



The R-to-MOSEK Optimization Interface

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The R Project

A fast open-source programming language for technical computing and graphics.

Highlights:

- One million users – Intel Capital, 2009
- The Comprehensive R Archive Network (3895 packages and counting)
- Direct interfaces to C and Fortran (as well as C++, Perl, Java, Python, Matlab, Excel etc.)
- Revolution Analytics
- RStudio

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MOSEK

A high performance software library designed to solve large-scale convex optimization problems.

Highlights:

- Linear constraints (LP)
- Second-order cone constraints (SOCP)
- Semidefinite constraints (SDP)
- Mixed-integer variables (MIP)
- Quadratic terms (QP and QCQP)
- Separable convex operators – $\log(x_i)$, $\exp(x_i)$, ...

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Rmosek

A simple interface from R to the optimizers of MOSEK.

Why?

- Some financial customers find R better than MATLAB.
- Replace unofficial implementations.

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Rmosek

A simple interface from R to the optimizers of MOSEK.

Why?

- Some financial customers find R better than MATLAB.
- Replace unofficial implementations.

Goals:

- Open-source
- Highly effective
- Integrate with what R users normally do

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A oneliner for Windows, MacOS and Linux:

```
install.packages( "Rmosek" , type="source" )
```

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A oneliner for Windows, MacOS and Linux:

```
install.packages("Rmosek", type="source")
```

Package overview:

```
library(help="Rmosek")
```

Major functions:

- `mosek(prob)`
- `mosek_write(prob, file)`
- `mosek_read(file)`

The problem is an R-variable.. Create as you like!

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$$\begin{array}{ll}
 \text{maximize} & 3x_1 + 1x_2 + 5x_3 + 1x_4 \\
 \text{subject to} & 3x_1 + 1x_2 + 2x_3 = 30, \\
 & 2x_1 + 1x_2 + 3x_3 + 1x_4 \geq 15, \\
 & 2x_2 + 3x_4 \leq 25, \\
 & 0 \leq x_1 \leq \infty, \\
 & 0 \leq x_2 \leq 10, \\
 & 0 \leq x_3 \leq \infty, \\
 & 0 \leq x_4 \leq \infty.
 \end{array}$$

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```
lo1 <- list()

## Objective
lo1$sense <- "max"
lo1$c      <- c(3, 1, 5, 1)

## Constraint matrix
lo1$A <- spMatrix(nrow = 3, ncol = 4,
  i = c(1, 2, 1, 2, 3, 1, 2, 2, 3),
  j = c(1, 1, 2, 2, 2, 3, 3, 4, 4),
  x = c(3, 2, 1, 1, 2, 2, 3, 1, 3)
)
```

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```
## Constraint bounds
lo1$bc <- cbind(c(30, 30),
                c(15, Inf),
                c(-Inf, 25))
```

```
## Variable bounds
lo1$bx <- cbind(c(0, Inf),
                c(0, 10),
                c(0, Inf),
                c(0, Inf))
```

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$$\begin{aligned}
 &\text{minimize} && \begin{bmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{bmatrix} \bullet \bar{x}_1 + x_1 \\
 &\text{subject to} && \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_1 &= 1.0, \\
 &&& \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_2 + x_3 &= 0.5, \\
 &&& (x_1, x_2, x_3) \in \mathbb{R}^3, \quad x_1 \geq \sqrt{x_2^2 + x_3^2}, \quad \bar{x}_1 \succeq 0.
 \end{aligned}$$

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$$\begin{aligned}
 &\text{minimize} && \begin{bmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{bmatrix} \bullet \bar{x}_1 + \mathbf{x}_1 \\
 &\text{subject to} && \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_1 &= 1.0, \\
 &&& \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_2 + x_3 &= 0.5, \\
 &&& (x_1, x_2, x_3) \in \mathbb{R}^3, \quad x_1 \geq \sqrt{x_2^2 + x_3^2}, \quad \bar{x}_1 \succeq 0.
 \end{aligned}$$

```
sdo1 <- list()
```

```
sdo1$sense <- "minimize"
```

```
sdo1$c <- c(1,0,0)
```

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$$\begin{aligned}
 &\text{minimize} && \begin{bmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{bmatrix} \bullet \bar{x}_1 + x_1 \\
 &\text{subject to} && \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \bullet \bar{x}_1 + \mathbf{x}_1 = 1.0, \\
 &&& \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \bullet \bar{x}_1 + \mathbf{x}_2 + \mathbf{x}_3 = 0.5, \\
 &&& (x_1, x_2, x_3) \in \mathbb{R}^3, \quad x_1 \geq \sqrt{x_2^2 + x_3^2}, \quad \bar{x}_1 \succeq 0.
 \end{aligned}$$

```

sdo1$A <- spMatrix(nrow=2, ncol=3,
  i = c(1,2,2),
  j = c(1,2,3),
  x = c(1,1,1)
)

```


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 &\text{subject to} && \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_1 &= 1.0, \\
 &&& \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_2 + x_3 &= 0.5, \\
 &&& (x_1, x_2, x_3) \in \mathbb{R}^3, \quad x_1 \geq \sqrt{x_2^2 + x_3^2}, \quad \bar{x}_1 \succeq 0.
 \end{aligned}$$

```

sdo1$bc <- cbind(c(1.0, 1.0),
                  c(0.5, 0.5))
    
```

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 &\text{subject to} && \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_1 &= 1.0, \\
 &&& \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_2 + x_3 &= 0.5, \\
 &&& (\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) \in \mathbb{R}^3, \quad x_1 \geq \sqrt{x_2^2 + x_3^2}, \quad \bar{x}_1 \succeq 0.
 \end{aligned}$$

```

sdo1$bx <- cbind(c(-Inf, Inf),
                  c(-Inf, Inf),
                  c(-Inf, Inf))
    
```

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 \end{aligned}$$

```

sdo1$cones <- cbind(
  list("quad", c(1,2,3))
)

```

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 &\text{subject to} && \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_1 &= 1.0, \\
 &&& \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_2 + x_3 &= 0.5, \\
 &&& (x_1, x_2, x_3) \in \mathbb{R}^3, \quad x_1 \geq \sqrt{x_2^2 + x_3^2}, \quad \bar{\mathbf{x}}_1 \succeq \mathbf{0}.
 \end{aligned}$$

```
sdol$bardim <- c(3)
```

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 &\text{minimize} && \begin{bmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{bmatrix} \bullet \bar{x}_1 + x_1 \\
 &\text{subject to} && \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_1 &= 1.0, \\
 &&& \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_2 + x_3 &= 0.5, \\
 &&& (x_1, x_2, x_3) \in \mathbb{R}^3, \quad x_1 \geq \sqrt{x_2^2 + x_3^2}, \quad \bar{x}_1 \succeq 0.
 \end{aligned}$$

```

# Block triplet format (SDPA)
sdo1$barc$j <- c(1,1,1,1,1)

sdo1$barc$k <- c(1,2,2,3,3)
sdo1$barc$l <- c(1,1,2,2,3)
sdo1$barc$v <- c(2,1,2,1,2)

```

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 &\text{subject to} && \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_1 &= 1.0, \\
 &&& \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \bullet \bar{x}_1 + x_2 + x_3 &= 0.5, \\
 &&& (x_1, x_2, x_3) \in \mathbb{R}^3, \quad x_1 \geq \sqrt{x_2^2 + x_3^2}, \quad \bar{x}_1 \succeq 0.
 \end{aligned}$$

```

# Block triplet format (SDPA)
sdo1$barA$i <- c(1,1,1, 2,2,2,2,2,2)
sdo1$barA$j <- c(1,1,1, 1,1,1,1,1,1)

sdo1$barA$k <- c(1,2,3, 1,2,3,2,3,3)
sdo1$barA$l <- c(1,2,3, 1,1,1,2,2,3)
sdo1$barA$v <- c(1,1,1, 1,1,1,1,1,1)

```

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$$\begin{array}{ll}\text{minimize} & \frac{1}{2}x^T Q x + c^T x + e^T v + k \\ \text{subject to} & -v \leq x \leq v,\end{array}$$

```
rr <- mosek_read( "probQP.opf" )
```

```
qp <- rr$prob
```

```
rqp <- mosek( qp )
```


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$$\begin{array}{ll} \text{minimize} & \frac{1}{2}x^T Qx + c^T x + e^T v + k \\ \text{subject to} & -v \leq x \leq v, \end{array}$$

ITE	PFEAS	DFEAS	POBJ	TIME
8	1.3e-14	3.1e-08	4.213156702e+03	74.84
9	1.2e-14	2.5e-09	4.213153627e+03	82.54
10	1.1e-14	1.3e-10	4.213153292e+03	90.21

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Cholesky Factorization: $Q = A^T A$

$$\begin{array}{ll} \text{minimize} & \frac{1}{2}w^T w + c^T x + e^T v + k \\ \text{subject to} & -v \leq x \leq v, \\ & Ax = w, \end{array}$$

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Cholesky Factorization: $Q = A^T A$

$$\begin{array}{ll} \text{minimize} & \frac{1}{2}w^T w + c^T x + e^T v + k \\ \text{subject to} & -v \leq x \leq v, \\ & Ax = w, \end{array}$$

```
Q <- sparseMatrix(qp$qobj$i,  
                  qp$qobj$j,  
                  x = qp$qobj$v,  
                  dims = c(nx,nx),  
                  symmetric = TRUE)  
  
A <- chol(Q)
```

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Cholesky Factorization: $Q = A^T A$

$$\begin{array}{ll} \text{minimize} & \frac{1}{2}w^T w + c^T x + e^T v + k \\ \text{subject to} & -v \leq x \leq v, \\ & Ax = w, \end{array}$$

```
sqp <- qp
```

```
sqp$qobj <- list(i = 1:nx,
                 j = 1:nx,
                 v = rep(1, nx))
```

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Cholesky Factorization: $Q = A^T A$

$$\begin{array}{ll} \text{minimize} & \frac{1}{2}w^T w + c^T x + e^T v + k \\ \text{subject to} & -v \leq x \leq v, \\ & Ax = w, \end{array}$$

```
eye    <- Diagonal(nx)
zeros  <- Matrix(0, nrow=nx, ncol=nx)

sqp$A  <- rBind(
  sqp$A,
  cBind(-eye, A, zeros)
)
```

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Cholesky Factorization: $Q = A^T A$

$$\begin{array}{ll} \text{minimize} & \frac{1}{2}w^T w + c^T x + e^T v + k \\ \text{subject to} & -v \leq x \leq v, \\ & Ax = w, \end{array}$$

ITE	PFEAS	DFEAS	POBJ	TIME
17	1.1e-06	1.5e-08	4.213156615e+03	29.76
18	1.0e-06	1.5e-08	4.213156490e+03	31.36
19	8.0e-06	4.8e-10	4.213153446e+03	32.99

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$$\begin{array}{ll}
 \text{minimize} & \frac{1}{2}p + c^T x + e^T v + k \\
 \text{subject to} & -v \leq x \leq v, \\
 & Ax = w, \\
 & 2qp \geq \|w\|_2^2 \\
 & q = \frac{1}{2}
 \end{array}$$

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$$\begin{array}{ll}
 \text{minimize} & \frac{1}{2}p + c^T x + e^T v + k \\
 \text{subject to} & -v \leq x \leq v, \\
 & Ax = w, \\
 & 2qp \geq \|w\|_2^2 \\
 & q = \frac{1}{2}
 \end{array}$$

```
cqp$obj <- NULL
```


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 \text{minimize} & \frac{1}{2}p + c^T x + e^T v + k \\
 \text{subject to} & -v \leq x \leq v, \\
 & Ax = w, \\
 & 2qp \geq \|w\|_2^2 \\
 & q = \frac{1}{2}
 \end{array}$$

```
cqp$c <- c(0.0, 0.5, cqp$c)
```

```
cqp$bx <- cBind( c(0.5, 0.5),
                  c(0.0, Inf),
                  cqp$bx
                  )
```

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$$\begin{array}{ll}
 \text{minimize} & \frac{1}{2}p + c^T x + e^T v + k \\
 \text{subject to} & -v \leq x \leq v, \\
 & Ax = w, \\
 & 2qp \geq \|w\|_2^2 \\
 & q = \frac{1}{2}
 \end{array}$$

```
cqp$cones <- cBind(
  list("rquad", 1:(2+nw))
)
```

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$$\begin{array}{ll}
 \text{minimize} & \frac{1}{2}p + c^T x + e^T v + k \\
 \text{subject to} & -v \leq x \leq v, \\
 & Ax = w, \\
 & 2qp \geq \|w\|_2^2 \\
 & q = \frac{1}{2}
 \end{array}$$

ITE	PFEAS	DFEAS	POBJ	TIME
10	8.7e-07	1.7e-06	4.213161986e+03	6.67
11	1.3e-07	2.5e-07	4.213154532e+03	7.17
12	1.2e-07	2.5e-07	4.213154518e+03	7.66

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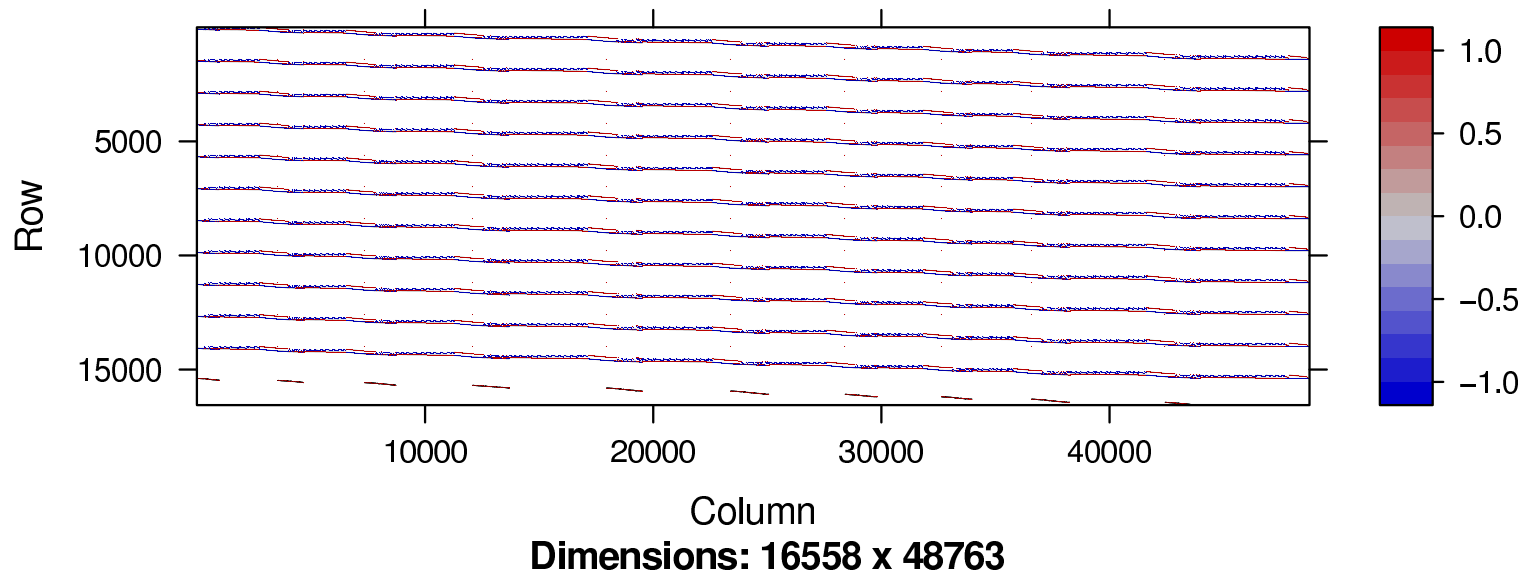
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```
r <- mosek_read("network.mps")  
problem <- r$prob
```

```
pdf("picture.pdf")  
print(image(problem$A))  
dev.off()
```



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Normalize a data-matrix:
(each column of X is a time series)

```
N <- nrow(X)
mu_r <- matrix(1, nrow=N) %*% colMeans(X)
Xbar <- 1/sqrt(N-1) * (X - mu_r)
```

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Find the QR factorization:

```
factor <- qr(Xbar)
```

```
Q <- qr.Q(factor)
```

```
R <- qr.R(factor)
```

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Save results and read them back in:

```
R <- as.matrix(  
  read.csv( "data/qr-r.csv" , header=FALSE )  
)
```


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$$\begin{array}{ll} \text{minimize} & r^T(w^0 + x) - \lambda \|R(w^0 + x)\| \\ \text{subject to} & e^T x = 0. \end{array}$$

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$$\begin{array}{ll} \text{minimize} & r^T(w^0 + x) - \lambda \|R(w^0 + x)\| \\ \text{subject to} & e^T x = 0. \end{array}$$

```
plot(RISK, RET, 'l',  
      xlim = c(0, 0.1),  
      ylim = c(tlow, thigh))
```

```
points(SHARPE_RISK, SHARPE_RET)
```

```
lines(x = c(0,1),  
      y = rf*sum(w0) + SHARPE_RATIO*c(0,1))
```

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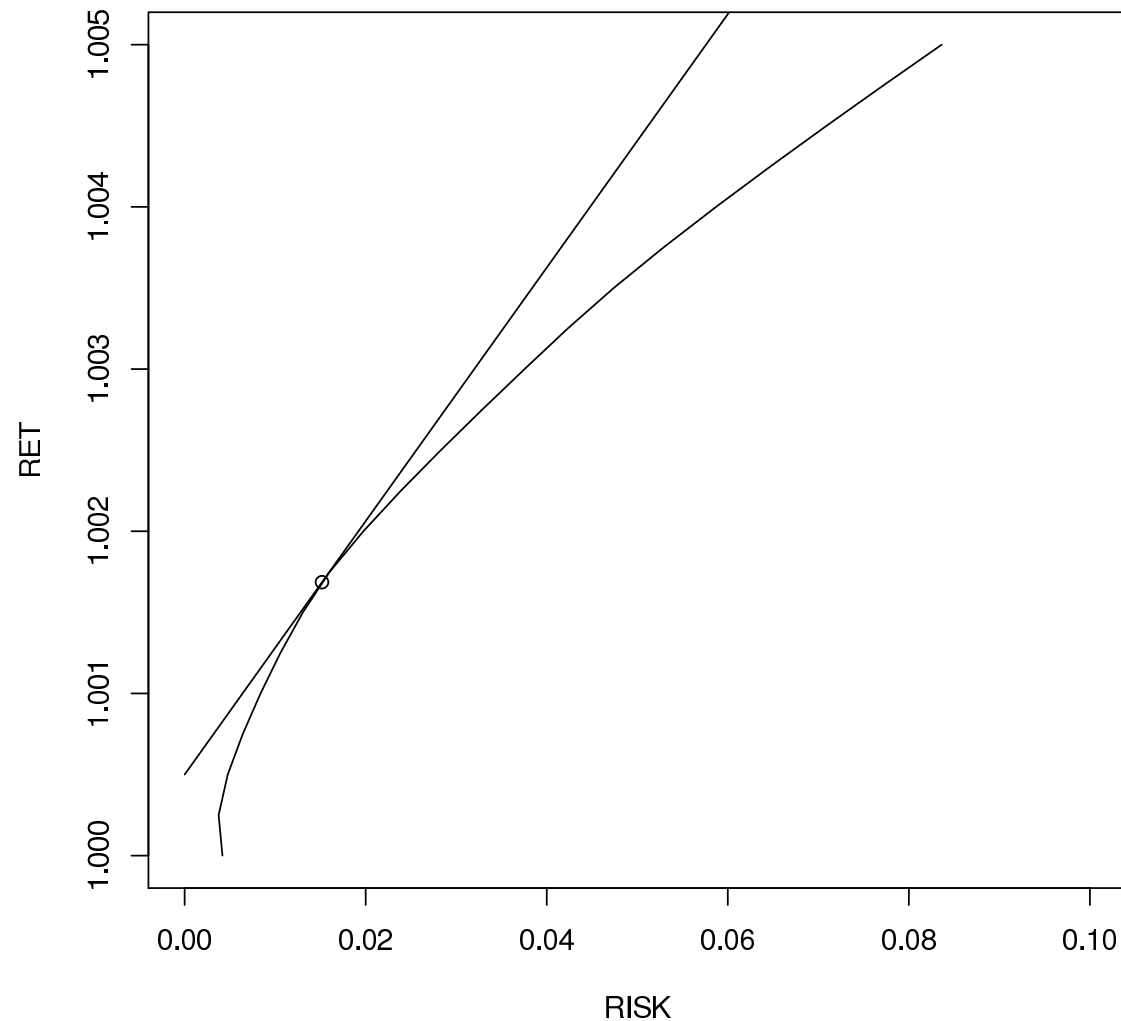
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$$\begin{array}{ll} \text{minimize} & r^T(w^0 + x) - \lambda \|R(w^0 + x)\| \\ \text{subject to} & e^T x = 0. \end{array}$$



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Multiple knapsack problem:

- One pool of (weight, value)-items
- N knapsacks of limited capacity
- Maximize profit

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Multiple knapsack problem:

- One pool of (weight, value)-items
- N knapsacks of limited capacity
- Maximize profit

One subproblem per knapsack (MILP):

- Select items for the individual knapsack
- Maximize reduced-cost-profit

One master (LP):

- Select item-disjoint solutions for each knapsack
- Maximize profit

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TASK: Update and solve subproblems in parallel.

```
>> require("doMC")
Loading required package: doMC
Loading required package: foreach
Loading required package: iterators
Loading required package: codetools
Loading required package: multicore

>> registerDoMC(cores = 16)
```

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```
mosek_clean( )
```

```
SPsols <-  
foreach(i = 1:nsack) %dopar% {  
  prob <- updateObj(SP[[i]],  
                    itemProfits,  
                    itemDuals,  
                    sackDuals[i],  
                    MPfeasibility)  
  
  sol <- getSolution(prob)  
  
  if (isImproving(prob, sol)) {  
    return(sol)  
  } else {  
    return(NULL)  
  }  
}
```


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200 knapsacks of:

- $\text{capacity} = 1, 2, \dots, 200$

200 items of:

- $\text{weight} = 1, 2, \dots, 200$
- $\text{profit} = 200, 199, \dots, 1$

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200 knapsacks of:

- capacity = 1, 2, ..., 200

200 items of:

- weight = 1, 2, ..., 200
- profit = 200, 199, ..., 1

Optimality in **449** seconds:

- **37600** subproblems solved (**17117** columns added)
- **187** master problems solved

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200 knapsacks of:

- capacity = 1, 2, ..., 200

200 items of:

- weight = 1, 2, ..., 200
- profit = 200, 199, ..., 1

Optimality in **449** seconds:

- **37600** subproblems solved (**17117** columns added)
- **187** master problems solved

Average speed: **84 problems per second.**

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The R project

- r-project.org

Home of MOSEK ApS

- mosek.com

Need help? MOSEK Google Group!

- groups.google.com/group/mosek

The Rmosek introduction page

- [cran.r-project.org/web/packages/Rmosek/
index.html](http://cran.r-project.org/web/packages/Rmosek/index.html)

The Rmosek development site

- rmosek.r-forge.r-project.org

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Goals:

- ✓ Open-source
- ✓ Highly effective
- ✓ Integrate with what R users normally do

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Thank you!