

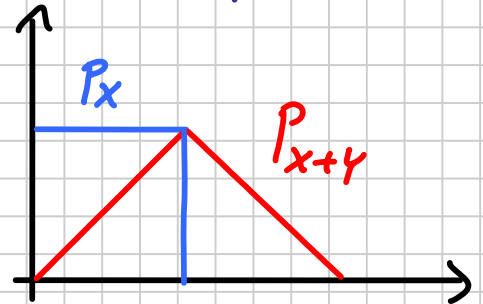
Vorübergehend englischsprachige Notizen, übernommen von der Vorlesung "Computer Simulations in Statistical Physics": [http://komet337.physik.uni-mainz.de/Bluemer/lectures\\_WS2006](http://komet337.physik.uni-mainz.de/Bluemer/lectures_WS2006)

Full probability distribution of sum/average?

$$\begin{aligned}
 P_{X+Y}(z) &= \int dx p_x(x) \int dy p_y(y) \delta(z - (x+y)) \\
 &= \int dx p_x(x) p_y(z-x)
 \end{aligned}$$

specifically for uniformly distributed  $X, Y$ :

$$P_{X+Y}(z) \begin{cases} z & \text{for } 0 \leq z \leq 1 \\ 2-z & \text{" } 1 \leq z \leq 2 \\ 0 & \text{else} \end{cases}$$



Look at higher order cumulants  $\rightarrow$  normal distribution

$$P_x(x) \xrightarrow{N \rightarrow \infty} \frac{1}{\sigma_x \sqrt{2\pi}} \exp\left[-\frac{1}{2} \left(\frac{\bar{x} - \langle x \rangle}{\sigma_x}\right)^2\right]; \quad \sigma_{\bar{x}} = \frac{\sigma}{\sqrt{N}}$$

central limit theorem (uncorrelated case!)

Now: application to simulation / measurements

again: random variables  $X_i$  with same distribution, non necessarily uncorrelated

Problem: probability distribution, mean, variance unknown. Best unbiased estimates?

easy:  $\langle X \rangle \approx \bar{X} = \frac{1}{N} \sum_{i=1}^N X_i$

$\bar{X}$  is (best) unbiased estimator for  $\langle X \rangle$  since

$$\langle \bar{X} \rangle = \frac{1}{N} \sum_{i=1}^N \underbrace{\langle X_i \rangle}_{=\langle X \rangle} = \langle X \rangle$$

(also unbiased:  $\tilde{X} = \sum_{i=1}^N a_i X_i$  with  $\sum_{i=1}^N a_i = 1$ )  
 (also estimator:  $\tilde{\tilde{X}} = \frac{1}{N-1} \sum_{i=1}^N X_i$  (since  $\langle \tilde{\tilde{X}} \rangle = \frac{N}{N-1} \langle X \rangle \xrightarrow{N \rightarrow \infty} \langle X \rangle$ )

difficult: best estimator for error of  $\bar{X}$ ?

(i) unbiased estimator for  $\sigma_x$ ?

$$\begin{aligned} \left\langle \sum_{i=1}^N (X_i - \bar{X})^2 \right\rangle &= N \langle (X_i - \bar{X})^2 \rangle \\ &= N \left\langle \left( X_i - \frac{1}{N} \sum_{j=1}^N X_j \right)^2 \right\rangle \\ &= N \left\langle \left[ (X_i - \langle X \rangle) - \frac{1}{N} \sum_{j=1}^N (X_j - \langle X \rangle) \right]^2 \right\rangle \\ &= N \left\langle \left[ \left(1 - \frac{1}{N}\right) (X_i - \langle X \rangle) - \frac{1}{N} \sum_{\substack{j=1 \\ j \neq i}}^N (X_j - \langle X \rangle) \right]^2 \right\rangle \\ &\stackrel{\text{no corr.}}{=} N \left[ \left(1 - \frac{1}{N}\right)^2 \langle (X_i - \langle X \rangle)^2 \rangle + \frac{1}{N^2} \sum_{\substack{j=1 \\ j \neq i}}^N \langle (X_j - \langle X \rangle)^2 \rangle \right] \\ &= \frac{(N-1)^2}{N} \sigma^2 + \frac{N-1}{N} \sigma^2 \\ &= \sigma^2 \frac{[(N-1)+1](N-1)}{N} = (N-1) \sigma^2 \end{aligned}$$

Thus,  $\sigma_{est}^2 = \frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2$  is an unbiased estimator for the variance  $\sigma^2$  if the data is uncorrelated.

(Above estimator systematically underestimates  $\sigma^2$  for autocorrelated data ( $\frac{1}{N}$ -effect). Better:  $\sigma_{est,imp}^2 = \frac{\sum (x_i - \bar{x})^2}{N - \gamma}$ .)

Step (ii) variance of mean value  $\bar{x}$ :

$$\begin{aligned} \langle (\bar{x} - \langle x \rangle)^2 \rangle &= \langle \left( \frac{1}{N} \sum_{i=1}^N x_i - \langle x \rangle \right)^2 \rangle \\ &= \langle \left[ \frac{1}{N} \sum_{i=1}^N (x_i - \langle x \rangle) \right]^2 \rangle \\ &= \frac{1}{N^2} \sum_{i=1}^N \sum_{j=1}^N \langle (x_i - \langle x \rangle) (x_j - \langle x \rangle) \rangle \\ &\quad \left( \text{gen. Def: } C_{A,B} = \frac{\langle (A - \langle A \rangle) (B - \langle B \rangle) \rangle}{\sigma_A \sigma_B} \right) \\ &\quad \text{covariance} \\ &= \frac{\sigma^2}{N^2} \sum_{i=1}^N \sum_{j=1}^N C_{i,j} \\ &\quad \text{translation} \\ &\quad \approx \frac{\sigma^2}{N^2} \sum_{i=1}^N \sum_{j=1}^N C_{i-j} \quad \begin{matrix} k=i-j \\ j=i-k \end{matrix} \\ &= \frac{\sigma^2}{N^2} \sum_{i=1}^N \sum_{k=i-N}^i C_k \\ &\quad \begin{matrix} C_k \approx 0 \\ \text{for } |k| > k \ll N \end{matrix} \\ &\quad \approx \frac{\sigma^2}{N^2} \sum_{i=1}^N \sum_{k=-\infty}^{\infty} C_k = \frac{\sigma^2}{N} \gamma \end{aligned}$$

autocorrelation time  $\gamma = \sum_{k=-\infty}^{\infty} C_k = \sum_{k=1}^{\infty} (C_k + C_{-k}) + 2 \sum_{k=1}^{\infty} C_k$

$$C_k \approx \frac{\frac{1}{N-k-1} \sum_{i=1}^{N-k} (x_i - \bar{x})(x_{i+k} - \bar{x})}{\frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2}$$

auto correlation  
function

**Beispiel Datenanalyse:** Vorgegeben sind 6 Zeitreihen ([http://komet337.physik.uni-mainz.de/Bluemer/lectures\\_SS2007#Datenanalyse](http://komet337.physik.uni-mainz.de/Bluemer/lectures_SS2007#Datenanalyse)), diese sollen analysiert werden, z.B. mit dem Statistik-Tool stats\_v1\_4.

**Mittelwert mit Fehler:**

```
prompt> stats_v1_4 -a < data_set1_1000.dat
```

Average: 5.7610558, variance: 0.6616717, error: 0.025735851

Korrelation time: 1.105017, corrected error: 0.027053472, transient: -2.6028452e-05

```
prompt> stats_v1_4 -a < data_set1_10000.dat
```

Average: 5.7484945, variance: 0.64816593, error: 0.0080512779

Korrelation time: 1, corrected error: 0.0080512779, transient: 1.0067773e-06

**Autokorrelationsfunktion:**

```
prompt> stats_v1_4 -c < data_set1_10000.dat | head -11
```

# Autocorrelation function: (i,c(i))

0 1.000000

1 -0.006890

2 0.004378

3 0.001233

4 -0.005135

5 -0.010988

6 0.013733

7 -0.013954

8 0.006322

9 0.001616

10 -0.005462