

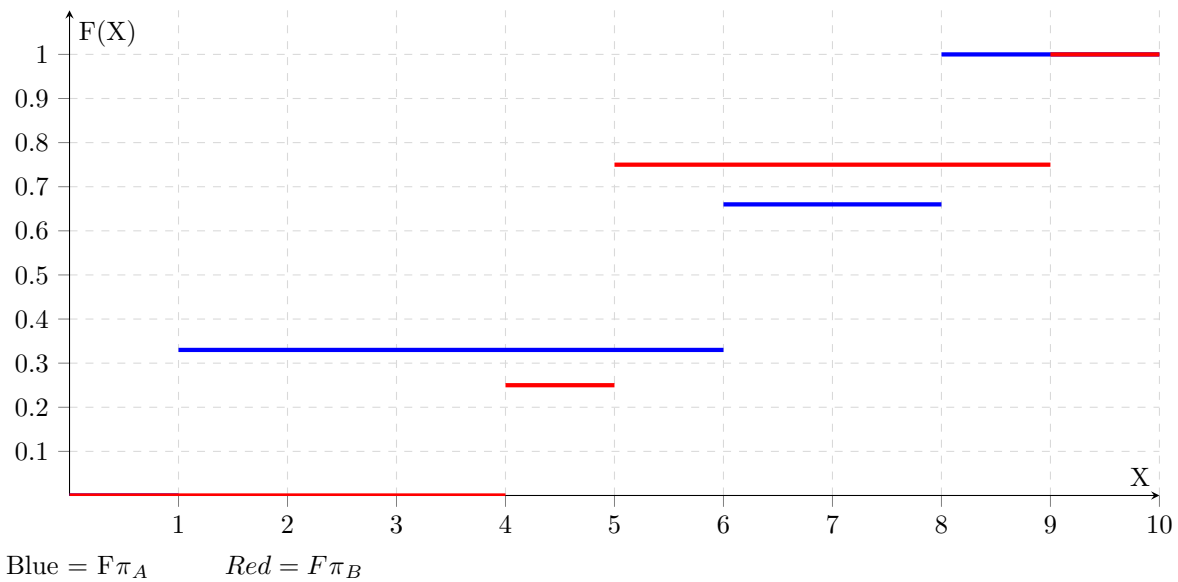
# Lecture 9

Econ139

February 19, 2019

## 1 Second Order Stochastic Dominance(SSD)

$\tilde{X}$	$\pi_A$	$\pi_B$	$F_{\pi_A}$	$F_{\pi_B}$	$\int F_{\pi_A}(t) - F_{\pi_B}(t)$
1	0.33	0	0.33	0	0
4	0	0.25	0.33	0.25	-0.99
5	0	0.5	0.33	0.75	-1.07
6	0.33	0	0.66	0.75	-0.65
8	0.34	0	1	0.75	-0.47
9	0	0.25	1	1	-0.72

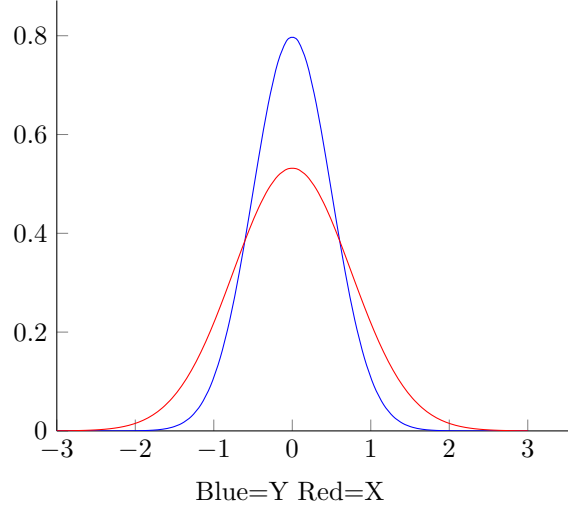


**Definition:** given preamble, we say that  $\tilde{X}_{\pi_A}$  SSD  $\tilde{X}_{\pi_B}$  iff  $\int_{-\infty}^t F_{\pi_B}(Z) - F_{\pi_A}(Z) dz \leq 0 \forall t$

- Asset B SSD because it is negative at every point.

**Theorem:** Given preamble, then  $\tilde{X}_{\pi_B}$  SSD  $\tilde{X}_{\pi_A}$  iff  $E_{\pi_B}[U(\tilde{X})] > E_{\pi_A}[U(\tilde{X})]$  for all increasing and concave U.

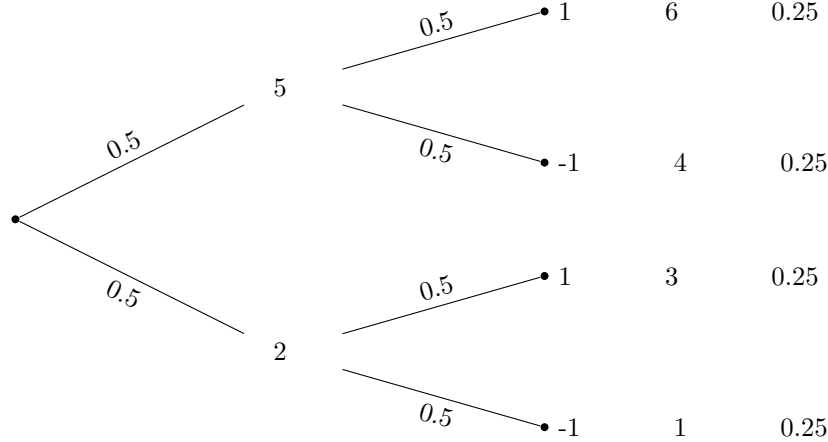
## 2 Mean Preserving Spread



Y is a mean preserving spread of X  
 $Y = X + Z$  where  $E[Z] = 0$   
 $E[Y] = E[X]$   
 $\theta_y^2 > \theta_x^2$

$\tilde{X}$	$\pi_Z$	$\pi_A$	$F_{\pi_B}$
-1	0.5	0	0
1	0.5	0	0.25
2	0	0.5	0
3	0	0	0.25
4	0	0	0.25
5	0	0.5	0
6	0	0	0.25

$$E_{\pi_A}[\tilde{X}] = 3.5, E_{\pi_B}[\tilde{X}] = 3.5, \theta_{\pi_A}^2 = 2.25, \theta_{\pi_B}^2 = 3.25, E_{\pi_Z}[\tilde{X}] = 0$$



**Definition:** Given preamble, we say that  $\tilde{X}_{\pi_B}$  is a mean preserving spread (MPS) of  $\tilde{X}_{\pi_A}$  if there exists a PDF  $\pi_Z$  over  $\tilde{X}$  such that  $E_{\pi_Z}[\tilde{X}] = 0$  and  $\tilde{X}_{\pi_B} = \tilde{X}_{\pi_A} + \tilde{X}_{\pi_Z}$

**Theorem:** Given preamble, if  $E_{\pi_B}[\tilde{X}] = E_{\pi_A}[\tilde{X}]$ , the following statements are equivalent:

- (i)  $\tilde{X}_{\pi_A}$  SSD  $\tilde{X}_{\pi_B}$
- (ii)  $\tilde{X}_{\pi_B}$  is a MPS of  $\tilde{X}_{\pi_A}$

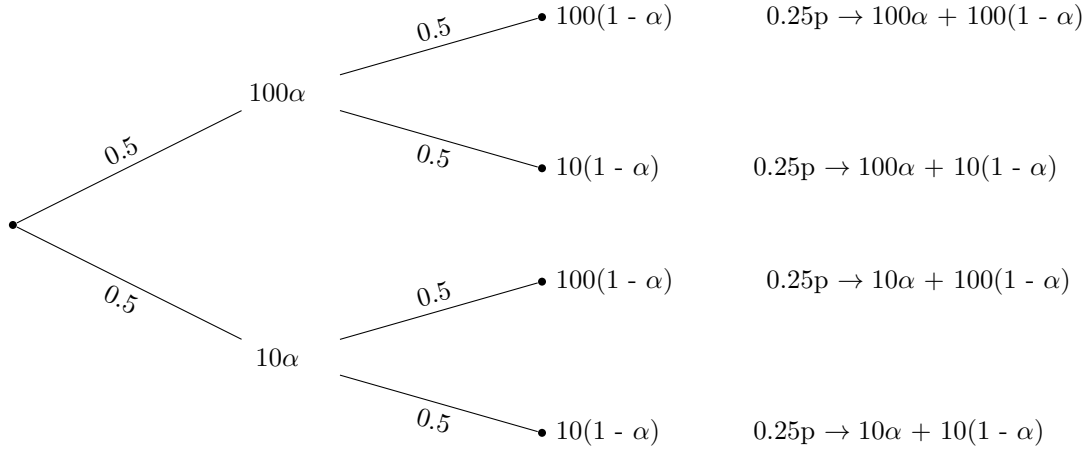
### 3 Diversification

**Consider:** A set of  $n$  lotteries given  $\tilde{x}_1, \dots, \tilde{x}_n$  that are assumed to be independent and identically distributed.

*feasible strategy:* Characterized by a vector,  $A = (\alpha_1, \dots, \alpha_n)$ , where  $\alpha_i$  is one's share in lottery  $i$  where  $\sum_{i \in 1}^n \alpha_i = 1$ . This gives a random payoff  $\tilde{y} = \sum_{i \in 1}^n \alpha_i \tilde{x}_i$

**Proposition:** The random payoff  $\tilde{Y}_D$  generated by the "perfect" diversification strategy  $D = (\frac{1}{n}, \dots, \frac{1}{n})$  SSD the random payoff generated by any other strategy.

**Example:** Two lotteries:  $\tilde{x}_i = (10, 100)$  with  $\pi_i = (0.5, 0.5)$  and  $i = 1, 2$



With perfect diversification:  $D = (\frac{1}{2}, \frac{1}{2})$

w/ prob.	payoff	$E[\tilde{Y}_D] = 55; \theta^2(\tilde{Y}_D) = 101.25,$
0.25	100	
0.25	55	
0.25	55	
0.25	10	

With different strategy:  $A = (\frac{1}{3}, \frac{2}{3})$

w/ prob.	payoff	$E[\tilde{Y}_D] = 55; \theta^2(\tilde{Y}_D) = 102.5,$
0.25	100	
0.25	40	
0.25	70	
0.25	10	

We see the second strategy has the same mean but a higher variance. Thus, the first strategy of perfect diversification SSD the second strategy.

## 4 Investment in Risky Asset II

Say we have the following:

- individual:
  - initial wealth  $w_0$
  - utility function  $u_1$  with  $u' > 0$ ,  $u'' < 0$
- two possible investments:
  - (I) bond that pays  $(1 + r_f)$  for every dollar invested
  - (II) risky asset with uncertain return  $\tilde{r}$
- let  $a$  represent the number of dollars invested in the risky asset
- future wealth:  $\tilde{w}_1 = (w_0 - a)(1 + r_f) + a(1 + \tilde{r}) = w_0(1 + r_f) + a(\tilde{r} - r_f)$

We will maximize expected utility:

$$\max_a E(u(w_1^2))$$

$$\max_a E[u(w_0(1 + r_f) + a(\tilde{r} - r_f))]$$

We will take the FOC (derivative with respect to  $a$ ):

$$E[u'(w_0(1 + r_f) + a(\tilde{r} - r_f))(\tilde{r} - r_f)] = 0$$

**Theorem:** Assume  $u' > 0$  and  $u'' < 0$  and  $a^* = \operatorname{argmax}_a E[u(\tilde{w}_1)]$ . Then:

- (I) For  $a^* > 0 \leftrightarrow E[\tilde{r}] > r_f$
- (II) For  $a^* = 0 \leftrightarrow E[\tilde{r}] = r_f$
- (III) For  $a^* < 0 \leftrightarrow E[\tilde{r}] < r_f$

Following this theorem, we can make assumptions about our expected return per investment.

**Proof:**

Let  $g(a) = E[u(w_0(1 + r_f) + a(\tilde{r} - r_f))]$

Then  $g'(a) = E[u'(w_0(1 + r_f) + a(\tilde{r} - r_f))(\tilde{r} - r_f)]$  and  $g''(a) = E[u''(w_0(1 + r_f) + a(\tilde{r} - r_f))(\tilde{r} - r_f)]$

Observe since  $u'' < 0$ ,  $g''(a) < 0$  implies  $g'(a)$  is strictly decreasing in  $a$

$$\begin{aligned} g'(0) &= E[u'(w_0(1 + r_f))(\tilde{r} - r_f)] \\ &= u'(w_0(1 + r_f))E[\tilde{r} - r_f] \\ &= u'(w_0(1 + r_f))(E[\tilde{r}] - r_f) \end{aligned}$$

We know the first term is positive, so this implies  $g'(0)$  has the same sign as  $E[\tilde{r}] - r_f$

*Summary:*

- (I)  $g'(a)$  is strictly decreasing
- (II)  $g'(0)$  has the sign of  $E[\tilde{r}] - r_f$
- (III)  $g'(a^*) = 0$  because of FOC

$$\text{Now } a^* = 0 \text{ iff } E[\tilde{r}] = r_f$$

To prove forward  $\rightarrow$ :

$$\begin{aligned} &g'(a^*) = 0 \text{ and } a^* = 0 \\ \rightarrow g'(0) = 0 &\text{ since } g'(0) \text{ has the same sign as } E[\tilde{r}] - r_f \\ &\rightarrow E[\tilde{r}] - r_f = 0 \\ &\rightarrow E[\tilde{r}] = r_f \end{aligned}$$

To prove backwards  $\leftarrow$ :

$$\begin{aligned} & \text{Suppose } E[\tilde{r}] = r_f \\ & E[\tilde{r}] - r_f = 0 \\ \rightarrow g'(0) &= 0 \text{ since } g'(a) \text{ is strictly decreasing} \\ \rightarrow g'(a^*) &= 0 \text{ from FOC} \\ \rightarrow a^* &= 0 \end{aligned}$$