# Repairing solutions

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#### 1 Introduction

There are several approaches for including constraints into heuristics, see Chapter 12 of Gilli et al. [2011]. The notes in this vignette give some examples for simple repair mechanisms. These can be called in DEopt, GAopt and PSopt through the repair function; in LSopt/TAopt, they could be included in the neighbourhood function.

```
> set.seed(112233)
> options(digits = 3)
```

## 2 Upper and lower limits

Suppose the solution x is to satisfy all(x >= lo) and all(x <= up), with lo and up being vectors of length(x).

#### 2.1 Setting values to the boundaries

One strategy is to replace elements of x that violate a constraint with the boundary value. Such a repair function can be implemented very concisely. An example:

```
> up <- rep(1, 4L)
> lo <- rep(0, 4L)
> x <- rnorm(4L)
> x

[1] 2.127 -0.380  0.167  1.600
```

Three of the elements of x actually violate the constraints.

```
> repair1a <- function(x,up,lo) pmin(up,pmax(lo,x))
> x
[1] 2.127 -0.380  0.167  1.600

> repair1a(x, up, lo)
[1] 1.000 0.000 0.167 1.000
```

We see that indeed all values greater than 1 are replaced with 1, and those smaller than 0 become 0. Two other possibilities that achieve the same result:

```
> repair1b <- function(x, up, lo) {
    ii <- x > up
    x[ii] <- up[ii]
    ii <- x < lo
    x[ii] <- lo[ii]
    y</pre>
```

The function repair1b uses comparisons to replace only the relevant elements in x. The function repair1c uses the 'trick' that

$$pmax(x,y) = \frac{x+y}{2} + \left| \frac{x-y}{2} \right|,$$

$$pmin(x,y) = \frac{x+y}{2} - \left| \frac{x-y}{2} \right|.$$

> repair1a(x, up, lo)

```
[1] 1.000 0.000 0.167 1.000
```

> repair1b(x, up, lo)

```
[1] 1.000 0.000 0.167 1.000
```

> repair1c(x, up, lo)

```
[1] 1.000 0.000 0.167 1.000
```

- > trials <- 10000L
- > strials <- seq\_len(trials)
- > system.time(for(i in strials) y1 <- repair1a(x, up, lo))</pre>

```
user system elapsed 0.184 0.000 0.183
```

> system.time(for(i in strials) y2 <- repair1b(x, up, lo))

```
user system elapsed 0.068 0.000 0.071
```

> system.time(for(i in strials) y3 <- repair1c(x, up, lo))

```
user system elapsed 0.040 0.000 0.041
```

The third of these functions would also work on matrices if up or lo were scalars.

```
> X <- array(rnorm(25L), dim = c(5L, 5L))
> X
```

```
[,1] [,2] [,3] [,4] [,5]

[1,] 0.1962 0.434 -2.155 -1.5881 -1.029

[2,] 0.2284 1.231 0.975 0.0682 1.818

[3,] -1.1492 0.580 -0.711 -0.4457 -1.315

[4,] -0.0712 0.246 0.628 1.4662 0.511

[5,] -0.5619 0.388 -0.136 -0.8412 1.337
```

> repair1c(X, up = 0.5, lo = -0.5)

```
[,1] [,2] [,3] [,4] [,5]

[1,] 0.1962 0.434 -0.500 -0.5000 -0.5

[2,] 0.2284 0.500 0.500 0.0682 0.5

[3,] -0.5000 0.500 -0.500 -0.4457 -0.5

[4,] -0.0712 0.246 0.500 0.5000 0.5

[5,] -0.5000 0.388 -0.136 -0.5000 0.5
```

The considerable speedup comes at a price, of course, since there is no checking (eg, for NA values) in repair1b and repair1c. We could also define new functions pmin2 and pmax2.

```
> pmax2 < -function(x1, x2) ((x1 + x2) + abs(x1 - x2)) / 2
> pmin2 <- function(x1, x2) ( (x1 + x2) - abs(x1 - x2) ) / 2
A test follows.
> x1 <- rnorm(100L)
> x2 <- rnorm(100L)
> t1 <- system.time(for (i in strials) z1 <- pmax(x1,x2) )
> t2 <- system.time(for (i in strials) z2 <- pmax2(x1,x2))
> t1[[3L]]/t2[[3L]] ## speedup
[1] 3.97
> all.equal(z1, z2)
[1] TRUE
> t1 <- system.time(for (i in strials) z1 <- pmin(x1,x2) )</pre>
> t2 <- system.time(for (i in strials) z2 <- pmin2(x1,x2))
> t1[[3L]]/t2[[3L]] ## speedup
[1] 4.09
> all.equal(z1, z2)
[1] TRUE
```

One downside of this repair mechanism is that a solution may quickly become stuck at the boundaries (but of course, in some cases this is exactly what we want).

### 2.2 Reflecting values into the feasible range

The function repair2 reflects a value that is too large or too small around the boundary. It restricts the change in a variable x[i] to the range up[i] - lo[i].

```
> repair2 <- function(x,up,lo) {
    done <- TRUE
    e <- sum(x - up + abs(x - up) + lo - x + abs(lo - x))
    if (e > 1e-12) done <- FALSE
    r <- up - lo
    while (!done) {
        adjU <- x - up
        adjU <- adjU + abs(adjU)
        adjU <- adjU + r - abs(adjU - r)</pre>
```

```
adjL <- lo - x
         adjL <- adjL + abs(adjL)
         adjL \leftarrow adjL + r - abs(adjL - r)
         x \leftarrow x - (adjU - adjL)/2
         e \leftarrow sum(x - up + abs(x - up) + lo - x + abs(lo - x))
         if (e < 1e-12) done <- TRUE
     }
     Х
}
> x
     2.127 -0.380 0.167 1.600
[1]
> repair2(x, up, lo)
[1] 0.873 0.380 0.167 0.600
> system.time(for (i in strials) y4 <- repair2(x,up,lo))
   user system elapsed
  0.284
           0.000
                    0.282
```

#### 2.3 Adjusting a cardinality limit

Let x be a logical vector.

```
> T <- 20L
> x <- logical(T)
> x[runif(T) < 0.4] <- TRUE
> x

[1] FALSE TRUE TRUE FALSE TRUE TRUE FALSE FALSE
```

Suppose we want to impose a minimum and maximum cardinality, kmin and kmax.

```
> kmax <- 5L
> kmin <- 3L
```

We could use an approach like the following (for the definition of resample, see ?sample):

```
> printK <- function(x)</pre>
    cat(paste(ifelse(x, "o", "."), collapse = ""),
        "-- cardinality", sum(x), "\n")
For kmax:
> for (i in 1:10) {
    if (i==1L) printK(x)
    x1 <- repairK(x, kmax, kmin)</pre>
    printK(x1)
}
.oo.oo.....oo..o -- cardinality 8
.oo.o.....oo..... -- cardinality 5
.o...o -- cardinality 5
.o..o....o...o...o -- cardinality 5
..o.oo.....o -- cardinality 5
.o...o....oo..o... -- cardinality 5
....oo.....oo..o... -- cardinality 5
.o..oo.....o -- cardinality 5
.oo..o.....oo..... -- cardinality 5
.oo..o....o... -- cardinality 5
For kmin:
> x <- logical(T); x[10L] <- TRUE
> for (i in 1:10) {
    if (i==1L) printK(x)
    x1 <- repairK(x, kmax, kmin)</pre>
    printK(x1)
..... -- cardinality 1
..... -- cardinality 3
o.....o....o.... -- cardinality 3
...o....o..o..... -- cardinality 3
..... -- cardinality 3
....o....oo...... -- cardinality 3
....o...oo..... -- cardinality 3
.....o.o......o -- cardinality 3
.....o.o..o..... -- cardinality 3
....o...o...o.... -- cardinality 3
....o....o.... -- cardinality 3
```

#### References

Manfred Gilli, Dietmar Maringer, and Enrico Schumann. *Numerical Methods and Optimization in Finance*. Elsevier, 2011.