### **Random Vectors**

Mohadeseh Shafiei Kafraj<sup>1</sup>

<sup>1</sup>Gatsby Computational Neuroscience Unit University College London

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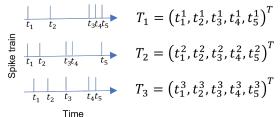
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## Random Vectors: When are they useful?







$$X = (X_1, ..., X_n)^T$$
: A random vector

 $F_x(x)$ : Probability \*Distribution\* Function (\*PDF\*)

 $f_x(x)$ : Probability \*Density\* function (\*pdf\*)



By definition, probability distribution function (PDF) is:

$$F_X(x) = P[X_1 \le x_1, ..., X_n \le x_n]$$

 $x = (x_1, ..., x_n)$  we get:

$$F_X(x) = P[X \le x]$$

we associate the events:

 $X \leq \infty$  with the certain event,  $F_x(\infty) = 1$ , and

 $X \leq -\infty$  with the impossible event,  $F_x(-\infty) = 0$ .



The probability \*density\* function is defined as:

$$f_{x}(x) = \frac{\partial^{n} F_{x}(x)}{\partial x_{1} ... \partial x_{n}}$$

Equivalently we could have defined it as:

$$f_{x}(x) = \lim_{\Delta x_{1} \to 0, \dots, \Delta x_{n} \to 0} \frac{P[x_{1} < X_{1} \le x_{1} + \Delta x_{1}, \dots, x_{n} < X_{n} \le x_{1} + \Delta x_{n}]}{\Delta x_{1} \dots \Delta x_{n}}$$

Therefore,

$$f_x(x)\Delta x_1...\Delta x_n \simeq P[x_1 < X_1 \le x_1 + \Delta x_1, ..., x_n < X_n \le x_1 + \Delta x_n]$$



pdf is defined as:

$$f_x(x) = \frac{\partial^n F_x(x)}{\partial x_1 ... \partial x_n}$$

if we integrate the equation, we obtain:

$$F_{x}(x) = \int_{-\infty}^{x_{1}} ... \int_{-\infty}^{x_{n}} f_{x}(x') dx'_{1} ... dx'_{n} = \int_{-\infty}^{x} f_{x}(x') dx'_{1}$$

more generally:

$$P[B] = \int_{x \in B} f_x(x') dx'$$
, where  $B \subset R^N$ 



Continued ...



## **Expectation Vectors and Covariance Matrices**

The expectation of the vector  $X = (X_1, ..., X_n)^T$  is a vector  $\mu$  whose elements are given by

$$\mu_i = \int_{-\infty}^{\infty} ... \int_{-\infty}^{\infty} x_i f_x(x_1, ..., x_n) dx_1 ... dx_n.$$



## Properties of Covariance Matrices



### The Multidimensional Gaussian Law



## Distribution of the Sample Mean



### Conditional Gaussian distributions



# Marginal Gaussian distributions

