### Multivariate probabilty and statistics

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#### Overview

This presentation is  $\dots$ 

#### Random variables

Random variable is...

#### Random vectors

Random vector is...

### Probability density

(Function, Examples)

## Marginal and joint probabilities

(Definition, examples)

### Conditional probabilities

(continuous case, discrete case)

# Independence

definition

### Expectation

(continuous, discrete (maybe otherway around would be more intuitive?))

#### Variance and covariance

(dim = 1, covariance matrix)

### Independence and covariance

the relation

# Bayesian inference

Baye's rule

### Non informative priors

definition

Bayesian inference as a incremental learning process

Explanation



(statistical model, estimation, ilkelihood & log-lkelihood)

## Maximum likelihood and maximum posteriori

(definition, definition)



Examples from the template on the next slides:

### Tables and Figures

- Use tabular for basic tables see Table 1, for example.
- You can upload a figure (JPEG, PNG or PDF) using the files menu.
- ▶ To include it in your document, use the includegraphics command (see the comment below in the source code).

| Item    | Quantity |
|---------|----------|
| Widgets | 42       |
| Gadgets | 13       |

Table 1: An example table.

#### Readable Mathematics

Let  $X_1, X_2, \ldots, X_n$  be a sequence of independent and identically distributed random variables with  $\mathsf{E}[X_i] = \mu$  and  $\mathsf{Var}[X_i] = \sigma^2 < \infty$ , and let

$$S_n = \frac{X_1 + X_2 + \dots + X_n}{n} = \frac{1}{n} \sum_{i=1}^{n} X_i$$

denote their mean. Then as n approaches infinity, the random variables  $\sqrt{n}(S_n - \mu)$  converge in distribution to a normal  $\mathcal{N}(0, \sigma^2)$ .