



R news and tutorials contributed by (750) R bloggers

- [Home](#)
- [About](#)
- [RSS](#)
- [add your blog!](#)
- [Learn R](#)
- [R jobs](#)
- [Contact us](#)

Welcome!



Here you will find daily **news and tutorials about R**, contributed by over 750 bloggers.

There are many ways to **follow us** - [By e-mail:](#)

[On Facebook:](#)



If you are an R blogger yourself you are invited to [add your own R content feed to this site](#) (**Non-English** R bloggers should add themselves- [here](#))

[Jobs for R-users](#)

- [R Developer](#)
- [postdoc in psychiatry: machine learning in human genomics](#)
- [Lead Quantitative Developer](#)
- [Research Data Analyst @ Arlington, Virginia, U.S.](#)
- [Market](#)



Never miss an update!
Subscribe to R-bloggers to receive e-mails with the latest R posts.
 (You will not see this message again.)

[Click here to close](#) (This popup will not appear again)

[Mobility for RSG](#)

Popular Searches

Recent Posts

- [Why R 2018 Winners](#)
- [Extracting a Reference Grid of your Data for Machine Learning Models Visualization](#)
- [#19: Intel MKL in Debian / Ubuntu follow-up](#)
- [Statistics Sunday: Converting Between Effect Sizes for Meta-Analysis](#)
- [Let R/Python send messages when the algorithms are done training](#)
- [Forecasting my weight with R](#)
- [A useful forecast combination benchmark](#)
- [A primer in using Java from R – part 1](#)
- [future.apply – Parallelize Any. Base R Apply. Function](#)
- [Thanks for Reading!](#)
- [A guide to working with character data in R](#)
- [Using DataCamp's Autograder to Teach R](#)
- [Melt and cast the shape of your data.frame – Exercises](#)
- [Creating Slopegraphs with R](#)
- [Parallelizing Linear Regression or Using Multiple Sources](#)

Never miss an update!

Subscribe to R-bloggers to receive e-mails with the latest R posts.
(You will not see this message again.)

[Click here to close](#) (This popup will not appear again)



Other sites

- [Jobs for R-users](#)
- [SAS blogs](#)

Black-Litterman Model

November 15, 2011

By [systematicinvestor](#)

Share

(This article was first published on [Systematic Investor » R](#), and kindly contributed to [R-bloggers](#).)

f Share

🐦 Tweet

The [Black-Litterman Model](#) was created by Fisher Black and Robert Litterman in 1992 to resolve shortcomings of traditional Markovitz mean-variance asset allocation model. It addresses following two items:

- Lack of diversification of portfolios on the mean-variance efficient frontier.
- Instability of portfolios on the mean-variance efficient frontier: small changes in the input assumptions often lead to very different efficient portfolios.

I recommend a very good non-technical introduction to [The Black-Litterman Model, An Introduction for the Practitioner by T. Idzorek \(2009\)](#).

I will take the country allocation example presented in [The Intuition Behind Black-Litterman Model Portfolios by G. He, R. Litterman \(1999\)](#) paper and update it using current market data.

First, I need market capitalization data for each country to compute equilibrium portfolio. I found following two sources of capitalization data:

- World Development Indicators database at the [World Databank](#). First select countries, for series type in "capitalization", and last choose years.
- [World Federation of Exchanges](#).

I will use market capitalization data from World Databank.

```
1 # load Systematic Investor Toolbox
2 setInternet2(TRUE)
3 source(gzcon(url('https://github.com/systemat
4
5 #-----
6 # Visualize Market Capitalization History
7 #-----
8
9 hist.caps = aa.test.hist.capitalization()
10 hist.caps.weight = hist.caps/rowSums(hist
11
12 # Plot Transition of Market Cap Weights i
13 plot.transition.map(hist.caps.weight, ind
14
15 # Plot History for each Country's Market
16 layout(matrix(1:9, nrow = 3, byrow=T) )
17 col = plot.colors(ncol(hist.caps))
18 for(i in 1:ncol(hist.caps)) {
19   plot(hist.caps[,i], type='l', lwd=5,
20 }
```

There is a major shift in weights between Japan and USA from 1988 to 2010. In 1988 Japan represented 47% and

Never miss an update!
Subscribe to R-bloggers to receive
 e-mails with the latest R posts.
 (You will not see this message again.)

Your E-mail...

[Click here to close](#) (This popup will not appear again)

shift was driven by inflow of capital to USA, the Japanese capitalization was pretty stable in time, as can be observed from time series plot for each country.

Second, I need historical prices series for each country to compute covariance matrix. I will use historical data from [Yahoo Fiance](#):

Australia [EWA](#)

Canada [EWC](#)

France [EWQ](#)

Germany [EWG](#)

Japan [EWJ](#)

U.K. [EWU](#)

USA [SPY](#)

The first step of the Black-Litterman model is to find implied equilibrium returns using reverse optimization.

$$\Pi = \delta \Sigma w_{eq}$$

where Π are equilibrium returns, δ is risk aversion, Σ is covariance matrix, and w_{eq} are market capitalization weights. The risk aversion parameter can be estimated from historical data by dividing the excess market portfolio return by its variance.

Never miss an update!
Subscribe to R-bloggers to receive
 e-mails with the latest R posts.
 (You will not see this message again.)

Your E-mail...

[Click here to close \(This popup will not appear again\)](#)

```

1  # Use reverse optimization to compute the vec
2  bl.compute.eqret <- function
3  (
4    risk.aversion, # Risk Aversion
5    cov,           # Covariance matrix
6    cap.weight,    # Market Capitalization W
7    risk.free = 0  # Risk Free Interest Rate
8  )
9  {
10   return( risk.aversion * cov %*% cap.weigh
11 }
12
13 #-----
14 # Compute Risk Aversion, prepare Black-Li
15 #-----
16 ia = aa.test.create.ia.country()
17
18 # compute Risk Aversion
19 risk.aversion = bl.compute.risk.aversion(
20
21 # the latest market capitalization weight
22 cap.weight = last(hist.caps.weight)
23
24 # create Black-Litterman input assumption
25 ia.bl = ia
26 ia.bl$expected.return = bl.compute.eqret(
27
28 # Plot market capitalization weights and
29 layout( matrix(c(1,1,2,3), nrow=2, byrow=
30 pie(coredata(cap.weight), paste(colnames(
31   main = paste('Country Market Capitali
32   , col=plota.colors(ia$n))
33
34 plot.ia(ia.bl, T)

```

Next, let's compare the efficient frontier created using historical input assumptions and Black-Litterman input assumptions

```

1  #-----
2  # Create Efficient Frontier(s)
3  #-----
4  n = ia$n
5
6  # -1 <= x.i <= 1
7  constraints = new.constraints(n, lb = 0, ub =
8
9  # SUM x.i = 1
10 constraints = add.constraints(rep(1, n), 1, t
11
12 # create Efficient Frontier(s)

```



```

14 ef.risk.bl = portopt(ia.bl, constraints, 50,
15
16 # Plot multiple Efficient Frontiers and Trans
17 layout( matrix(1:4, nrow = 2) )
18 plot.ef(ia, list(ef.risk), portfolio.risk, T,
19 plot.ef(ia.bl, list(ef.risk.bl), portfolio.ri

```

Comparing the transition maps, the Black-Litterman efficient portfolios are well diversified. Efficient portfolios have allocation to all asset classes at various risk levels. By its construction, the Black-Litterman model is well suited to address the diversification problems.

The Black-Litterman model also introduces a mechanism to incorporate investor's views into the input assumptions in such a way that small changes in the input assumptions will NOT lead to very different efficient portfolios. The Black-Litterman model adjusts expected returns and covariance:

Formula does not parse

where P is Views pick matrix, and Q Views mean vector. The Black-Litterman model assumes that views are $N \sim (Q, P)$.

```

1 bl.compute.posterior <- function
2 (
3   mu,          # Equilibrium returns
4   cov,         # Covariance matrix
5   pmat=NULL,   # Views pick matrix
6   qmat=NULL,   # Views mean vector
7   tau=0.025    # Measure of uncertainty of t
8 )
9 {
10  out = list()
11  omega = diag(c(1, diag(tau * pmat %%% cov
12
13  temp = solve(solve(tau * cov) + t(pmat) %
14  out$cov = cov + temp
15
16  out$expected.return = temp %%% (solve(tau
17  return(out)
18 }
19
20 #-----
21 # Create Views
22 #-----
23 temp = matrix(rep(0, n), nrow = 1)
24 colnames(temp) = ia$symbols
25
26 # Relative View
27 # Japan will outperform UK by 2%
28 temp[, 'Japan'] = 1
29 temp[, 'UK'] = -1
30
31 pmat = temp
32 qmat = c(0.02)
33
34 # Absolute View
35 # Australia's expected return is 12%
36 temp[] = 0
37 temp[, 'Australia'] = 1
38
39 pmat = rbind(pmat, temp)
40 qmat = c(qmat, 0.12)
41
42 # compute posterior distribution paramete
43 post = bl.compute.posterior(ia.bl$expecte
44
45 # create Black-Litterman input assumption
46 ia.bl.view = ia.bl
47   ia.bl.view$expected.return = post$exp
48   ia.bl.view$cov = post$cov
49   ia.bl.view$risk = sqrt(diag(ia.bl.vie
50
51 # create efficient frontier(s)
52 ef.risk.bl.view = portopt(ia.bl.view, con
53
54 # Plot multiple Efficient Frontiers and T
55 layout( matrix(1:4, nrow = 2) )
56 plot.ef(ia.bl, list(ef.risk.bl), portfoli

```

Never miss an update!
Subscribe to R-bloggers to receive
 e-mails with the latest R posts.
 (You will not see this message again.)

[Click here to close \(This popup will not appear again\)](#)



Comparing the transition maps, the Black-Litterman + Views efficient portfolios have more allocation to Japan and Australia, as expected. The portfolios are well diversified and are not drastically different from the Black-Litterman efficient portfolios.

The Black-Litterman model provides an elegant way to resolve shortcomings of traditional Markovitz mean-variance asset allocation model based on historical input assumptions. It addresses following two items:

- Lack of diversification of portfolios on the mean-variance efficient frontier. The Black-Litterman model uses equilibrium returns implied from the current market capitalization weights to construct well diversified portfolios.
- Instability of portfolios on the mean-variance efficient frontier. The Black-Litterman model introduces a mechanism to incorporate investor's views into the input assumptions in such a way that small changes in the input assumptions will NOT lead to very different efficient portfolios.

I highly recommend exploring and reading following articles and websites for better understanding of the Black-Litterman model:

- [The Intuition Behind Black-Litterman Model Portfolios by G. He, R. Litterman \(1999\)](#)
- [AllocationADVISOR and The Black-Litterman Model by T. Idzorek \(2004\)](#)
- [A STEP-BY-STEP GUIDE TO THE BLACK-LITTERMAN MODEL by T. Idzorek \(2005\)](#)
- [The Intuition Behind Black-Litterman Model Portfolios by G. He, R. Litterman \(1999\)](#)
- [A STEP-BY-STEP GUIDE TO THE BLACK-LITTERMAN MODEL by T. IDZOREK \(2002\)](#)
- [The Black-Litterman Model and Alternative Investments by M. Odo](#)
- [Incorporating Trading Strategies in the Black-Litterman Framework by F. FABOZZI, S. FOCARDI, P. KOLM \(2006\)](#)
- [Jay Walters published two papers on the The Black-Litterman Model: "The Black-Litterman Model In Detail" and "The Factor Tau in the Black-Litterman Model"](#)
- [Jay Walters also gathered a collection of Implementations of the Black-Litterman Model at his site.](#)
- [Beyond Black-Litterman in Practice: A Five-Step Recipe to Input Views on Non-Normal Markets by A. Meucci \(2005\) accompanied by Matlab code.](#)
- [Fully Flexible Views: Theory and Practice by A. Meucci \(2008\) accompanied by Matlab code.](#)

To view the complete source code for this example, please have a look at the [aa.black.litterman.test\(\) function in aa.test.r at github](#).

Never miss an update!
Subscribe to R-bloggers to receive
 e-mails with the latest R posts.
 (You will not see this message again.)

[Click here to close](#) (This popup will not appear again)

 Share

 Tweet

To leave a comment for the author, please follow the link and comment on their blog: [Systematic Investor » R](#).

R-bloggers.com offers [daily e-mail updates](#) about R news and [tutorials](#) on topics such as: [Data science](#), [Big Data](#), [R jobs](#), visualization ([ggplot2](#), [Boxplots](#), [maps](#), [animation](#)), programming ([RStudio](#), [Sweave](#), [LaTeX](#), [SQL](#), [Eclipse](#), [git](#), [hadoop](#), [Web Scraping](#)) statistics ([regression](#), [PCA](#), [time series](#), [trading](#)) and more...



If you got this far, why not **subscribe for updates** from the site? Choose your flavor: [e-mail](#), [twitter](#), [RSS](#), or [facebook](#)...

[Share](#)

Tags: [Asset Allocation](#), [Portfolio Construction](#), [R](#)

Comments are closed.

Search R-bloggers

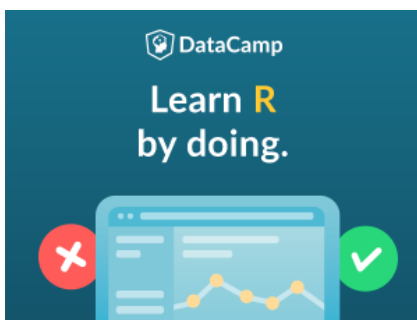
Recent popular posts

- [future.apply - Parallelize Any Base R Apply Function](#)
- [A primer in using Java from R - part 1](#)
- [Let R/Python send messages when the algorithms are done training](#)
- [Forecasting my weight with R](#)
- [Why R 2018 Winners](#)

Most visited articles of the week

1. [How to write the first for loop in R](#)
2. [Installing R packages](#)
3. [Using apply, sapply, lapply in R](#)
4. [R – Sorting a data frame by the contents of a column](#)
5. [How to perform a Logistic Regression in R](#)
6. [How to Make a Histogram with Basic R](#)
7. [Tutorials for learning R](#)
8. [How to Make a Histogram with ggplot2](#)
9. [Creating Slopegraphs with R](#)

Sponsors



MACHINE LEARNING WORKSHOP
 MAX KUHN
 August 23 & 24, 2018 | Washington, DC
[LEARN MORE AND REGISTER](#)

McKinsey Analytics
Online Hackathon
 July 20-22, 2018

Beginner's Guide to
Spatial, Temporal and Spatial-Temporal Ecological Data Analysis with R-INLA
 Zuur, Ieno, Saveliev

Register for free updates
 in all areas of statistics

STATISTICS VIEWS
 Bringing Statistics Together

WILEY

DATASCIENCE GO
 Conference 2018

JOIN OVER 400 DATA SCIENTISTS & TECH EXPERTS
 To help you boost your career

[Reserve your Seat Today](#)

EARLY BIRD PRICE! 55% OFF



Quantide: statistical consulting and training

Don't Miss the Leading Data Science Conference

Save 60% Ends Friday

ODSC[®] West 2018
 San Francisco | Oct 31 - Nov 3

Don't Miss Europe's Leading Data Science Conference

Save 55% Ends Friday

ODSC[®] Europe 2018
 London | September 19-22

R Courses for Professionals
 Download R templates for machine learning

The fastest way to learn data science!

[VIEW SNEAK PEAK](#)

DATA SOCIETY

R-Training
 September | Stuttgart

eoda
 eoda GmbH

Never miss an update!
Subscribe to R-bloggers to receive e-mails with the latest R posts.
 (You will not see this message again.)

[Click here to close \(This popup will not appear again\)](#)



- [R Developer](#)
- [postdoc in psychiatry: machine learning in human genomics](#)
- [Lead Quantitative Developer](#)
- [Research Data Analyst @ Arlington, Virginia, U.S.](#)
- [Market Research Analyst: Mobility for RSG](#)
- [Data Scientist @ New Delhi, India](#)
- [Data Scientist/Programmer @ Milwaukee, Wisconsin, U.S.](#)

[Full list of contributing R-bloggers](#)

[R-bloggers](#) was founded by [Tal Galili](#), with gratitude to the [R](#) community.

Is powered by [WordPress](#) using a [bavotasan.com](#) design.

Copyright © 2018 **R-bloggers**. All Rights Reserved. [Terms and Conditions](#)

Never miss an update!
Subscribe to R-bloggers to receive
e-mails with the latest R posts.
(You will not see this message again.)



[Click here to close](#) (This popup will not appear again)

