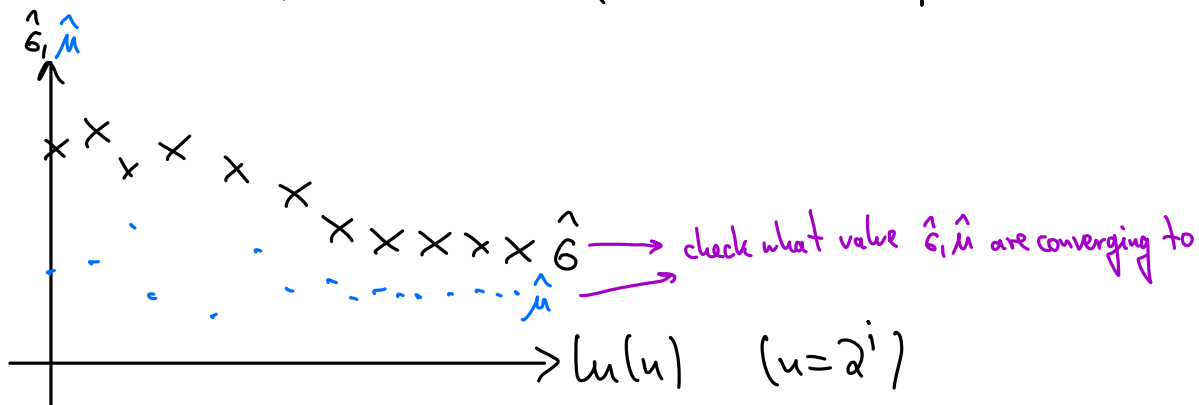


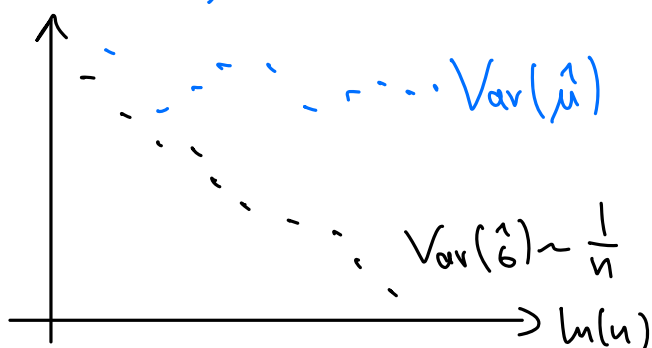
Prof. Dr. Soeren Petrat

Homework:

a) one realization of GBM, size  $2^k$ then estimate  $\hat{\mu}, \hat{\sigma}$  for every  $2^i$ -th sample point,  $i=0, \dots, k-1$ 

python: "semilogx(...)" for plot with logarithmic x-axis

b) ensemble of GBMs with some parameters

 $\hookrightarrow \text{Var}(\hat{\sigma}), \text{Var}(\hat{\mu})$  $\ln \text{Var}(\hat{\sigma}), \ln \text{Var}(\hat{\mu})$   $\uparrow$  ensemble variance

c) "Backtracking"

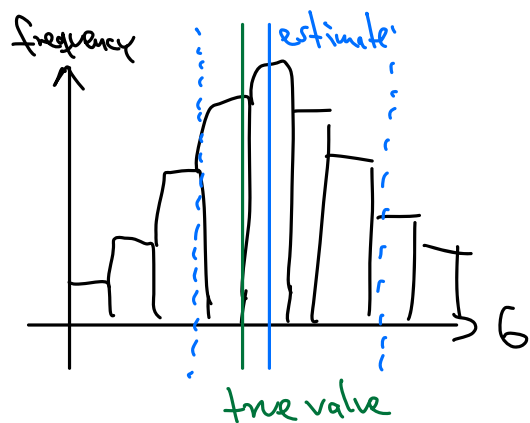
• given a single time series from part a)  $\rightarrow$  compute  $\hat{\mu}, \hat{\sigma}$

- generate ensemble of GBMs with these parameters

- compute  $\text{Var}(\hat{\mu}), \text{Var}(\hat{\sigma})$

=> test how reliable estimate was

python: `hist(sigma-distribution, number of bins, histtype = 'stepfilled')`



← very thin for  $\sigma$   
but wide for  $\mu$  →

d), e), f) consider some noise sources:

- periodic noise:  $S_{\text{per}} = S + c, \sqrt{\Delta t} \overset{\leftarrow \text{GBM}}{\sin(2\pi f \text{ arange}(N+1))}$

- Gaussian noise:  $S_{\text{Gauss}} = S + c, \sqrt{\Delta t} \text{ normal}(0, 1, N+1)$

- how does the noise change estimates for  $\hat{\mu}, \hat{\sigma}$ ?

- normality?

- independence?