



COMPUTATIONAL FINANCE & RISK MANAGEMENT

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UNIVERSITY *of* WASHINGTON

Department of Applied Mathematics

# More Monte Carlo and `mcsim(. )` in blotter R Package

CFRM 522

Introduction to Trading Strategies (018a)

## Lecture References

- Brian Peterson: Monte Carlo analysis for strategy  
[http://braverock.com/brian/monte\\_carlo.html](http://braverock.com/brian/monte_carlo.html)
- Documentation for the `mcsim( . )` function in blotter help
  - <https://opensourcequant.wordpress.com/2017/08/09/monte-carlo-for-your-portfolio-pl/>
- Jaekle & Tomasini, Ch 4, § 2 (Monte Carlo Analysis) (Previous Reading Assignment)
- Aronson, Ch 5: Hypothesis Tests and Confidence Intervals (Reading Assignment)
- Jasen Mackie & Brian G. Peterson: Round Turn Trade Simulation (pdf – available on Canvas), 2018 (`mcsim( . )`)

# Different forms of Monte Carlo Simulation for Strategy Development

- Equity Curve Simulation
- Weights/Positions/Portfolio
- Round Turn Trades
- Resampled Market Data

# Equity Curve Simulation

- Used in texts such as Jaekle & Tomasini, Aronson, and others
- Also implemented in a few commercial backtesters
- This type of simulation takes the equity curve either for a specific instrument or for the entire portfolio and samples from the equity curve
- Almost always done daily. For intraday strategies, it makes sense to consider the daily returns as a synthetic return series, from which inference may be attempted
- Equity curve Monte Carlo is implemented as `mcsim(.)` in blotter
  - Daily P&L
  - Trade (transaction) P&L

# Equity Curve Simulation

- To replicate the method of sampling from trades as done in Jaekle & Tomasini:
  - Use `mcsim(.)` with parameter `use = 'txns'`
  - This means the individual transaction P&L values will be rearranged
  - Put `replacement = FALSE` for simulation *without* replacement, as is done in the text
  - Put `replacement = TRUE` for simulation *with* replacement

# Equity Curve Simulation

- Sampling without replacement (as done in text)
  - Simulation without replacement will yield an equity curve which has the same mean and cumulative P&L as the original equity curve, but which has re-ordered the individual daily returns
  - This model is not as useful as with replacement (Peterson)
  - J&T: Exactly duplicates the probability distribution of the input sequence
- Sampling with replacement
  - Allows individual daily returns to be reused (with replacement) in the simulation.
  - It will create a more varied path, and is more useful for confidence intervals and other standard statistical inference, eg on the potential paths of a Drawdown.
  - J&T: Introduces more randomness – more representative of expected behavior in the future

# Equity Curve Simulation

- Upside: One can make a reasonable argument that resampling from daily returns is a valid *risk measurement* or *statistical inference tool*
- Downside: Cannot accept that the evolution of a trade would be the same if we start it at a different point

- Resampled market data is commonly used for evaluating portfolio risk
- It is much harder to use resampled market data to evaluate trading strategies; the usual caveats of i.i.d. (and non-normality) assumptions apply
- It is possible to construct synthetic return series using a bootstrap, or to construct synthetic returns using a multi-moment or other fitted distribution.
  - One can then turn this synthetic return series into a synthetic price series
  - With this synthetic price series, one can apply the strategy to the synthetic price series
  - The data is 'fake', so one can't use the cash P&L from the synthetic data in any real way
  - One can likely (if careful) draw some statistical inference about robustness of the strategy, whether the rules work as intended, and a few other things
  - It is easy, however, to confuse the possible statistical inference with actual possible strategy returns



## Having said all that...

- The literature is very thin on these methodologies, and their implications
- Opportunities exist for more research in this area
- Some potential goals would include rigorously defining all the given methods, constructing a full literature survey, comparing the strengths and weaknesses for given input data, and providing some guidance on which methods to use and why

# Implementation Example

- Use the Long-short Bollinger Band reversal strategy code from Section 015, through

```
initEq <- 0    # Set to 0 for Monte Carlo analysis
applyStrategy(strategy = stratName,
               portfolios=portName,parameters=list(sd=SD,n=N))

updatePortf(Portfolio = portName)
updateAcct(name = acctName)
updateEndEq(Account = acctName)
```

# Implementation Example

- Then, append the following:

```
# 1st do replacement = FALSE a la J & T book (use = 'txns'):
bbsim <- mcsim(Portfolio = portName, Account = acctName, n = 1000,
               replacement = FALSE, use = 'txns')

quantile(bbsim)
summary(bbsim)

plot(bbsim, normalize=FALSE)

hist(bbsim, normalize = FALSE)

# bbsimRep: sample WITH Replacement
bbsimRep <- mcsim(Portfolio = portName, Account = acctName, n = 1000,
                  replacement = TRUE, use = 'txns')

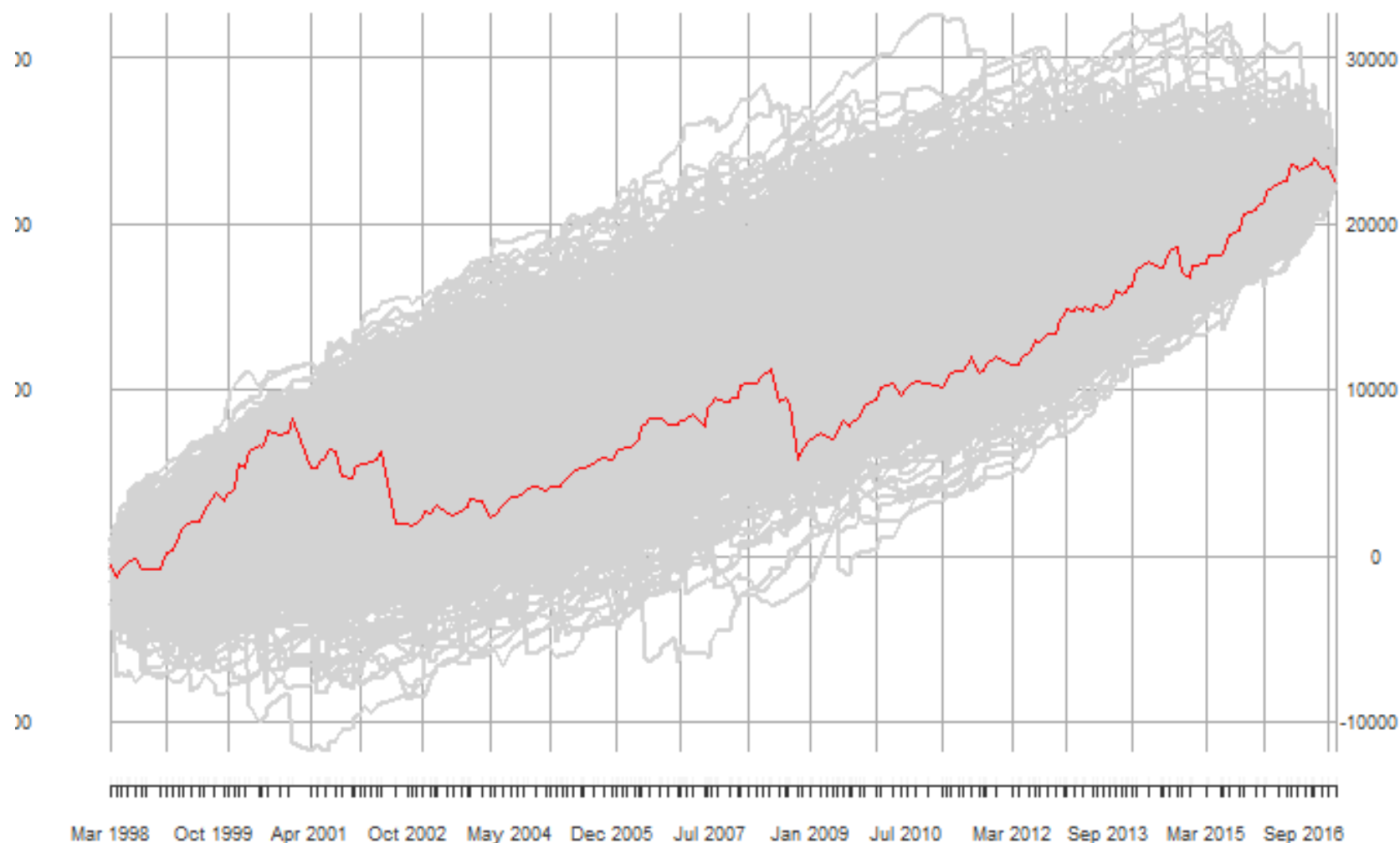
quantile(bbsimRep)
summary(bbsimRep)

plot(bbsimRep, normalize=FALSE)
hist(bbsimRep, normalize = FALSE)
```

# Compare plots of without vs with replacement

1000 replicates without replacement and block length 1

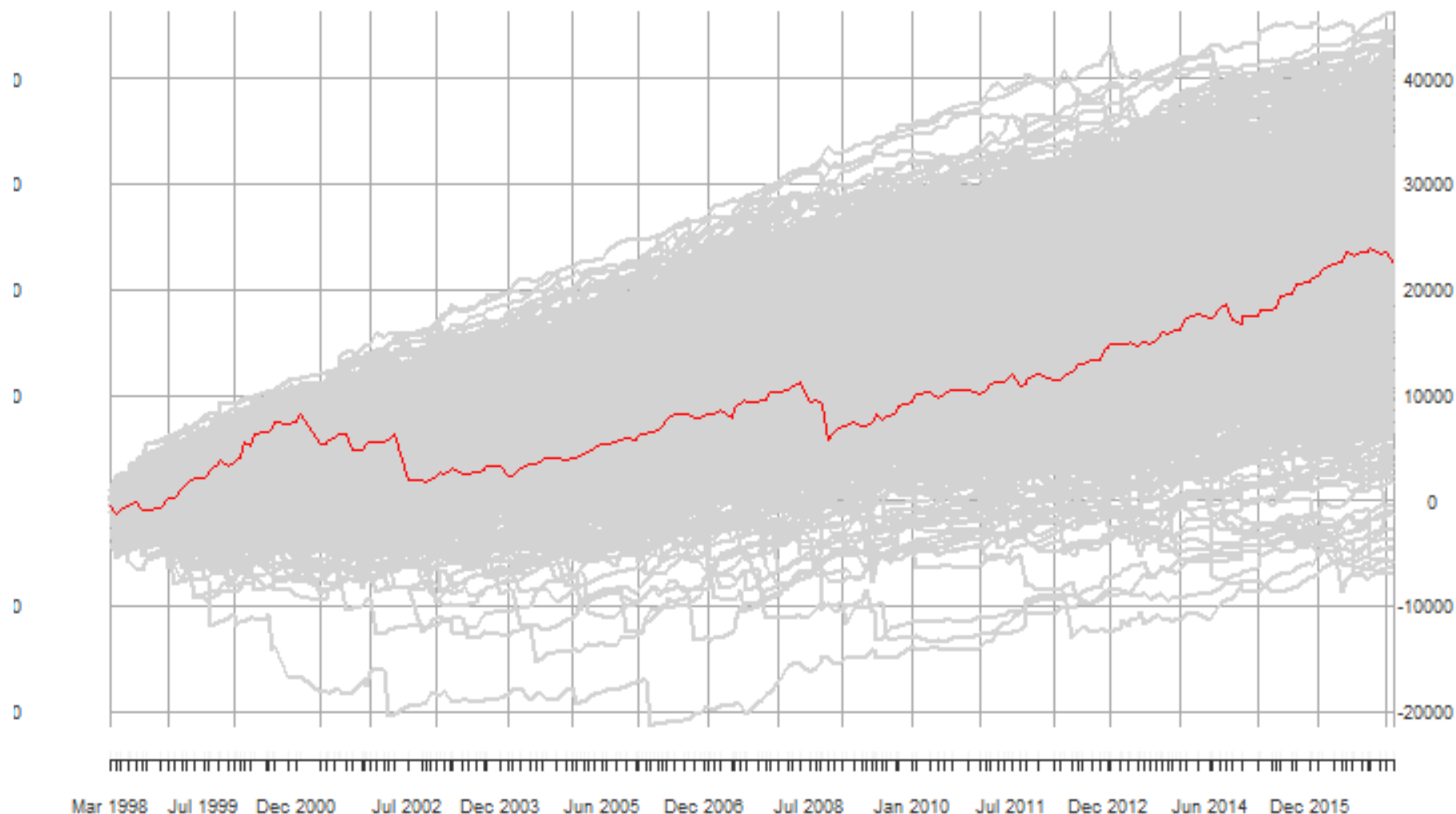
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# Compare plots of without vs with replacement

1000 replicates with replacement and block length 1

1998-03-06 / 2017-03-15



# More Results, with replacement (normalized)

```
> quantile(bbsimRep)
```

0%	25%	50%	75%	100%
-4.399028e-03	-7.886864e-06	1.912950e-04	3.495302e-04	1.420225e-03

# More Results, with replacement (normalized)

```
> summary(bbsimRep)
```

	Sample Mean	Sample Median	Backtest	Lower CI
mean	0.0001067514	0.0001084382	0.0001055970	2.477312e-05
median	0.0001866789	0.0001911880	0.0001915528	1.440206e-04
stddev	0.0005874713	0.0005892130	0.0006053529	3.898585e-04
maxDD	-0.0059030003	-0.0053135753	-0.0064269159	-1.096739e-02
sharpe	0.1972262435	0.1891255507	0.0006053529	-4.136289e-03

	Upper CI	Std. Error
mean	0.0001887297	4.182643e-05
median	0.0002293371	2.176481e-05
stddev	0.0007850842	1.008248e-04
maxDD	-0.0008386129	2.583919e-03
sharpe	0.3985887761	1.027379e-01

**Mean max DD = -0.5903%**

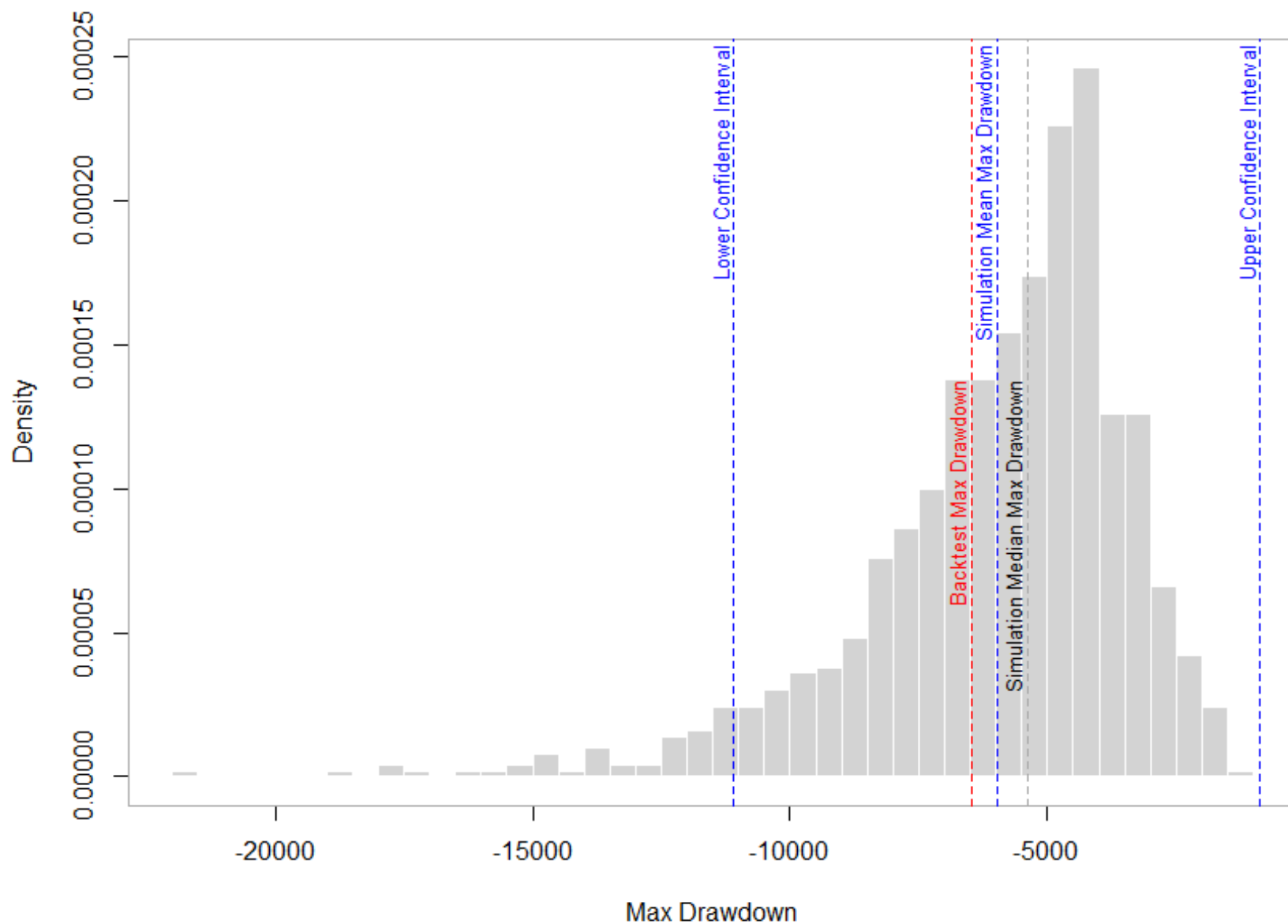
**Worst case max DD = -1.0967%**

- `hist(bbsimRep, normalize = FALSE)`
- Gives multiple histograms with respect to different metrics; eg,
  - Max Drawdown
  - Mean Return
  - Volatility
  - etc



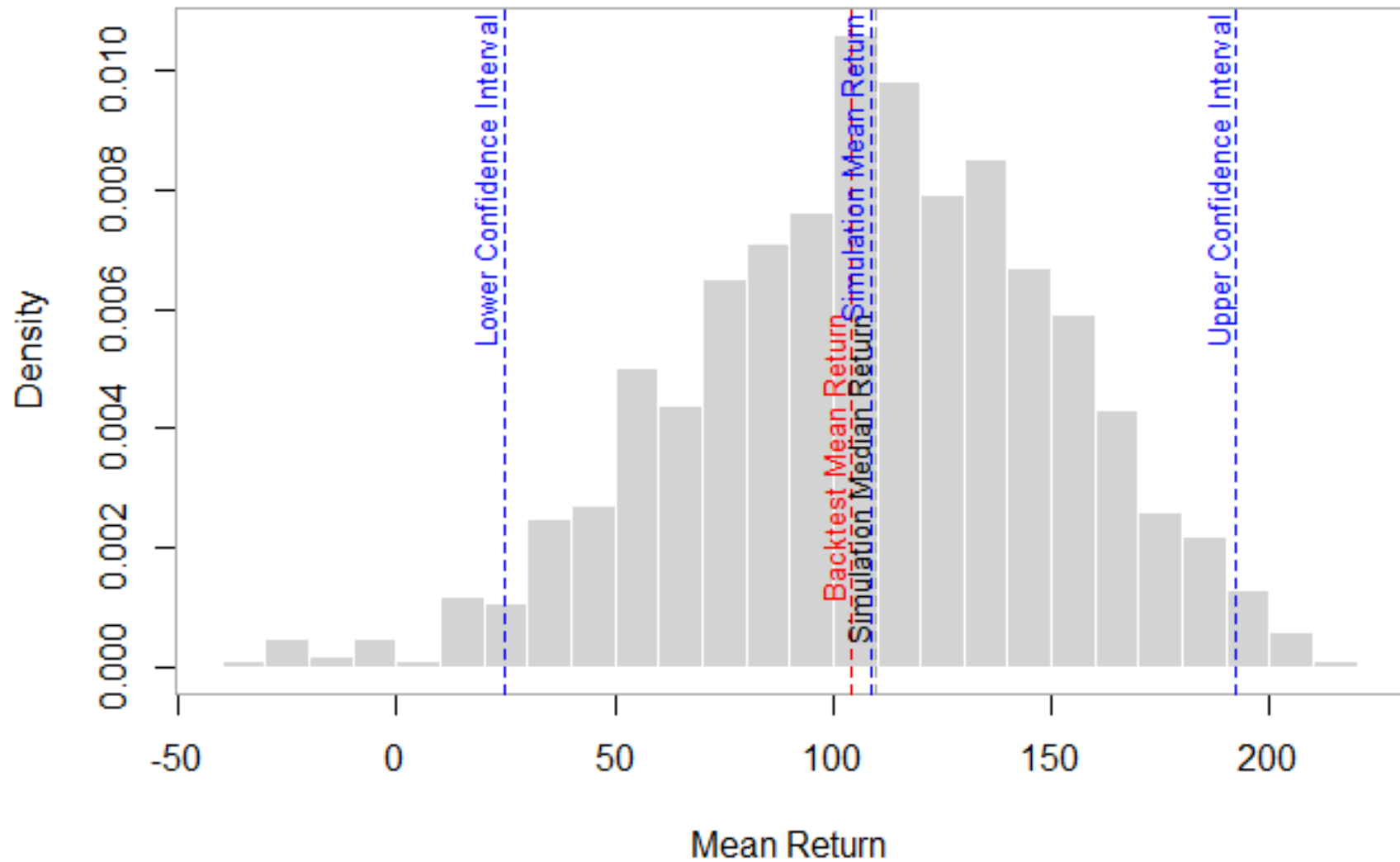
# More Results, with replacement

maxDrawdown distribution of 1000 replicates with replacement using block length 1 and 0.95 confidence interval



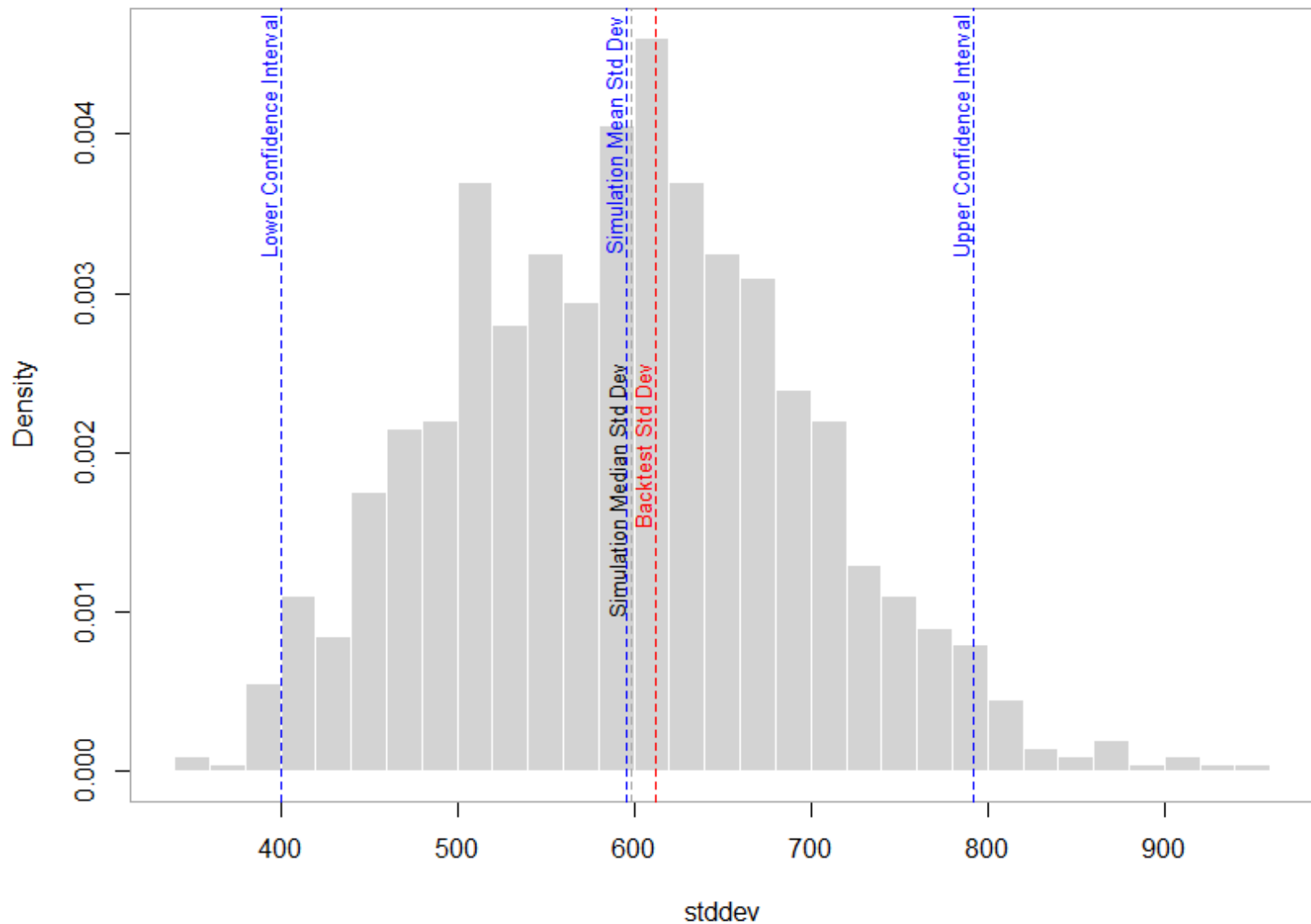
# More Results, with replacement

Mean distribution of 1000 replicates with replacement using block length 1 and 0.95 confidence interval



# More Results, with replacement

Std Dev distribution of 1000 replicates with replacement using block length 1 and 0.95 confidence interval





- txnsim is a newer feature in blotter
- mcsim is merely a re-ordering of the P&L observations
  - Daily
  - Trade P&L
- txnsim samples from the observed characteristics of the original strategy
  - How long was the strategy long, short, and flat?
  - What were the number of levels building up a trade?
  - What was the max position size etc?
  - Using those as constraints, resample the strategy such that the random versions of the strategy mimic the structure of the original
  - These are random, however, and therefore exhibit no skill
  - Can compare the original backtest graphically to test skill vs luck

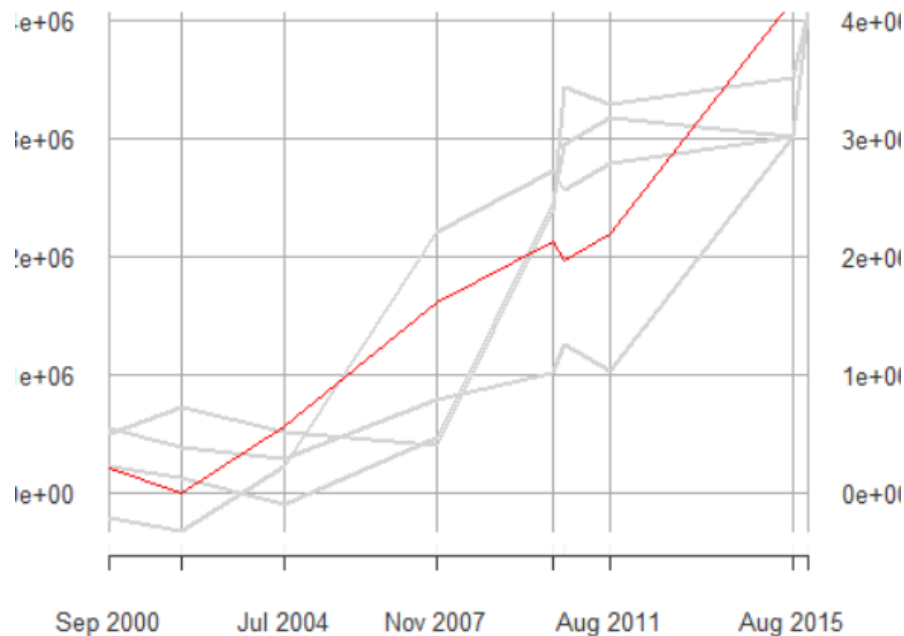
# Round Turn Trades and txnsim

- More generally, trade resampling is really about sampling the properties of the trade, primarily quantity, direction, and duration
- A round turn trade may consist of multiple transactions
  - Building up to a position
  - Unwinding a position
- It is also important to sample flat periods, if any exist
- The goal is to create a 'random' strategy that trades with similar behavior to the strategy that you are evaluating
- This type of sampling can also account for trade definitions other than flat to flat (as in the text), where a strategy levels into or out of a position
- This method is most useful for analyzing skill versus luck, because all of the synthetic/sampled results are from random strategies that trade in a similar manner, on the same data, as the strategy we are trying to analyze

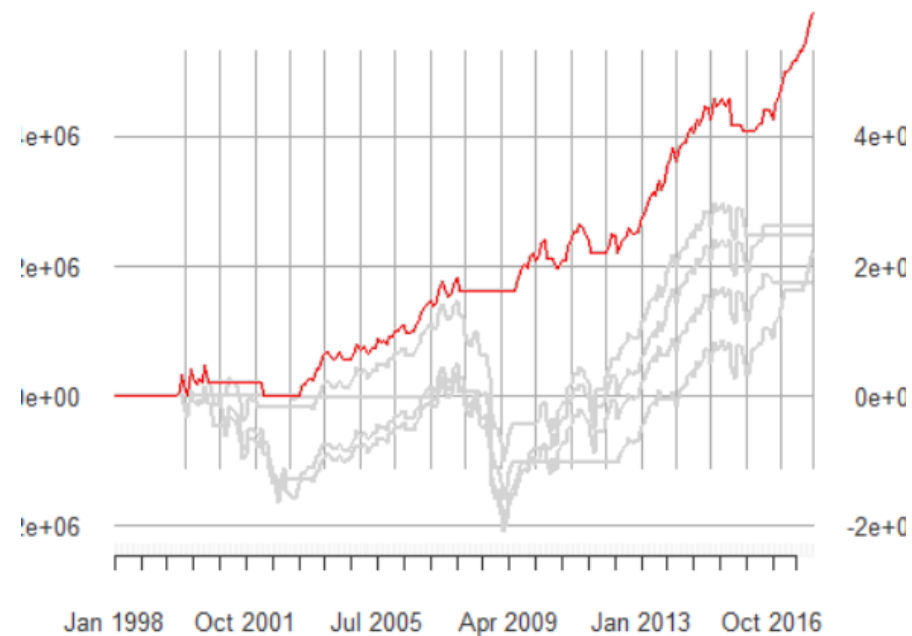
# txnsim Example

- Faber strategy, with four simulations
- Compare mcsim vs txnsim, without replacement:

4 replicates without replacement and block length



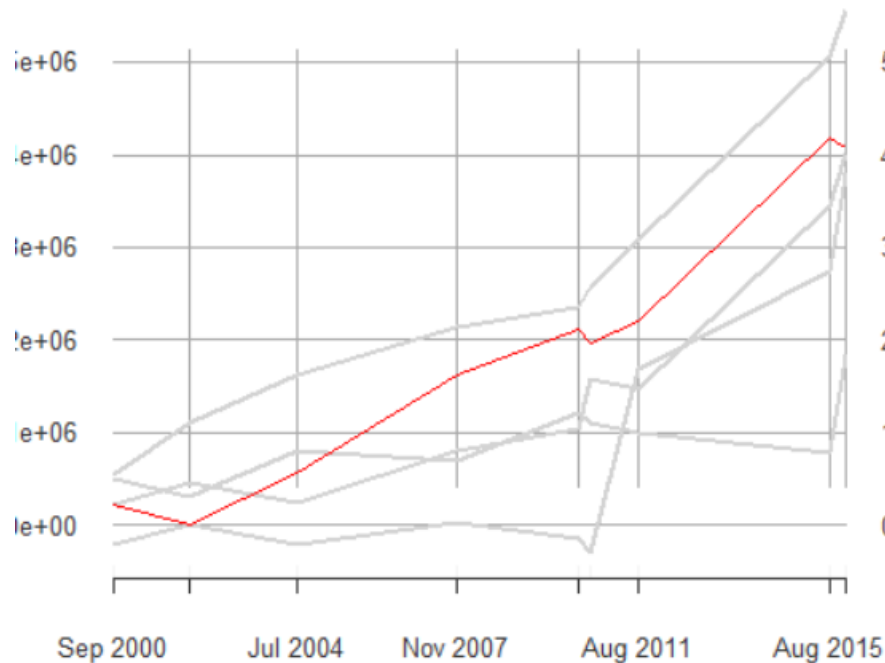
longtermDH txnsim cumulative P&L 4 reps. with re



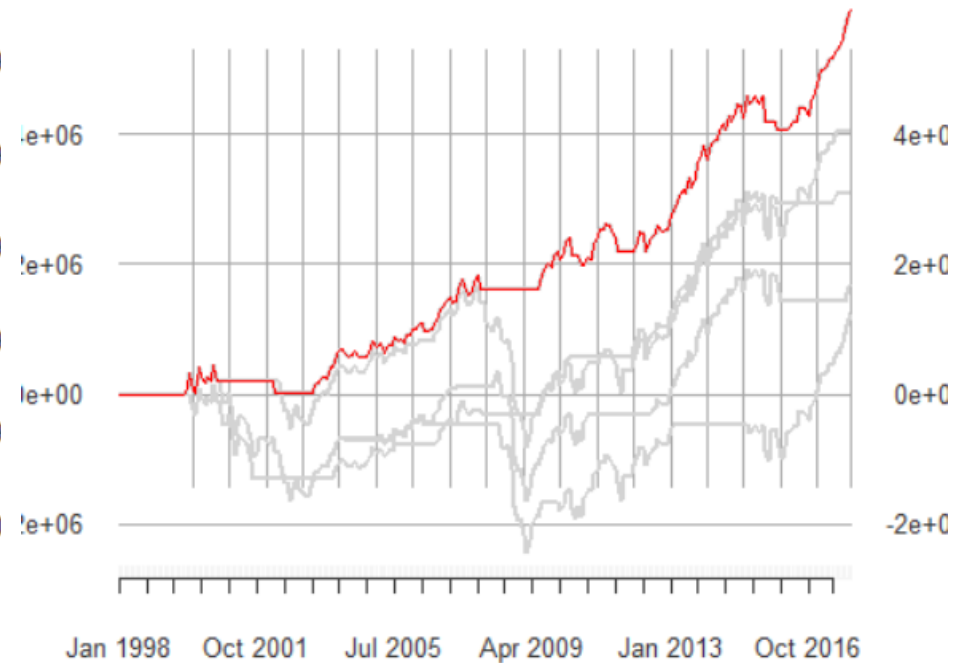
# txnsim Example

- Faber strategy, with four simulations
- Compare mcsim vs txnsim, with replacement:

4 replicates with replacement and block length 1



longtermDH txnsim cumulative P&L 4 reps with replacement





- txnsim also has
  - Summary of confidence intervals
  - Cumulative P&L distributions (quantile(.))
  - Histograms of various performance measures with CI's overlaid
- See example: CFRM522\_018a\_FaberStrategy\_mcsim\_txnsim.R

# Summary

- Overview of Monte Carlo methods
- `mcsim(.)` examples
- Generating results and plots
- Sampling without replacement not as useful as with replacement; `mcsim(.)` assumes daily statistics and will not be as realistic in the latter case
- `txnsim(.)` examples
- Generating results and plots
- Simulations better reflect characteristics of the strategy rather than just rearranging P&L values
- Can be a useful tool for visualizing skill vs luck