

Multivariate Normal dist (MVN):

$$Z_i \sim \text{NID}(0,1), i=1, \dots, n$$

$$\text{const } a_{ij}, i=1, \dots, m, j=1, \dots, n \text{ and } \mu_i, i=1, \dots, m$$

$$[] \quad X_i = a_{i1}Z_1 + \dots + a_{in}Z_n + \mu_i, i=1, \dots, m$$

then the r.v's X_1, \dots, X_m are said to have a MVN.

$$E(X_i) = \mu_i \quad ; \quad V(X_i) = \sum_{j=1}^n a_{ij}^2$$

$$\text{mgf of } X_1, \dots, X_m$$

$$\phi(t_1, \dots, t_m) = E(e^{t_1 X_1 + \dots + t_m X_m})$$

$$E\left(\sum_{i=1}^m t_i X_i\right) = \sum_{i=1}^m t_i \mu_i$$

$$V\left(\sum_{i=1}^m t_i X_i\right) = \text{Cov}\left(\sum_{i=1}^m t_i X_i, \sum_{j=1}^m t_j X_j\right)$$

$$= \sum_{i=1}^m \sum_{j=1}^m t_i t_j \text{Cov}(X_i, X_j)$$

$$\phi(t_1, \dots, t_m) = e^{\sum_{i=1}^m t_i \mu_i + \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^m t_i t_j \text{Cov}(X_i, X_j)}$$

joint dist's of X_1, \dots, X_m is completely determined
from the knowledge of $\mu_i = E(X_i)$ and $\text{Cov}(X_i, X_j)$
 $i, j = 1, \dots, m$

Gaussian Process is normal process.

A SP $(X(t), t \geq 0)$ is Gaussian process if $X(t_1), \dots, X(t_n)$ has MVN $\forall t_1, \dots, t_n$.

→ $(B(t), t \geq 0)$ BM process

$$\left. \begin{array}{l} B(t_1) \\ B(t_2) = B(t_1) + \overbrace{(B(t_2) - B(t_1))}^{\text{indep}} \\ \vdots \end{array} \right\} \Rightarrow \text{BM is a Gaussian process}$$

SBM/Weiner process $W(t)$ is Gaussian process
 $W(t) \sim N(0, t)$ $E(W(t)) = 0$

Für $s < t$ $V(W(t)) = t$

$$\text{Cov}(W(s), W(t)) = \text{Cov}(W(s), W(t) - W(s) + W(s))$$

$$= \text{Cov}(W(s), W(t) - W(s)) + \text{Cov}(W(s), W(s))$$

$$= 0 + \underbrace{V(W(s))}_x = 0 + s$$

$$\text{Cov}(W(s), W(t)) = s \wedge t \leq \min(s, t)$$

$$\wedge \rightarrow \min$$

$$V \rightarrow \max$$