Custom Moment and Objective Functions

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Abstract

The purpose of this vignette is to demonstrate how to write and use custom moment functions and custom objective functions to solve complex optimization problems.

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1 Getting Started

1.1 Load Packages

Load the necessary packages.

```
library(PortfolioAnalytics)

## Loading required package: zoo

##

## Attaching package: 'zoo'

##

## The following objects are masked from 'package:base':
```

```
##
## as.Date, as.Date.numeric
##
## Loading required package: xts
## Loading required package: PerformanceAnalytics
##
## Attaching package: 'PerformanceAnalytics'
##
## The following object is masked from 'package:graphics':
##
## legend
```

1.2 Data

The edhec data set from the PerformanceAnalytics package will be used as data for the following examples.

```
data(edhec)

# Use the first 4 columns in edhec for a returns object
R <- edhec[, 1:4]
colnames(R) <- c("CA", "CTAG", "DS", "EM")
head(R, 5)

## CA CTAG DS EM
## 1997-01-31 0.0119 0.0393 0.0178 0.0791
## 1997-02-28 0.0123 0.0298 0.0122 0.0525
## 1997-03-31 0.0078 -0.0021 -0.0012 -0.0120
## 1997-04-30 0.0086 -0.0170 0.0030 0.0119
## 1997-05-31 0.0156 -0.0015 0.0233 0.0315

# Get a character vector of the fund names
funds <- colnames(R)</pre>
```

2 Setting the Portfolio Moments

The Portfolio Analytics framework to estimate solutions to constrained optimization problems is implemented in such a way that the moments of the returns are calculated only once and then used in lower level optimization functions. The set.portfolio.moments function computes the first, second, third, and fourth moments depending on the objective function(s) in the portfolio object. For example, if the third and fourth moments do not need to be calculated for a given objective, then set.portfolio.moments will try to detect this and not compute those moments. Currently, set.portfolio.moments implements methods to compute moments based on sample estimates, higher moments from fitting a statistical factor model based on the work of Kris Boudt (Boudt et al., 2014), the Black Litterman model (Meucci, 2008b), and the Fully Flexible Framework based on the work of Attilio Meucci (Meucci, 2008a).

```
# Construct initial portfolio with basic constraints.
init.portf <- portfolio.spec(assets=funds)
init.portf <- add.constraint(portfolio=init.portf, type="full_investment")
init.portf <- add.constraint(portfolio=init.portf, type="long_only")

# Portfolio with standard deviation as an objective

SD.portf <- add.objective(portfolio=init.portf, type="risk", name="StdDev")

# Portfolio with expected shortfall as an objective

ES.portf <- add.objective(portfolio=init.portf, type="risk", name="ES")</pre>
```

Here we see the names of the list object that is returned by set.portfolio.moments.

```
sd.moments <- set.portfolio.moments(R, SD.portf)
names(sd.moments)

## [1] "mu" "sigma"

es.moments <- set.portfolio.moments(R, ES.portf)
names(es.moments)

## [1] "mu" "sigma" "m3" "m4"</pre>
```

3 Custom Moment Functions

In many cases for constrained optimization problems, one may want to estimate moments for a specific use case or further extend the idea of set.portfolio.moments. A user defined custom moment function can have any arbitrary named arguments. However, arguments named R for the asset returns and portfolio for the portfolio object will be detected automatically and handled in an efficient manner. Because of this, it is strongly encouraged to use R for the asset returns object and portfolio for the portfolio object.

The moment function should return a named list object where the elements represent the moments:

```
$mu first moment; expected returns vector
$sigma second moment; covariance matrix
$m3 third moment; coskewness matrix
$m4 fourth moment; cokurtosis matrix
```

The lower level optimization functions expect an object with the structure described above. List elements with the names mu, sigma, m3, and m4 are matched automatically and handled in an efficient manner.

Here we define a function to estimate the covariance matrix using a robust method.

```
sigma.robust <- function(R){
  out <- list()
  set.seed(1234)
  out$sigma <- MASS::cov.rob(R, method="mcd")$cov
  return(out)
}</pre>
```

Now we can use the custom moment function in optimize.portfolio to estimate the solution to the minimum standard deviation portfolio.

```
## Default solver: ROI_NULL.
##
## Attaching package: 'ROI'
##
## The following object is masked from 'package:PortfolioAnalytics':
##
##
     objective
opt.sd
```

Here we extract the weights and compute the portfolio standard deviation to verify that the the robust estimate of the covariance matrix was used in the optimization.

```
weights <- extractWeights(opt.sd)
sigma <- sigma.robust(R)$sigma

sqrt(t(weights) %*% sigma %*% weights)

## [,1]
## [1,] 0.008646

extractObjectiveMeasures(opt.sd)$StdDev</pre>
```

```
## StdDev
## 0.008646
```

4 Custom Objective Functions

A key feature of PortfolioAnalytics is that the name for an objective can be any valid Rfunction. PortfolioAnalytics was designed to be flexible and modular, and custom objective functions are a key example of this.

Here we define a very simple function to compute annualized standard deviation for monthly data that we will use as an objective function.

```
pasd <- function(R, weights, sigma, N=36){
   R <- tail(R, N)
   tmp.sd <- sqrt(as.numeric(t(weights) %*% sigma %*% weights))
   sqrt(12) * tmp.sd
}</pre>
```

A few guidelines should be followed for defining a custom objective function.

- The objective function must return a single value for the optimizer to minimize.
- It is strongly encouraged to use the following argument names in the objective function:

```
R for the asset returns
weights for the portfolio weights
```

These argument names are detected automatically and handled in an efficient manner. Any other arguments for the objective function can be for the moments or passed in through the arguments list in the objective.

For our pasd function, we need custom moments function to return a named list with sigma as an element. We can use the sigma.robust function we defined in the previous section. Here we construct a portfolio with our pasd function as an objective to minimize.

```
# Construct initial portfolio with basic constraints.
pasd.portf <- portfolio.spec(assets=funds)
pasd.portf <- add.constraint(portfolio=pasd.portf, type="full_investment")
pasd.portf <- add.constraint(portfolio=pasd.portf, type="long_only")</pre>
```

Now we can run the optimization to estimate a solution to our optimization problem.

```
opt.pasd <- optimize.portfolio(R, pasd.portf,</pre>
                             optimize_method="DEoptim",
                             search_size=5000, trace=TRUE, traceDE=0,
                             momentFUN="sigma.robust")
##
## DEoptim package
## Differential Evolution algorithm in R
## Authors: D. Ardia, K. Mullen, B. Peterson and J. Ulrich
##
## Leverage constraint min_sum and max_sum are restrictive,
##
               consider relaxing. e.g. 'full_investment' constraint should be min_sum=0.99
and max\_sum=1.01
## [1] 0.566 0.112 0.322 0.000
opt.pasd
## **********
## PortfolioAnalytics Optimization
## *************
##
## Call:
## optimize.portfolio(R = R, portfolio = pasd.portf, optimize_method = "DEoptim",
      search_size = 5000, trace = TRUE, traceDE = 0, momentFUN = "sigma.robust")
##
##
## Optimal Weights:
     CA CTAG
## 0.566 0.112 0.322 0.000
```

##

```
## Objective Measures:
## pasd
## 0.03043
```

We now consider an example with a more complicated objective function. Our objective to maximize the fourth order expansion of the Constant Relative Risk Aversion (CRRA) expected utility function as in (Boudt et al., 2014).

$$EU_{\lambda}(w) = -\frac{\lambda}{2}m_{(2)}(w) + \frac{\lambda(\lambda+1)}{6}m_{(3)}(w) - \frac{\lambda(\lambda+1)(\lambda+2)}{24}m_{(4)}(w)$$

Here we define a function to compute CRRA estimate. Note how we define the function to use sigma, m3, and m4 as arguments that will use the output from a custom moment function. We could compute the moments inside this function, but re-computing the moments potentially tens of thousands of times (i.e. at each iteration) can be very compute intensive.

```
CRRA <- function(R, weights, lambda, sigma, m3, m4){
    weights <- matrix(weights, ncol=1)
    M2.w <- t(weights) %*% sigma %*% weights
    M3.w <- t(weights) %*% m3 %*% (weights %x% weights)
    M4.w <- t(weights) %*% m4 %*% (weights %x% weights %x% weights)
    term1 <- (1 / 2) * lambda * M2.w
    term2 <- (1 / 6) * lambda * (lambda + 1) * M3.w
    term3 <- (1 / 24) * lambda * (lambda + 1) * (lambda + 2) * M4.w
    out <- -term1 + term2 - term3
    out
}</pre>
```

We now define the custom moment function to compute the moments for the objective function.

```
crra.moments <- function(R, ...){
  out <- list()
  out$sigma <- cov(R)
  out$m3 <- PerformanceAnalytics:::M3.MM(R)
  out$m4 <- PerformanceAnalytics:::M4.MM(R)
  out
}</pre>
```

Finally, we set up the portfolio and run the optimization using our custom moment function and objective function to maximize CRRA. Note that type="return" is used to maximize an objective function.

```
opt.crra <- optimize.portfolio(R, crra.portf, optimize_method="DEoptim",</pre>
                               search_size=5000, trace=TRUE, traceDE=0,
                               momentFUN="crra.moments")
## [1] 0.1940 0.3600 0.3860 0.0539
opt.crra
## ***********
## PortfolioAnalytics Optimization
## ***********
##
## Call:
## optimize.portfolio(R = R, portfolio = crra.portf, optimize_method = "DEoptim",
##
      search_size = 5000, trace = TRUE, traceDE = 0, momentFUN = "crra.moments")
##
## Optimal Weights:
      CA
          CTAG
                   DS
                          EM
##
## 0.1940 0.3600 0.3860 0.0539
##
```

```
## Objective Measures:
## CRRA
## -0.001083
```

PortfolioAnalytics supports several methods to estimate moments as well as user defined moment functions. The name of the objective must be the name of a valid Rfunction and PortfolioAnalytics integrates well with PerformanceAnalytics to utilize several of the risk measure functions such as StdDev and ES. Because an objective function can be a valid Rfunction, user defined objective functions are supported. The modular framework of PortfolioAnalytics allows one to easily define custom moment functions and objective functions as valid Rfunctions to solve complex and specialized objective functions.

References

- K. Boudt, W. Lu, and B. Peeters. Higher order comments of multifactor models and asset allocation. June 2014. URL http://papers.srn.com/sol3/papers.cfm?abstract_id=2409603.
- A. Meucci. Fully flexible views: Theory and practice. *Journal of Risk*, 21(10):97–102, 2008a. URL http://papers.ssrn.com/sol3/papers.cfm?abstract_id=1213325.
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