

Stat 134 lec 40

Warmup 10:00-10:10

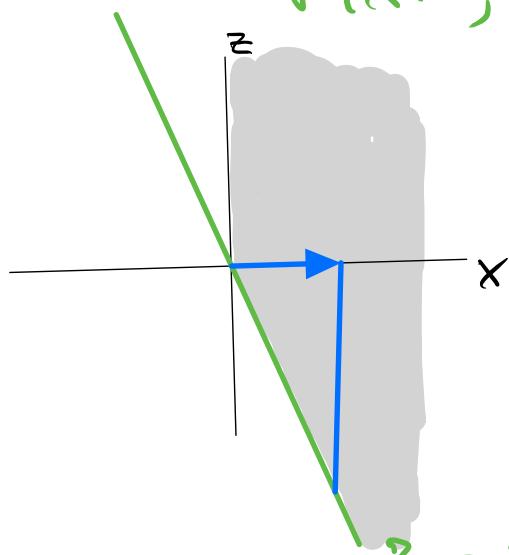
Let X, Y are std. bivariate normal, $\rho > 0$

Find $P(X > 0, Y > 0)$

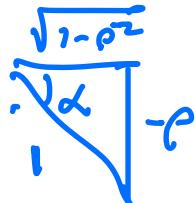
Hint

$$P(X > 0, Y > 0) = P(X > 0, \rho X + \sqrt{1-\rho^2} Z > 0)$$

$$= P(X > 0, Z > -\frac{\rho}{\sqrt{1-\rho^2}} X)$$



note X and Z are uncorrelated
so joint(X, Z) is a bell over X, Z with rotational symmetry.



$$Z = -\frac{\rho}{\sqrt{1-\rho^2}} X$$

$$\tan \alpha = \frac{-\rho}{\sqrt{1-\rho^2}} \Rightarrow \alpha = \tan^{-1}\left(\frac{-\rho}{\sqrt{1-\rho^2}}\right)$$

$$P(X > 0, Z > -\frac{\rho}{\sqrt{1-\rho^2}} X) = \boxed{\frac{90 + |\alpha|}{360}}$$

Last time

- RRR week schedule
- | | |
|---|----------|
| M Review (post questions on b-courses), | } in SLC |
| W OFF | |
| F Review (post questions on b-courses) | |

Sec 6.5. Bivariate Normal

Defⁿ (Standard Bivariate Normal Distribution)

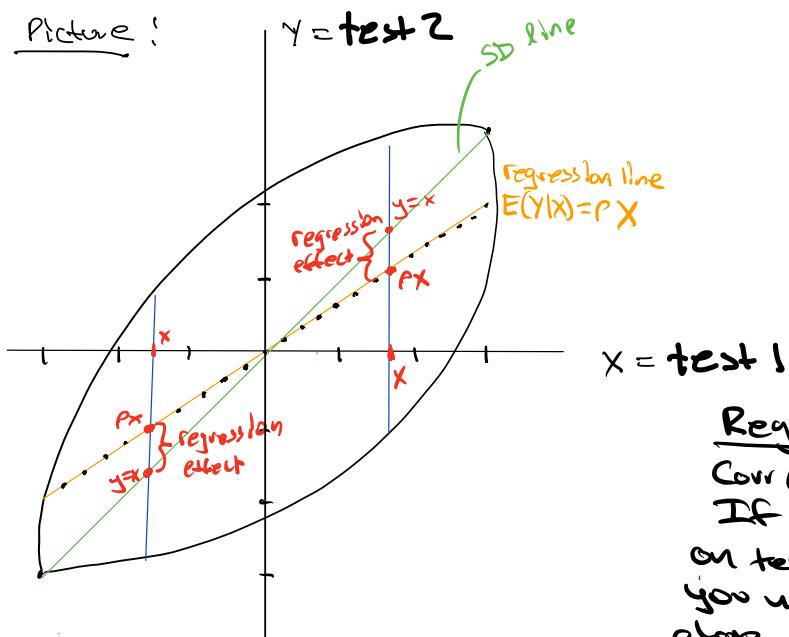
let $X, Z \sim N(0,1)$, $-1 \leq \rho \leq 1$

$$Y = \rho X + \sqrt{1-\rho^2} Z \sim N(0,1)$$

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

regression effect

Picture:



Regression effect,
 $\text{Cov}(\text{test 1}, \text{test 2}) = .6$
 If 1 SD above mean
 on test 1 then on average
 you will be less than 1 SD
 above average on test 2.
 (regression line is less steep
 than SD line).

Today (1) MGF of bivariate normal

(2) sec 6.5 Properties of bivariate normal

① MGF of bivariate normal

The single and bivariate MGF is defined as:

$$M_Y(t) = E(e^{tY}) \quad \text{single variable MGF}$$

$$M_{(X,Y)}(s,t) = E(e^{sX+tY}) \quad \text{bivariate MGF}$$

Show that $M_{(X,Y)}(s,t) = M_X(s) M_Y(t)$ iff
 X, Y are independent,

$$\begin{aligned} M_{(X,Y)}(s,t) &= E(e^{sX+tY}) \\ &= E(e^{sx} \cdot e^{ty}) \\ &\stackrel{\text{iff } X, Y \text{ indep.}}{\sim} = E(e^{sx}) E(e^{ty}) \\ &= M_X(s) \cdot M_Y(t) \end{aligned}$$

$$E(AB) = E(A)E(B)$$

iff A, B
are indep.

Have
 $A = e^{sx}$
 $B = e^{ty}$

Thm Let (X, Y) be standard bivariate normal.

The MGF of (X, Y) is

$$M_{(X,Y)}(s,t) = e^{\frac{s^2}{2} + \frac{t^2}{2} + st\rho}$$

$$\begin{aligned} \text{Pf/ } M_{X,Y}(s,t) &= E[e^{sx+ty}] \\ &= E[e^{sx+t(\rho X + \sqrt{1-\rho^2}Z)}] \\ &= E[e^{(s+t\rho)x} \cdot e^{t\sqrt{1-\rho^2}z}] \\ &\stackrel{\text{independence}}{=} E[e^{(s+t\rho)x}] E[e^{t\sqrt{1-\rho^2}z}] \\ &= M_X(s+t\rho) \cdot M_Z(t\sqrt{1-\rho^2}) \\ &= e^{\frac{(s+t\rho)^2}{2}} \cdot e^{\frac{t^2\sqrt{1-\rho^2}}{2}} \end{aligned}$$

Recall that $X \sim N(0,1)$ so $M_X(a) = e^{\frac{a^2}{2}}$

$$\begin{aligned} &= e^{\frac{(s+t\rho)^2}{2}} \cdot e^{\frac{(t\sqrt{1-\rho^2})^2}{2}} \\ &= e^{\frac{s^2}{2} + st\rho + \frac{t^2\rho^2}{2}} \cdot e^{\frac{t^2 - t^2\rho^2}{2}} \\ &= e^{\boxed{\frac{s^2}{2} + \frac{t^2}{2} + st\rho}} \end{aligned}$$

Finish proof.

② Properties of Bivariate Normal

Recall that if 2 RVs X, Y are independent, $\text{Cov}(X, Y) = 0$
 $\Rightarrow \text{Corr}(X, Y) = 0$

However the converse is not true in general. ($\text{Corr}(X, Y) = 0 \not\Rightarrow X, Y \text{ indep.}$)

Let $X \sim N(0, 1)$

$$Y = X^2$$

X and Y are dependent

we show X, Y are uncorrelated,

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

$$= E(X^3) - E(X)E(X^2)$$

Real world example

$$x = \text{height}$$

$$y = (\text{height})^2$$

If you make a scatter plot of height vs $(\text{height})^2$ it will look uncorrelated even though height and $(\text{height})^2$ are dependent

Find $E(X^3)$

$$= \int_{-\infty}^{\infty} x^3 \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$\int_{-\infty}^{\infty} x^3 \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = 0$ since $\frac{x^3}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$ is an odd function

OR

$$M_X(t) = e^{t^2/2}$$

$$M_X'''(0) = t e^{t^2/2} + 2t e^{t^2/2} + t^3 e^{t^2/2} \Big|_{t=0} = 0$$

$$\Rightarrow E(X^3) = 0$$

$$\Rightarrow \text{Corr}(X, Y) = 0$$

D

Thm If (X, Y) is ^{std} bivariate normal then

$\rho = \text{Corr}(X, Y) = 0$ iff X, Y are independent.

Pf/ From earlier we know

for any joint distribution (X, Y) ,

$M_{(X,Y)}(s,t) = M_X(s) M_Y(t)$ iff X, Y are independent.

Since (X, Y) is ^{std} bivariate normal,

$$M_{(X,Y)}(s,t) = e^{\frac{s^2}{2} + \frac{t^2}{2} + st\rho}$$

$$\begin{aligned} X, Y \text{ indep } M_{(X,Y)}(s,t) &= M_X(s) M_Y(t) \Rightarrow e^{st\rho} = 1 \\ &\quad " " \frac{s^2}{2} " " \frac{t^2}{2} \Rightarrow (\rho = 0) \\ &\quad " " \frac{s^2}{2} + \frac{t^2}{2} + st\rho \quad e^{\frac{s^2}{2}} \cdot e^{\frac{t^2}{2}} \end{aligned}$$

$$\begin{aligned} \text{Conversely, if } \rho = 0 \\ e^{\frac{s^2}{2} + \frac{t^2}{2} + st\rho} &= e^{\frac{s^2}{2} + \frac{t^2}{2}} = e^{\frac{s^2}{2}} \cdot e^{\frac{t^2}{2}} \Rightarrow X, Y \text{ indep.} \\ M_{(X,Y)}(s,t) &= M_X(s) M_Y(t) \end{aligned}$$

Recall from Lec 30 that the sum of independent normal random variables is normal.

What about dependent normal random variables?

recall,

$$M_Y(t) = E(e^{tY}) = M_{tY}(1)$$

single variable MGF

$$M_{(X,Y)}(s,t) = E(e^{sX+tY}) = M_{sX+tY}(1)$$

bivariate MGF

recall,

$$Z \sim N(\mu, \sigma^2) \text{ iff } M_Z(\omega) = e^{\mu\omega} e^{\frac{\sigma^2 \omega^2}{2}}$$

Thm Let $X, Y \sim N(0, 1)$ and $\text{Corr}(X, Y) = \rho$.

(X, Y) is std bivariate normal iff

$sX + tY$ is normal for all constants s, t .

Pf

\Rightarrow : Suppose X, Y std bivariate normal.
(i.e. $Y = \rho X + \sqrt{1-\rho^2} Z$ for $X, Z \text{ iid } N(0, 1)$)

$$\begin{aligned} sX + tY &= sX + t(\rho X + \sqrt{1-\rho^2} Z) \\ &= (s + t\rho)X + t\sqrt{1-\rho^2} Z \end{aligned}$$

is normal since X, Z are indep normals

\Leftarrow : Suppose $sX + tY$ is normal,

Note $E(sX + tY) = sE(X) + tE(Y) = 0$

$$\begin{aligned} \text{Var}(sX + tY) &= \text{Var}(sX) + \text{Var}(tY) + 2\text{Cov}(sX, tY) \\ &= s^2 + t^2 + 2st\rho \end{aligned}$$

$$2st\text{Cov}(X, Y)$$

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

Hence, $sX + tY \sim N(0, s^2 + t^2 + 2st\rho)$

Recall,

$$\text{For } Z \sim N(\mu, \sigma^2) \Rightarrow M_Z^{(k)} = e^{\mu k} e^{\sigma^2 \frac{k^2}{2}}$$

$$\begin{aligned} \text{Then } M_{(X,Y)}^{(s,t)} &= M_{sX+tY}^{(1)} = e^{(s^2 + t^2 + 2st\rho)\frac{1}{2}} \\ &\quad \text{normal} \qquad \leftarrow \text{MF of } (X, Y) \\ &= e^{\frac{s^2}{2} + \frac{t^2}{2} + st\rho} \end{aligned}$$

$\Rightarrow (X, Y)$ is standard bivariate normal \square

We don't need to restrict ourselves to $X, Y \sim N(0, 1)$

Corollary — proof at end of lecture

$$\text{Let } U \sim N(\mu_U, \sigma_U^2)$$

$$V \sim N(\mu_V, \sigma_V^2)$$

$$\text{and } \text{Corr}(U, V) = \rho$$

$$(U, V) \sim BV(\mu_U, \mu_V, \sigma_U^2, \sigma_V^2, \rho) \text{ if}$$

$sU + tV$ is normal for all constants s, t .

Appendix

Corollary

Let $U \sim N(\mu_U, \sigma_U^2)$
 $V \sim N(\mu_V, \sigma_V^2)$

and $\text{Corr}(U, V) = \rho$

$(U, V) \sim BV(\mu_U, \mu_V, \sigma_U^2, \sigma_V^2, \rho)$ iff
 $sU + tV$ is normal for all constants s, t .

Pf/ let $X = \frac{U - \mu_U}{\sigma_U}$ ($U = \mu_U + X\sigma_U$)

$Y = \frac{V - \mu_V}{\sigma_V}$ ($V = \mu_V + Y\sigma_V$)

$(U, V) \sim BV(\mu_U, \mu_V, \sigma_U^2, \sigma_V^2, \rho)$

iff $(X, Y) \sim BV(0, 0, 1, 1, \rho)$

$$\begin{aligned} sU + tV &= s(\mu_U + X\sigma_U) + t(\mu_V + Y\sigma_V) \\ &= \underset{a}{s\sigma_U}X + \underset{b}{t\sigma_V}Y + \underset{\text{constant}}{s\mu_U + t\mu_V} \end{aligned}$$

this is normal for all constants s, t

if $\alpha X + \beta Y$ is normal for all constants α, β ,

□

