

## Recap: Multi-dimensional Gaussian density

- $\mathbf{X} = (X_1, \dots, X_n)^T$  are said to be jointly Gaussian if

$$f_{\mathbf{X}}(\mathbf{x}) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma|^{\frac{1}{2}}} e^{-\frac{1}{2}(\mathbf{x}-\boldsymbol{\mu})^T \Sigma^{-1}(\mathbf{x}-\boldsymbol{\mu})}$$

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- ▶ When  $X, Y$  are jointly Gaussian, the joint density is given by

$$f_{XY}(x, y) = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} e^{-\frac{1}{2(1-\rho^2)}\left(\frac{(x-\mu_x)^2}{\sigma_x^2} + \frac{(y-\mu_y)^2}{\sigma_y^2} - \frac{2\rho(x-\mu_x)(y-\mu_y)}{\sigma_x\sigma_y}\right)}$$

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- ▶ If  $X_1, \dots, X_n$  are jointly Gaussian and  $\mathbf{A}$  is a  $k \times n$  matrix of rank  $k$ , then,  $\mathbf{Y} = \mathbf{A}\mathbf{X}$  is jointly gaussian



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- ▶ **Minkowski's Inequality:**

$$(E|X + Y|^r)^{\frac{1}{r}} \leq (E|X|^r)^{\frac{1}{r}} + (E|Y|^r)^{\frac{1}{r}}$$

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## ► Hoeffding Inequality $X_i$ iid, $X_i \in [a, b]$ , $\forall i$ and $EX_i = \mu$

$$P \left[ \left| \sum_{i=1}^n X_i - n\mu \right| \geq \epsilon \right] \leq 2e^{-\frac{2\epsilon^2}{n(b-a)}}, \quad \epsilon > 0$$

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$$\Rightarrow \lim_{n \rightarrow \infty} P\left[\left|\frac{S_n}{n} - \mu\right| \geq \epsilon\right] = 0, \quad \forall \epsilon > 0$$

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- ▶ We only need marginal distributions of individual  $X_n$  to decide whether a sequence converges to a constant in probability

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- ▶ Hence,  $X_n \xrightarrow{P} 0$

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  3.  $X_n Y_n \xrightarrow{P} XY$
  4.  $g(X_n) \xrightarrow{P} g(X)$  where  $g$  is a continuous function from  $\mathbb{R}$  to  $\mathbb{R}$ .



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► Hence, we can conclude  $P[|X - Y| > c] = 0, \forall c > 0$  and hence  $P[X = Y] = 1$ .

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► All the other properties listed can be proved by following similar arguments



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- ▶ Almost sure convergence is a stronger mode of convergence



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- ▶ We will first look at limits of general sequences of events.



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$$x \in [0, 1] \Rightarrow x \in \cup_{k=n}^{\infty} A_k, \forall n \Rightarrow x \in \cap_{n=1}^{\infty} \cup_{k=n}^{\infty} A_k \Rightarrow x \in \limsup A_n$$

- ▶ Given any  $\epsilon > 0$ ,  $1 + \epsilon \notin [0, 1 + \frac{1}{n})$  if  $\epsilon > \frac{1}{n}$  or  $n > \frac{1}{\epsilon}$ .
- ▶ Hence, given any  $\epsilon > 0$ ,  $\exists n$  such that  $1 + \epsilon \notin \cup_{k=n}^{\infty} A_k$ .

## example

- ▶ Consider the sets  $A_n = [0, 1 + \frac{(-1)^n}{n})$

The sequence is

$$[0, 0), \left[0, 1 + \frac{1}{2}\right), \left[0, 1 - \frac{1}{3}\right), \left[0, 1 + \frac{1}{4}\right) \dots$$

- ▶ Guess:  $\limsup A_n = [0, 1]$  and  $\liminf A_n = [0, 1]$
- ▶ First note that  $[0, 1 + \frac{1}{n+1}) \subset \cup_{k=n}^{\infty} A_k \subset [0, 1 + \frac{1}{n})$ .  
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- ▶ This proves  $\liminf A_n = [0, 1)$
- ▶ Since  $\limsup A_n \neq \liminf A_n$ , this sequence does not have a limit

►  $X_n \xrightarrow{a.s.} X$  iff

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- ▶ We look at an important result that allows us to do this

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► This completes proof of first part of Borel-Cantelli lemma

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- This finally gives us

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- ▶ By Borel-Cantelli lemma

$$\sum_{k=1}^{\infty} P(A_k) < \infty \Rightarrow P(\limsup A_k) = 0 \Rightarrow X_k \xrightarrow{a.s.} X$$

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