

1179: Probability
Lecture 23 — MGFs, Covariance, and
Properties of Bivariate Normal

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December 8, 2021

Announcements

- ▶ HW3 is now available on E3!
 - ▶ Due on 12/17 (Friday), 1pm

Quick Review

$$f_{V_1, V_2}(v_1, v_2) = \frac{1}{\det(A)} \cdot f_{U_1, U_2}\left(A^{-1} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}\right)$$

U_1, U_2, f_{U_1, U_2}

- ▶ Linear transformation of 2 random variables? $\begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = \underbrace{\begin{bmatrix} a & b \\ c & d \end{bmatrix}}_A \begin{bmatrix} U_1 \\ U_2 \end{bmatrix}$
- ▶ How to find the distribution of $X + Y$? (X, Y independent)
 - "Convolution Theorem" $\underbrace{\quad}_{Z}$

$$p_Z(z) = \sum_{\text{all } x} p_X(x) \cdot p_Y(z-x)$$

This Lecture

1. Moment Generating Functions

2. Covariance and Correlation Coefficient

3. Nice Properties of Bivariate Normal

- Reading material: Chapter 10.2-10.5 and 11.1

Recall: There are still a few remaining questions about bivariate normal...

(Q1) Is X_2 a normal random variable? What is the PDF?

MGF and sum of independent random variables

(Q2) What is “ ρ ” in the joint PDF of bivariate normal?

Covariance and correlation coefficient

(Q3) Why is bivariate normal useful? Any nice properties?

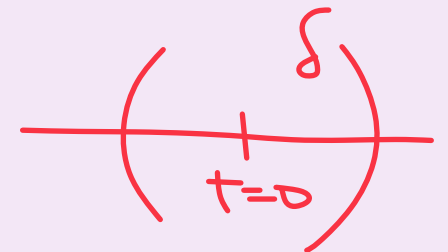
4 nice properties

(Q1) Moment Generating Functions

Moment Generating Function (Formally)

- **Moment Generating Function (MGF):** For a random variable X , define

$$\underline{M_X(t)} = \underline{E[e^{tX}]}, t \in \mathbb{R}$$



If there exists $\delta > 0$ such that $M_X(t) < \infty$ for all $t \in (-\delta, \delta)$, then $M_X(t)$ is called the moment generating function of X

- **Remark:** If X is discrete with PMF $p_X(x)$, then

$$\underline{M_X(t)} = \sum_{\text{all } x} \underline{p_X(x)} \cdot \underline{e^{tx}}$$

- **Remark:** If X is continuous with PDF $f_X(x)$, then

$$M_X(t) = \int_{-\infty}^{+\infty} f_X(x) \cdot e^{tx} dx$$

Nice Properties of MGF?

► Let X_1, X_2 be two random variables:

1. _____ Suppose $M_{X_1}(t) = M_{X_2}(t)$, for all $t \in \mathbb{R}$. Do X_1 and X_2 always have the same distribution (i.e., the same CDF)?
2. _____ Could we find moments $E[X_1^n]$ by using $M_{X_1}(t)$?
3. _____ Suppose X_1, X_2 are independent. Could we express $M_{X_1+X_2}(t)$ in terms of $M_{X_1}(t), M_{X_2}(t)$?

Nice Property (I): MGF Uniqueness Theorem

- ▶ **MGF Uniqueness Theorem:** Let X_1 and X_2 be two random variables with MGFs $M_{X_1}(t)$ and $M_{X_2}(t)$, respectively. If $M_{X_1}(t) = M_{X_2}(t)$ for all t in some interval $(-\alpha, \alpha)$, then X_1 and X_2 follow the same distribution, i.e.

$$\underline{P(X_1 \leq u)} = \underline{P(X_2 \leq u)}, \text{ for all } u \in \mathbb{R}$$

- ▶ **Remark:** More details in the following reference
 - ▶ J. H. Curtiss, “A note on the theory of moment generating functions,” 1942
 - ▶ https://projecteuclid.org/download/pdf_1/euclid.aoms/1177731541

Example: Find CDF from MGF

- ▶ **Example:** Suppose the MGF of a random variable X is

$$M_X(t) = \frac{1}{6}e^{-2t} + \frac{1}{3}e^{-t} + \frac{1}{4}e^t + \frac{1}{4}e^{2t}$$

$$M_X(t) = E[e^{tX}]$$

- ▶ **Question:** $P(|X| \leq 1) = ?$

$$p_X(x) = \begin{cases} \frac{1}{6} & , \text{ if } x = -2 \\ \frac{1}{3} & , \text{ if } x = -1 \\ \frac{1}{4} & , \text{ if } x = 1 \\ \frac{1}{4} & , \text{ if } x = 2 \\ 0 & , \text{ otherwise} \end{cases}$$

$$= \begin{cases} \sum_{\text{all } x} \underline{p_X(x) \cdot e^{tx}} \\ \int f_X(x) \cdot e^{tx} dx \end{cases}$$

$$P(|X| \leq 1) = \frac{1}{3} + \frac{1}{4} = \frac{7}{12}$$

(By Uniqueness Theorem)

MGF of Special Random Variables

Distribution	Moment-generating function $M_X(t)$
Degenerate δ_a	e^{ta}
Bernoulli $P(X=1)=p$	$1 - p + pe^t$
Geometric $(1-p)^{k-1}p$	$\frac{pe^t}{1 - (1-p)e^t}$ $\forall t < -\ln(1-p)$
Binomial $B(n, p)$	$(1 - p + pe^t)^n$
Negative Binomial $NB(r, p)$	$\frac{(1-p)^r}{(1-pe^t)^r}$
Poisson $Pois(\lambda)$	$e^{\lambda(e^t-1)}$
Uniform (continuous) $U(a, b)$	$\frac{e^{tb} - e^{ta}}{t(b-a)}$
Uniform (discrete) $DU(a, b)$	$\frac{e^{at} - e^{(b+1)t}}{(b-a+1)(1-e^t)}$
Laplace $L(\mu, b)$	$\frac{e^{t\mu}}{1 - b^2 t^2}, t < 1/b$
Normal $N(\mu, \sigma^2)$	$e^{t\mu + \frac{1}{2}\sigma^2 t^2}$

(By Uniqueness Theorem)

► Example: If $M_X(t) = \frac{1}{2}e^{0 \cdot t} + \frac{1}{2}e^t$, then what kind of r.v. is X ?

$$X \sim \text{Bernoulli}(p = \frac{1}{2})$$

► Example: If $M_Z(t) = e^{2t^2 - t}$, then what kind of r.v. is Z ?

$$Z \sim \mathcal{N}(\mu = -1, \sigma^2 = 4)$$

Nice Property (II): From Sum to Product

- **MGF and Sum of 2 Independent Random Variables:** Given 2 independent random variables X_1, X_2 with MGFs $M_{X_1}(t)$ and $M_{X_2}(t)$, the MGF of $X_1 + X_2$ is

$$M_{X_1+X_2}(t) = M_{X_1}(t) \cdot M_{X_2}(t)$$

$$E[XY] = E[X] \cdot E[Y]$$

- **Proof:**
$$\begin{aligned} M_{X_1+X_2}(t) &= E[e^{t(X_1+X_2)}] \\ &= E[e^{tX_1} \cdot e^{tX_2}] \\ &\stackrel{\text{independence of } X_1, X_2}{=} E[e^{tX_1}] \cdot E[e^{tX_2}] \\ &= M_{X_1}(t) \cdot M_{X_2}(t). \end{aligned}$$

Example: MGF of Sum of 2 Normal R.V.s

► **Example:** $X_1 \sim \mathcal{N}(\mu_1, \sigma_1^2)$ and $X_2 \sim \mathcal{N}(\mu_2, \sigma_2^2)$

► X_1 and X_2 are assumed to be independent

► **Question:** What is the MGF of $X_1 + X_2$? What is the PDF of $X_1 + X_2$?

$$M_{X_1}(t) = e^{t \cdot \mu_1 + \frac{1}{2} \sigma_1^2 t^2}$$

$$M_{X_2}(t) = e^{t \cdot \mu_2 + \frac{1}{2} \sigma_2^2 t^2}$$

$$M_{X_1+X_2}(t) = M_{X_1}(t) \cdot M_{X_2}(t) = e$$

$$X_1 + X_2 \sim \mathcal{N}(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$$

$X_1 \sim \mathcal{N}(\mu_1, \sigma_1^2)$
 $X_2 \sim \mathcal{N}(\mu_2, \sigma_2^2)$
 \vdots
 $X_n \sim \mathcal{N}(\mu_n, \sigma_n^2)$
 X_1, X_2, \dots, X_n are independent

$(X_1 + X_2 + \dots + X_n) \sim \mathcal{N}\left(\sum_{i=1}^n \mu_i, \sum_{i=1}^n \sigma_i^2\right)$

$t(\mu_1 + \mu_2) + \frac{1}{2}(\sigma_1^2 + \sigma_2^2)t^2$

Nice Property (III): Why Is $M_X(t)$ Called the Moment Generating Function?

- **Recall:** What is the “ n -th moment” of X ?

$M_X(t)$ needs to be differentiable around $t=0$

- **Use MGF to Find Moments:** Let X be a random variable with MGF $M_X(t)$. Then, for every $n \in \mathbb{N}$, we have

$$E[X^n] = \frac{d^n}{dt^n} M_X(t) \Big|_{t=0}$$

(Suppose X is discrete)

- **Proof:**

$$\frac{d}{dt} M_X(t) \Big|_{t=0} = \frac{d}{dt} E[e^{tX}] \Big|_{t=0} = \frac{d}{dt} \left(\sum_{\text{all } x} P_X(x) \cdot e^{tx} \right) \Big|_{t=0}$$

$$= \sum_{\text{all } x} P_X(x) \cdot \frac{d}{dt} e^{tx} \Big|_{t=0} = \sum_{\text{all } x} P_X(x) \cdot x \cdot e^{tx} \Big|_{t=0}$$

$$= E[X]$$

$$\frac{d^n}{dt^n} M_X(t) \Big|_{t=0} = \sum_{\text{all } x} P_X(x) \cdot \frac{d^n}{dt^n} e^{tx} \Big|_{t=0} = \sum_{\text{all } x} P_X(x) \cdot x^n \cdot e^{tx} \Big|_{t=0} = E[X^n]$$

Example: Moments of $\text{Exp}(\lambda)$

► **Example:** Suppose $X \sim \text{Exp}(\lambda)$

► What is the MGF of X ?

► Use MGF to verify that $E[X] = \frac{1}{\lambda}$ and $\text{Var}[X] = \frac{1}{\lambda^2}$?

$$M_X(t) = E[e^{tX}] = \int_{-\infty}^{+\infty} f_X(x) \cdot e^{tx} dx = \int_0^{+\infty} \lambda e^{-\lambda x} \cdot e^{tx} dx = \frac{\lambda}{-(\lambda-t)} \left[e^{-(\lambda-t)x} \right]_0^{+\infty} = \frac{\lambda}{\lambda-t}, \text{ if } \lambda > t$$

$$E[X] = \left. \frac{d}{dt} \left(\frac{\lambda}{\lambda-t} \right) \right|_{t=0} = \left. \frac{\lambda}{(\lambda-t)^2} \right|_{t=0} = \frac{\lambda}{\lambda^2} = \frac{1}{\lambda}$$

$$E[X^2] = \left. \frac{d^2}{dt^2} \left(\frac{\lambda}{\lambda-t} \right) \right|_{t=0} = \left. \frac{2\lambda}{(\lambda-t)^3} \right|_{t=0} = \frac{2}{\lambda^2}$$

$$\text{Var}[X] = E[X^2] - (E[X])^2 = \frac{1}{\lambda^2}$$

$= \begin{cases} \frac{\lambda}{\lambda-t}, & \text{if } \lambda > t \\ \text{does not exist}, & \text{else} \end{cases}$

" ρ "

(Q2) Covariance

Motivating Example for “Covariance”

- ▶ **Example:** Bus #2 (NCTU - Mackay - Train Station)
 - ▶ X = traveling time from NCTU to Mackay
 - ▶ Y = traveling time from Mackay to Train Station
 - ▶ We want to know $\text{Var}[X + Y]$
 - ▶ **Question:** Given $\text{Var}[X]$ and $\text{Var}[Y]$, can we get $\text{Var}[X + Y]$?



Covariance and Where to Find Them

► Property:

$$\text{Var}[aX + bY] = a^2 \text{Var}[X] + b^2 \text{Var}[Y] + 2ab \overbrace{E[(X - E[X])(Y - E[Y])]}^{\text{"Covariance"}}$$

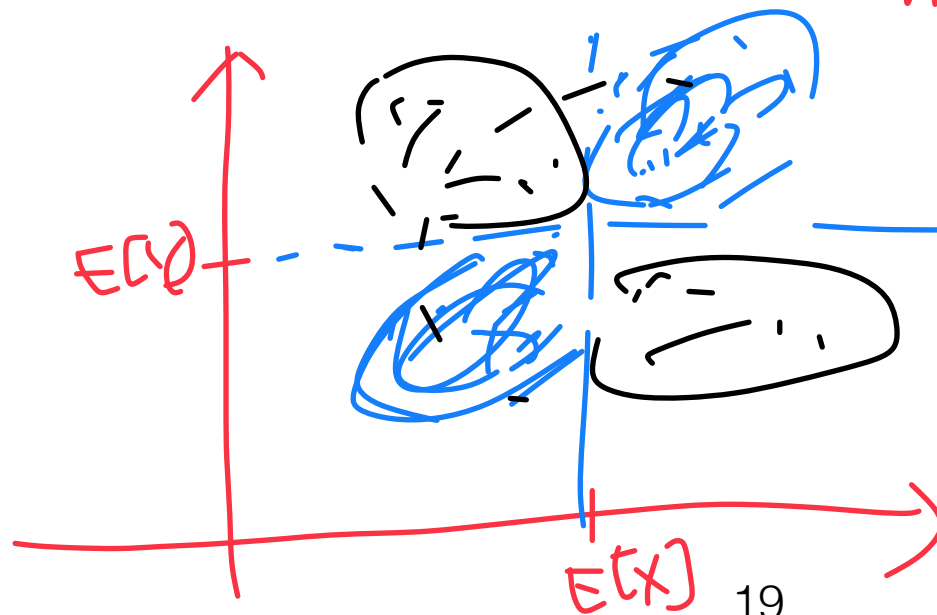
$$\begin{aligned} \text{Var}[aX + bY] &= E\left[\left(\underbrace{(aX + bY)}_Z - E[aX + bY]\right)^2\right] \\ &= E\left[\left((aX - E[aX]) + (bY - E[bY])\right)^2\right] \\ &= E\left[\underbrace{(aX - E[aX])^2} + \underbrace{(bY - E[bY])^2} + 2 \cdot \underbrace{(aX - E[aX])(bY - E[bY])}\right] \end{aligned}$$

Covariance

- **Covariance:** Let X, Y be two random variables. Then, the covariance of X and Y is defined as

$$\text{Cov}(X, Y) = E[(X - E[X])(Y - E[Y])]$$

- $\text{Cov}(X, X)$ = $\text{Var}[X]$
- $\text{Cov}(X, Y) = 0$: X, Y are said to be uncorrelated
- $\text{Cov}(X, Y) > 0$: X, Y are said to be positively correlated
- $\text{Cov}(X, Y) < 0$: X, Y are said to be negatively correlated
- **Intuition:**



Another Expression of Covariance $E[XY] = E[X]E[Y]$

- Let X, Y be two random variables. Then, the covariance of X and Y can also be written as

$$\text{Cov}(X, Y) = \underline{E[XY]} - \underline{E[X]E[Y]}$$

- Question: How to show this?

$$\text{Cov}(X, Y) = \underline{E[(X - E[X])(Y - E[Y])]}$$

$$= E[XY - E[X] \cdot Y - X \cdot E[Y] + E[X]E[Y]]$$

$$= E[XY] - E[X]E[Y]$$

$$P(X \in A, Y \in B) = P(X \in A)P(Y \in B)$$

- Question: If X, Y are independent, then $\text{Cov}(X, Y) = \underline{0}$

- Question: How about the converse argument? $\text{Cov}(X, Y) = 0 \Rightarrow X, Y \text{ independent}$

Example: Uncorrelated \nRightarrow Independence

- ▶ **Example:** The pair of random variables (X, Y) takes the values $(1,0)$, $(0,1)$, $(-1,0)$, $(0,-1)$, each with probability $\frac{1}{4}$

- ▶ $\text{Cov}(X, Y) = ?$
- ▶ Are X, Y independent?

$$\text{Cov}(X, Y) = \frac{E[XY]}{0} - \frac{E[X]}{0} \cdot \frac{E[Y]}{0} = 0$$

$$P_{X,Y}(x,y) = \begin{cases} \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \end{cases} \quad \begin{matrix} \text{if } (x,y) = (1,0) \\ (0,1) \\ (-1,0) \\ (0,-1) \end{matrix}$$

Find some events A, B , show $P(X \in A, Y \in B) \neq P(X \in A) \cdot P(Y \in B)$

$$A = \{1\}$$
$$B = \{1\}$$

$$P(X \in A, Y \in B) = 0$$

$$P(X \in A) = \frac{1}{4}$$

$$P(Y \in B) = \frac{1}{4}$$

A Property of Covariance

► Property:

$$(\text{Cov}(X, Y))^2 \leq \text{Var}[X] \cdot \text{Var}[Y]$$

► Question: How to show this?

Any Issue With Covariance?

- ▶ **Example:** Bus #2 (NCTU - Mackay - Train Station)
 - ▶ From NCTU to Mackay: X minutes
 - ▶ From Mackay to Train Station: Y minutes
 - ▶ **Question:** $\text{Cov}(X, Y) = ?$
 - ▶ **Question:** What if time is measured in “seconds”? Any change in the covariance?



Covariance is Sensitive to the Units

- ▶ **Property:** $\text{Cov}(aX, aY) = a^2 \cdot \text{Cov}(X, Y)$
 - ▶ a : scaling factor due to change of unit
- ▶ **Question:** Any suggested solution?

Correlation Coefficient

- ▶ **Correlation Coefficient:** Let X, Y be two random variables with finite variance $\sigma_X^2 > 0$ and $\sigma_Y^2 > 0$. Then, the correlation coefficient of X and Y is defined as

$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

- ▶ **Question:** Do we have $\rho(X, Y) = \rho(aX, aY)$, for any $a \neq 0$?

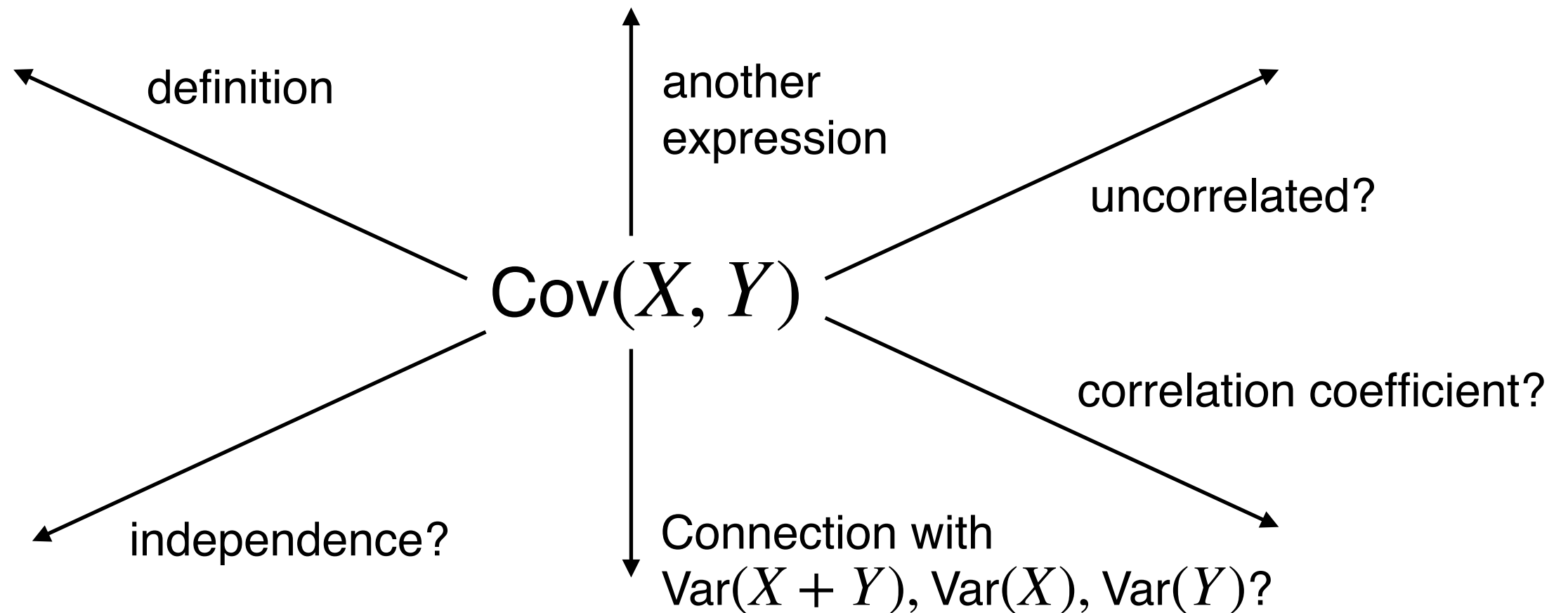
A Property of Correlation Coefficient

► **Property:**

$$-1 \leq \rho(X, Y) \leq 1$$

► **Question:** How to prove this?

A Brief Summary of Covariance



(Q3) Nice Properties of Bivariate Normal

Properties of Bivariate Normal R.V.

- Suppose the joint PDF of X_1, X_2 is bivariate normal as

$$f_{X_1X_2}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp \left[-\frac{\left(\frac{(x_1 - \mu_1)^2}{\sigma_1^2} - 2\rho \frac{(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2} \right)}{2(1 - \rho^2)} \right]$$

Then we have:

1. Marginal: $X_1 \sim \mathcal{N}(\mu_1, \sigma_1^2)$ and $X_2 \sim \mathcal{N}(\mu_2, \sigma_2^2)$
2. Conditional: $X_2 | X_1 = x_1 \sim \mathcal{N}\left(\mu_2 + \frac{\rho\sigma_2(x_1 - \mu_1)}{\sigma_1}, (1 - \rho^2)\sigma_2^2\right)$
3. Correlation coefficient: $\rho(X_1, X_2) = \rho$
4. If X_1, X_2 are uncorrelated ($\rho = 0$), then X_1, X_2 are independent

1. Marginal: $X_1 \sim \mathcal{N}(\mu_1, \sigma_1^2)$ and $X_2 \sim \mathcal{N}(\mu_2, \sigma_2^2)$

$$f_{X_1 X_2}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp \left[-\frac{\left(\frac{(x_1 - \mu_1)^2}{\sigma_1^2} - 2\rho \frac{(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2} \right)}{2(1 - \rho^2)} \right]$$
$$\frac{\left(\frac{(x_1 - \mu_1)^2}{\sigma_1^2} - 2\rho \frac{(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2} \right)}{2(1 - \rho^2)} = \frac{(x_1 - \mu_1)^2}{2\sigma_1^2} + \frac{1}{2\sigma_2^2} \left(\frac{(x_2 - \mu_2) - \rho(x_1 - \mu_1)}{\sqrt{1 - \rho^2}} \right)^2$$

Take X_1 for example (X_2 would be similar)

$$f_{X_1}(x_1) =$$

2. Conditional: $X_2 | X_1 = x_1 \sim \mathcal{N}\left(\mu_2 + \frac{\rho\sigma_2(x_1 - \mu_1)}{\sigma_1}, (1 - \rho^2)\sigma_2^2\right)$

$$f_{X_1 X_2}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp\left[-\frac{\left(\frac{(x_1 - \mu_1)^2}{\sigma_1^2} - 2\rho\frac{(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2}\right)}{2(1 - \rho^2)}\right]$$

$$f_{X_1}(x_1) = \frac{1}{\sqrt{2\pi}\sigma_1} \exp\left[-\frac{(x_1 - \mu_1)^2}{2\sigma_1^2}\right]$$

$$f_{X_2|X_1}(x_2 | x_1) =$$

3. Correlation Coefficient: $\rho(X_1, X_2) = \rho$

$$f_{X_1X_2}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp \left[-\frac{\left(\frac{(x_1 - \mu_1)^2}{\sigma_1^2} - 2\rho \frac{(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2} \right)}{2(1 - \rho^2)} \right]$$

$$\text{Cov}(X_1, X_2) = E[(X_1 - \mu_1)(X_2 - \mu_2)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x_1 - \mu_1)(x_2 - \mu_2) f_{X_1X_2}(x_1, x_2) dx_1 dx_2$$

Hint: $f_{X_2|X_1} = \frac{f_{X_1X_2}}{f_{X_1}} \Rightarrow f_{X_1X_2} = f_{X_2|X_1} f_{X_1}$

$$\text{Cov}(X_1, X_2) =$$

4. Uncorrelated ($\rho = 0$) Implies Independence

$$f_{X_1X_2}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp \left[-\frac{\left(\frac{(x_1 - \mu_1)^2}{\sigma_1^2} - 2\rho \frac{(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2} \right)}{2(1 - \rho^2)} \right]$$

► **If $\rho = 0$:**

$$f_{X_1X_2}(x_1, x_2) =$$

Final Remark: X_1, X_2 Normal $\not\Rightarrow X_1, X_2$ Bivariate Normal

- ▶ **Example:** Let Y and Z be two independent standard normal r.v.s
 - ▶ $X_1 = |Y| \cdot \text{sign}(Z)$
 - ▶ $X_2 = Y$
- ▶ **Question:**
 - ▶ Are X_1 and X_2 normal?
 - ▶ Are X_1 and X_2 bivariate normal?