

# Warmup

Let  $X, Z$  iid  $N(0, 1)$

$$Y = \rho X + \sqrt{1-\rho^2} Z \quad \text{where } -1 \leq \rho \leq 1$$

← Greek letter rho

- (1) What distribution is  $Y$  (include parameters)?
- (2) What is  $\text{Corr}(X, Y)$

$Y$  is a linear combination of independent normals which is normal.

$$E(Y) = E(\rho X) + E(\sqrt{1-\rho^2} Z) = \rho \underbrace{E(X)}_0 + \sqrt{1-\rho^2} \underbrace{E(Z)}_0 = 0$$

$$\begin{aligned} \text{Var}(Y) &= \text{Var}(\rho X) + \text{Var}(\sqrt{1-\rho^2} Z) \\ &= \rho^2 \underbrace{\text{Var}(X)}_1 + (1-\rho^2) \underbrace{\text{Var}(Z)}_1 = 1 \end{aligned}$$

$$\Rightarrow Y \sim N(0, 1)$$

$$\begin{aligned} \text{Corr}(X, Y) &= \frac{\text{Cov}(X, Y)}{\underbrace{\text{SD}(X)}_1 \underbrace{\text{SD}(Y)}_1} = \text{Cov}(X, Y) = \text{Cov}(X, \rho X + \sqrt{1-\rho^2} Z) \\ &= \rho \underbrace{\text{Cov}(X, X)}_{\text{Var}(X)=1} + \sqrt{1-\rho^2} \underbrace{\text{Cov}(X, Z)}_0 \\ &= \rho \end{aligned}$$

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

We will see  $(X, Y) \sim \text{BVN}$

## Announce next

### Schedule

W sec 6.5

F sec 6.5

M review + OH

W off

F GSI review + OH

Th Final exam

← will post review materials  
and practice final.

### Last time

$$\text{let } X = \frac{U - \mu_U}{\sigma_U} \quad \text{--- } U \text{ in std units}$$

$$Y = \frac{V - \mu_V}{\sigma_V} \quad \text{--- } V \text{ in std units}$$

$$\text{Corr}(X, Y) = \text{Corr}(U, V)$$

### Today

Sec 6.5 Bivariate Normal

## Sec 6.5 Bivariate Normal

Def<sup>n</sup> (Standard Bivariate Normal Distribution)

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

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

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

We call the joint distribution  $(X, Y)$  the  
Standard bivariate normal with  $\text{Corr}(X, Y) = \rho$

Written  $(X, Y) \sim \text{BVN}(0, 0, 1, 1, \rho)$



$$(X, Y) \sim \text{BVN}(0, 0, 1, 1, \rho)$$

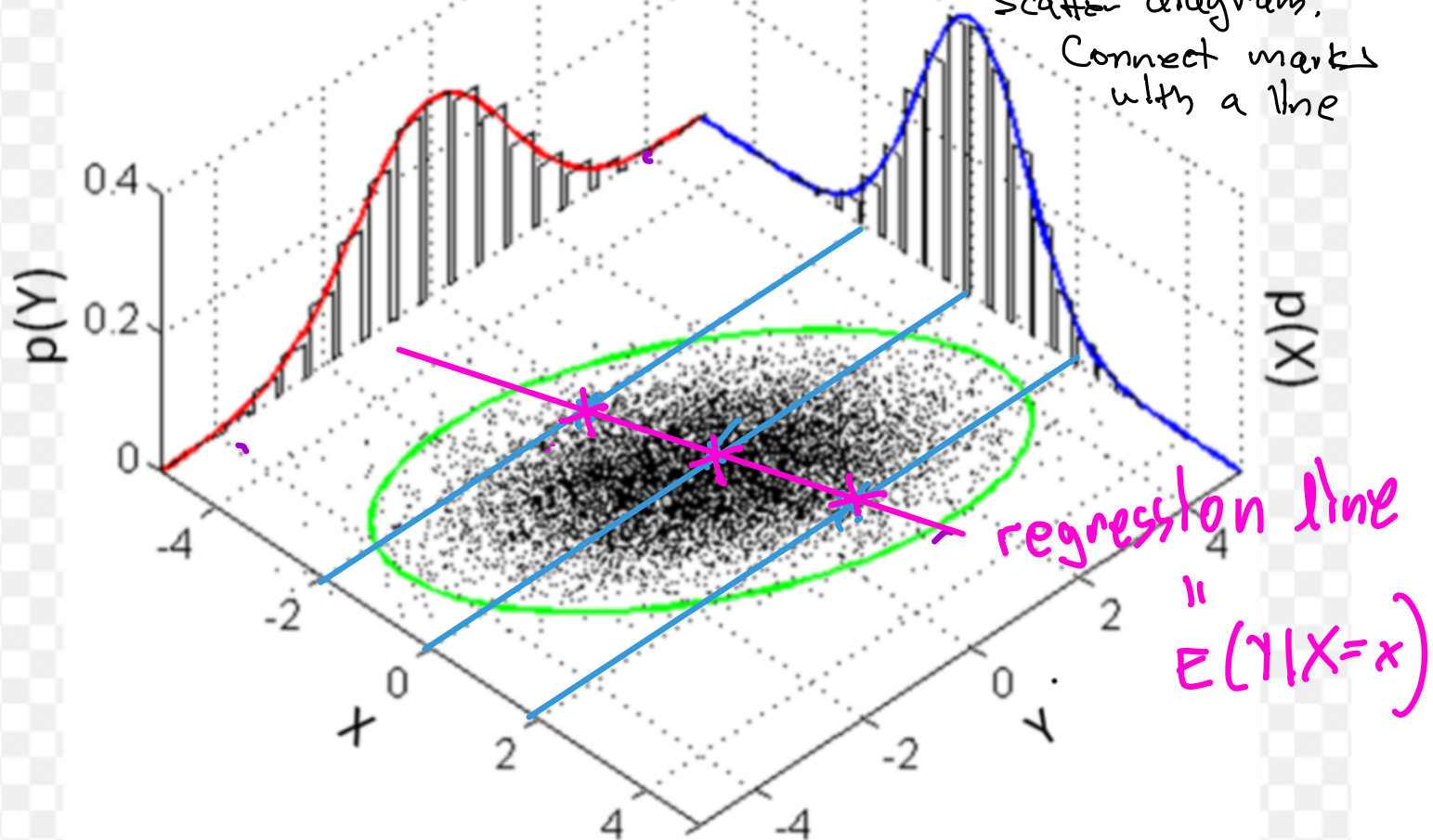
The scatter diagram of BVN always, with  $\rho \neq 0$  looks like  
an upwards or downwards facing football:

## Picture

$$\rho > 0$$

For  $X = -2, 0, 2$

mark the average  
y value in the  
scatter diagram.  
Connect marks  
with a line



when  $(X, Y)$  is standard BVN, the best fitting line through your data is approximately  $E(Y|X=x) = \rho x$ .

We call  $E(Y|X=x) = \rho x$  the regression line

Note The regression line is actually the best fitting line through your data but with many points this is approx  $E(Y|X=x)$

1/2x

Let  $x, z \stackrel{iid}{\sim} N(0,1)$ ,  $-1 \leq \rho \leq 1$

Let  $y = \rho x + \sqrt{1-\rho^2} z$

a) Show that  $Y|X=x$  is normally distributed,

b) Find  $E(Y|X=x) = E(\rho x + \sqrt{1-\rho^2} z | x)$

c) Find  $\text{Var}(Y|X=x) = \text{Var}(\rho x + \sqrt{1-\rho^2} z | x)$

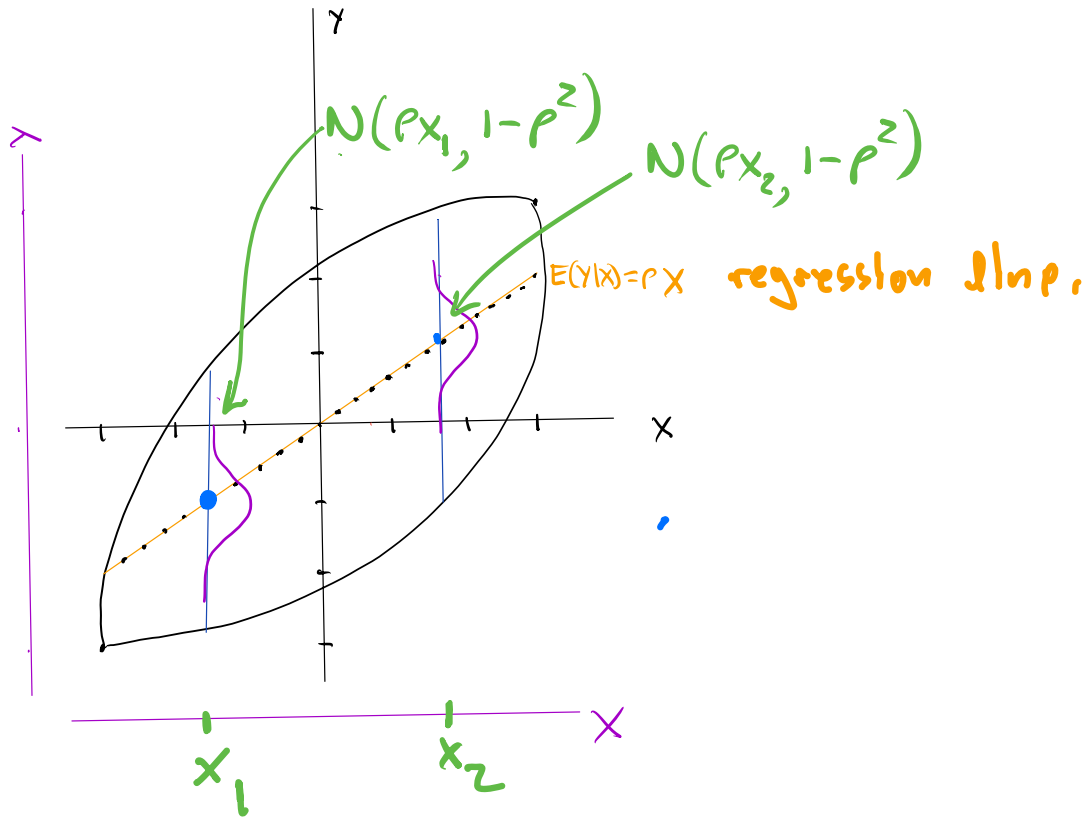
$$\begin{aligned} \text{a) } Y|X=x &= \rho X + \sqrt{1-\rho^2} z \mid X=x = \rho X|X=x + \sqrt{1-\rho^2} z|X=x \\ &= \rho x + \sqrt{1-\rho^2} z \leftarrow \text{this is normal.} \end{aligned}$$

$$\text{b) } E(Y|X=x) = E(\rho x + \sqrt{1-\rho^2} z) = \rho x$$

$$\begin{aligned} \text{c) } \text{Var}(Y|X=x) &= \text{Var}(\rho x + \sqrt{1-\rho^2} z) = 1 - \rho^2 \\ &= \text{Var}(\rho x) + \text{Var}(\sqrt{1-\rho^2} z) \end{aligned}$$

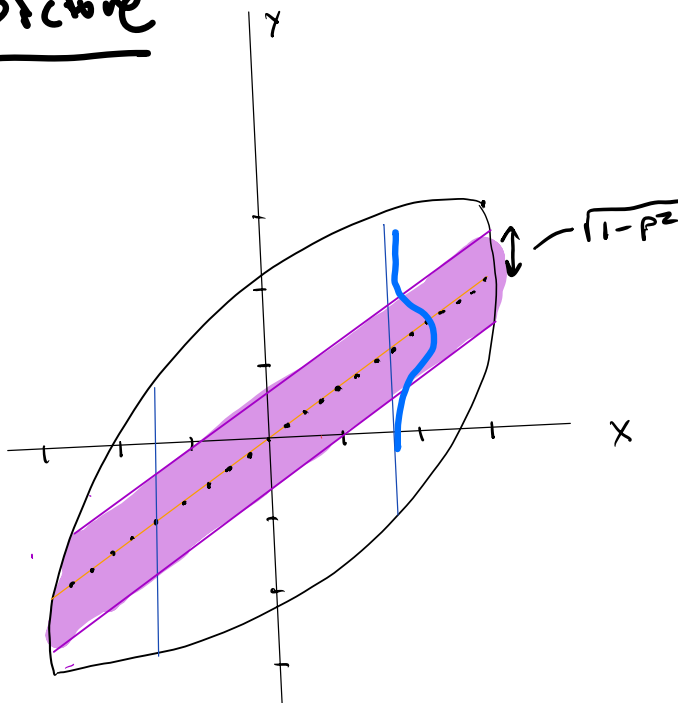
$$\Rightarrow Y|X=x \sim N(\rho x, 1 - \rho^2)$$

We call the line  $E(Y|X=x) = \rho X$  the regression line  
Picture



What is the chance your data is more than 1 SD outside  
of the regression line?  $\frac{1}{3}$

Picture



For every  $x$ , the chance  
you are within 1SD of  
the mean is  $\frac{2}{3}$   
Hence  $\frac{1}{3} = 1 - \frac{2}{3}$

## Def<sup>n</sup> (Bivariate Normal Distribution)

Random variables  $U$  and  $V$  have bivariate normal distribution with parameters  $\mu_U, \mu_V, \sigma_U^2, \sigma_V^2, \rho$  iff the standardized variables

$$X = \frac{U - \mu_U}{\sigma_U}$$

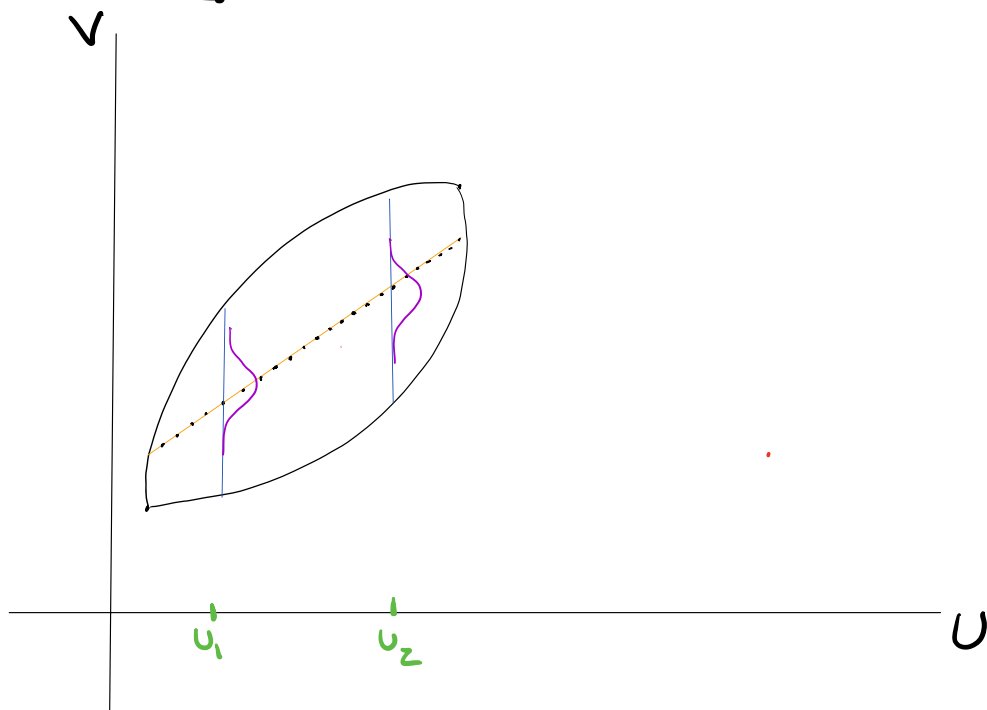
$$Y = \frac{V - \mu_V}{\sigma_V}$$

have std. bivariate normal distribution with corr  $\rho$ .  
Then  $\rho = \text{Corr}(X, Y) = \text{Corr}(U, V)$ .

We write  $(U, V) \sim \text{BVN}(\mu_U, \mu_V, \sigma_U^2, \sigma_V^2, \rho)$

See the appendix for the equation of the regression line  $E(U|V=v)$ .

Picture



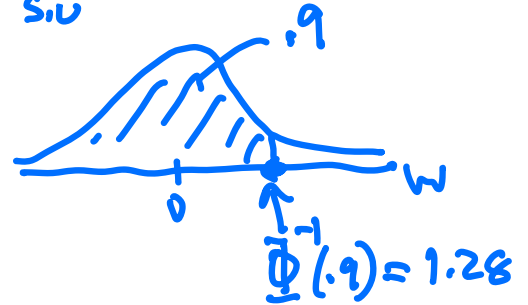
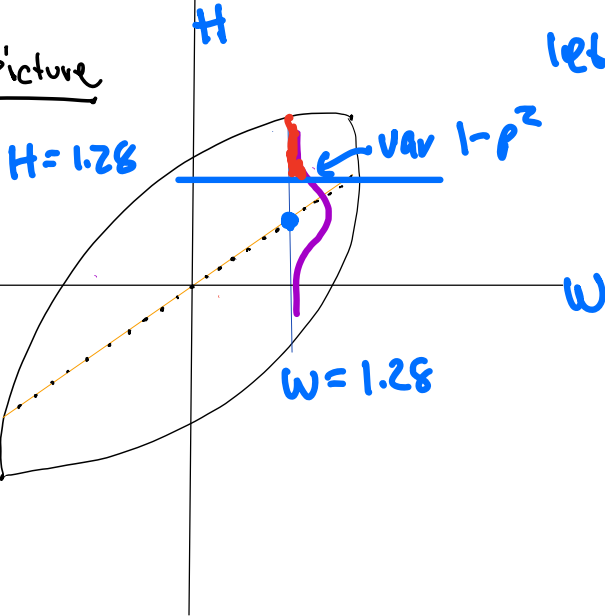
110x

3. Heights and weights of a large group of people follow a bivariate normal distribution, with correlation 0.75. Of the people in the 90th percentile of weights, about what percentage are above the 90th percentile of heights?

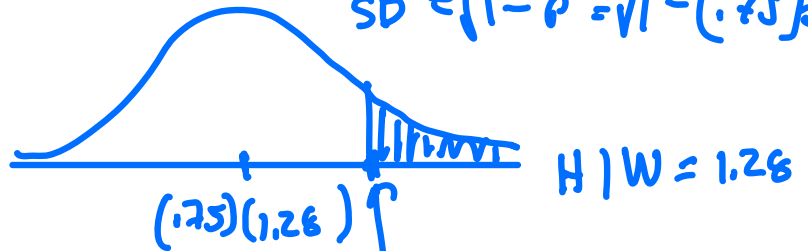
Fact  $\Phi^{-1}(.9) = 1.28$

Picture

let  $W \approx$  weight in s.u.  
 $H \approx$  height in s.u.



$$SD = \sqrt{1 - \rho^2} = \sqrt{1 - (.75)^2} = .66$$



$$1 - \Phi\left(\frac{1.28 - (.75)(1.28)}{.66}\right)$$



1. Here is a summary of Pre-SAT and SAT scores of a large group of students.

$X =$	PSAT scores:	average: 1200	SD: 100
$Y =$	SAT scores:	average: 1300	SD: 90
	correlation: 0.6		

Assume the data are approximately bivariate normal in distribution.

Of the students who scored 1000 on the PSAT, about what percentage scored above average on the SAT?

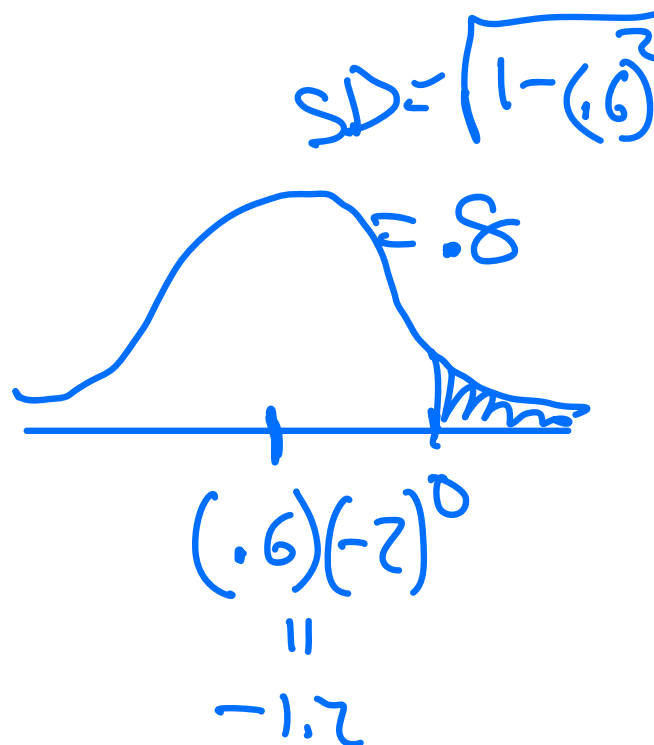
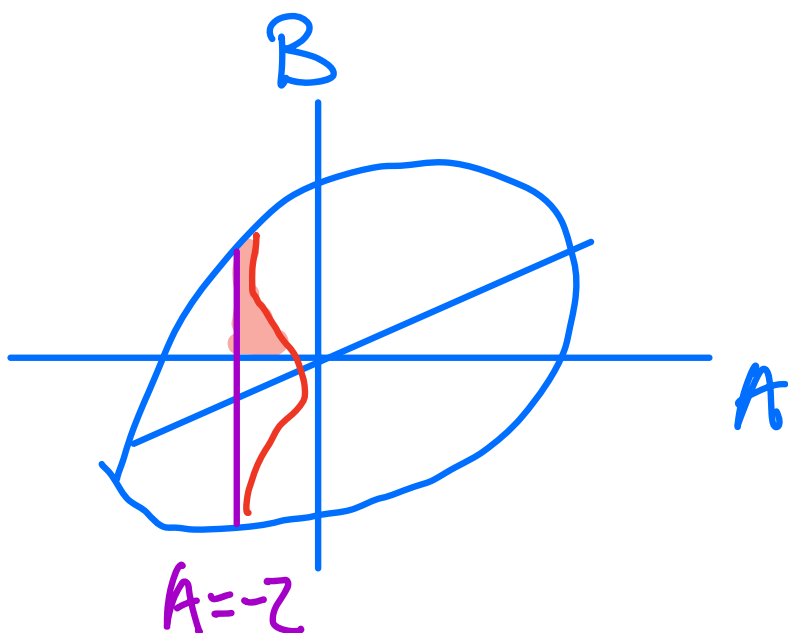
$$\text{Find } P(Y > 1300 | X = 1000)$$

$$\text{let } B = Y \text{ in s.d.}$$

$$A = X \text{ in s.d.}$$

$$\text{Find } P\left(B > \frac{1300 - 1300}{90} \mid A = \frac{1000 - 1200}{100}\right)$$

$$= P(B > 0 | A = -2)$$



$$1 - \Phi\left(\frac{0 - (-1.2)}{0.8}\right)$$

## Appendix

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

The regression line of  $(U, V)$  is

$$E(V|U) = \underbrace{\left(\frac{\sigma_V}{\sigma_U} \rho\right)}_m U + \underbrace{\mu_V - \frac{\sigma_V}{\sigma_U} \rho \mu_U}_b$$

Pl/

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

$$X = \frac{U - \mu_U}{\sigma_U}$$
$$Y = \frac{V - \mu_V}{\sigma_V}$$

$E(Y|X) = \rho X$  is regression line in s.v.

$$\begin{aligned} E(Y|X) &= E\left(\frac{V - \mu_V}{\sigma_V} \mid \frac{U - \mu_U}{\sigma_U}\right) \\ &= E\left(\frac{V - \mu_V}{\sigma_V} \mid U\right) \\ &= \frac{E(V|U) - \mu_V}{\sigma_V} \end{aligned}$$

$$\text{so } \frac{E(V|U) - \mu_V}{\sigma_V} = \rho \frac{U - \mu_U}{\sigma_U}$$

$$\Leftrightarrow E(V|U) - \mu_V = \frac{\sigma_V}{\sigma_U} \rho (U - \mu_U)$$

(=)

$$E(V|U) = \underbrace{\left( \frac{a}{c} \rho \right)}_m U + \underbrace{\mu_V - \frac{a}{c} \rho \mu_U}_b$$

