

We don't have to reduce the probability to zero:

$$L = \begin{pmatrix} \$10 & \$50 & \$110 \\ \frac{1}{4} & \frac{1}{2} = \frac{5}{10} & \frac{1}{4} \end{pmatrix}$$

Take away some of the probability of \$50, say  $\frac{3}{10}$  and spread it between a lower amount, say \$15, and a higher amount, say \$90:

$$M = \begin{pmatrix} \$10 & \$15 & \$50 & \$90 & \$110 \\ \frac{1}{4} & r & \frac{5}{10} - \frac{3}{10} = \frac{2}{10} & s & \frac{1}{4} \end{pmatrix}$$

For this to be a mean preserving spread we need

$$(1) \quad r + s = \frac{3}{10}$$

$$(2) \quad \frac{3}{10} 50 = 15r + 90s$$

in computation of  $E[M] = \frac{2}{10} 50 + r15 + s90$

in computation of  $E[L] = \frac{5}{10} 50 = \frac{2}{10} 50 + \frac{3}{10} 50$

$$M = \begin{pmatrix} \$10 & \$15 & \$50 & \$90 & \$110 \\ \frac{1}{4} & \frac{4}{25} & \frac{2}{10} & \frac{7}{50} & \frac{1}{4} \end{pmatrix}$$

solution is  $r = \frac{4}{25}, s = \frac{7}{50}$

need these to be equal in order to have  $E[L] = E[M]$

Write  $L >_{SSD} M$  to mean that  $L$  dominates  $M$  in the sense of **second-order stochastic dominance**.

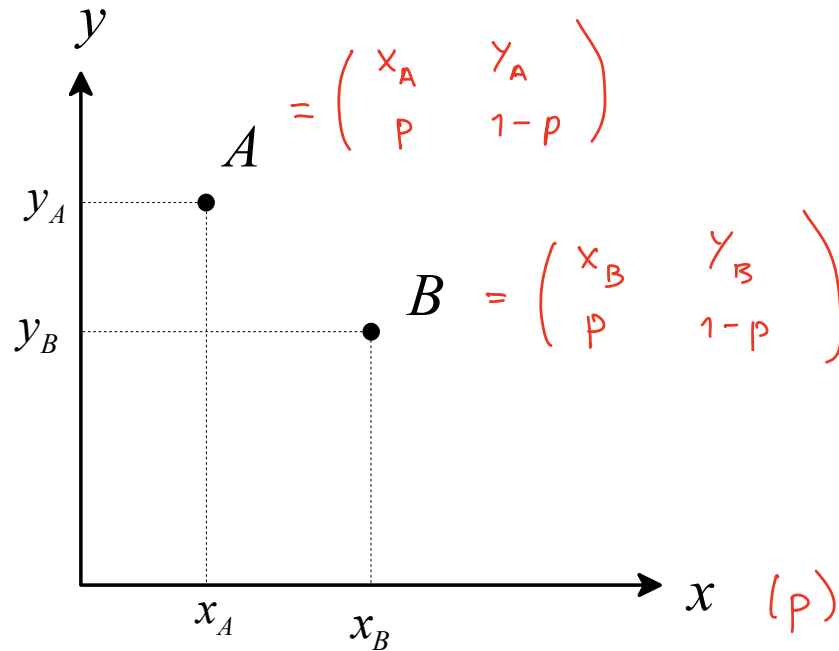
**Definition.**  $L >_{SSD} M$  if  $M$  can be obtained from  $L$  by a finite sequence of mean-preserving spreads, that is, if there is a sequence of money lotteries  $\langle L_1, L_2, \dots, L_m \rangle$  (with  $m \geq 2$ ) such that:

- (1)  $L_1 = L$ ,
- (2)  $L_m = M$
- (3) for every  $i = 1, \dots, m-1$ ,  $L_i \rightarrow_{MPS} L_{i+1}$

**Theorem.**  $L >_{SSD} M$  if and only if  $\mathbb{E}[U(L)] > \mathbb{E}[U(M)]$  for every strictly increasing and strictly concave utility function  $U$ .

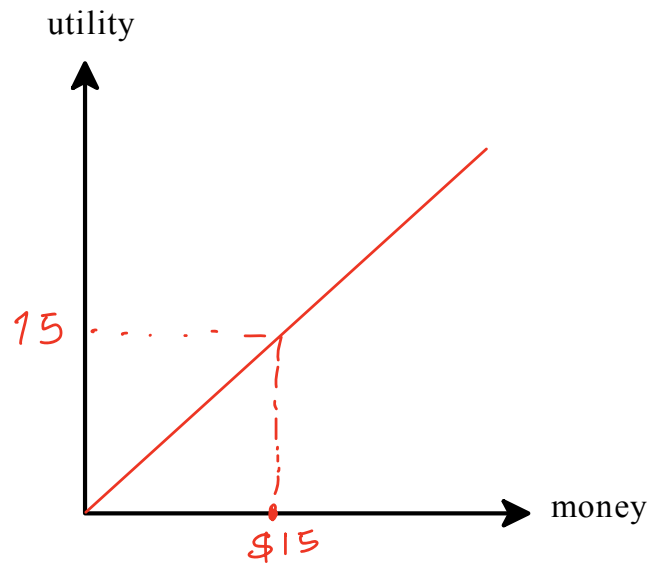
# BINARY LOTTERIES

Lotteries of the form  $\begin{pmatrix} \$x & \$y \\ p & 1-p \end{pmatrix}$  with  $p$  fixed and  $x$  and  $y$  allowed to vary.



We want to draw indifference curves in this diagram.

## Case 1: risk-neutral agent



$$U(x) = x \quad \text{identity function}$$

$$\text{more generally } V(x) = ax + b \quad \text{with } a > 0$$

Let  $A$  and  $B$  be such that  $\mathbb{E}[U(A)] = \mathbb{E}[U(B)]$ :

$$A = \begin{pmatrix} x_A & y_A \\ p & 1-p \end{pmatrix}$$

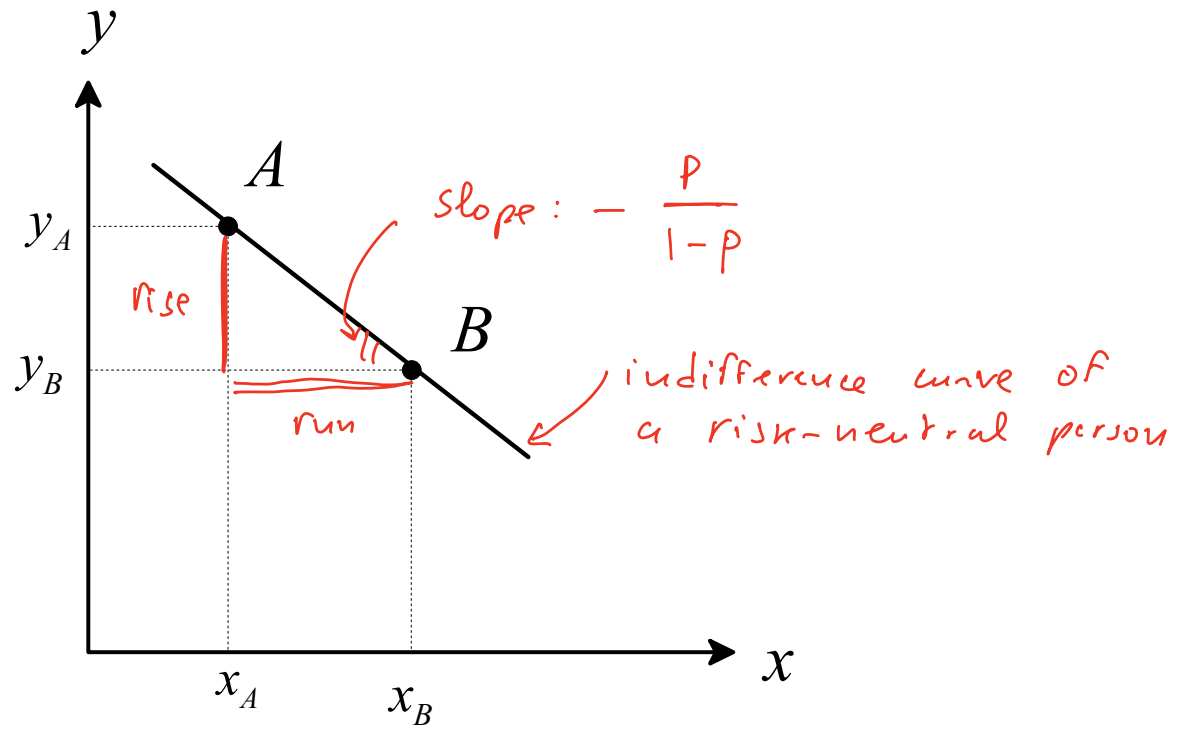
$$B = \begin{pmatrix} x_B & y_B \\ p & 1-p \end{pmatrix}$$

$$\begin{array}{l} \swarrow = E[A] \\ \searrow = E[B] \end{array}$$

$$px_A + (1-p)y_A = px_B + (1-p)y_B$$

$$(1-p)(y_A - y_B) = -p(x_A - x_B)$$

$$\frac{\overbrace{y_A - y_B}^{\text{rise}}}{\underbrace{x_A - x_B}_{\text{run}}} = -\frac{p}{1-p}$$



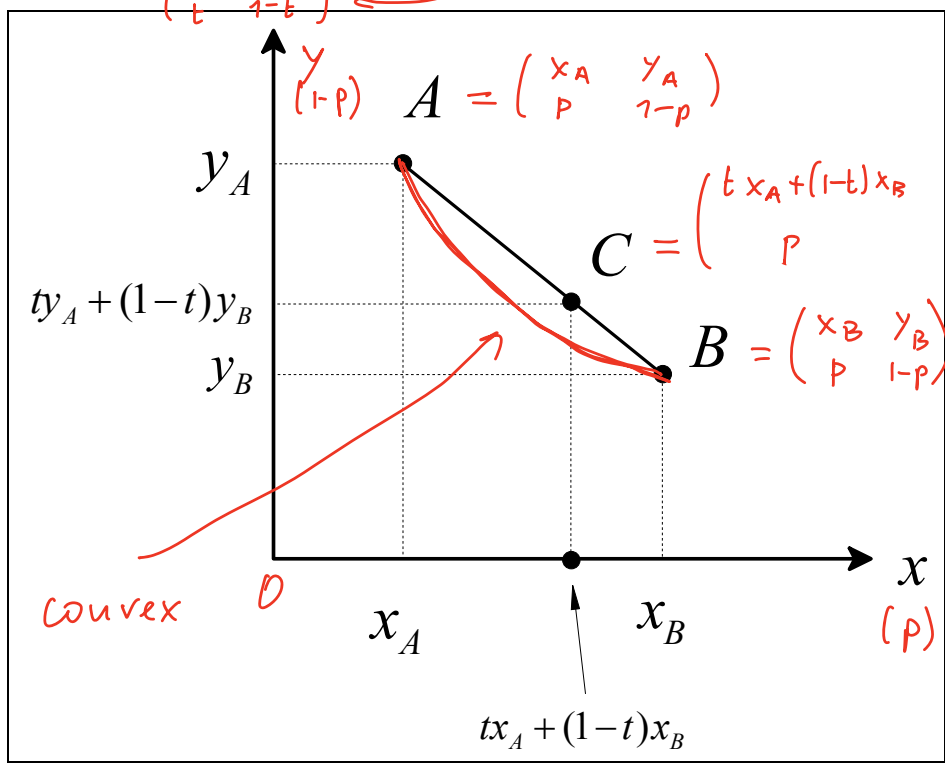
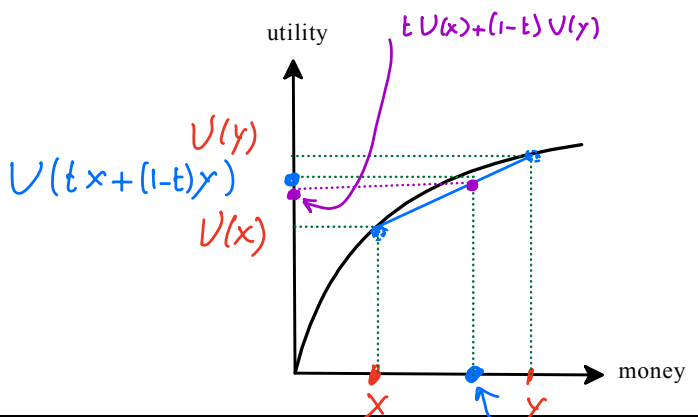
## Case 2: risk-averse agent

fix a  $0 < t < 1$   
 $tx + (1-t)y$

$U(m)$  is strictly concave: if for all  $m_1, m_2$  (with  $0 \leq m_1 < m_2$ ) and all  $t \in (0, 1)$

$$U(t m_1 + (1-t) m_2) > t U(m_1) + (1-t) U(m_2)$$

utility of expected value of lottery  $\begin{pmatrix} m_1 & m_2 \\ t & 1-t \end{pmatrix}$       expected utility of lottery  $y$



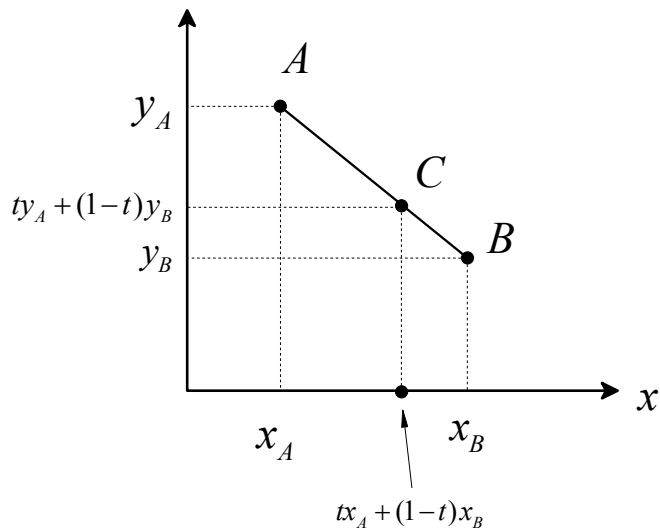
Suppose that  $\mathbb{E}[U(A)] = \mathbb{E}[U(B)]$

$\begin{pmatrix} tx_A + (1-t)x_B & ty_A + (1-t)y_B \\ p & 1-p \end{pmatrix}$

Want to show that  $\mathbb{E}[U(C)] > \mathbb{E}[U(A)] (= \mathbb{E}[U(B)])$

$$[tU(x_A) + (1-t)U(x_B)]$$

by definition of concavity  $\wedge$



$$E[U(C)] = p \underbrace{U(tx_A + (1-t)x_B)} + (1-p) \underbrace{U(ty_A + (1-t)y_B)}$$

$$\checkmark$$

$$t U(y_A) + (1-t) U(y_B)$$

$$E[U(C)] =$$

$$E[U(C)] > p [t U(x_A) + (1-t) U(x_B)]$$

$$+ (1-p) [t U(y_A) + (1-t) U(y_B)]$$

$$= p [t U(x_A) + (1-p) t U(y_A)] + p (1-t) U(x_B) + (1-p) (1-t) U(y_B)$$

$$t [p U(x_A) + (1-p) U(y_A)] + (1-t) \cdot [p U(x_B) + (1-p) U(y_B)]$$

$$E[U(A)]$$

$$E[U(B)]$$

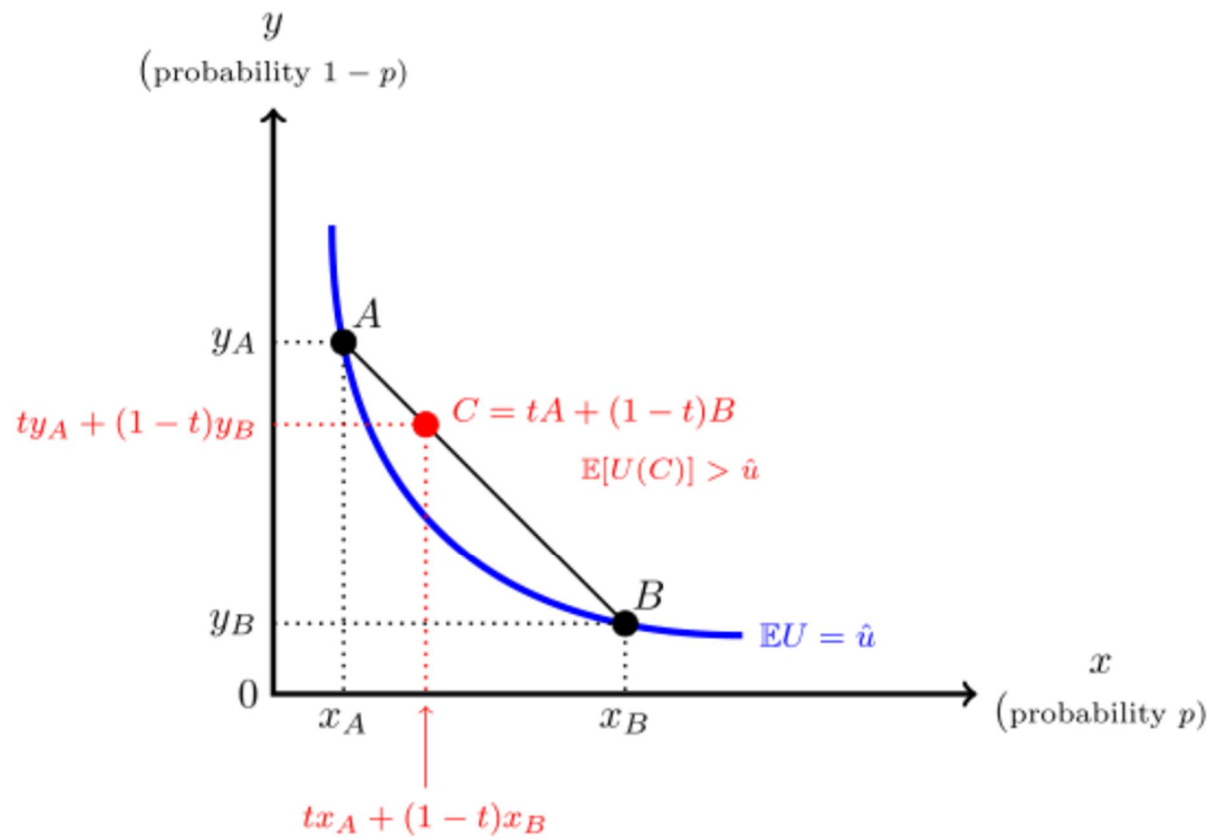
$$E[U(A)]$$

||

$$E[U(A)]$$

The indifference curve must lie below the straight-line segment joining A and B.

$$= E[U(A)]$$





## Case 2: risk-loving agent

