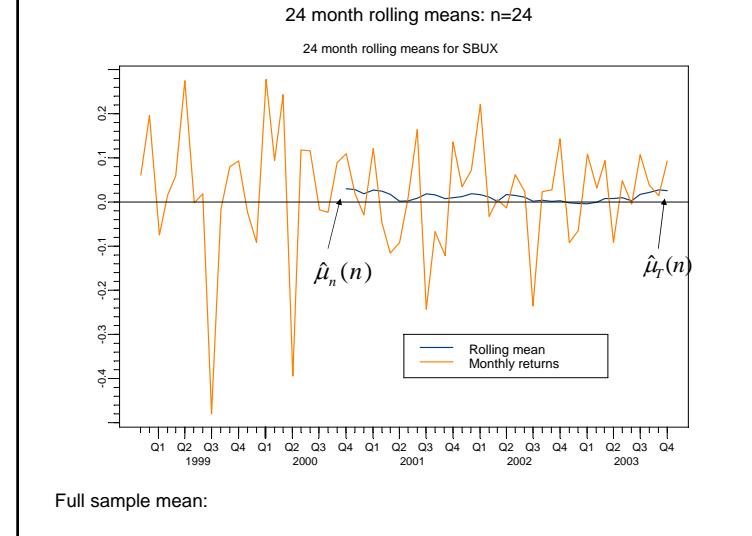


```

# compute rolling means over 24 month windows

> roll.muhat =
aggregateSeries(si.ts[, "sbux"], moving=24,
+ adj=1, FUN=mean)
> class(roll.muhat)
[1] "timeSeries"
> roll.muhat[1:5]
  Positions      sbux
Oct 2000  0.03009224
Nov 2000  0.02837211
Dec 2000  0.01897018
Jan 2001  0.02713285
Feb 2001  0.02453311

```

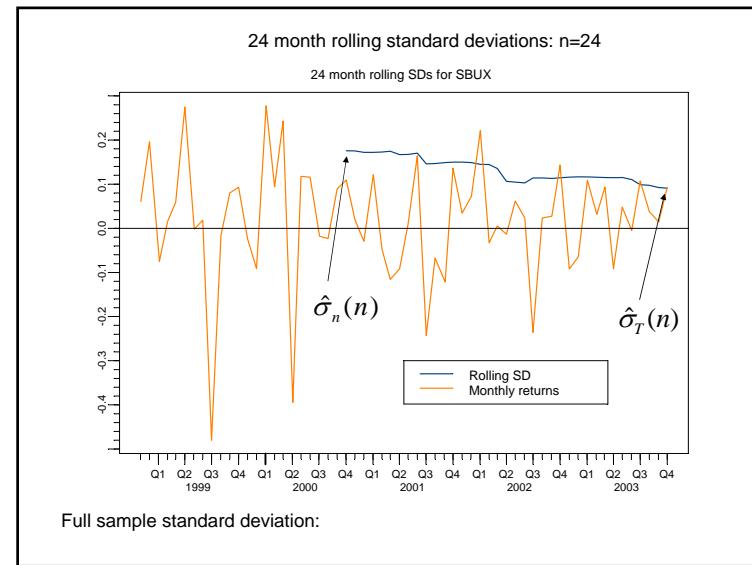


```

# compute rolling standard deviations over
# 24 month windows

> roll.sigmahat =
aggregateSeries(si.ts[, "sbux"], moving=24,
+ adj=1, FUN=stdev)
> roll.sigmahat[1:5]
  Positions      sbux
Oct 2000  0.1755684
Nov 2000  0.1754568
Dec 2000  0.1720834
Jan 2001  0.1720927
Feb 2001  0.1727502

```

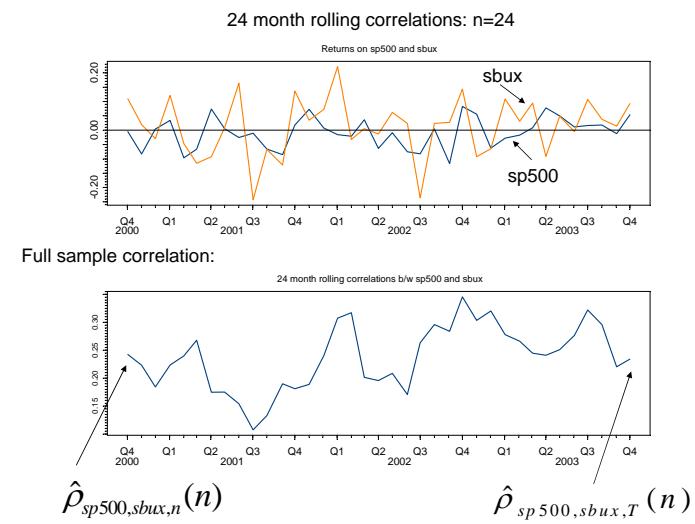


```

# function to compute pairwise correlation
rhohat = function(x) {
  cor(x)[1,2]
}

# compute rolling correlations b/w sp500 and
# sbux
> roll.rhohat =
aggregateSeries(si.ts[,c("sp500","sbux")],
+               moving=24,adj=1,together=T,
+               FUN=rhohat)
> roll.rhohat[1:5]
Positions      1
Oct 2000  0.2423265
Nov 2000  0.2233655
Dec 2000  0.1842506
Jan 2001  0.2234648
Feb 2001  0.2397327

```



```

# compute rolling estimates of alpha and beta from si
# model
# use lsfit function instead of lm function
roll.si = function(x) {
  ans = lsfit(x[,2],x[,1])
  ans$coeff
}
> roll.fit =
aggregateSeries(si.ts[,c("sbux","sp500")],
+               moving=24,adj=1,together=T,
+               colnames=c("alpha","beta"),
+               FUN=roll.si)
> roll.fit[1:5]
Positions      alpha      beta
Oct 2000  0.01917849 0.9953474
Nov 2000  0.02399980 0.8583382
Dec 2000  0.01684467 0.7138994
Jan 2001  0.02476536 0.86999310
Feb 2001  0.02448659 0.8592915

```

