

# ROY (1951) model

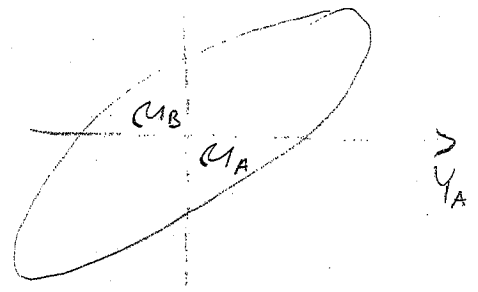
- self-selection
- choosing professions
- observed distr of earnings determined by choice

EXAMPLE:

$Y_{iA}$  ..... hunter } if  $Y_{iA} > Y_{iB} \rightarrow$  hunter  
 $Y_{iB}$  ..... fisher } if  $Y_{iB} < Y_{iA} \rightarrow$  fisher

Assume joint normal distr. of  $(Y_A, Y_B) \sim \text{means } (\mu_A, \mu_B)$

$$\sim \text{COV} \begin{bmatrix} \sigma_A^2 & \sigma_{AB} \\ \sigma_{AB} & \sigma_B^2 \end{bmatrix}$$



Question:

- $E(Y_A | \text{hunter}) = ?$
- $E(Y_B | \text{fisher}) = ?$
- $E(Y_A | \text{fisher}) = ?$
- $E(Y_B | \text{hunter}) = ?$

Define:  $u_A = Y_A - \mu_A$  (will become  $X_{B_A}$ );  $u_B = Y_B - \mu_B$  ( $X_{B_B}$ );  $\sigma^2 = \text{VAR}(u_A - u_B)$

$z = \frac{u_A - u_B}{\sigma}$ ;  $u = \frac{u_B - u_A}{\sigma}$ ;  $\sigma_{AU} = \text{COV}(u_A, u)$   
 $\sigma_{BU} = \text{COV}(u_B, u)$

$$\begin{aligned}
 E(Y_A | \text{hunter}) &= E(Y_A | Y_A - Y_B > 0) = E(Y_A | \frac{u_A - u_B}{\sigma} > \frac{u_B - u_A}{\sigma}) \\
 &= E(Y_A | z > u) = \mu_A + \sigma_{AU} \frac{\phi(z)}{\Phi(z)}
 \end{aligned}$$

$$\begin{aligned}
 E(Y_B | \text{fisher}) &= E(Y_B | Y_A - Y_B < 0) = \dots = \\
 &= \mu_B + \sigma_{BU} \frac{\phi(z)}{1 - \Phi(z)}
 \end{aligned}$$

# Mean of truncated normal

## SIMPLE CASE

Let  $U_1 \sim N(0, 1)$

$E(U_1) = 0$

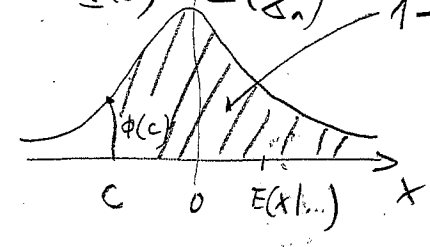
$E(U_1 | U_1 \geq c) = \frac{\phi(c)}{1 - \Phi(c)}$

$E(U_1 | U_1 \leq c) = -\frac{\phi(c)}{\Phi(c)}$

if  $U_n \sim N(0, \sigma_n^2)$

$\phi(c) = \frac{1}{\sigma_n} \phi\left(\frac{c}{\sigma_n}\right)$

$\Phi(c) = \Phi\left(\frac{c}{\sigma_n}\right)$       $1 - \Phi(c)$



## COMPLICATED CASE

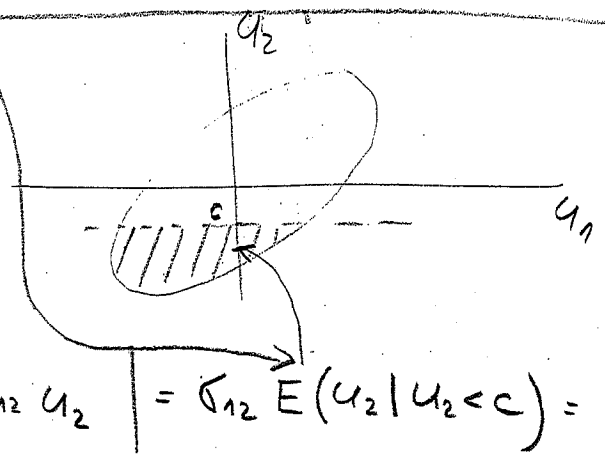
Let  $\begin{bmatrix} U_1 \\ U_2 \end{bmatrix} \sim \begin{bmatrix} 0 \\ 0 \end{bmatrix}; \begin{bmatrix} \sigma_1^2 & \sigma_{12} \\ \sigma_{12} & \sigma_2^2 \end{bmatrix}$

assume 1

$E(U_1 | U_2 < c) = ?$

$E(U_1 | U_2 < c) = | E(U_1 | U_2) = \sigma_{12} U_2 | = \sigma_{12} E(U_2 | U_2 < c) =$

$= -\sigma_{12} \frac{\phi(c)}{\Phi(c)}$



## SPECIAL COMPLICATED CASE

Let  $\begin{bmatrix} U_A \\ U_B - U_A \end{bmatrix} \sim \begin{pmatrix} 0 \\ 0 \end{pmatrix}; \begin{bmatrix} \sigma_A^2 & \sigma_{AU} \\ \sigma_{AU} & \sigma_U^2 \end{bmatrix}$      assume = 1

$E(U_A | U_B - U_A < c) = -\sigma_{AU} \frac{\phi(c)}{\Phi(c)} = \begin{cases} \sigma_{AU} = \text{COV}(U_A, U_B - U_A) = \\ = \frac{1}{N} \sum (U_A - 0)(U_B - U_A - 0) \\ = \frac{1}{N} \sum U_A U_B - \frac{1}{N} \sum U_A^2 = \\ = \sigma_{AB} - \sigma_A^2 \end{cases}$

CONSIDER POSSIBLE CASES (if  $\sigma \neq 1$ )

NOTE that

$$\sigma_{AU} = \frac{\sigma_{AB} - \sigma_A^2}{\sigma} \quad \sigma_{BU} = \frac{\sigma_B^2 - \sigma_{AB}}{\sigma} \quad \rho = \frac{\sigma_{AB}}{\sigma_A \sigma_B}$$

$$\sigma_{BU} - \sigma_{AU} > 0$$

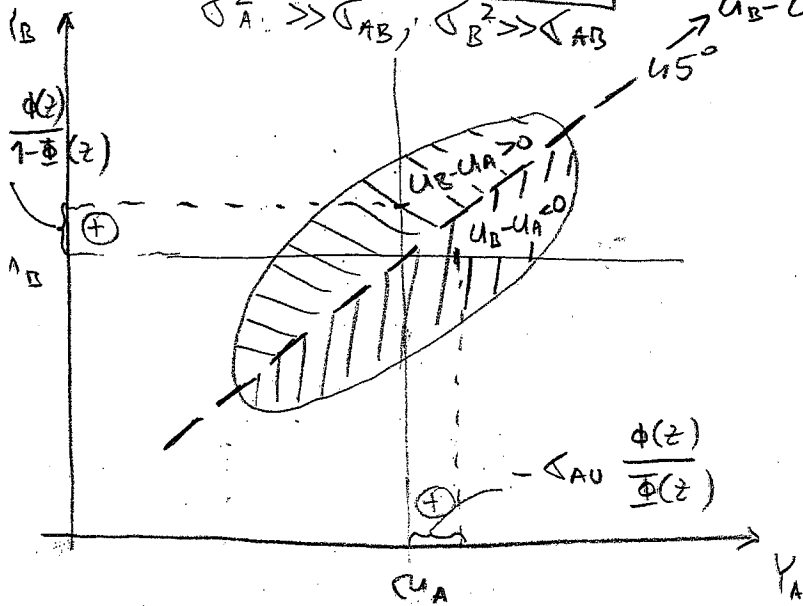
$$\sigma_B^2 - \sigma_{AB} - \sigma_{AB} + \sigma_A^2 > 0$$

$$\sigma_B^2 + \sigma_A^2 > 2\sigma_{AB}$$

$$\sigma_B^2 + \sigma_A^2 > 2\rho\sigma_A\sigma_B$$

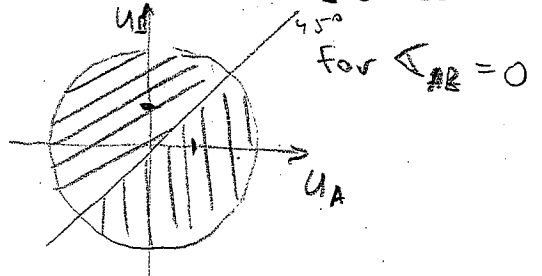
CASE I:  $\sigma_{AU} < 0, \sigma_{BU} > 0$

$$\sigma_A^2 \gg \sigma_{AB}, \sigma_B^2 \gg \sigma_{AB}$$



• Mean income of hunters is above  $\mu_A$  (all)

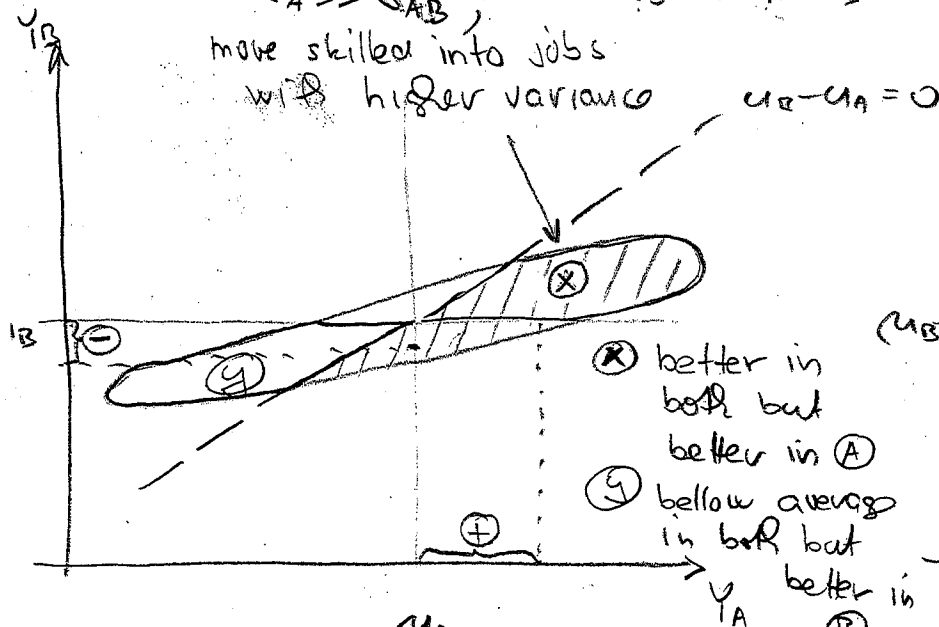
• Mean income of fishermen is above  $\mu_B$  (all)



CASE IIa:  $\sigma_{AU} < 0, \sigma_{BU} < 0$

$$\sigma_A^2 \gg \sigma_{AB}, \sigma_{AB} \gg \sigma_B^2 \Rightarrow \sigma_A^2 \gg \sigma_B^2$$

move skilled into jobs with higher variance



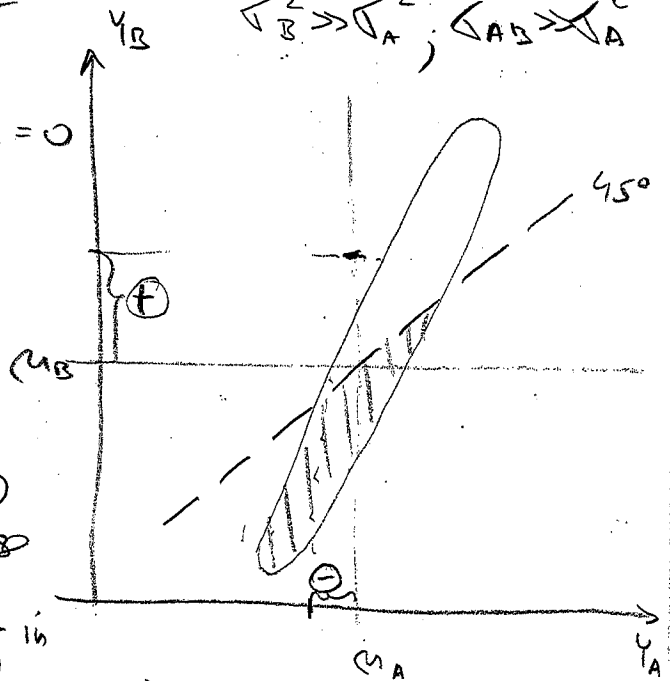
⊗ better in both but better in A

⊙ below average in both but better in

Mean income hunters  $> \mu_A$   
fisherman  $< \mu_B$

CASE IIb:  $\sigma_{AU} > 0, \sigma_{BU} > 0$

$$\sigma_B^2 \gg \sigma_A^2, \sigma_{AB} \gg \sigma_A^2$$



CASE III:  $\sigma_{AU} > 0, \sigma_{BU} < 0$

ALTERNATIVE CHOICES - CONTINUED

$$E(Y_B | Y_A > Y_B) = E(Y_B | u > z) = \mu_B - \sigma_{BU} \frac{\phi(z)}{\Phi(z)}$$

sign matters

↓  
would hunters choose fishing

$$E(Y_A | Y_B > Y_A) = E(Y_A | z < u) = \mu_A + \sigma_{AU} \frac{\phi(z)}{1 - \Phi(z)}$$

sign matters

↓  
would fishermen choose hunting

Observed difference in mean incomes:

$$E(\bar{Y}_A - \bar{Y}_B) = [\mu_A - \mu_B] - \left( \sigma_{AU} \frac{\phi(z)}{\Phi(z)} - \sigma_{BU} \frac{\phi(z)}{1 - \Phi(z)} \right)$$

$$\text{bias} = f(\sigma_{AU}, \sigma_{BU})$$

- $\sigma_{AU} < 0, \sigma_{BU} > 0$
- $\sigma_{AU} < 0, \sigma_{BU} < 0$
- $\sigma_{AU} > 0, \sigma_{BU} > 0$

CASE I: ambiguous bias  
 CASE IIa: upward bias  
 CASE IIb: downward bias