)>>, <<gset(201)>>, <<gset(201)>>,
 <<gset(201)>>)
```

The `sets` package provides support for visualizing the membership functions of generalized sets, and in particular fuzzy sets. For (fuzzy) multisets, the plot method produces a (grouped) barplot for the membership vector (see Figure 1, top left):

```
> X <- gset(c("A", "B"), list(1:2/2, 0.5))
> plot(X)
```

Characteristic function generators can directly be plotted using a default universe (see Figure 1, top right):

```
> plot(fuzzy_bell)
```

There is a plot method for tuples for visualizing a sequence of sets (see Figure 1, bottom left):

```
> plot(make_fuzzy_tuple(fuzzy_cone, 10), col = gray.colors(10))
```

Plots of several sets can be superposed using the line method (see Figure 1, bottom right):

```
> x <- fuzzy_normal_gset()
> y <- fuzzy_trapezoid_gset(corners = c(5, 10, 15, 17), height = c(0.7,
+ 1))
> plot(tuple(x, y), lty = 3)
> lines(x | y, col = 2)
> lines(gset_mean(x, y), col = 3, lty = 2)
```

## 5 Customizable Sets

Customizable sets extend generalized sets in two ways: First, users can control the way elements are matched, i.e., define equivalence classes of elements. Second, arbitrary iteration orders can be specified.

By default, sets and generalized sets use `identical()` to match elements which is maximal restrictive. Customizable sets can be used to obtain the behavior of `"=="` or `match()`.

```
> ## restore string quoting
> sets_options("quote", TRUE)
> ## default behavior of sets: matching of elements is very strict
> ## Note that on most systems, 3.3 - 2.2 != 1.1
> x <- set("1", 1L, 1, 3.3 - 2.2, 1.1)
> print(x)

{"1", 1L, 1, 1.1, 1.1}

> y <- set(1, 1.1, 2L, "2")
> print(y)

{"2", 2L, 1, 1.1}

> 1L %e% y

[1] FALSE

> set_union(x, y)
```

```

{"1", "2", 1L, 2L, 1, 1.1, 1.1}
> set_intersection(x, y)
{1, 1.1}
> set_complement(x, y)
{"2", 2L}
> ### Now use the more sloppy match()-function (i.e., `==`)
> ### Note that 1 == "1" == 1L ...
> X <- cset(x, matchfun = match)
> print(X)
{"1", 1.1}
> Y <- cset(y, matchfun = match)
> print(Y)
{"2", 1, 1.1}
> 1L %e% Y
[1] TRUE
> cset_union(X, Y)
{"1", "2", 1.1}
> cset_intersection(X, Y)
{"1", 1.1}
> cset_complement(X, Y)
{"2"}
> ## Same using all.equal().
> ## This is a non-vectorized predicate, so use make_matchfun
> ## to generate a vectorized version:
> FUN <- make_matchfun(function(x, y) isTRUE(all.equal(x, y)))
> X <- cset(x, matchfun = FUN)
> print(X)
{"1", 1L, 1.1}
> Y <- cset(y, matchfun = FUN)
> print(Y)
{"2", 2L, 1, 1.1}
> 1L %e% Y
[1] TRUE
> cset_union(X, Y)
{"1", "2", 1L, 2L, 1.1}

```

```
> cset_intersection(X, Y)
```

```
{1L, 1.1}
```

```
> cset_complement(X, Y)
```

```
{"2", 2L}
```

set\_options() can be used to conveniently switch the default match and/or order function if a number of cset objects need to be created.

```
> ### change default functions via sets_option
```

```
> sets_options("matchfun", match)
```

```
> cset(x)
```

```
{"1", 1.1}
```

```
> cset(y)
```

```
{"2", 1, 1.1}
```

```
> cset(1:3) <= cset(c(1,2,3))
```

```
[1] TRUE
```

```
> ### restore package defaults
```

```
> sets_options("matchfun", NULL)
```

In addition, an order function (or permutation index) can be specified for each set for changing the order in which iterators such as as.list() process the elements. The latter in particular influences the labeling and print methods for customizable sets. Sets and generalized sets have a canonical internal ordering which by default is also used for iterations. With customizable sets, a “natural” ordering of elements can be kept.

```
> ## simple example using a permutation index vector
```

```
> cset(letters[1:5], orderfun = 5:1)
```

```
{"e", "d", "c", "b", "a"}
```

```
> ### customized order function
```

```
> FUN <- function(x) order(as.character(x), decreasing = TRUE)
```

```
> Z <- cset(letters[1:5], orderfun = FUN)
```

```
> print(Z)
```

```
{"e", "d", "c", "b", "a"}
```

```
> as.character(Z)
```

```
[1] "a" "b" "c" "d" "e"
```

Note that converters for ordered factors keeps the order:

```
> o <- ordered(c("a", "b", "a"), levels = c("b", "a"))
```

```
> as.set(o)
```

```
{b, a}
```

```
> as.cset(o)
```

```
{a [1], b [2]}
```

Converter for other data types keep order if the elements are unique:

```
> as.cset(c("A", "quick", "brown", "fox"))
```

```
{"A", "quick", "brown", "fox"}
```

```
> as.cset(c("A", "quick", "brown", "fox", "quick"))
```

```
{"A" [1], "brown" [1], "fox" [1], "quick" [2]}
```

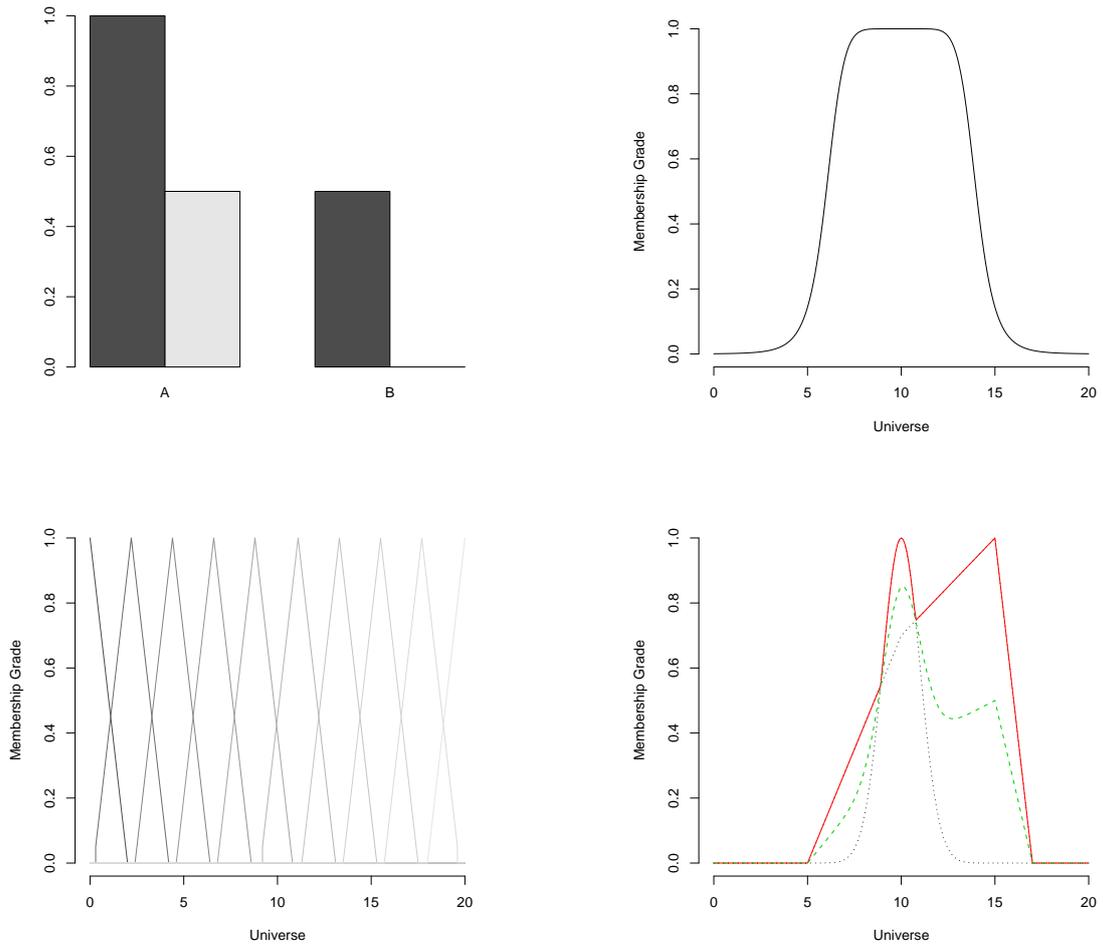


Figure 1: Membership plots for fuzzy sets. Top left: grouped barplot for a fuzzy multiset. Top right: graph of a bell curve. Bottom left: sequence of triangular functions. Bottom right: two combinations of a normal and a trapezoid function (dotted lines: basic shapes; solid (red) line: union; dashed (green) line: arithmetic mean).