

## MC-Simulation for pathes in Heston's stochastic volatility model

### References:

S. L. Heston, A closed-form solution for options with stochastic volatility ..., Review of Financial Studies 6, 327 (1993)

For applications on timeseries (instead of option prices) see: Yakovenko et al (2002),

Comparison between the probability distribution of returns in the Heston model and

empirical data for stock indexes at <http://lanl.arXiv.org/abs/cond-mat/0211050>  
[http://www2.physics.umd.edu/~yakovenk/personal\\_data/publications.html](http://www2.physics.umd.edu/~yakovenk/personal_data/publications.html)

Maple seems to be a little bit slow for that, but i always wanted to have it in a worksheet.

Have fun with it!

Axel

```
[ > restart; with(stats): # UseHardwareFloats := true:
The model is formulated in terms of stochastic differential equations, the SDEs
allow correlation between spot returns and volatility (both are Brownian
motions).

I do not use it, but as different notations are in use for the parameters, here
they are:
> 'dS[t]=mu*S[t]*dt + sqrt(v[t])*S[t]*dB[t]';
'dv[t]=kappa*(theta-v[t])*dt+sigma*sqrt(v[t])*dZ[t]';
'dZ[t]=rho*dB[t]+sqrt(1-rho^2)*dW[t]'; 'cov(dB[t],dZ[t])=rho*dt':


$$dS_t = \mu S_t dt + \sqrt{v_t} S_t dB_t$$


$$dv_t = \kappa (\theta - v_t) dt + \sigma \sqrt{v_t} dZ_t$$


$$dZ_t = \rho dB_t + \sqrt{1 - \rho^2} dW_t$$


Take a discretization for that
> 'S[k+1]='S[k] + mu*S[k]*dt + sqrt(v[k]*dt) *S[k]* B[k]';
'v[k+1]='v[k]+kappa*(theta-v[k])*dt+sigma*sqrt(v[k]*dt)*Z[k]';
'Z[k]='rho*B[k]+sqrt(1-rho^2)*W[k]';


$$S_{k+1} = S_k + \mu S_k dt + \sqrt{v_k dt} S_k B_k$$


$$v_{k+1} = v_k + \kappa (\theta - v_k) dt + \sigma \sqrt{v_k dt} Z_k$$


$$Z_k = \rho B_k + \sqrt{1 - \rho^2} W_k$$


Now choose model parameters and starting values ...
(for real markets one would need other parameter values)
> mu:= 0.025; # drift = rates
sigma:=0.40; # vol of vol
kappa:=2.00; # mean reversion
theta:=0.04; # long run variance
rho:= -0.50; # correlation
#mu:=0.025; sigma:=0.6; kappa:=0.24; theta:=0.02;rho:= -0.64;
``;

S0:=100.0; # start values
V0:= 0.04; # instant vol

mu := 0.025
sigma := 0.40
kappa := 2.00
theta := 0.04
rho := -0.50

S0 := 100.0
V0 := 0.04
```

□ ... and let it run:

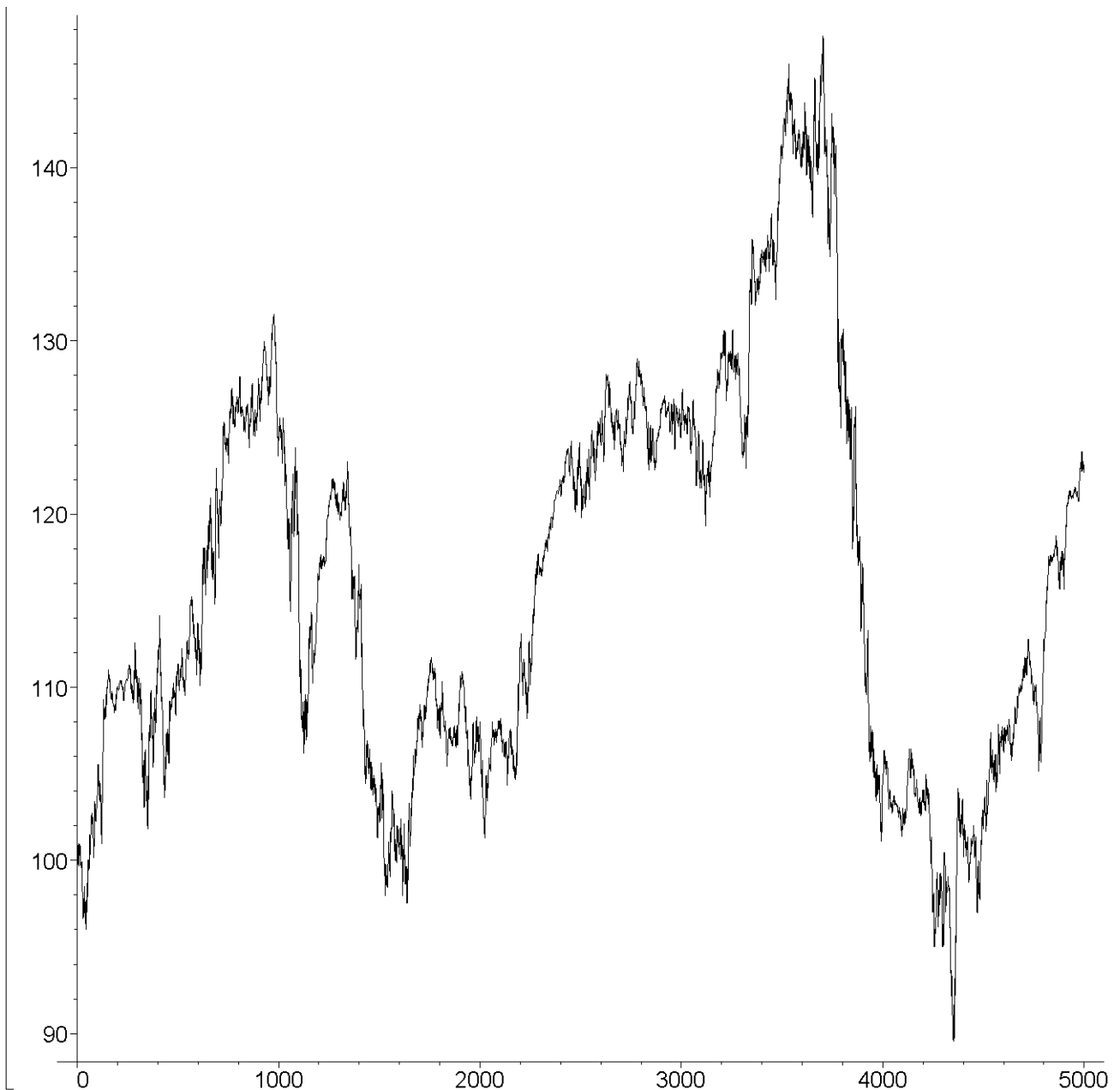
```
> N:=5000:      # number of steps
dt:=1/252.0:   # time step = step size

st:=time():
S:=array(1..N):
S[1]:=S0:
v_t:=V0:
randomize():
B:=[stats[random,normald[0,1]](N+1)]:
W:=[stats[random,normald[0,1]](N+1)]:
Z:= zip((x,y) -> evalf( rho*x+sqrt(1-rho^2)*y ),B,W):
`runtime random generation seconds`=time()-st;
for k from 1 to N-1 do
  S[k+1]:=evalf(S[k] + mu*S[k]*dt + sigma*sqrt(v_t*dt) *S[k]*B[k]):
  v_t:=abs(evalf(v_t+kappa*(theta-v_t)*dt+sigma*sqrt(v_t*dt)*Z[k]));
end do: k:='k':
`runtime total seconds`=time()-st; # 15 - 25 seconds
``;
`years`=evalf[3](N*dt);
plots[listplot](S);
```

runtime random generation seconds = 4.418

runtime total seconds = 6.120

years = 19.8



compare it with riskless rates

```
> `bonds`='S0*(1+mu*dt)^N'; %: evalf[3](%); `final spot`=evalf[3](S[N]); ``;
```

```
      bonds = S0 (1 + μ dt)N
```

```
      bonds = 164.
```

```
      final spot = 122.
```

statistics for the logarithmic returns:

skewed ( not equal 0 ) with excess kurtosis ( greater than 3 ) and a 'clustering' in the graph

```
> returns:=[seq(evalf(ln(S[k]/S[k-1])),k=2..N)]: k:='k':
```

```
  `mean`=describe[mean](returns);
```

```
  `variance`=describe[variance](returns); #
```

```
  `vola`=sqrt(describe[variance](returns));
```

```
  `skewness`=describe[skewness](returns);
```

```
  `kurtosis`=describe[kurtosis](returns);
```

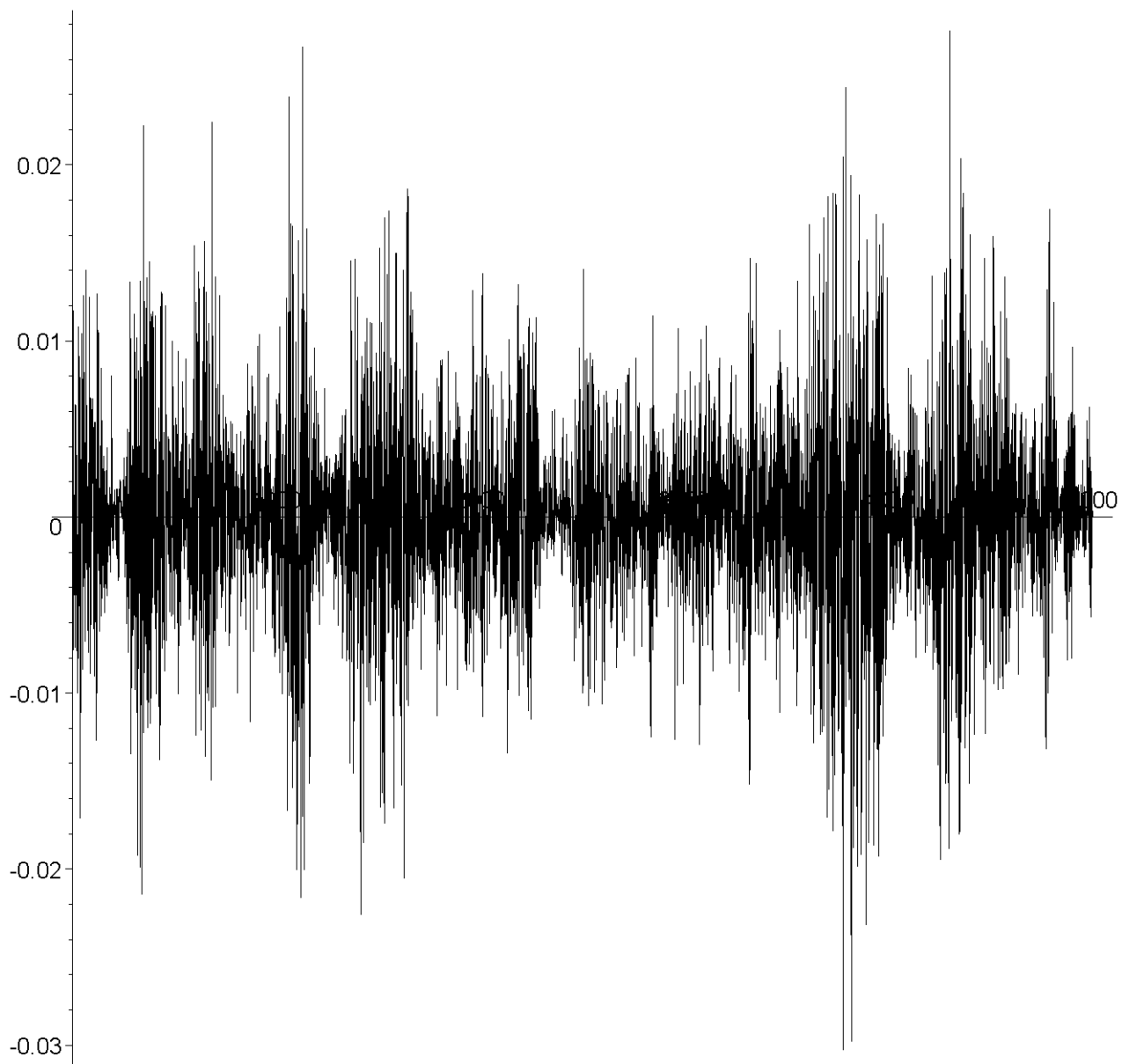
```
  plots[listplot](returns);
```

```
      mean = 0.00004046391882
```

```
      variance = 0.00002810221470
```

skewness = -0.05332529307

kurtosis = 5.383420543



```
[ >
[ Variant: Box-Muller
[ > N:=5000:      # number of steps
[ dt:=1/252.0:  # time step = step size

[ st:=time():
[ S:=array(1..N):
[ S[1]:=S0:
[ v_t:=V0:

[ # take 2 variates uniform and independently distributed in (0,1):
[ randomize(): U1:=[stats[random,uniform](N+1)]:
[ randomize(): U2:=[stats[random,uniform](N+1)]:
[ # use Box-Muller to get a 2 normal variates (mean=0, variance=1)
[ biv:=zip( (u1,u2) ->
[   evalf([sqrt(-2*ln(u1))*cos(2*Pi*u2),sqrt(-2*ln(u1))*sin(2*Pi*u2)]) ,
[ U1,U2):
[ #bivcor:=map( x->[x[1],rho*x[1]+sqrt(1-rho^2)*x[2]],biv):
```

```

# now correlated them as above
B:=map(x -> x[1],biv):
Z:=map(x -> evalf(rho*x[1]+sqrt(1-rho^2)*x[2]), biv):
`runtime random generation seconds`=time()-st;
for k from 1 to N-1 do
  S[k+1]:=evalf(S[k] + mu*S[k]*dt + sigma*sqrt(v_t*dt) *S[k]*B[k]):
  v_t:=abs(evalf(v_t+kappa*(theta-v_t)*dt+sigma*sqrt(v_t*dt)*Z[k]));
end do: k:='k':
`runtime seconds`=time()-st; # 15 - 25 seconds
``;
`years`=evalf[3](N*dt);
plots[listplot](S);

```

runtime random generation seconds = 17.130

runtime seconds = 19.185

years = 19.8

