



Efficient R programming the rolygon example

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- This brief tutorial illustrates how to combine S4 object oriented capabilities with function closures in order to develop classes with built in methods.
- In practice, we want to write highly reusable code in order to increase development and maintenance efficiency
- Finally, a great thank to Hadley Wickham for the great contribution of material and tutorials made available on the web and to Bill Venables and Stefano lacus for their kind support.



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As Wikipedia states: In Euclidean geometry, a regular polygon is a polygon that is equiangular (all angles are equal in measure) and equilateral (all sides have the same length). Square, pentagon, hexagon are regular polygons.

Within R we would like to have simple functions like:

> e1 <- heptagon(s = 1)
> plot(e1)





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> setClass("heptagon", representation(s = "numeric"))

Previous call is the result of a simple S4 methods and classes implementation:

```
> heptagon <- function(s){new("heptagon", s=s)}</pre>
> setMethod(f = "plot", signature = "heptagon",
          definition = function(x, y){
            object <- x
            s <- object@s
            n <- 7
            pi <- base::pi
            rho <- (2*pi)/n
            h <- .5*s*tan((pi/2)-(pi/n))
            r <- sqrt(h^2+(s/2)^2)
            sRho <- ifelse(n %% 2 == 0, (pi/2- rho/2), pi/2)</pre>
            cumRho <- cumsum(c(sRho, rep(rho, n)))</pre>
            cumRho <- ifelse(cumRho > 2*pi, cumRho-2*pi, cumRho)
            x \leq r * \cos(cum Rho)
            v <- r*sin(cumRho)</pre>
            par(ptv = "s")
            plot(x, y, type = "n", xlab = "", ylab = "")
            lines(x, y, col = "red", lwd = 2)
            points(0, 0, pch = 16, col = "red")
            grid()
            invisible(NULL)
```



With this mind set we would have to define a new class, function and method for each polygon ...

```
> setClass("penthagon", representation(s = "numeric"))
                                                                           p1 <- penthagon(s=1)
                                                                           plot(p1)
> penthagon <- function(s){new("penthagon", s=s)}</pre>
> setMethod(f = "plot", signature = "penthagon",
                                                                          2
          definition = function(x, y){
            object <- x
            s <- object@s
                                                                           8
            n <- 5
                                                                          9.9
                                                                          10
            invisible(NULL)
          3
)
                                                                             -15 -10 -05 00
                                                                                            0.5
                                                                                                  1.5
```

This could become a quite boring job. Moreover, despite the number of cases we may take into account, we are pretty much sure that sooner or later we will need something more ... a hendecagon (11 sides) or a enneadecagon (19 sides).



We could accept some compromises and write a generic schema:

```
> setClass("rolygon",
representation(n = "numeric", s = "numeric"))
> rolygon = function(n, s){new("rolygon", n= n, s=s)}
> setMethod(f = "plot", signature = "rolygon",
definition = function(x, y){
object = x
s = object@s
n = object@n
invisible(NULL)
> h11 = rolygon(n = 11, s = 1)
> plot(h11)
```

but this is not what we wanted we do want:

```
> h11 = hendecagon(s = 1)
> plot(h11)
```





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- Any time a function is called, a new environment is created, whose enclosure is the environment where the function is defined. The computation, as expressed by the body of the function, occurs in the newly created environment.
- Thus, whenever we call a function we have at least two environments: the environment the function was defined in and the environment where the function evaluation takes place.
- ★ By using this idea, we can define a function f() that returns a function g().
- As g() is created within the evaluation environment of f(), this last environment is the enclosure of g(). Therefore, g() remembers all symbols bound to that environment.



As a practical application of this idea let's consider a function f()that returns a function g():

```
> f <- function(x) {
   g = function(y){x+y}
   g
}</pre>
```

As g() is created within the evaluation environment of f(), g() "remembers" the value of x. Therefore we can define a simple function f1() that adds one to the given y argument

```
> f1 <- f(x = 1) ; f1(y = 3)
[1] 4
```

Note that f1() remembers the value of x The environment of f1() can be directly accessed and manipulated:

```
> ls(env=environment(f1))
[1] "g" "x"
> get("x", env=environment(f1))
[1] 1
```

```
> environment(f1)$x <- 0
> f1(1)
[1] 1
```

The same exercise apply to any fx() such as f99

```
> f99 <- f(99)
> f99(y = 1)
[1] 100
```



Function Closures

An other example consists of a simple estimate() that generate specific 1() functions for maximum likelihood estimates

```
estimate = function(dist, theta){
  estimate = function(x){
   neglik = function(theta = theta , x = x, log = T){
    args = c(list(x), as.list(theta), as.list(log))
   neglik = -sum(do.call(dist, args))
   neglik
  }
  optim(par = theta, fn = neglik , x = x)
  }
estimate}
```

Once we have estimate(), we can use it to define any 1() function as long as its d() exists

That is, we can now write a lnorm() that computes mle estimate as simply as:

```
lnorm = estimate("dnorm", theta = c(mean(x), sd(x))
```

lnorm() is a new function that can
be used as:

```
x = rnorm(100, 7 , 2)
lnorm(x)$par
[1] 6.803764 1.704783
```

Similarly, for a poison distribution:

```
lpois = estimate("dpois", theta = c(mean(x)))
events = rpois(1000, lambda = 22)
lpois(events)$par
[1] 22.02248
```



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- The combination of the two previous ideas allows quite interesting coding techniques.
- * We define a rolygon() function that returns a generic f() capable of generating specific regular polygons with plot method inherited from rolygon's environment:

```
> rolygon <- function(n) {
    # Define rolygon class
    setClass("rolygon",
    representation(n = "numeric", s = "numeric"))</pre>
```

```
# Define a plot method for
#object of class rolygon
setMethod(f = "plot", signature = "rolygon",
definition = function(x, y){
        object <- x
        s <- object@s
        n <- object@n
        ...
        invisible(NULL)
})
# Define a function that returns an object
# of class rolygon
f <- function(s){new("rolygon", n = n, s = s)}</pre>
```

```
# Return the newly created function
return(f)
```



}

Now, we can easily define any polygon we need with no extra coding

```
heptagon <- rolygon(n = 7)
e1 <- heptagon(1)
plot(e1)</pre>
```

circumference <- rolygon(n = 10^4)
plot(circumference(s = base::pi/10^4))</pre>





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