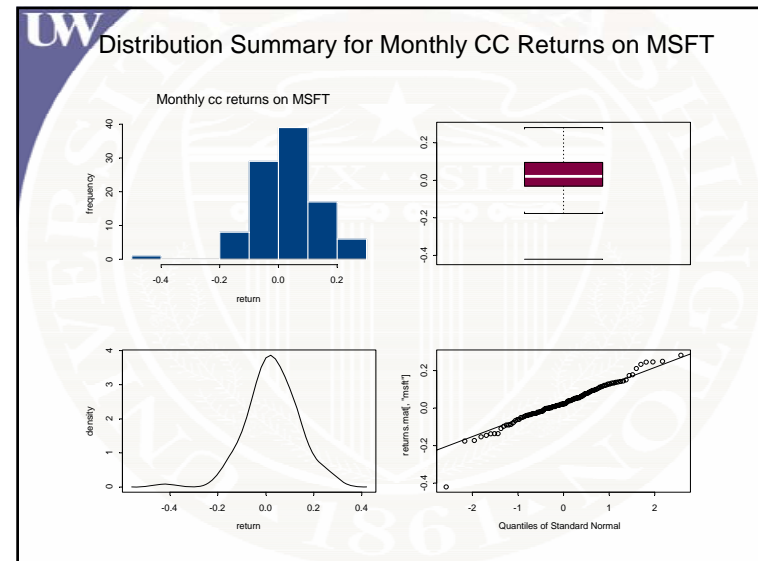
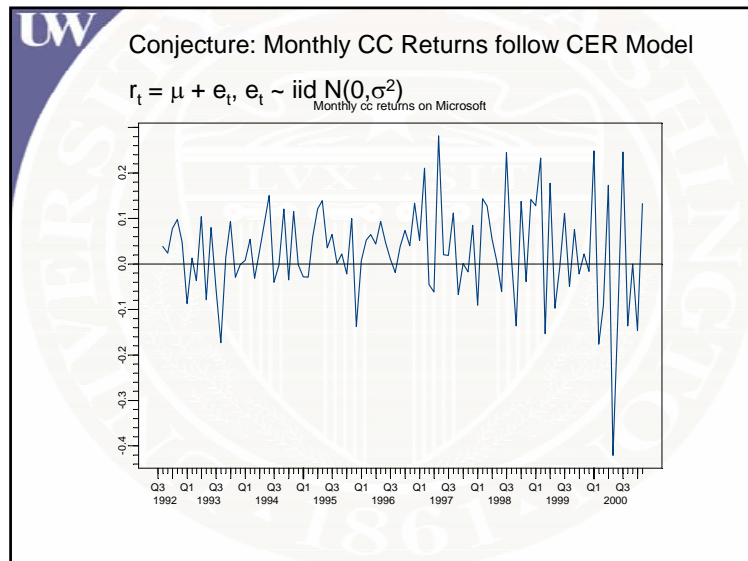
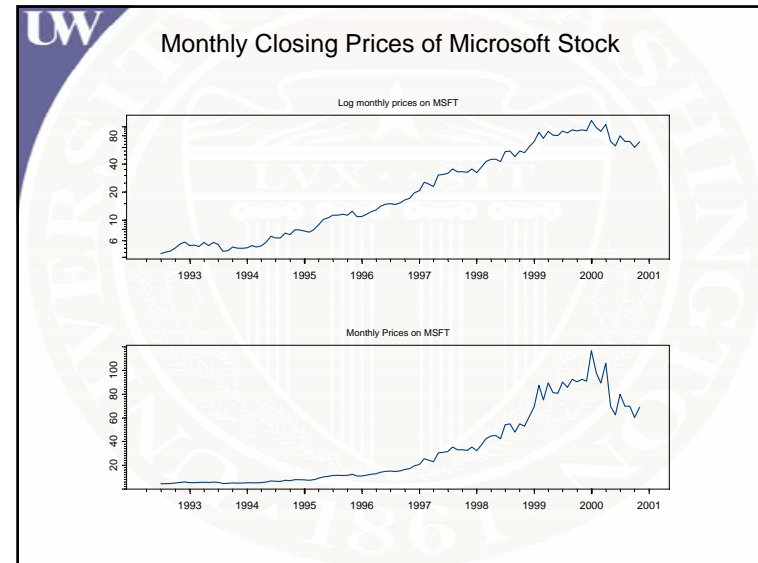
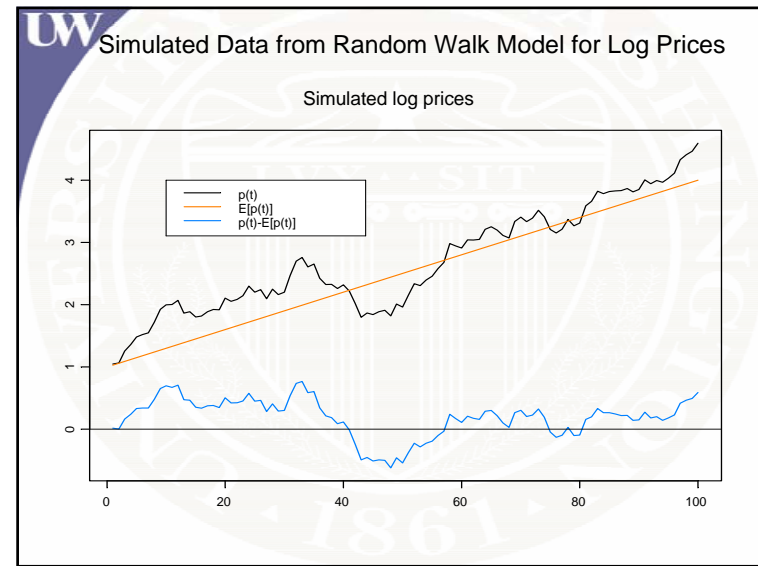
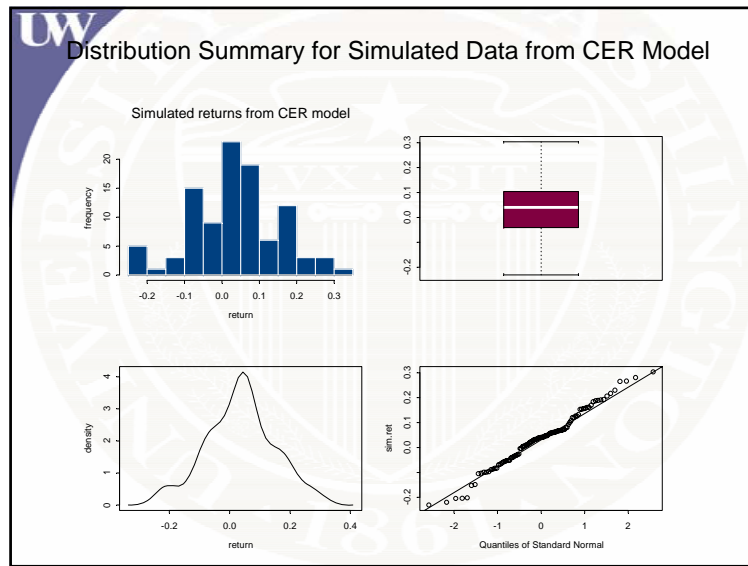
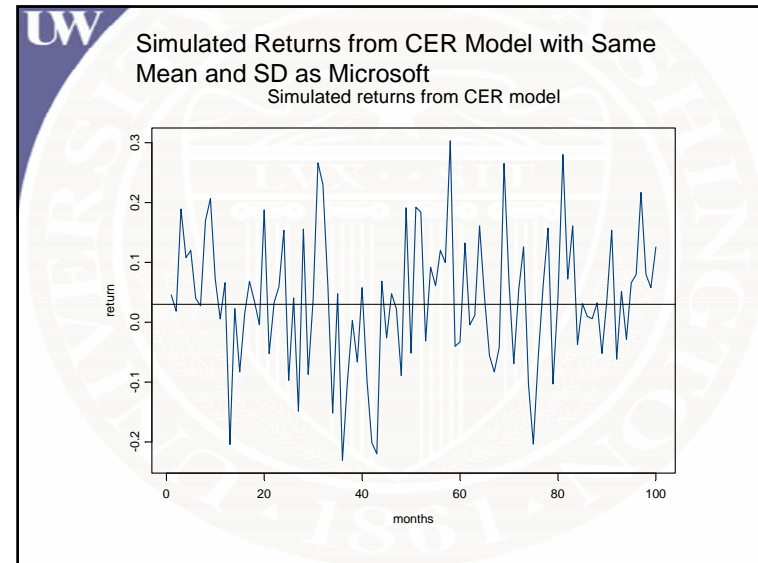
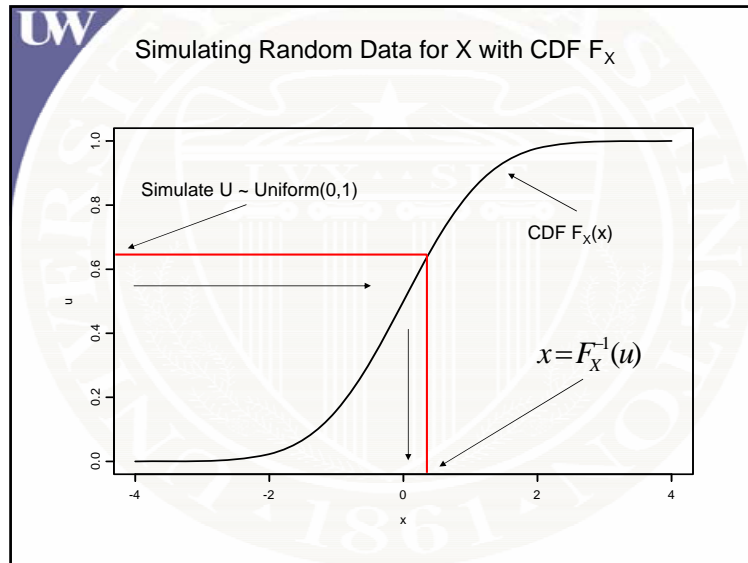


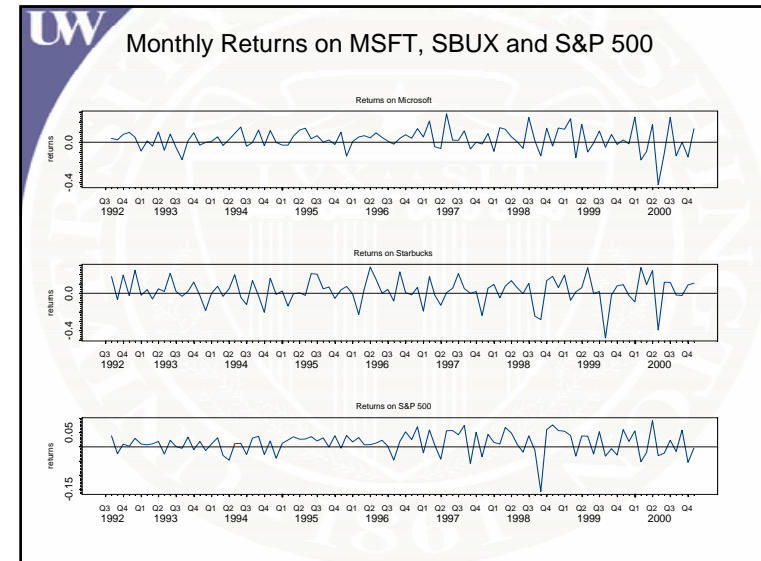
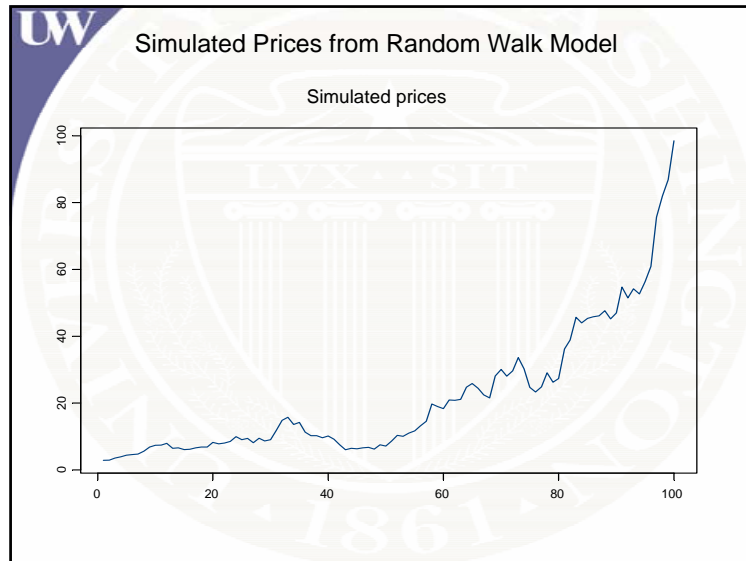
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# Constant Expected Return Model

Econ 424  
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Updated: October 24, 2006







### CER Model Estimates

```

> muhat.vals
  sbux  msft  sp500
0.02777 0.02756 0.01253

> sigma2hat.vals = colVars(returns.mat)
> sigma2hat.vals
  sbux  msft  sp500
0.01846 0.01141 0.001432

> sigmahat.vals = colStdevs(returns.mat)
> sigmahat.vals
  sbux  msft  sp500
0.1359 0.1068 0.03785

```

### CER Model Estimates

```

> cov.mat = var(returns.ts)
> cor.mat = cor(returns.ts)

> covhat.vals = cov.mat[lower.tri(cov.mat)]
> rhohat.vals = cor.mat[lower.tri(cor.mat)]
> names(covhat.vals) = names(rhohat.vals) =
+ c("sbux,msft","sbux,sp500","msft,sp500")

> covhat.vals
  sbux,msft  sbux,sp500  msft,sp500
0.00403  0.00215  0.00224

> rhohat.vals
  sbux,msft  sbux,sp500  msft,sp500
0.2777  0.4197  0.5551

```

## Estimated Standard Errors

```

> se.muhat = sigma2hat.vals/sqrt(nobs)
> rbind(muhat.vals,se.muhat)
      sbux      msft      sp500
muhat.vals 0.0277 0.0275 0.01253
se.muhat   0.0135 0.0106 0.00378

> se.sigma2hat = sigma2hat.vals/sqrt(nobs/2)
> rbind(sigma2hat.vals,se.sigma2hat)
      sbux      msft      sp500
sigma2hat.vals 0.01845 0.01141 0.00143
se.sigma2hat   0.00261 0.00161 0.00020

> se.sigmahat = sigma2hat.vals/sqrt(2*nobs)
> rbind(sigmahat.vals,se.sigmahat)
      sbux      msft      sp500
sigmahat.vals 0.1358 0.1068 0.0378
se.sigmahat   0.0096 0.0075 0.0026

```

## Estimated Standard Errors

```

> se.rhohat = (1-rhohat.vals^2)/sqrt(nobs)
> rbind(rhohat.vals,se.rhohat)
      sbux,msft  sbux,sp500  msft,sp500
rhohat.vals 0.2777 0.4197 0.5551
se.rhohat 0.0922 0.0823 0.0691

```

- ## Stylized Facts for the Estimation of CER Model Parameters
- The mean is not estimated very precisely
    - Large standard errors relative to size of mean estimates
  - Standard deviations and correlations are estimated more precisely than the mean

## 95% Confidence Intervals

```

> mu.lower = muhat.vals - 2*se.muhat
> mu.upper = muhat.vals + 2*se.muhat
> mu.width = mu.upper - mu.lower
> cbind(mu.lower,mu.upper,mu.width)
      mu.lower mu.upper mu.width
sbux 0.0006 0.0549 0.0543
msft 0.0061 0.0489 0.0427
sp500 0.0049 0.0201 0.0151

> sigma.lower = sigmahat.vals - 2*se.sigmahat
> sigma.upper = sigmahat.vals + 2*se.sigmahat
> sigma.width = sigma.upper - sigma.lower
> cbind(sigma.lower,sigma.upper,sigma.width)
      sigma.lower sigma.upper sigma.width
sbux 0.1166 0.1550 0.0384
msft 0.0917 0.1219 0.0302
sp500 0.0324 0.0431 0.0107

```

## 95% Confidence Intervals

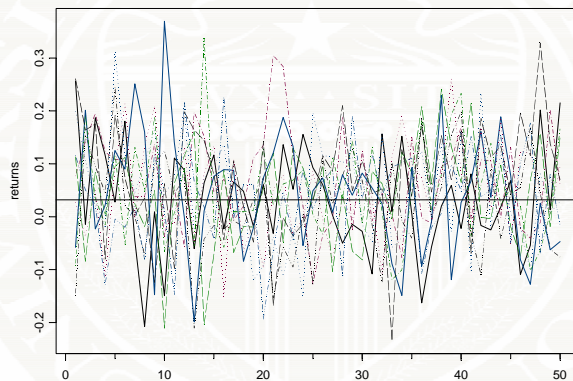
```
> rho.lower = rhohat.vals - 2*se.rhohat
> rho.upper = rhohat.vals + 2*se.rhohat
> rho.width = rho.upper - rho.lower
> cbind(rho.lower,rho.upper,rho.width)
```

	rho.lower	rho.upper	rho.width
sbux,msft	0.0931	0.4622	0.3691
sbux,sp500	0.2549	0.5845	0.3295
msft,sp500	0.4167	0.6934	0.2767

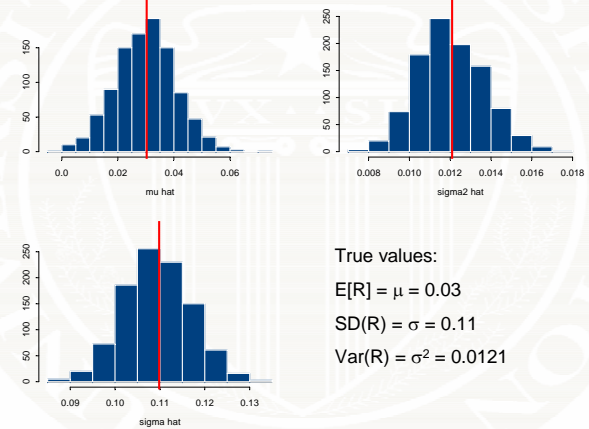
## Monte Carlo Simulation Loop

```
mu = 0.03
sd = 0.11
n.obs = 100
n.sim = 1000
set.seed(111)
sim.means = rep(0,n.sim) # initialize vectors
sim.vars = rep(0,n.sim)
sim.sds = rep(0,n.sim)
for (sim in 1:n.sim) {
  sim.ret = rnorm(n.obs,mean=mu,sd=sd)
  sim.means[sim] = mean(sim.ret)
  sim.vars[sim] = var(sim.ret)
  sim.sds[sim] = sqrt(sim.vars[sim])
}
```

10 simulated samples from CER model



Histograms of 1000 Monte Carlo Estimates



## Monte Carlo Evaluation of Bias

```
> mean(sim.means)      # true mean = 0.03
[1] 0.0298
> mean(sim.means) - mu # estimate of bias
[1] -0.0001

> mean(sim.vars)      # true variance = 0.012
[1] 0.0120
> mean(sim.vars) - sd^2 # estimate of bias
[1] 0.0020

> mean(sim.sds)      # true SD = 0.11
[1] 0.1093
> mean(sim.sds) - sd # estimate of bias
[1] 0.0093
```

## Monte Carlo Evaluation of Estimated Standard Error

```
> stdev(sim.means) # SD of mu estimates across 1000
[1] 0.0106 # Monte Carlo experiments
> se.muhat["msft"]
msft
0.0107 # SE estimate from formula

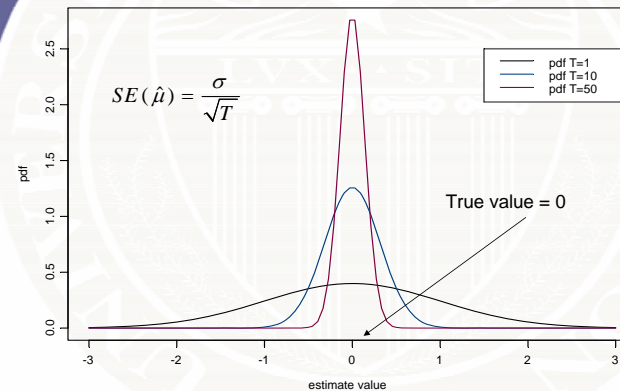
> stdev(sim.vars) # SD of sigma^2 estimates across 1000
[1] 0.0016 # Monte Carlo experiments
> se.sigma2hat["msft"]
msft
0.0016 # SE estimate from formula

> stdev(sim.sds) # SD of sigma estimates across 1000
[1] 0.0075 # Monte Carlo experiments
> se.sigmahat["msft"]
msft
0.0075 # SE estimate from formula
```

## Monte Carlo Evaluation of 95% Confidence Interval Coverage

```
mu = 0.03
sd = 0.11
n.sim = 1000
set.seed(111)
mu.lower = rep(0,n.sim) # initialize vectors
mu.upper = rep(0,n.sim)
for (sim in 1:n.sim) {
  sim.ret = rnorm(n.obs,mean=mu,sd=sd)
  mu.hat = mean(sim.ret)
  se.muhat = stdev(sim.ret)/sqrt(n.obs)
  mu.lower[sim] = mu.hat - 2*se.muhat
  mu.upper[sim] = mu.hat + 2*se.muhat
  in.interval = (mu >= mu.lower) & (mu <= mu.upper)
}
in.interval = (mu >= mu.lower) & (mu <= mu.upper)
sum(in.interval)/n.sim
> 0.934
```

## Graphical Illustration of Consistency



```

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Estimating VaR in CER Model

# estimate quantiles from CER model
> qhat.05 = muhat.vals + sigmahat.vals*qnorm(0.05)
> qhat.05
      sbux      msft      sp500
-0.19571 -0.14815 -0.049717

# estimate 5% VaR
> W0 = 100000
> VaRhat.05 = (exp(qhat.05)-1)*W0
> VaRhat.05
      sbux      msft      sp500
-17775 -13769 -4850.1

```

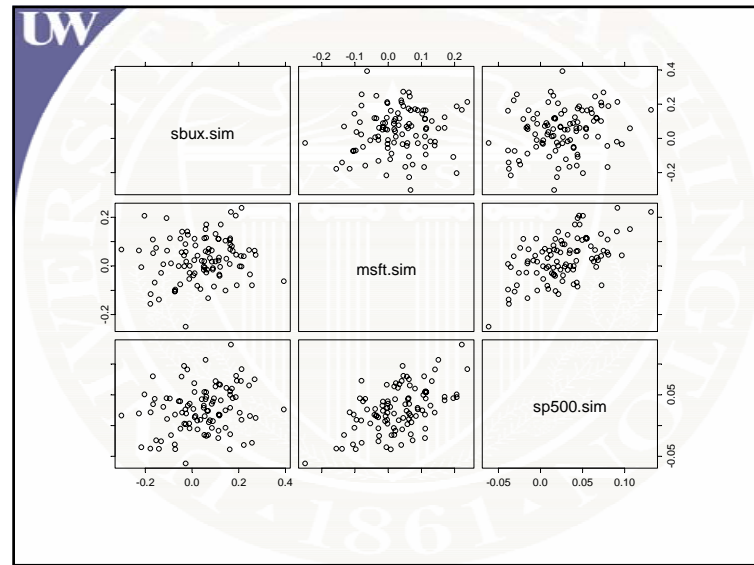
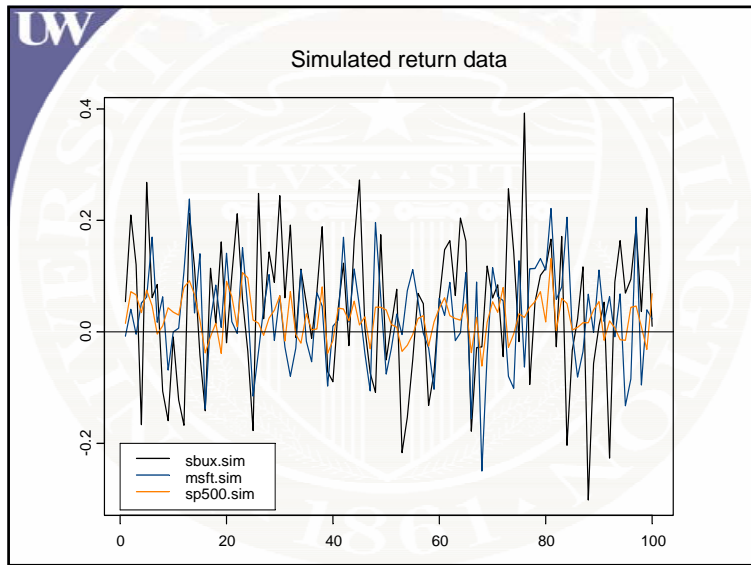
```

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Monte Carlo Simulation: 3 assets

# simulate data to match SBUX, MSFT, SP500
> muhat.vals
      sbux      msft      sp500
0.027768 0.027564 0.012534
> cov.mat
      sbux      msft      sp500
sbux 0.0184589 0.0040306 0.0021583
msft 0.0040306 0.0114112 0.0022442
sp500 0.0021583 0.0022442 0.0014323

> nobs = 100
> set.seed(123)
> sim.e = rmvnorm(nobs,mean=rep(0,3),cov=cov.mat)
> sim.ret = muhat.vals + sim.e
> colIds(sim.ret) =
+ paste(colIds(returns.mat), ".sim", sep="")

```



```

# generate 1000 samples from CER and compute correlations
n.obs = 100
n.sim = 1000
set.seed(111)
sim.corr = matrix(0,n.sim,3) # initialize vectors
colIds(sim.corr) =
c("sbux,msft","sbux,sp500","msft,sp500")

for (sim in 1:n.sim) {
  sim.ret = rmvnorm(n.obs,mean=muhat.vals,cov=cov.mat)
  cor.mat = cor(sim.ret)
  sim.corr[sim,] = cor.mat[lower.tri(cor.mat)]
}

```

