

# Bayesian Nash Equilibrium

Incomplete information game

→ structure of the game is not completely known

Harsanyi's approach :

players' preferences are determined

by realization of a random variable.

ex ante distribution of the r.v. → common knowledge

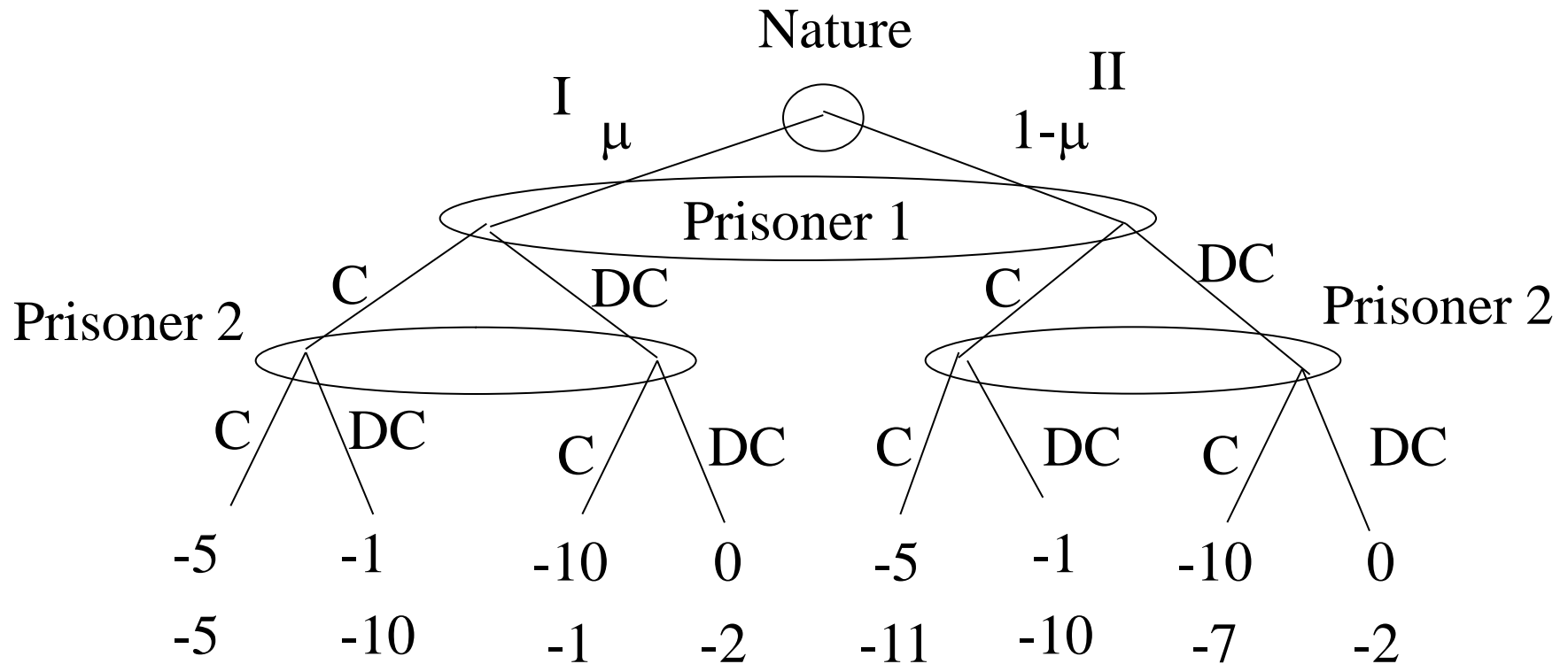
incomplete information game → complete information game  
w/ imperfect information

Nature chooses realization of r.v.

that determines player's preference type

Bayesian game

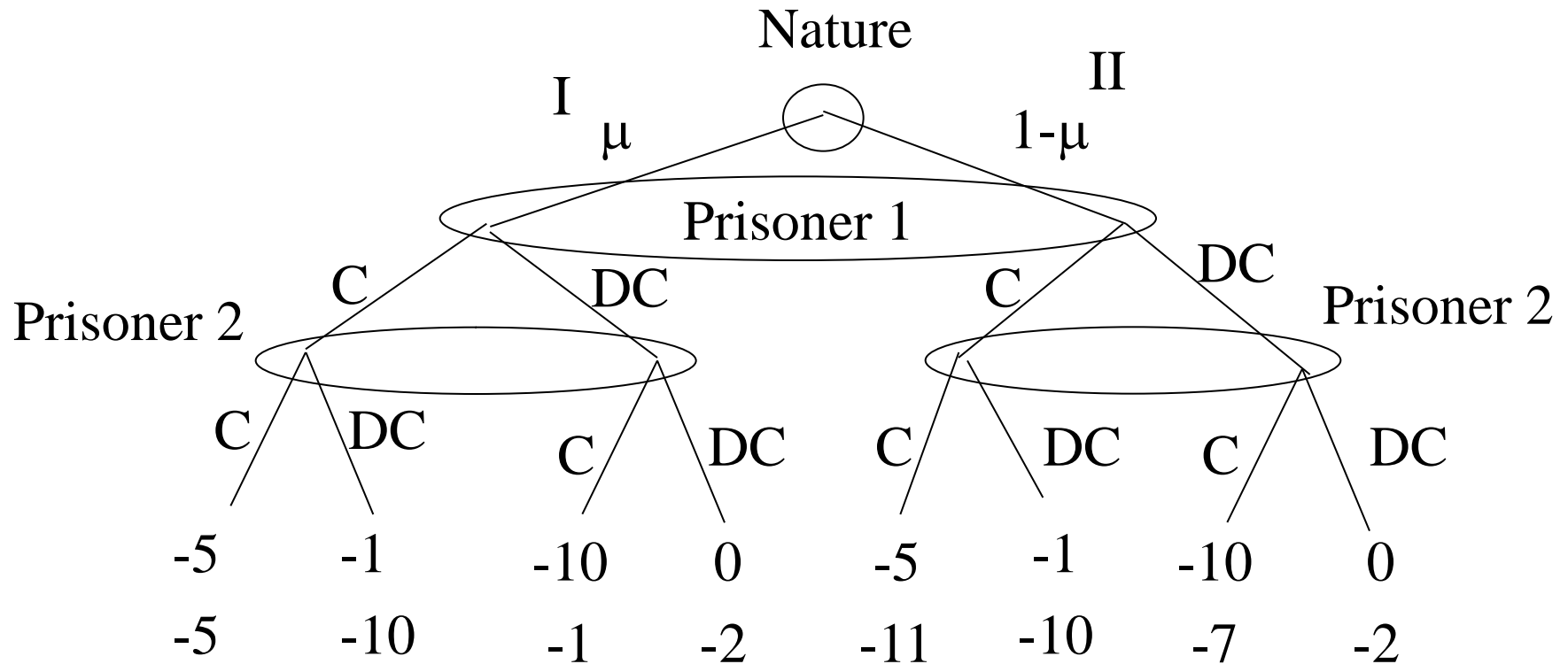
# Example 8.E.1 (Modification of DA's Brother)



	DC	C
DC	0, -2	-10, -1
C	-1, -10	-5, -5

	DC	C
DC	0, -2	-10, -7
C	-1, -10	-5, -11

## Example 8.E.1 (Modification of DA's Brother)



Prisoner 2 has four strategies: (C,C), (C, DC), (DC, C), (DC, DC)  
(type I, type II)

Prisoner 1 has two strategies: C, DC

# Bayesian Game

$[N=\{0,1,\dots,I\}, \{S_i\}, \{u_i\}, \Theta, F]$

$u_i : S(=S_1 \times \dots \times S_I) \times \Theta_i \rightarrow \mathfrak{R}$  utility function

$\Theta = \Theta_1 \times \dots \times \Theta_I$  set of profiles of types

$F : \Theta_1 \times \dots \times \Theta_I \rightarrow \mathfrak{R}$  joint distribution function

$S_i$  : set of actions (so far called strategies)

i's pure strategy  $\mathbf{s}_i : \Theta_i \rightarrow S_i$

set of i's pure strategies  $\mathbf{S}_i$

i's expected payoff under  $(\mathbf{s}_1, \dots, \mathbf{s}_I)$

$$\tilde{u}_i(\mathbf{s}_1, \dots, \mathbf{s}_I) = E_\theta[u_i(\mathbf{s}_1(\theta_1), \dots, \mathbf{s}_I(\theta_I), \theta_i)]$$

# Bayesian Nash Equilibrium

Definition 8.E.1:  $(\mathbf{s}_1, \dots, \mathbf{s}_I)$  is a Bayesian Nash equilibrium

for  $[N=\{0,1,\dots,I\}, \{S_i\}, \{u_i\}, \Theta, F]$

if  $u_i(\mathbf{s}_i, \mathbf{s}_{-i}) \geq u_i(\mathbf{s}'_i, \mathbf{s}_{-i}) \quad \forall \mathbf{s}'_i \in \mathbf{S}_i$

(i.e.,  $(\mathbf{s}_1, \dots, \mathbf{s}_I) \in \mathbf{S} = \mathbf{S}_1 \times \dots \times \mathbf{S}_I$

is a Nash eq. of  $[N=\{1,\dots,I\}, \{\mathbf{S}_i\}, \{u_i\}])$

Proposition 8.E.1:

$(\mathbf{s}_1, \dots, \mathbf{s}_I) \in \mathbf{S}$  is a Bayesian Nash eq.

for  $[N=\{0,1,\dots,I\}, \{S_i\}, \{u_i\}, \Theta, F]$

$\Leftrightarrow \quad \forall i = 1, \dots, I, \quad \forall \theta_i \in \Theta_i$  occurring with positive prob.

$E_{\theta_{-i}}[u_i(\mathbf{s}_i(\theta_i), \mathbf{s}_{-i}(\theta_{-i}), \theta_i) \mid \theta_i] \geq E_{\theta_{-i}}[u_i(\mathbf{s}'_i, \mathbf{s}_{-i}(\theta_{-i}), \theta_i) \mid \theta_i]$

$\forall \mathbf{s}'_i \in \mathbf{S}_i$

# Bayesian Nash Equilibrium

Proposition 8.E.1:

$(\mathbf{s}_1, \dots, \mathbf{s}_I) \in \mathbf{S}$  is a Bayesian Nash eq.

for  $[N=\{0,1,\dots,I\}, \{S_i\}, \{u_i\}, \Theta, F]$

$\Leftrightarrow \forall i = 1, \dots, I, \forall \theta_{-i} \in \Theta_{-i}$  occurring with positive prob.

$$E_{\theta_{-i}}[u_i(\mathbf{s}_i(\theta_{-i}), \mathbf{s}_{-i}(\theta_{-i}), \theta_{-i}) \mid \theta_{-i}] \geq E_{\theta_{-i}}[u_i(s'_i, \mathbf{s}_{-i}(\theta_{-i}), \theta_{-i}) \mid \theta_{-i}]$$

$$\forall s'_i \in S_i$$

Pf.:  $\rightarrow$ ) Suppose  $\exists i \in N, \exists \theta_{-i} \in \Theta_{-i}$  occurring with positive prob.  
and  $\exists s'_i \in S_i$  such that

$$E_{\theta_{-i}}[u_i(\mathbf{s}_i(\theta_{-i}), \mathbf{s}_{-i}(\theta_{-i}), \theta_{-i}) \mid \theta_{-i}] < E_{\theta_{-i}}[u_i(s'_i, \mathbf{s}_{-i}(\theta_{-i}), \theta_{-i}) \mid \theta_{-i}].$$

Let  $\mathbf{s}''_i \in \mathbf{S}_i$  be such that  $\mathbf{s}''_i(\theta_i) = \mathbf{s}_i(\theta_i) \forall \theta_i \neq \theta_{-i}$  and  $\mathbf{s}''_i(\theta_{-i}) = s'_i$ .

Then  $u_i(\mathbf{s}_i, \mathbf{s}_{-i}) < u_i(\mathbf{s}''_i, \mathbf{s}_{-i})$ .

# Bayesian Nash Equilibrium

Proposition 8.E.1:

$(\mathbf{s}_1, \dots, \mathbf{s}_I) \in \mathbf{S}$  is a Bayesian Nash eq.

for  $[N=\{1, \dots, I\}, \{S_i\}, \{u_i\}, \Theta, F]$

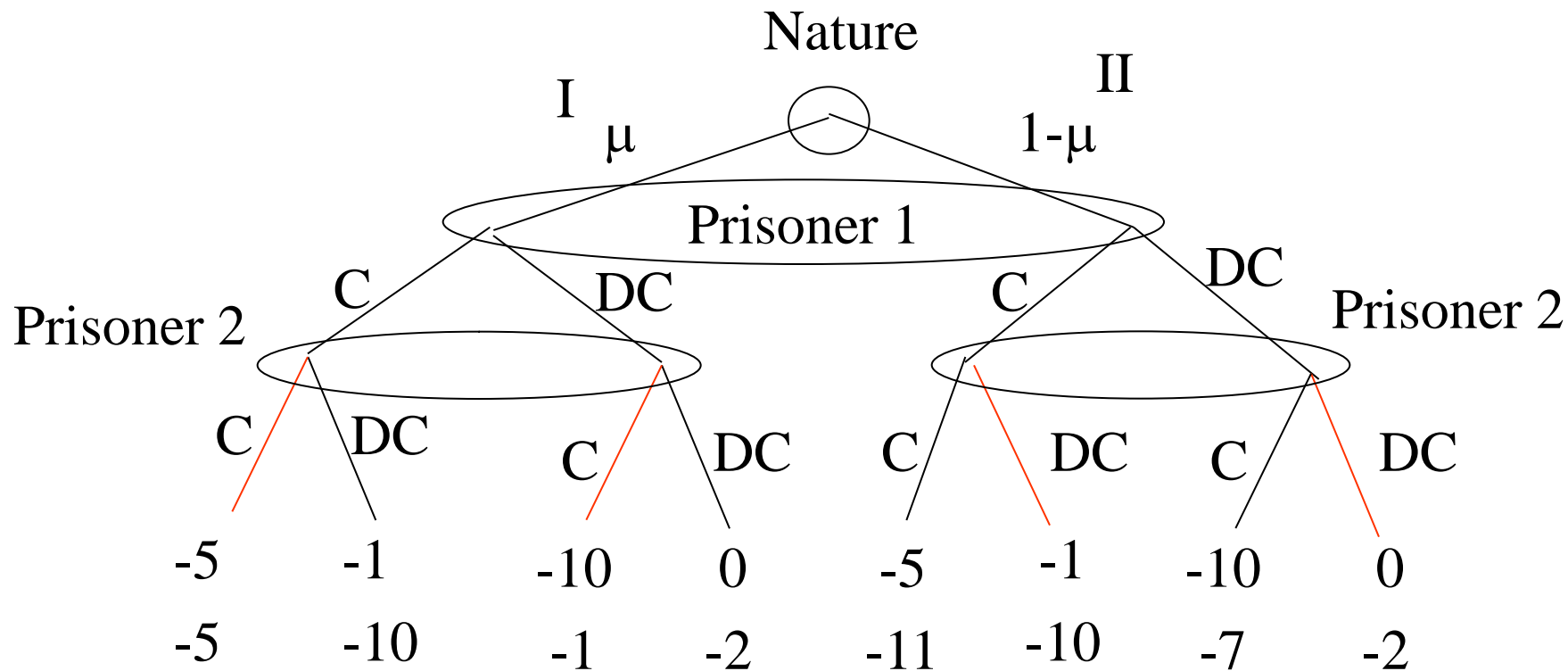
$\Leftrightarrow \forall i = 1, \dots, I, \forall \theta_{-i} \in \Theta_{-i}$  occurring with positive prob.

$$E_{\theta_{-i}}[u_i(\mathbf{s}_i(\theta_{-i}), \mathbf{s}_{-i}(\theta_{-i}), \theta_{-i}) \mid \theta_{-i}] \geq E_{\theta_{-i}}[u_i(s'_i, \mathbf{s}_{-i}(\theta_{-i}), \theta_{-i}) \mid \theta_{-i}]$$
$$\forall s'_i \in S_i$$

Pf.:  $\leftarrow$ ) Take any strategy  $\mathbf{s}'_i \in \mathbf{S}_i$ .

$$\begin{aligned} u_{-i}(\mathbf{s}_i, \mathbf{s}_{-i}) &= E_{\theta}[u_i(\mathbf{s}_i(\theta_i), \mathbf{s}_{-i}(\theta_{-i}), \theta_i)] \\ &= \sum_{\theta_i} \text{Prob}(\theta_i) E_{\theta_{-i}} [u_i(\mathbf{s}_i(\theta_i), \mathbf{s}_{-i}(\theta_{-i}), \theta_i) \mid \theta_i] \\ &\geq \sum_{\theta_i} \text{Prob}(\theta_i) E_{\theta_{-i}} [u_i(\mathbf{s}'_i(\theta_i), \mathbf{s}_{-i}(\theta_{-i}), \theta_i) \mid \theta_i] \\ &= E_{\theta}[u_i(\mathbf{s}'_i(\theta_i), \mathbf{s}_{-i}(\theta_{-i}), \theta_i)] = u_{-i}(\mathbf{s}'_i, \mathbf{s}_{-i}) \end{aligned}$$

# Example 8.E.1 (Modification of DA's Brother)



	DC	C
DC	0, -2	-10, -1
C	-1, -10	-5, -5

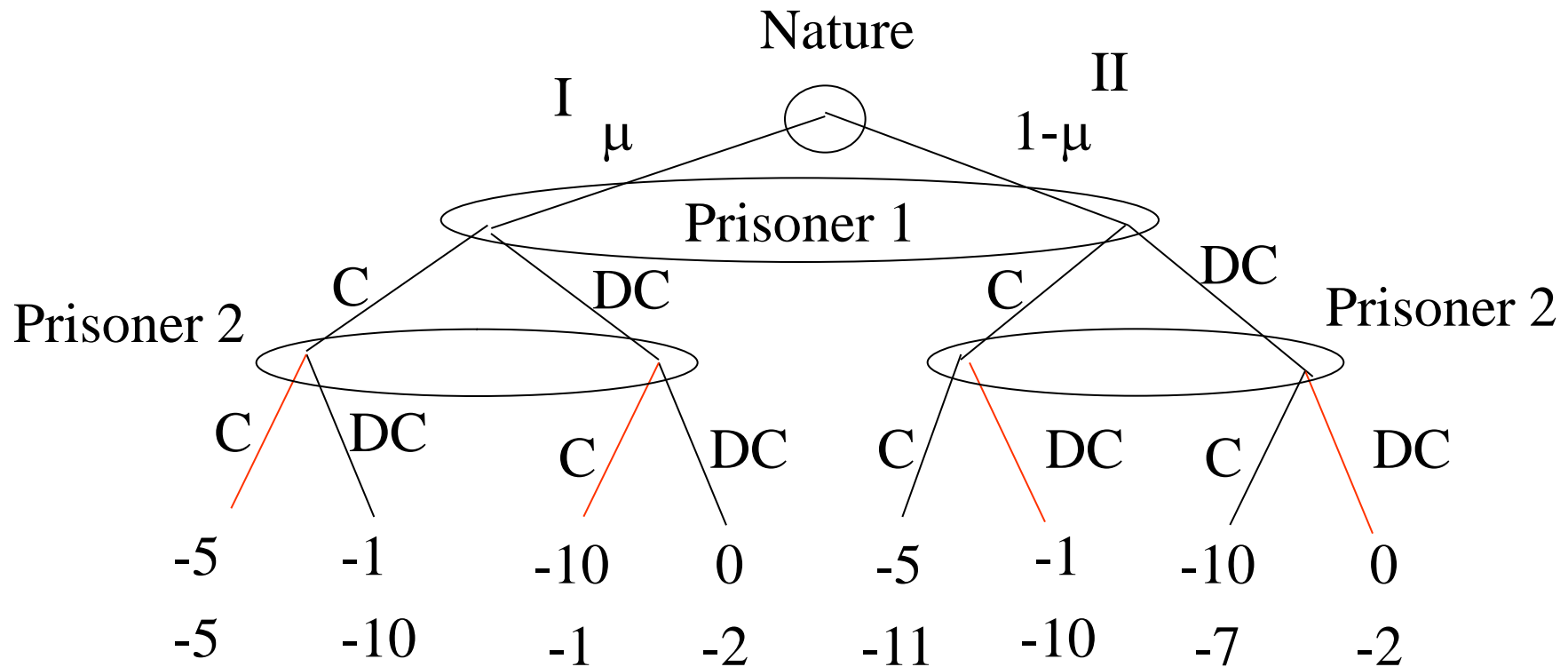
For 2, **C dom DC**

	DC	C
DC	0, -2	-10, -7
C	-1, -10	-5, -11

For 2, **DC dom C**



## Example 8.E.1 (Modification of DA's Brother)



For 1:

$$C \rightarrow -5\mu + (-1)(1-\mu) = -1-4\mu, \quad DC \rightarrow -10\mu + 0(1-\mu) = -10\mu$$

$$-1-4\mu > -10\mu \quad (\mu > 1/6) \rightarrow C$$

$$-1-4\mu < -10\mu \quad (\mu < 1/6) \rightarrow DC$$

$$-1-4\mu = -10\mu \quad (\mu = 1/6) \rightarrow C, DC \text{ indifferent}$$

## Example 8.E.2

AB R&D consortium : firm 1 and firm 2

new tech. Z cost  $c$  ( $0 < c < 1$ )

benefit depends on the type  $\theta_i$  (uniformly distributed on  $[0, 1]$ )  
and given by  $\theta_i^2$

Bayesian NE:

strategy:  $s_i(\theta_i) = 1$  if  $i$  develops Z;  $s_i(\theta_i) = 0$  if not

develop  $\rightarrow \theta_i^2 - c$ ; not develop  $\rightarrow \theta_i^2 \text{ Prob}(s_j(\theta_j) = 1)$

$i$  develops, i.e.  $s_i(\theta_i) = 1$

iff  $\theta_i \geq (c / (1 - \text{Prob}(s_j(\theta_j) = 1)))^{1/2}$  (cutoff rule)

Only one firm: develop iff  $\theta_i^2 \geq c$ , i.e.,  $\theta_i \geq c^{1/2}$

In the consortium: the cutoff value is larger  
possibility to free-ride

## Example 8.E.2

Bayesian NE:  $s_i(\theta_i) = 1$  if  $i$  develops  $Z$ ;  $s_i(\theta_i) = 0$  if not  
develop  $\rightarrow \theta_i^2 - c$ ; not develop  $\rightarrow -\theta_i^2$   $\text{Prob}(s_j(\theta_j) = 1)$   
 $i$  develops iff  $\theta_i \geq (c / (1 - \text{Prob}(s_j(\theta_j) = 1)))^{1/2}$  (cutoff rule)

$\theta_1^*, \theta_2^*$  cutoff values in Bayesian NE

$$\text{Prob}(s_j(\theta_j) = 1) = 1 - \theta_j^*$$

$$\theta_i^* = (c / \theta_j^{*2})^{1/2}, \text{ i.e., } \theta_i^{*2} \theta_j^* = c, \text{ similarly } \theta_i^* \theta_j^{*2} = c$$

$$\text{Thus } \theta_1^* = \theta_2^* = c^{1/3}$$

In the Bayesian NE,

$$\text{Prob}(\text{both develop}) = (1 - c^{1/3})^2$$

$$\text{Prob}(\text{one develops}) = 2c^{1/3}(1 - c^{1/3})$$

$$\text{Prob}(\text{neither develops}) = c^{2/3}$$

# Mixed Strategies and Bayesian NE

Complete information game with a unique NE

Introduce many different types of each player.

Realization of various players' types are statistically independent.  
all types of a player have identical preferences.

A (pure strategy) Bayesian NE

= a mixed strategy NE of the original game

R. Gibbons, Game Theory for Applied Economists

(Japanese translation), pp.150-153

Types are correlated. → correlated equilibrium

# Assignments

Problem Set 5 (due May 19)

Exercises (pp.262-266):

8.E.1 (Assume  $s < M < w$ .)

8.E.3

Reading Assignment:

Text, Chapter 8, pp.258-261