



# *Game Theory*

CO 456



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# Preface

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*Sibelius Peng*

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# Combinatorial games

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## 1.1 Impartial games

- <http://web.mit.edu/sp.268/www/nim.pdf>
- <https://ivv5hpp.uni-muenster.de/u/baysm/teaching/3u03/notes/14-games.pdf>

### Example: Game of Nim

We are given a collection of piles of chips. Two players play alternatively. On a player's turn, they remove at least 1 chip from a pile. First player who cannot move loses the game.

For example, we have three piles with 1, 1, 2 chips. Is there a winning strategy? In this case, there is one for the first player: Player I ( $p_1$ ) removes the pile of 2 chips. This forces  $p_2$  to move a pile of 1 chip.  $p_1$  removes the last chip.  $p_2$  has no move and loses the game. In this case,  $p_1$  has a winning strategy, so this is a **winning game** or **winning position**.

Now let's look at another example with two piles of 5 chips each. Regardless of what  $p_1$  does,  $p_2$  can make the same move on the other pile.  $p_1$  loses. If  $p_1$  loses regardless of their move (i.e.,  $p_2$  has a winning strategy), then this is a **losing game** or **losing position**.

What if we have two piles have unequal sizes? say 5, 7.  $p_1$  moves to equalize the chip count (remove 2 from the pile of 7).  $p_2$  then loses, this is a winning game.

### Lemma 1.1

In instances of Nim with two piles of  $n, m$  chips, it is a winning game if and only if  $m \neq n$ .

Solving Nim with only two piles is easy, but what about games with more than two piles? This is more complicated.

Nim is an example of an **impartial game**. Conditions required for an impartial game:

1. There are 2 players, player I and player II.
2. There are several positions, with a starting position.
3. A player performs one of a set of allowable moves, which depends only on the current position, and not on the player whose turn it is. ("impartial") Each possible move generates an option.
4. The players move alternately.
5. There is complete information.

6. There are no chance moves.
7. The first player with no available move loses.
8. The rules guarantee that games end.

**Example: Not an impartial game**

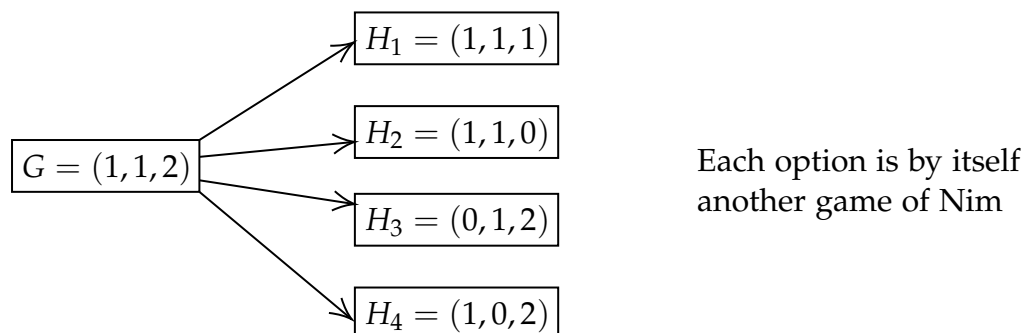
Tic-tac-toe: violates 7.

Chess: violates 3, since players can only move their own pieces.

Monopoly: violates 6. Poker: violates 5.

**Example:**

Let  $G = (1, 1, 2)$  be a Nim game. There are 4 possible moves (hence 4 possible options):



**Note:**

We can define an impartial game by its position and options recursively.

### simpler game

A game  $H$  that is reachable from game  $G$  by a sequence of allowable moves is **simpler** than  $G$ .

Other impartial games:

1. Subtraction game: We have one pile of  $n$  chips. A valid move is taking away 1, 2, or 3 chips. The first player who cannot move loses.



2. Rook game: We have an  $m \times n$  chess board, and a rook in position  $(i, j)$ . A valid move is moving the rook any number of spaces left or up. The first player who cannot move loses.



3. Green hackenbush game: We have a graph and the floor. The graph is attached to the floor at some vertices. A move consists of removing an edge of the graph, and any part of the graph not connected to the floor is removed. The first player who cannot move loses.



**Spoiler** A main result we will prove is that all impartial games are essentially like a Nim game.

### Lemma 1.2

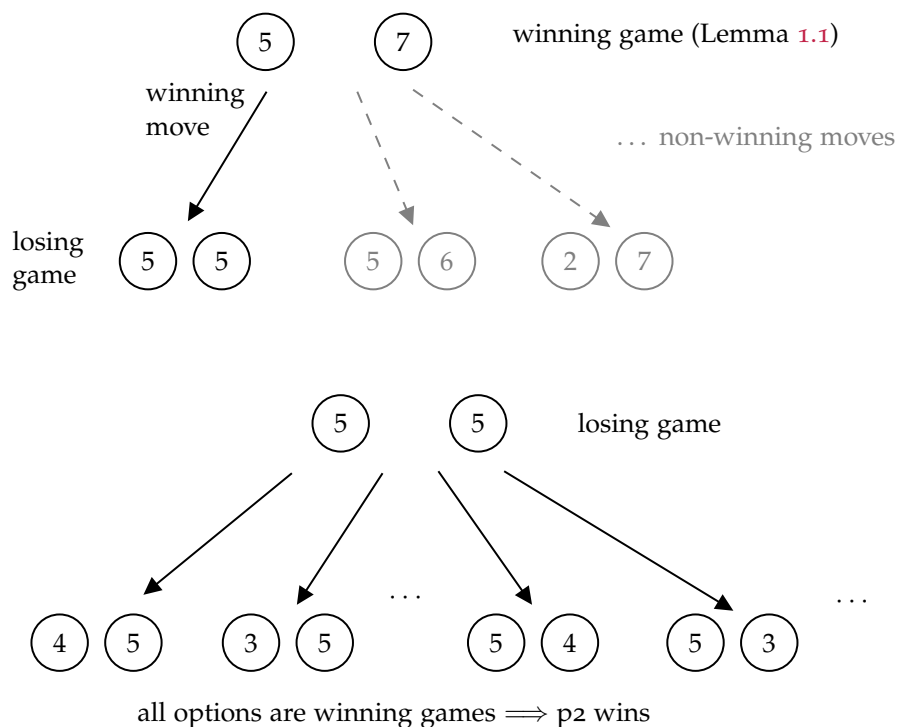
In any impartial game  $G$ , either player I or player II has a winning strategy.

#### Proof:

We prove by induction on the simplicity of  $G$ . If  $G$  has no allowable moves, then  $p_1$  loses, so  $p_2$  has a winning strategy. Assume  $G$  has allowable moves and the lemma holds for games simpler than  $G$ . Among all options of  $G$ , if  $p_1$  has a winning strategy in one of them, then  $p_1$  moves to that option and wins. Otherwise,  $p_2$  has a winning strategy for all options. So regardless of  $p_1$ 's move,  $p_2$  wins.  $\square$

So every impartial game is either a winning game ( $p_1$  has a winning strategy) or a losing game ( $p_2$  has a winning strategy).

#### Example: Nim



#### Note:

We assume players play perfectly. If there is a winning move, then they will take it.

## 1.2 Equivalent games

### game sums

Let  $G$  and  $H$  be two games with options  $G_1, \dots, G_m$  and  $H_1, \dots, H_n$  respectively. We define  $G + H$  as the games with options

$$G_1 + H, \dots, G_m + H, G + H_1, \dots, G + H_n.$$

#### Example:

We denote  $*n$  to be a game of Nim with one pile of  $n$  chips. Then  $*1 + *1 + *2$  is the game with 3 piles of 1, 1, 2 chips.

#### Example:

If we denote  $\#2$  to be the subtraction game with  $n$  chips, then  $*5 + \#7$  is a game where a move consists of either removing at least 1 chip from the pile of 5 (Nim game), or removing 1, 2 or 3 chips from the pile of 7 (subtraction game).

### Lemma 1.3

Let  $\mathcal{G}$  be the set of all impartial games. Then for all  $G, H, J \in \mathcal{G}$ ,

1.  $G + H \in \mathcal{G}$  (closure)
2.  $(G + H) + J = G + (H + J)$  (associative)
3. There exists an identity  $0 \in \mathcal{G}$  (game with no options) where  $G + 0 = 0 + G = G$
4.  $G + H = H + G$  (symmetric)

#### Note:

This is an abelian group except the inverse element.

### equivalent game

Two games  $G, H$  are **equivalent** if for any game  $J$ ,  $G + J$  and  $H + J$  have the same outcome (i.e., either both are winning games, or both are losing games).

Notation:  $G \equiv H$ .

#### Example:

$*3 \equiv *3$  since  $*3 + J$  is the same game as  $*3 + J$  for any  $J$ , so they have the same outcome.

$*3 \not\equiv *4$  since  $*3 + *3$  is a losing game, but  $*4 + *3$  is a winning game from Lemma 1.1.

### Lemma 1.4

$*n \equiv *m$  if and only if  $n = m$ .

**Lemma 1.5**

The relation  $\equiv$  is an equivalence relation. That is, for all  $G, H, K \in \mathcal{G}$ ,

1.  $G \equiv G$  (reflexive)
2.  $G \equiv H$  if and only if  $H \equiv G$  (symmetric)
3. If  $G \equiv H$  and  $H \equiv K$ , then  $G \equiv K$  (transitive).

**Exercise:**

Prove that if  $G \equiv H$ , then  $G + J \equiv H + J$  for any game  $J$ .

Note that the definition above only says they have the same outcome. To prove that they are equivalent, one needs to add another game on both sides to show they have the same outcome.

Nim with one pile  $*n$  is a losing game if and only if  $n = 0$ .

**Theorem 1.6**

$G$  is a losing game if and only if  $G \equiv *0$ .

**Proof:**

$\Leftarrow$  If  $G \equiv *0$ , then  $G + *0$  has the same outcome as  $*0 + *0$ . But  $*0$  is a losing game, so  $G$  is a losing game.

$\Rightarrow$  Suppose  $J$  is a losing game. (We want to show  $G \equiv *0$ , meaning  $G + J$  and  $*0 + J \equiv J$  have the same outcome.)

1. Suppose  $J$  is a losing game. (We want to show that  $G + J$  is a losing game.)

We will prove “If  $G$  and  $J$  are losing games, then  $G + J$  is a losing game” by induction on the simplicity of  $G + J$ . When  $G + J$  has no options, then  $G, J$  both have no options, so  $G, J, G + J$  are all losing games.

Suppose  $G + J$  has some options. Then  $p_1$  makes a move on  $G$  or  $J$ . WLOG say  $p_1$  makes a move in  $G$ , and results in  $G' + J$ . Since  $G$  is a losing game,  $G'$  is a winning game. So  $p_2$  makes a winning move from  $G'$  to  $G''$ , and this results in  $G'' + J$ . Then  $G''$  is a losing game, so by induction,  $G'' + J$  is a losing game for  $p_1$ . So  $p_1$  loses, and  $G + J$  is a losing game.

2. Suppose  $J$  is a winning game. Then  $J$  has a winning move to  $J'$ . So  $p_1$  moves from  $G + J$  to  $G + J'$ . Now both  $G, J'$  are losing games, so by case 1,  $G + J'$  is a losing game. So  $p_2$  loses, meaning  $p_1$  wins, so  $G + J$  is a winning game.

□

**Corollary 1.7**

If  $G$  is a losing game, then  $J$  and  $J + G$  have the same outcome for any game  $J$ .

**Proof:**

Since  $G$  is a losing game,  $G \equiv *0$  by Theorem 1.6. Then  $J + G \equiv J + *0 \equiv J$  (previous exercise + Lemma 1.3). So  $J$  and  $G + J$  have the same outcome. □

**Example:**

1. Recall  $*5 + *5$  and  $*7 + *7$  are losing games. Then Corollary 1.7 says  $*5 + *5 + *7 + *7$  is also a losing game. ( $p_1$  moves in either  $*5 + *5$  or  $*7 + *7$ . Then  $p_2$  makes a winning move from the same part, equalizing piles.)



2.  $\underbrace{*1 + *1 + *2}_{\text{winning}} + \underbrace{*5 + *5}_{\text{losing}}$ . Corollary 1.7 implies this is a winning game.

(p1 makes a winning move in  $*1 + *1 + *2$ , therefore we have  $\underbrace{*1 + *1}_{\text{losing}} + \underbrace{*5 + *5}_{\text{losing}}$ . p2 loses.)

### Lemma 1.8: Copycat principle

For any game  $G$ ,  $G + G \equiv *0$ .

#### Proof:

Induction on the simplicity of  $G$ . When  $G$  has no options,  $G + G$  has no options, so  $G + G \equiv *0$  by Theorem 1.6. Suppose  $G$  has options, and WLOG suppose p1 moves from  $G + G$  to  $G' + G$ . Then p2 can move to  $G' + G'$ . By induction,  $G' + G' \equiv *0$ , so it is a losing game for p1. Therefore,  $G + G$  is a losing game, and  $G + G \equiv *0$ .  $\square$

### Lemma 1.9

$G \equiv H$  if and only if  $G + H \equiv *0$ .

#### Proof:

$\Rightarrow$  From  $G \equiv H$ , we add  $H$  to both sides to get  $G + H \equiv H + H \equiv *0$  by the copycat principle.

$\Leftarrow$  From  $G + H \equiv *0$ , we add  $H$  to both sides to get  $G + H + H \equiv *0 + H \equiv H$ . But  $G + G + G \equiv G + *0 \equiv G$  by the copycat principle. So  $G \equiv H$ .  $\square$

#### Example:

$*1 + *2 + *3$  is a losing game, so  $*1 + *2 + *3 \equiv *0$ . By Lemma 1.9,  $*1 + *2 \equiv *3$ , or  $*1 + *3 \equiv *2$ .

Another way to prove game equivalence is by showing that they have equivalent options.

### Lemma 1.10

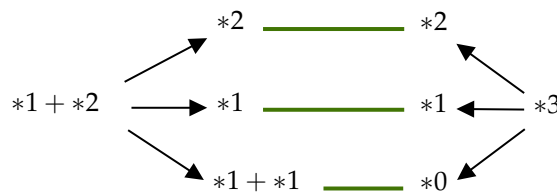
If the options of  $G$  are equivalent to options of  $H$ , then  $G \equiv H$ . (More precisely: There is a bijection between options of  $G$  and  $H$  where paired options are equivalent.)

#### Proof:

It suffices to show that  $G + H \equiv *0$  by Lemma 1.9, i.e.,  $G + H$  is a losing game. This is true when  $G, H$  both have no options. Suppose  $G, H$  have options, and suppose WLOG p1 moves to  $G'H$ . By assumption, there exists an options of  $H$ , say  $H'$ , such that  $H' \equiv G'$ . So p2 can move to  $G' + H'$ . Since  $G' \equiv H'$ ,  $G' + H' \equiv *0$  by Lemma 1.9. So  $G' + H'$  is a losing game for p1. Hence  $G + H$  is a losing game.  $\square$

#### Example:

We can show  $*1 + *2 \equiv *3$  using Lemma 1.10.



#### Note:

The converse is false.

### 1.3 Nim and numbers

**Goal** Show that every Nim game is equivalent to a Nim game with a single pile.

#### number

If  $G$  is a game such that  $G \equiv *n$  for some  $n$ , then  $n$  is the **number** of  $G$ .

#### Example:

Any losing game has number 0 by Theorem 1.6.

#### Exercise:

Show that the notion of a number is well-defined. That is it is not possible for a game to have more than one number.

#### Theorem 1.11

Suppose  $n = 2^{a_1} + 2^{a_2} + \dots$  where  $a_1 > a_2 > \dots$ , then  $*n \equiv *2^{a_1} + *2^{a_2} + \dots$

#### Example:

$11 = 2^3 + 2^1 + 2^0$ ,  $13 = 2^3 + 2^2 + 2^0$ . Using this theorem,  $*11 \equiv *2^3 + *2^1 + *2^0$  and  $*13 \equiv *2^3 + *2^2 + *2^0$ . Then

$$\begin{aligned}
 *11 + *13 &\equiv (*2^3 + *2^1 + *2^0) + (*2^3 + *2^2 + *2^0) \\
 &\equiv (*2^3 + *2^3) + *2^2 + *2^1 + (*2^0 + *2^0) \quad \text{by assoc'y and commu'y} \\
 &\equiv *0 + *2^2 + *2^1 + *0 \quad \text{by copycat principle} \\
 &\equiv *2^2 + *2^1 \\
 &\equiv *(2^2 + 2^1) \\
 &\equiv *6
 \end{aligned}$$

So the number of  $*11 + *13$  is 6.

In general, how can we find the number for  $*b_1 + *b_2 + \dots + *b_n$ ? Look for binary expansions of each  $b_i$ . Copycat principle cancels any pair of identical powers of 2. So we look for powers of 2's that appear in odd number of expansions of the  $b_i$ 's.

Use binary numbers: 11 in binary is 1011, 13 in binary is 1101. Take the XOR operation. We do normal addition except we do not carry over.

$$\begin{array}{r}
 1011 \\
 \oplus 1101 \\
 \hline
 0110
 \end{array}$$

and 0110 is 6. So  $11 \oplus 13 = 6$ .

#### Example:

Consider  $*25 + *21 + *11$ . In binary they are 11001, 10101, 01011.

$$\begin{array}{r}
 11001 \\
 10101 \\
 \oplus 01011 \\
 \hline
 00111
 \end{array}$$

and 00111 is 7. So  $*25 + *21 + *11 \equiv *7$ . (The number is 7)

#### Corollary 1.12

$$*b_1 + *b_2 + \dots + *b_n \equiv *(b_1 \oplus b_2 \oplus \dots \oplus b_n).$$

This shows that every Nim game has a number.

## Winning strategy for Nim

**Example:**

$*11 + *13 \equiv *6$ . This is a winning game. How to find a winning move? Want to move a game equivalent to  $*0$ . Add  $*6$  to both sides:  $*11 + *13 + *6 \equiv *6 + *6 \equiv *0$  (copycat principle).

Consider  $*11 + (*13 + *6)$ . We see  $13 \oplus 6 = 11$ . So this is equivalent to  $*11 + *11$ , a losing game. Winning move: remove 2 chips from the pile of 13.

**Example:**

$*25 + *13 + *11 \equiv *7$ . Add  $*7$  to both sides. Consider  $*25 + (*21 + *7) + *11$ . We see  $21 \oplus 7 = 18$ , so this is equivalent to  $*25 + *18 + *11$ . Winning move: remove 3 chips from the pile of 21.

Why did we pair  $*7$  with  $*21$  instead of  $*25$  or  $*11$ ?  $25 \oplus 7 = 31$ ,  $11 \oplus 7 = 12$ . This means that we are adding 6 chips to 25, or adding 1 chip to 11. Not allowed in Nim.

### Lemma 1.13

If  $*b_1 + \dots + *b_n \equiv *s$  where  $s > 0$ , then there exists some  $b_i$  where  $b_i \oplus s < b_i$ .

Idea: Look for the largest power of 2 in  $s$ .

$$\begin{array}{r}
 *25 + *21 + *11 \equiv *7 \\
 \oplus \quad \begin{array}{r}
 11001 \quad 25 \\
 10101 \quad 21 \\
 01011 \quad 11 \\
 \hline
 00111 \quad 7
 \end{array}
 \end{array}$$

$21 \oplus 7: 4 \text{ is subtracted from } 21$   
 $25 \oplus 7 \text{ or } 11 \oplus 7: 4 \text{ is added}$

$\begin{array}{c} \uparrow \uparrow \uparrow \\ 4 \ 2 \ 1 \end{array} \leftarrow 4 > 2 + 1 \longrightarrow \begin{array}{l} \oplus \text{ reduces } 21 \\ \oplus \text{ increases } 25 \text{ or } 11 \end{array}$

**Proof:**

Suppose  $s = 2^{a_1} + 2^{a_2} + \dots$  where  $a_1 > a_2 > \dots$ . Then  $2^{a_1}$  appears in the binary expansions of  $b_1, \dots, b_n$  an odd number of times. Let  $b_i$  be one of them. Suppose  $*b_i + *s \equiv *t$  for some  $t$ . Since  $2^{a_1}$  is in the binary expansions of  $b_i$  and  $s$ ,  $2^{a_1}$  is not in the binary expansion of  $t$ . For  $2^{a_2}, 2^{a_3}, \dots$ , at worse none of them are in the binary expansion of  $b_i$ , so all of them are in the binary expansion of  $t$ . So

$$t \leq b_i - 2^{a_1} + 2^{a_2} + 2^{a_3} + \dots < b_i \quad \text{since } 2^{a_1} > 2^{a_2} + 2^{a_3} + \dots$$

□

Finding winning moves in a winning Nim game: Say a game has number  $s$ . Look at the largest power of 2 in the binary expansion of  $s$ . Pair it up with any pile  $*b_i$  containing this power of 2. Then  $s \oplus b_i < b_i$ . So a winning move is taking away  $b_i - (s \oplus b_i)$  chips from the pile  $*b_i$ .

Now we wish to prove Theorem 1.11. The proof uses the following lemma:

### Lemma 1.14

Let  $0 \leq p, q < 2^a$ , and suppose Theorem 1.11 hold for all values less than  $2^a$ . Then  $p \oplus q < 2^a$ .

*Illustration for the proof of Theorem 1.11.* Consider  $*7$ .  $7 = 4 + 2 + 1$ . Want to prove  $*7 \equiv *4 + \underbrace{*2 + *1}_{\equiv *3 \text{ by induction}}$

Options of  $*7$ :  $*0, *1, \dots, *6$

Options of  $*4 + *3$ : (1) Move on  $*4$       (2) Move on  $*3$

$$\begin{array}{l}
 (1) \quad \left. \begin{array}{l} *0 + *3 \equiv *3 \\ *1 + *3 \equiv *2 \\ *2 + *3 \equiv *1 \\ *3 + *3 \equiv *0 \\ < 4 \quad < 4 \Rightarrow < 4 \end{array} \right\} \text{distinct} \\
 \text{by Lemma 1.14}
 \end{array}$$

$$\begin{array}{l}
 (2) \quad \begin{array}{l} *4 + *2 \equiv *6 \\ *4 + *1 \equiv *5 \\ *4 + *0 \equiv *4 \end{array} \\
 \text{binary expansion do not have 4} \\
 \text{each power of 2 appears at most once} \\
 \Rightarrow \text{apply induction}
 \end{array}$$

**Proof of Theorem 1.11:**

We prove by induction on  $n$ .

When  $n = 1$ ,  $n = 2^0$  and  $*1 \equiv *2^0$ . Suppose  $n = 2^{a_1} + 2^{a_2} + \dots$  where  $a_1 > a_2 > \dots$ . Let  $q = n - 2^{a_1} = 2^{a_2} + 2^{a_3} + \dots$

If  $q = 0$ , then  $n = 2^{a_1}$ , so  $*n \equiv 2^{a_1}$ .

Assume  $q \geq 1$ . Since  $q < n$ , by induction,  $*q \equiv *2^{a_2} + *2^{a_3} + \dots$ . It remains to show that  $*n \equiv *2^{a_1} + *q$ . The options of  $*n$  are  $*0, *1, \dots, *(n-1)$ . The options of  $*2^{a_1} + *q$  can be partitioned into 2 types.

1. Consider options of the form  $*i + *q$  where  $0 \leq i < 2^{a_1}$ . Since  $i, q < n$ , by induction, the theorem holds for  $i, q$ . So  $*i, *q$  are equivalent to sums of Nim piles by their binary expansions. Using arguments from Corollary 1.12,  $*i + *q \equiv *r_i$  where  $r_i = i \oplus q$ . Since  $i, q < 2^{a_1}, r_i < 2^{a_1}$  by Lemma 1.14. So  $0 \leq r_0, r_1, \dots, r_{2^{a_1}-1} < 2^{a_1}$ .

(We now show that these  $r_i$ 's are distinct.) Suppose  $r_i = r_j$  for some  $i, j$ . Then  $*r_i \equiv *r_j$ , so  $*i + *q \equiv *j + *q$ . Adding  $*q$  on both sides, we get  $*i \equiv *j$  (copycat principle), so  $i = j$ . So the  $r_i$ 's are distinct.

Also there are  $2^{a_1}$  of these  $r_i$ 's, and there are  $2^{a_1}$  possible values (0 to  $2^{a_1} - 1$ ). By Pigeonhole principle, for each  $0 \leq j < 2^{a_1} - 1$ , there is one  $r_i$  with  $r_i = j$ . So the options of this type are equivalent to  $\{ *0, *1, \dots, *(2^{a_1} - 1) \}$ .

2. Consider options of the form  $*2^{a_1} + *i$  where  $0 \leq i < q$ . Suppose  $i = 2^{b_1} + 2^{b_2} + \dots$  where  $b_1 > b_2 > \dots$ . Then no  $b_i$  is equal to  $a_1$  since  $i < q = 2^{a_2} + \dots$ . So  $2^{a_1} + 2^{b_1} + \dots$  is a sum of distinct powers of 2. Then

$$\begin{aligned}
 *2^{a_1} + *i &\equiv *2^{a_1} + *2^{b_1} + \dots \quad \text{by applying induction on } i \\
 &\equiv *(2^{a_1} + 2^{b_1} + 2^{b_2} + \dots) \quad \text{by applying induction on } 2^{a_1} + i \\
 &\equiv *(2^{a_1} + i)
 \end{aligned}$$

Since  $0 \leq i < q$ , the options of this type are equivalent to

$$\{ *2^{a_1}, *(2^{a_1} + 1), \dots, \underbrace{*(2^{a_1} + q - 1)}_{n-1} \}.$$

Combining the two types of options, we see that the options of  $*2^{a_1} + *q$  are equivalent to the options of  $*n$ . So  $*2^{a_1} + *q \equiv *n$ .  $\square$

## 1.4 Sprague-Grundy theorem

So far: All Nim games are equivalent to a Nim game of a single pile. Goal: Extend this to all impartial games.

### Poker nim

Being equivalent does not mean that they play the same way.

**Example:**

$$*11 + *13 \equiv *6.$$

We move to  $*11 + *11 \equiv *0$  by removing 2 chips from  $*13$ . RHS remove 6 chips.

There are other moves, say we move to  $*11 + *8 \equiv *15$ . We remove 5 chips from  $*13$ . RHS adding 9 chips.

Or, starting with  $*11 + *11 \equiv *0$ , any move on  $*11 + *11$  will increase  $*0$ .

A variation on Nim: Poker nim consists of a regular Nim game plus a bag of  $B$  chips. We now allow regular Nim moves and adding  $B' \leq B$  chips to one pile. Example:  $*3 + *4 \rightarrow *53 + *4$ .

How does this change the game of Nim?

Nothing. Say we face a losing game, so any regular Nim move would lead to a loss. In poker nim, we now add some chips to one pile. The opposing player will simply remove the chips we placed, and nothing changed.

When we say that a game is equivalent to a Nim game with one pile, it is actually a game is equivalent to a Nim game with one pile, it is actually a game of poker nim with one pile.

**Mex**

Suppose a game  $G$  has options equivalent to  $*0, *1, *2, *5, *10, *25$ . We claim that  $G$  is equivalent to  $*3$ . The options of  $*3$ , which are  $*0, *1, *2$ , are all available. If we add chips to  $*3$ , then the opposing player can remove them to get back to  $*3$ . How do we get 3?

**mex( $S$ )**

Given a set of non-negative integers  $S$ ,  $\text{mex}(S)$  is the smallest non-negative integer not in  $S$ .  
“minimum excluded integer”

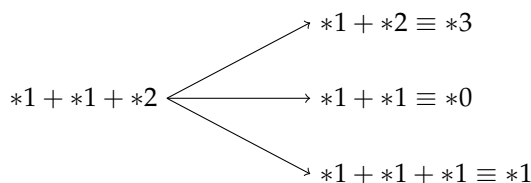
**Example:**

$$\text{mex}(\{0, 1, 2, 5, 15, 25\}) = 3.$$

The mex function is the critical link between any impartial games and Nim games.

**Theorem 1.15**

Let  $G$  be an impartial game, and let  $S$  be the set of integers  $n$  such that there exists an option of  $G$  equivalent to  $*n$ . Then  $G \equiv *(\text{mex}(S))$ .

**Example:**

By theorem,  $*1 + *1 + *2 \equiv *(\text{mex}(\{0, 1, 3\})) \equiv *2$ .

**Exercise:**

A game cannot be equivalent to one of its options.

**Proof of Theorem 1.15:**

Let  $m = \text{mex}(S)$ . It suffices to show that  $G + *m \equiv *0$ .

1. Suppose we move to  $G + *m'$  where  $m' < m$ . Since  $m = \text{mex}(S)$ , there exists an option  $G'$  of  $G$  such that  $G' \equiv *m'$ . p2 moves to  $G' + *m'$ , which is a losing game since  $G' \equiv *m'$ . So  $G + *m$  is a losing game for p1, and  $G + *m \equiv *0$ .
2. Suppose we move to  $G' + *m$ , where  $G'$  is an option of  $G$ . Then  $G' \equiv *k$  for some  $k \in S$ . So  $G' + *m \equiv *k + *m \not\equiv *0$  since  $k \neq \text{mex}(S)$ . So  $G' + *m$  is a winning game for p2. Then  $G + *m$  is a losing game for p1, so  $G + *m \equiv *0$ .

□

**Theorem 1.16: Sprague-Grundy Theorem**

Any impartial game  $G$  is equivalent to a poker nim game  $*n$  for some  $n$ .

**Proof (slightly sketchy):**

If  $G$  has no options, then  $G \equiv *0$ . Suppose  $G$  has options  $G_1, \dots, G_k$ . By induction,  $G_i \equiv *n_i$  for some  $n_i$ . By Theorem 1.15,  $G \equiv *(\text{mex}(\{n_1, \dots, n_k\}))$ . □

So any impartial game has a number.

Finding numbers is recursive: Games with no options have number 0. Move backwards and use mex to determine other numbers.

**Example: Rock game**

	1	2	3	4	5	
1	*0	*1	*2	*3	*4	
2	*1	*0	*3	*2	*5	
3	*2	*3	*0	*1	*6	
4	*3	*2	*1	*0	R	← *7

(4,5)

Winning move: move to (4,4), an options with number 0.

This is like a 2-pile Nim game.

**Example: Subtraction game (remove 1,2, or 3 chips)**

Let  $s_n$  be the number of a subtraction game with  $n$  chips. Then  $s_n = \text{mex}(\{s_{n-1}, s_{n-2}, s_{n-3}\})$  (if they exist)


$n$	0	1	2	3	4	5	6	7	8	9	10	11	12	...
$s_n$	0	1	2	3	0	1	2	3	0	1	2	3	0	...

Losing game if and only if  $n \equiv 0 \pmod{4}$ . When  $n \not\equiv 0 \pmod{4}$ , the winning move is remove just enough chips to the next multiple of 4.

**Example:**

Subtraction game with removing 2, 5, or 6 chips Then  $s_n = \text{mex}(\{s_{n-2}, s_{n-5}, s_{n-6}\})$  (if they exist)

$n$	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	...
$s_n$	0	0	1	1	0	2	1	3	0	2	1	0	0	1	1	...

  
 repeats (not proved here)

Losing game if and only if  $n \equiv 0, 1, 4, 8 \pmod{11}$ . Winning move from 9: move to 4.

**Example: Combining games**

Let  $G$  be the rook game at  $(4, 2)$ . Let  $H$  be the second subtraction with  $n = 7$ .

Then  $G \equiv *2, H \equiv *3$ , so  $G + H \equiv *2 + *3 \equiv *1$ . Winning game.

Winning move:

- From  $H$ ,  $3 \oplus 1 = 2$ . Move to  $*2$ . Remove 2 chips in the subtraction game.
- From  $G$ ,  $2 \oplus 1 = 3$ . Move to  $*3$ . Move to  $(4, 1)$  or  $(3, 2)$ .

# 2

## Strategic games

### Example: Prisoner's dilemma

Game show version: 2 players won \$10,000. They each need to make a final decision: "share" or "steal".

- If both pick "share", then they each win \$5,000.
- If one picks "steal" and the other picks "share", then the one who picks "steal" gets \$10,000, the other gets nothing.
- If both pick "steal", then they both get a consolation price with \$10.

How would players behave? The benefit a player receives is dependent on their own decision and the decisions of other players.

### strategic game

A **strategic game** is defined by specifying a set  $N = \{1, \dots, n\}$  of players, and for each player  $i \in N$ , then there is a set of possible strategies  $s_i$  to play, and a utility function:  $u_i : s_1 \times \dots \times s_n \rightarrow \mathbb{R}$ .

### Example:

With prisoner's dilemma above,  $s_1 = s_2 = \{\text{share}, \text{steal}\}$ . Samples of the utility functions:  $u_1(\text{share}, \text{share}) = 5000, u_2(\text{steal}, \text{share}) = 0$ . We can summarize the utility functions in a payoff table.

		PII	
		share	steal
PI	share	5k, 5k	0, 10k
	steal	10k, 0	10, 10

Each cell records the utilities of PI, PII in this order given the strategies played in that row (PI) and column (PII).

Assumptions about strategic games;

1. All players are rational and selfish (want to maximize their own utility).
2. All players have knowledge of all game parameters.
3. All players move simultaneously.
4. Player  $i$  plays a strategy  $s_i \in S_i$ , this forms a strategy profile  $s = (s_1, \dots, s_n) \in S_1 \times \dots \times S_n$ . Player  $i$  earns  $u_i(s)$ .



Given a strategic game, what are we looking for? One answer is we want to know how are the players expected to behave?

## Resolving prisoner's dilemma

Recall the payoff table from a previous example. What would a rational and selfish player choose to play?

1. If you know that the other player chooses to "share", then choosing "share" gives 5k, choosing "steal" gives 10k. Steal is better.
2. If you know that the other player chooses "steal", then choosing "share" gives 0, choosing "steal" gives 10. Steal is better.

In both cases, it is better to steal than to share. So we expect both players to choose "steal".

This is an example of a **strictly dominating strategy**: regardless of how other players behave, this strategy gives the best utility over all other possible strategies. If a strictly dominating strategy exists, then we expect the players to play it.

In this case, playing a strictly dominating strategy "steal" yields very little benefit. They could get more if there is some cooperation (both share). So even though we expect strictly dominating strategy is played, it might not have the best "social welfare" (the overall utility of the players).

## 2.1 Nash equilibrium

There are many games with "no" strictly dominating strategies.

**Example: Bach or Stravinsky?**

Two players want to go to a concert. Player I likes Bach, player II likes Stravinsky, but they both prefer to be with each other. Payoff table:

		PII	
		Bach	Stravinsky
PI	Bach	2, 1	0, 0
	Stravinsky	0, 0	1, 2

No strict dominating strategy exists.

What do we expect to happen? If both choose "Bach", then there is no reason for one player to switch their strategy (which gives utility 0). Similar if both choose "Stravinsky".

These are steady states, which we call **Nash equilibria**: a strategy profile where no player is incentivized to change strategy.

## Mixed strategies

There are many games with no Nash equilibria.

**Example: Rock paper scissors**

R beats S, S beats P, P beats R. Utility 1 if they win, -1 if they lose, 0 if they tie.

		PII		
		R	P	S
PI	R	0, 0	-1, 1	1, -1
	P	1, -1	0, 0	-1, 1
	S	-1, 1	1, -1	0, 0

"No" NE exist: regardless what they play, someone is incentivized to switch strategy so that they

win.

How would we expect players to play this? Randomly, probability  $\frac{1}{3}$  each. This is a **mixed strategy**. It is also a NE, there is no incentive to change to a different probability distribution.

### Nash's Theorem

Every strategic game with finite number of strategies has a Nash equilibrium (could be mixed strategies).

### Notation

Recall: Strategic game is defined by

- Players  $N = \{1, \dots, n\}$ .
- Strategy set  $S_i$  for player  $i$ .
- Utility for player  $i$ :  $u_i : s_1 \times \dots \times s_n \rightarrow \mathbb{R}$ . A strategy profile is a vector  $s = (s_1, \dots, s_n) \in S_1 \times \dots \times S_n$  which records what the players played.

Let  $S = S_1 \times S_n$  be the set of all strategy profiles. We will often compare the utilities of a player's strategies when we fix the strategies of the remaining players. Let  $S_{-i}$  be the set of all strategy profiles of all players except player  $i$  (we drop  $S_i$  from the cartesian product  $S_1 \times \dots \times S_n$ ). If  $s \in S$ , then the profile obtained from  $s$  by dropping  $s_i$  is denoted  $s_{-i} \in S_{-i}$ . If player  $i$  switches their strategy from  $s_i$  to  $s'_i$ , then the new strategy profile is denoted  $(s'_i, s_{-i}) \in S$ .

### Nash equilibrium

A strategy profile  $s^* \in S$  is a **Nash equilibrium** if  $u_i(s^*) \geq u_i(s'_i, s_{-i}^*)$  for all  $s'_i \in S_i$  and for all  $i \in N$ .

#### Example: Prisoner's dilemma

		PII	
		share	steal
PI	share	5k, 5k	0, 10k
	steal	10k, 0	10, 10

Let  $s^* = (\text{steal}, \text{steal})$ .

From PI:  $u_1(s^*) = 10$ ,  $u_1(\underbrace{\text{share}}_{s'_1}, \underbrace{\text{steal}}_{s'_{-1}}) = 0 < u_1(s^*)$ .

Similar for PII. So  $s^*$  is a NE.

#### Example: Guess 2/3 average game

3 players, a positive integer  $k$ . Each player simultaneously pick an integer from  $\{1, \dots, k\}$ , producing the strategy profile  $s = (s_1, s_2, s_3)$ . There is \$1 which is split among all players whose choices are closest to  $\frac{2}{3}$  of the 3 numbers. Other players get \$0.

If  $s = (5, 2, 4)$ , then the average is  $\frac{11}{3}$ , and  $\frac{2}{3}$  average is  $\frac{22}{9} = 2 + \frac{4}{9}$ . p2 is the closest, so  $u_2(s) = 1$ ,  $u_1(s) = u_3(s) = 0$ . Is  $s$  a NE? No. If p1 switches to 2, the  $u_1(2, s_{-1}) = u_1(2, 2, 4) = \frac{1}{2}$ . ( $\frac{2}{3}$  average is  $\frac{16}{9}$ , closer to 2 than 4).

Is there a NE? Idea: Lowering the guess generally pulls the  $\frac{2}{3}$  average closer. Try  $(1, 1, 1)$ . If a player switches to  $t \geq 2$ , then the  $\frac{2}{3}$  average is  $\frac{4+2t}{9} = \frac{4}{9} + \frac{2}{9}t$ , which is closer to 1 than  $t$ .

Prove that  $(1, 1, 1)$  is the only NE of this game.

### 2.1.1 Best response function

For a NE, a player does not want to switch. If you fix the strategies of the remaining players, then you play a strategy that maximizes utility for yourself, i.e., it is a “best response” to the fixed strategies.

#### best response function

Player  $i$ 's **best response function** for  $s_{-i} \in S_{-i}$  is given by

$$B_i(s_{-i}) = \{s'_i \in S_i : \underbrace{u_i(s'_i, s_{-i})}_{\text{utility of a best response}} \geq \underbrace{u_i(s_i, s_{-i})}_{\text{utility of all possible responses to } s_{-i}} \quad \forall s_i \in S_i\}.$$

**Example: Prisoner's dilemma**

$$B_1(\text{share}) = \{\text{steal}\}, \quad B_1(\text{steal}) = \{\text{steal}\}.$$

**Example: 2/3 average game**

$$B_1(5,5) = \{1, 2, 3, 4\} \quad u_1(x, 5, 5) = \begin{cases} 1 & x < 5 \\ 1/3 & x = 5 \\ 0 & x > 5 \end{cases} \quad \text{best response}$$

If  $s^*$  is a NE, then each player  $i$  must have played a best response to  $s_{-i}^*$ . Changing  $s_i^*$  cannot increase utility for  $i$ . Converse is also true.

#### Lemma 2.1

$s^* \in S$  is a Nash equilibrium if and only if  $s_i^* \in B_i(s_{-i}^*)$  for all  $i \in N$ .

This lemma helps us find NE by looking for strategies in the BRF.

**Example:**

		PII	
		share	steal
PI	share	5k, 5k	0, 10k <sup>o</sup>
	steal	10k*, 0	10*, 10 <sup>o</sup> → These are best responses to each other. So this is a NE

$$\begin{array}{ll} B_1(\text{share}) = \{\text{steal}\} & B_1(\text{steal}) = \{\text{steal}\} \quad * \\ B_2(\text{share}) = \{\text{steal}\} & B_2(\text{steal}) = \{\text{steal}\} \quad \circ \end{array}$$

**Example: Arbitrary game**

		PII		
		X	Y	Z
PI	A	1, 2 <sup>o</sup>	2*, 1	1*, 0
	B	2*, 1 <sup>o</sup>	0, 1 <sup>o</sup>	0, 0
	C	0, 1	0, 0	1*, 2 <sup>o</sup>

$$\begin{array}{lll} B_1(X) = \{B\} & B_1(Y) = \{A\} & B_1(Z) = \{A, C\} \quad * \\ B_2(A) = \{X\} & B_2(B) = \{X, Y\} & B_2(C) = \{Z\} \quad \circ \end{array}$$

NE are  $(B, X)$  and  $(C, Z)$ , as they are best responses to each other. The rest are not NE as one is not a best response to the other.

## 2.2 Cournot's oligopoly model

We have a set  $N = \{1, \dots, n\}$  of  $n$  firms producing a single type of goods sold on the common market. Each firm  $i$  needs to decide the number of units of goods  $q_i$  to produce. (variables)

Production cost is  $C_i(q_i)$  where  $C_i$  is a given increasing function.

Given a strategy profile  $q = (q_1, \dots, q_n)$ , a unit of the goods sell for the price of  $P(q)$ , where  $P$  is a given non-increasing function on  $\sum_i q_i$  (more goods in the market = low price)

The utility of firm  $i$  in the strategy profile  $q$  is  $u_i(q) = \underbrace{q_i P(q)}_{\text{revenue for selling } q_i \text{ units}} - \underbrace{C_i(q_i)}_{\text{production cost}}$

Szidarovszky and Yakowitz proved that a Nash equilibrium always exists under some continuity and differentiability assumptions on  $P, C$ .

### Special case: linear costs and prices

Suppose we assume  $C_i(q_i) = cq_i, \forall i \in N$  (the cost is linear, same unit cost  $c$  for all firms).  $P(q) = \max\{0, \alpha - \sum_j q_j\}$  (prices starts at  $\alpha$ , decreases 1 for each unit produced, min price 0) where  $0 < c < \alpha$ .

Utility is

$$u_i(q) = q_i P(q) - C_i(q_i) = \begin{cases} q_i(\alpha - c - \sum_j q_j) & \alpha - \sum_j q_j \geq 0 \\ -cq_i & \alpha - \sum_j q_j < 0 \end{cases}$$

When is it possible to make a profit? When  $\alpha - c - \sum_j q_j > 0$ . Separate  $q_i$  from the sum:  $\alpha - c - q_i - \sum_{j \neq i} q_j > 0$ . So  $q_i < \alpha - c - \sum_{j \neq i} q_j$ . Does not make sense for  $q_i$  if  $\text{RHS} \leq 0$ , so assume  $\text{RHS} > 0$ .

The utility is  $q_i(\alpha - c - q_i - \sum_{j \neq i} q_j)$ . Treating  $q_i$  as the variable, this utility is maximized when  $q_i = (\alpha - c - \sum_{j \neq i} q_j)/2$ . So the best response function for firm  $i$  given the production of other firms  $q_{-i}$  is

$$B_i(q_{-i}) = \begin{cases} \left\{ (\alpha - c - \sum_{j \neq i} q_j)/2 \right\} & \alpha - c - \sum_j q_j > 0 \\ \{0\} & \text{otherwise} \end{cases}$$

### Two-firm case

Suppose we simplify to 2 firms. Suppose  $q^* = (q_1^*, q_2^*)$  is a Nash equilibrium. By Lemma 2.1, a player's choice must be the best response to the other player's choice. So  $q_1^* \in B_1(q_2^*)$  and  $q_2^* \in B_2(q_1^*)$ .

Verify that we may assume  $q_1^*, q_2^* > 0$ . Then  $q_1^* = (\alpha - c - q_2^*)/2$  and  $q_2^* = (\alpha - c - q_1^*)/2$ .

Solving this gives  $q_1^* = q_2^* = (\alpha - c)/3$ . This is the amount we expect each firm to produce at equilibrium.

Price at equilibrium:  $P(q^*) = \alpha - q_1^* - q_2^* = \alpha - \frac{2}{3}(\alpha - c) = \frac{\alpha}{3} + \frac{2c}{3}$ .

Profit at equilibrium:  $u_i(q^*) = q_i^*(\alpha - c - q_1^* - q_2^*) = (\alpha - c)^2/9$ .

#### Note:

1. Suppose the two firms can collude, and together they produce  $Q$  units total. Total profit is  $Q(\alpha - c - Q)$ , which is maximized at  $Q = (\alpha - c)/2$ . The profit is  $\left(\frac{\alpha-c}{2}\right)(\alpha - c - \frac{\alpha-c}{2}) = (\alpha - c)^2/4$ . Each firm gets  $\frac{(\alpha-c)^2}{8} > \frac{(\alpha-c)^2}{9}$ .
2. In the general case with  $n$  firms, if  $q^*$  is a NE, then  $q_i^* = (\alpha - c - \sum_{j \neq i} q_j^*)/2$ . Solving this system gives  $q_j^* = \frac{\alpha-c}{n+1}$ . Price is

$$P(q^*) = \alpha - \sum_j q_j^* = \alpha - \frac{n}{n+1}(\alpha - c) = \frac{1}{n+1}\alpha + \frac{n}{n+1}c$$



**Example: Facility location game**

Two firms are each given a permit to open one store in one of 6 towns along a high way. Firm I can open in A, C or E, firm II can open in B, D or F. Assume towns are equally spaced and equally populated. Customers in a town will go to the closest store. Where to open stores?

		Firm II		
		B	D	F
Firm I	A	1, 5	2, 4	3, 3
	C	4, 2	3, 3	4, 2
	E	3, 3	2, 4	5, 1

Firm I, A is strictly dominated by C.  
Firm II, F is strictly dominated by D.  
Eliminate these two strategies.

		Firm II	
		B	D
Firm I	C	4, 2	3, 3
	E	3, 3	2, 4

Firm I, E is strictly dominated by C.  
Firm II, B is strictly dominated by D.  
Eliminate these two strategies.

		Firm II
		D
Firm I	C	3, 3

(C, D) is a NE.

Note: Extend this to 1000 towns with alternating options. The two ends are strictly dominated by the centre towns. Eliminate them to get 998 towns. Repeat. End with the two towns in the centre as NE.

**Results in IESDS****Theorem 2.4**

Suppose  $G$  is a strategic game. If IESDS ends with only one strategy profile  $s^*$ , then  $s^*$  is the unique Nash equilibrium of  $G$ .

This is a consequence of the following result.

**Theorem 2.5**

Let  $H$  be a strategic game where  $s_i$  is a strictly dominated strategy for player  $i$ . Let  $G'$  be obtained from  $G$  by removing  $s_i$  from  $S_i$ . Then  $s^*$  is a Nash equilibrium of  $G$  if and only if  $s^*$  is a Nash equilibrium of  $G'$ .

**Proof Sketch:**

Suppose  $s^*$  is a NE of  $G$ . Since  $s_i$  is strictly dominated, it cannot appear in  $s^*$  (Lemma 2.3). So  $s^*$  is a valid strategy profile in  $G'$ . If  $s^*$  is not a NE of  $G'$ , then a player can deviate to get a higher utility. However, all strategies in  $G'$  are available in  $G$ , so such a player can do it in  $G$  as well. This contradicts  $s^*$  is a NE of  $G$ .

Suppose  $s^*$  is a NE of  $G'$ . Suppose  $s^*$  is not a NE of  $G$ . Then a player can deviate to get a higher utility. This can be replicated in  $G'$  (which results in a contradiction) unless it is player  $i$  switching to strategy  $s_i$  (the only strategy in  $G$  not in  $G'$ ). Then player  $i$  could switch to the strategy that strictly dominates  $s_i$  (available in  $G'$ ) to get a higher utility in  $G'$ , contradicting  $s^*$  is a NE in  $G'$ .  $\square$

### 2.3.2 Weak dominance

#### weak dominance

For two strategies  $s_i^{(1)}, s_i^{(2)} \in S_i$  for player  $i$ , we say that  $s_i^{(1)}$  **weakly dominates**  $s_i^{(2)}$  if for all  $s_{-i} \in S_{-i}$ ,  $u_i(s_i^{(1)}, s_{-i}) \geq u_i(s_i^{(2)}, s_{-i})$ , and this inequality is strict for at least one  $s_{-i} \in S_{-i}$ .

If some strategy weakly dominates  $s_i$ , then  $s_i$  is **weakly dominated**.

If  $s_i$  weakly dominates all strategies  $s'_i \in S_i \setminus \{s_i\}$ , then  $s_i$  is a **weakly dominating strategy**.

Example:

		PII		
		X	Y	Z
PI	A	3, 3	1, 1	4, 1
	B	2, 1	0, 1	3, 1

Z is weakly dominated by X,  $u_2(A, X) > u_2(A, Z)$  and  $u_2(B, X) \geq u_2(B, Z)$ . Z is not weakly dominated by Y, no strict inequality.

### Iterated elimination of weakly dominated strategies (IEWDS)

Remove weakly dominated strategies until there is only one strategy profile.

Example:

Z and Y are weakly dominated by X above. Eliminating them gives

	X
A	3, 3
B	2, 1

A weakly dominates B.

	X
A	3, 3

$(A, X)$  is a NE.

#### Theorem 2.6

Suppose  $G$  is a strategy game. If IEWDS ends with only one strategy profile  $s^*$ , then  $s^*$  is a Nash equilibrium of  $G$ .

Note:

Compared with Theorem 2.4, here we can no longer claim that the NE is unique. A different sequence of eliminations can result in a different NE.

Exercise:

	X	Y	Z
A	1, 1	1, 0	2, 1
B	1, 1	0, 0	0, 0
C	0, 0	0, 0	1, 1

Show that two different applications of IEWDS here could end with two different profiles.

**Key difference** Unlike strictly dominated strategies, weakly dominated strategies can appear in a NE.

Some NE cannot found through IEWDS, e.g., *Bach or Stravinsky* has no weakly dominated strategies.

Just like strictly dominating strategies, weakly dominating strategies are good to play.

**Lemma 2.7**

If for all players  $i$ ,  $s_i^*$  is a weakly dominating strategy, then  $s^*$  is a Nash equilibrium.

**2.4 Auctions**

*Set up of an auction:* A seller puts one item up for an auction. Potential buyers put in bids to buy the item. Seller decides who wins (usually highest bidder) and the prices they pay.

*Typical auction:* Open bid auction. Buyers bid repeatedly until no one else bids. Highest bid wins and pays their bid price. Another type: Closed bid auctions. Each buyer submits one secret bid to the seller. (Easier to analyze).

*First price auction:* Highest bid wins, winner pays their bid. For example, 3 bidders: 150, 100, 200, pays 200. Does this simulate an open auction? No, in the open auction setting, the winner will bid slightly over 150 and win, so they pay  $\sim 150$ .

*Second price auction:* Highest bid wins, winner pays 2nd highest bid. For example, 3 bidders: 150, 100, 200, pays 150. We will analyze second price closed bid auction.

**Set up**

We have buyers  $N = \{1, \dots, n\}$ . Buyer  $i$  thinks the item has value  $v_i$  "valuation". Suppose buyer  $i$  submits the bid  $b_i$ , giving strategy profile  $b = (b_1, \dots, b_n)$ . The winner is the buyer who submits the highest bid, pays price equal to the second highest bid. If there is a tie, then the winner is the buyer with the lowest index  $i$  among all tied buyers.

Given a strategy profile  $b$ , the utility for buyer  $i$  is

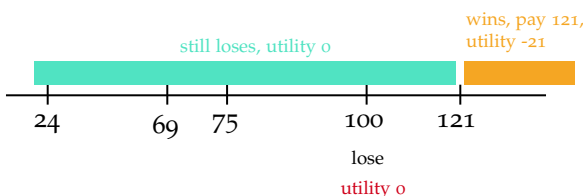
$$u_i(b) = \begin{cases} v_i - \max_{j \neq i} b_j & i \text{ wins in } b \\ 0 & \text{otherwise} \end{cases}$$

Suppose your valuation of the item is 100. Would you bid anything other than 100?

(1) Say your bid wins



(2) Say your bid loses



utility does not increase if you bid anything else

**Theorem 2.8**

In the second price auction,  $v_i$  is a weakly dominating strategy for player  $i \in N$ .

**Proof:**

We first show that  $u_i(v_i, b_{-i}) \geq u_i(b_i, b_{-i})$  for all  $b_i \in S_i$  and  $b_{-i} \in S_{-i}$ . 2 cases.

1.  $v_i$  is a winning bid in  $(v_i, b_{-i})$ . Let  $b_j$  be the second highest bid (could equal  $v_i$ ). The utility



for player  $i$  is  $u_i(v_i, b_{-i}) = v_i - b_j \geq 0$ . Suppose player  $i$  changes their bid to  $b_i$ .

If  $b_i > b_j$  or ( $b_i = b_j$  and  $i < j$ ), then  $b_i$  is still the winning bid in  $(b_j, b_{-i})$ . Payment is  $b_j$ , so utility remains the same. Otherwise,  $b_i$  is a losing bid, so the utility is 0, which is at most  $u_i(b_i, b_{-i})$ .

So  $u_i(v_i, b_{-i}) \geq u_i(b_i, b_{-i})$  for any  $b_i$ .

2.  $v_i$  is a losing bid in  $(v_i, b_{-i})$ . Let  $b_j$  be the winning bid (so  $b_j \geq v_i$ ). The utility for player  $i$  is  $u_i(v_i, b_{-i}) = 0$ . Suppose player  $i$  changes their bid to  $b_i$ .

If  $b_i < b_j$  or ( $b_i = b_j$  and  $i > j$ ), then  $b_i$  is still a losing bid in  $(b_i, b_{-i})$ . Utility is still 0. Otherwise,  $b_i$  is a winning bid, with payment  $b_j$ . The utility is  $u_i(b_i, b_{-i}) = v_i - b_j \leq 0$  (since  $b_j \geq v_i$ ). So  $u_i(v_i, b_{-i}) \geq u_i(b_i, b_{-i})$  for any  $b_i$ .

In both cases, bidding  $v_i$  gives the highest utility among all possible bids of player  $i$ .

We still need to show that for all  $b_i \neq v_i$ , there exists  $s_{-i} \in S_{-i}$  such that  $u_i(v_i, b_{-i}) > u_i(b_i, b_{-i})$ . Two cases:

1. Suppose  $b_i < v_i$ . Let  $k$  be in  $b_i < k < v_i$ . Set  $b_j = k$  for all  $j \neq i$ .

When  $v_i$  is played against  $b_{-i}$ , player  $i$  wins ( $v_i > k$ ) and pays  $k$ . Utility  $u_i(v_i, b_{-i}) = v_i - k > 0$ . When  $b_i$  is played against  $b_{-i}$ , player  $i$  loses ( $b_i < k$ ) and utility  $u_i(b_i, b_{-i}) = 0$ . So  $u_i(v_i, b_{-i}) > u_i(b_i, b_{-i})$ .

2. Suppose  $b_i > v_i$ . Let  $k$  be in  $v_i < k < b_i$ . Set  $b_j = k$  for all  $j \neq i$ .

When  $v_i$  is played against  $b_{-i}$ , player  $i$  loses ( $v_i < k$ ) and utility  $u_i(v_i, b_{-i}) = 0$ . When  $b_i$  is played against  $b_{-i}$ , player  $i$  wins ( $b_i > k$ ) and pays  $k$ . Utility  $u_i(b_i, b_{-i}) = v_i - k < 0$ . So  $u_i(v_i, b_{-i}) > u_i(b_i, b_{-i})$ .

Therefore, playing  $v_i$  is a weakly dominating strategy.  $\square$

#### Note:

The way we play this game does not depend on knowing how other players value the item. So it is easy to play: simply bid your valuation.

#### Exercise:

Suppose buyer 1 has highest valuation  $v_1$ , and buyer 2 has second highest valuation  $v_2$ , then  $(v_2, v_1, 0, 0, \dots, 0)$  is a NE.

## 2.5 Mixed strategies

### Example: Matching pennies

Two players each has a penny. They simultaneously show heads or tails. If they match, then player I gains the penny from player II. If they don't match, then player II gets the penny from player I.

		PII	
		H	T
PI	H	1, -1	-1, 1
	T	-1, 1	1, -1

There's no Nash equilibrium here (in the way NE has been described so far). Allow players to play this probabilistically. For example, PI might play H  $\frac{1}{3}$  of the time, and play T  $\frac{2}{3}$  of the time. PII might play  $\frac{3}{4}$  on H,  $\frac{1}{4}$  on T.

Is there an equilibrium here? If p1 plays  $\frac{1}{3}$ H,  $\frac{2}{3}$ T, then p2 wants to play H more often than T. Then p1 wants to play H more often than T. Then p2 wants to play T more often than H, ... etc. Seems that it is stable only if both players play  $\frac{1}{2}$ H,  $\frac{1}{2}$ T.

**mixed strategy**

A **mixed strategy** for player  $i$  is a vector  $x_i \in \mathbb{R}_+^{S_i}$  such that  $\sum_{s \in S_i} x_s^i = 1$ . The set of all mixed strategies for player  $i$  is denoted  $\Delta^i$ .

**mixed strategy profile**

A **mixed strategy profile** is a vector  $x = (x^1, \dots, x^n)$  where  $x^i \in \Delta^i$  is a mixed strategy for player  $i$ . The set of all mixed strategy profiles is denoted  $\Delta = \Delta^1 \times \dots \times \Delta^n$ . The mixed strategy profile with player  $i$  removed is  $x^{-i} \in \Delta^{-i}$ .

**Note:**

- If we play a strategy with probability 1, then it is a **pure strategy** (this is the way we play previously).
- As convention for this course, we use  $s$ 's to represent pure strategies,  $x$ 's to represent mixed strategies.

**Example:**

In matching pennies, if we order the pure strategies in the order H, T, then we had

$$x^1 = (x_H^1, x_T^1) = \left(\frac{1}{3}, \frac{2}{3}\right), x^2 = (x_H^2, x_T^2) = \left(\frac{3}{4}, \frac{1}{4}\right)$$

as mixed strategies. The strategy profile is  $x = (x^1, x^2) = \left(\left(\frac{1}{3}, \frac{2}{3}\right), \left(\frac{3}{4}, \frac{1}{4}\right)\right)$ .

Why mixed strategies?

1. Introduce unpredictability in games that are played repeatedly. Examples: In penalty kicks, you do not always kick to the same side; in politics, you do not always want to make major announcements on Tuesdays. Then the oppositions and preempt you on their announcements on Mondays.
2. Think of a player as representing a population, with probability of a strategy being proportional to the portion of the population who prefer it. Example: Say 55% like donkeys and 45% like elephants, perhaps there will be more donkeys in zoos.

**Utility**

We will use expected value as utility.

**Example:**

		PII	
		H	T
PI	H	1, -1	-1, 1
	T	-1, 1	1, -1

$$x^1 = \left(\frac{1}{3}, \frac{2}{3}\right), \quad x^2 = \left(\frac{3}{4}, \frac{1}{4}\right)$$

Two cases for  $p_1$ :

1. If  $p_1$  plays  $H$  as pure strategy, then  $\frac{3}{4}$  chance we get 1,  $\frac{1}{4}$  chance we get -1. We expect to get  $\frac{3}{4} \cdot 1 + \frac{1}{4} \cdot (-1) = \frac{1}{2}$ .
2. If  $p_1$  plays  $T$  as pure strategy, then  $\frac{3}{4}$  chance we get -1,  $\frac{1}{4}$  chance we get 1. We expect to get  $\frac{3}{4} \cdot (-1) + \frac{1}{4} \cdot 1 = -\frac{1}{2}$ .

Overall,  $p_1$  plays  $H$   $\frac{1}{3}$  of the time and  $T$   $\frac{2}{3}$  of the time. So the expected utility is  $\frac{1}{3} \cdot \left(\frac{1}{2}\right) + \frac{2}{3} \cdot \left(-\frac{1}{2}\right) = -\frac{1}{6}$ .

**expected utility of a pure strategy**

We are given a strategy profile  $x = (x^1, \dots, x^n) \in \Delta$ . The **expected utility of a pure strategy**  $s_i \in S_i$  for player  $i$  is

$$u_i(s_i, x^{-i}) = \sum_{s_{-i} \in S_{-i}} \underbrace{u_i(s_i, s_{-i})}_{\text{utility of playing } s_i} \prod_{j \neq i} \underbrace{x_{s_j}^j}_{\text{probability that the remaining players play } s_{-i}}$$

where  $u_i(s_i, x^{-i})$  is the utility from the pure strategy game.

**expected utility**

The **expected utility** of player  $i$  in  $x$  is

$$u_i(x) = \sum_{s_i \in S_i} \underbrace{x_{s_i}^i}_{\text{prob. that } p_i \text{ plays } s_i} \underbrace{u_i(s_i, x^{-i})}_{\text{utility } p_i \text{ gets for playing } s_i}$$

**Example:**

For matching pennies above,  $u_1(H, x^2) = \frac{1}{2}$ ,  $u_1(T, x^2) = -\frac{1}{2}$ ,  $u_1(x) = -\frac{1}{6}$

**Example:**

Suppose 3 players each make a choice between  $A$  and  $B$ . A \$1 prize is split among players who pick the majority choice. Suppose  $x^1 = (p, 1-p)$ ,  $x^2 = (\frac{1}{2}, \frac{1}{2})$ ,  $x^3 = (\frac{2}{5}, \frac{3}{5})$ . What is the expected utility for  $p_1$ ?

When  $p_1$  plays  $A$ , there are 4 cases:

1.  $u_1(A, A, A) = \frac{1}{3}$ . The probability that this happens is  $x_A^2 \cdot x_A^3 = (\frac{1}{2})(\frac{2}{5}) = \frac{1}{5}$ .
2.  $u_1(A, A, B) = \frac{1}{2}$ . The probability that this happens is  $x_A^2 \cdot x_B^3 = (\frac{1}{2})(\frac{3}{5}) = \frac{3}{10}$ .
3.  $u_1(A, B, A) = \frac{1}{2}$ . The probability that this happens is  $x_B^2 \cdot x_A^3 = (\frac{1}{2})(\frac{2}{5}) = \frac{1}{5}$ .
4.  $u_1(A, B, B) = 0$ . Does not matter.

Utility for playing  $A$  is  $u_1(A, x^{-1}) = (\frac{1}{5})(\frac{1}{3}) + (\frac{3}{10})(\frac{1}{2}) + (\frac{1}{5})(\frac{1}{2}) + 0 = \frac{19}{60}$

And  $u_1(B, x^{-1}) = \frac{7}{20}$ . Then expected utility for  $p_1$  is  $u_1(x) = p \cdot \frac{19}{60} + (1-p) \cdot \frac{7}{20} = \frac{7}{20} - \frac{1}{15}p$ .

It would make sense to pick  $p = 0$ , so  $p_1$  always plays  $B$ . ( $p_3$  is more likely to pick  $B$ , letting us form a majority more often.)

**2.5.1 Mixed equilibria****mixed Nash equilibrium**

A mixed strategy profile  $\bar{x} \in \Delta$  is a **mixed Nash equilibrium** if for each player  $i \in N$ ,  $u_i(\bar{x}) \geq u_i(x^i, \bar{x}^{-i})$  for all  $x^i \in \Delta^i$ .

We often omit the word “mixed”, so it is also a Nash equilibrium.

**best response function**

Given a profile  $\bar{x}^{-i} \in \Delta^{-i}$ , the **best response function** for player  $i$ ,  $B_i(\bar{x}^{-i})$ , is the set of all mixed strategies of player  $i$  that have maximum utility against  $\bar{x}^{-i}$ , i.e.,

$$B_i(\bar{x}^{-i}) = \left\{ \bar{x}^i \in \Delta^i : u_i(\bar{x}^i, \bar{x}^{-i}) \geq u_i(x_i, \bar{x}^{-i}) \quad \forall x_i \in \Delta^i \right\}$$

**Proposition 2.9**

$\bar{x} = (\bar{x}^1, \dots, \bar{x}^n) \in \Delta$  is a Nash equilibrium if and only if  $\bar{x}^i \in B_i(\bar{x}^{-i})$  for all  $i \in N$ .

**Example: Matching pennies**

		PII	
		H	T
PI	H	1, -1	-1, 1
	T	-1, 1	1, -1

Suppose  $x^1 = (p, 1-p)$  and  $x^2 = (q, 1-q)$ .

For p1, the expected utility for playing  $H$  is  $q \cdot 1 + (1-q) \cdot (-1) = 2q - 1$ . The expected utility for playing  $T$  is  $q \cdot (-1) + (1-q) \cdot 1 = 1 - 2q$ . Utility for p1 is  $p(2q - 1) + (1-p)(1 - 2q) = p(-2 + 4q) + (1 - 2q)$ .

Given  $q$ , which  $p$  maximizes this utility?  $1 - 2q$  is constant, so we maximize  $p(-2 + 4q)$ . 3 cases:

1. If  $q < \frac{1}{2}$ , then  $-2 + 4q < 0$ . So we maximize with  $p = 0$ .
2. If  $q = \frac{1}{2}$ , then  $-2 + 4q = 0$ . Then any  $p$  maximizes it, so  $p \in [0, 1]$ .
3. If  $q > \frac{1}{2}$ , then  $-2 + 4q > 0$ . Maximize with  $p = 1$ .

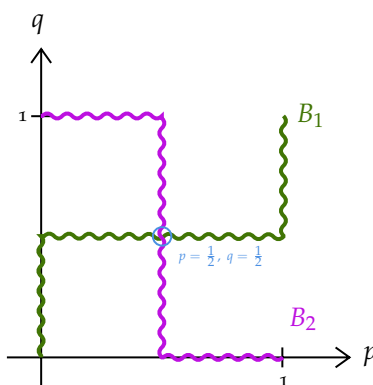
BRF for p1:

$$B_1(x^2) = \begin{cases} \{(0, 1)\} & q < \frac{1}{2} \\ \{(p, 1-p) : p \in [0, 1]\} & q = \frac{1}{2} \\ \{(1, 0)\} & q > \frac{1}{2} \end{cases}$$

Similarly, for p2, the utility is  $q(2 - 4p) + (2p - 1)$ . Divide cases with  $p = \frac{1}{2}$ . Then

$$B_2(x^1) = \begin{cases} \{(1, 0)\} & p < \frac{1}{2} \\ \{(q, 1-q) : q \in [0, 1]\} & p = \frac{1}{2} \\ \{(0, 1)\} & p > \frac{1}{2} \end{cases}$$

We look for  $p, q$  such that  $x^1, x^2$  are best responses to each other. Draw  $B_1, B_2$  on a “graph”.



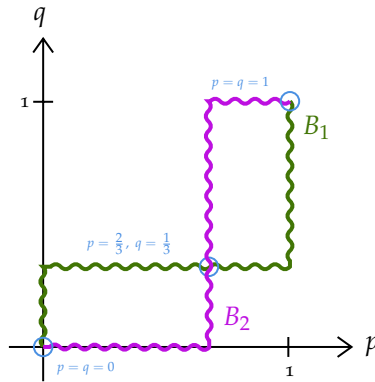
The intersection is where they are best responses simultaneously, hence a Nash equilibrium.  $x^1 = (1/2, 1/2)$ ,  $x^2 = (1/2, 1/2)$  and  $(x^1, x^2)$  is a NE.

Example: Bach or Stravinsky

		PII	
		B	S
PI	B	2, 1	0, 0
	S	0, 0	1, 2

Suppose  $x^1 = (p, 1-p)$ ,  $x^2 = (q, 1-q)$ . We have

$$B_1(x^2) = \begin{cases} \{(0, 1)\} & q < \frac{1}{3} \\ \{(p, 1-p) : p \in [0, 1]\} & q = \frac{1}{3} \\ \{(1, 0)\} & q > \frac{1}{3} \end{cases} \quad B_2(x^1) = \begin{cases} \{(0, 1)\} & p < \frac{2}{3} \\ \{(q, 1-q) : q \in [0, 1]\} & p = \frac{2}{3} \\ \{(1, 0)\} & p > \frac{2}{3} \end{cases}$$



3 NE: 2 pure strategies  $((0, 1), (0, 1))$  and  $((1, 0), (1, 0))$ . 1 mixed strategy  $((\frac{2}{3}, \frac{1}{3}), (\frac{1}{3}, \frac{2}{3}))$

## 2.5.2 Support characterization

Suppose  $\bar{x}^{-i}$  is fixed. Which  $x^i \in \Delta^i$  maximizes  $u_i(x^i, \bar{x}^{-i})$ ? Write a LP:

$$\begin{aligned} \max \quad & \sum_{s \in S_i} x_s^i u_i(s, \bar{x}^{-i}) \\ \text{s.t.} \quad & \sum_{s \in S_i} x_s^i = 1 \\ & x^i \geq 0 \end{aligned} \tag{P}$$

Variables:  $x_s^i$  for each  $s \in S_i$ . What is the dual? One dual variable  $y$ .

$$\begin{aligned} \min \quad & y \\ \text{s.t.} \quad & y \geq u_i(s, \bar{x}^{-i}) \quad \text{for all } s \in S_i \end{aligned} \tag{D}$$

(P) is feasible (set  $x^i$  to be any probability distribution). (D) is feasible (set  $y$  to be max value of  $u_i(s, \bar{x}^{-i})$ ). Therefore, (P) and (D) both have optimal solutions, and their optimal values are equal.

(D) is easy to solve:  $y = \max_{s \in S_i} u_i(s, \bar{x}^{-i})$ , maximum utility when pure strategies are played against  $\bar{x}^{-i}$ . (P) also has optimal value  $y$ . So the maximum utility of all mixed strategies is equal to the max utility of pure strategies.

Complementary slackness conditions:  $x_s^i = 0$  or  $y = u_i(s, \bar{x}^{-i})$  for all  $s \in S_i$ . Equivalently,  $x_s^i > 0$  implies  $y = u_i(s, \bar{x}^{-i})$ . Translation: only pure strategies with maximum utility could have positive probabilities in a best response.

**Theorem 2.10: Support characterization**

Given  $\bar{x}^{-i} \in \Delta^{-i}$ , a mixed strategy  $x^i \in B_i(\bar{x}^{-i})$  if and only if  $x_s^i > 0$  implies  $s \in S_i$  is a pure strategy of maximum utility against  $\bar{x}^{-i}$ .

**support**

For a mixed strategy  $x^i \in \Delta^i$ , the **support** is the set of strategies with positive probability in  $x^i$ .

Rephrasing of Theorem 2.10:  $x^i$  is in the BRF if and only if the support of  $x^i$  are strategies with maximum utility.

Example: Bach or Stravinsky

		PII	
		B	S
PI	B	2, 1	0, 0
	S	0, 0	1, 2

Suppose p2 plays  $x^2 = (q, 1 - q)$ . The utilities of p1 using pure strategies are:  $u_1(B, x^2) = 2q$ ,  $u_1(S, x^2) = 1 - q$ . Depending on  $q$ , the strategies with maximum utility are different.

1. If  $2q < 1 - q$ , then  $q < \frac{1}{3}$ , and  $B$  is not in the support and gets probability 0. BRF  $\{(0, 1)\}$ .
2. If  $2q = 1 - q$ , then  $q = \frac{1}{3}$ , and both  $B, S$  could be in the support. Any combination works, so BRF  $\{(p, 1 - p) : p \in [0, 1]\}$ .
3. If  $2q > 1 - q$ , then  $q > \frac{1}{3}$ , and  $S$  is not in the support. BRF  $\{(1, 0)\}$ .

This matches the BRF we calculated previously.

Example:

Consider a 2-player game with this payoff table. Suppose p2 plays  $x^2 = (0, \frac{1}{3}, \frac{2}{3})$ . What is  $B_1(x^2)$ ?

	D	E	F
A	2, 2	3, 3	1, 1
B	3, 1	0, 4	2, 1
C	3, 4	5, 1	0, 7

$$u_1(A, x^2) = 0 + \frac{1}{3} \cdot 3 + \frac{2}{3} \cdot 1 = \frac{5}{3}$$

$$u_1(B, x^2) = 0 + 0 + \frac{2}{3} \cdot 2 = \frac{4}{3}$$

$$u_1(C, x^2) = 0 + \frac{1}{3} \cdot 5 + 0 = \frac{5}{3}$$

By support characterization,  $x_B^1 = 0$ . Any distribution over  $x_A^1$  and  $x_C^1$  works.

So  $B_1(x^2) = \{(p, 0, 1 - p) : p \in [0, 1]\}$ .

The maximum utility for p1 is  $p \cdot \frac{5}{3} + (1 - p) \cdot \frac{5}{3} = \frac{5}{3}$ , which is equal to the max utility for a pure strategy.

Any strategy in  $B_1(x^2)$  maximizes utility for p1. Which of these maximizes utility for p2? This will give a NE.

Suppose  $x^1 = (p, 0, 1 - p)$ . Calculate the utilities for p2:  $u_2(D, x^1) = 4 - 2p$ ,  $u_2(E, x^1) = 1 + 2p$ ,  $u_2(F, x^1) = 7 - 6p$ . If  $x^2 = (0, \frac{1}{3}, \frac{2}{3})$  is in the best response, then  $E, F$  must have maximum utility.  $1 + 2p = 7 - 6p$ , so  $p = \frac{3}{4}$ . Utility for  $E, F$  is  $\frac{5}{2}$ . Utility for  $D$  is also  $\frac{5}{2}$ , so indeed  $E, F$  have max utility. (So does  $D$ , but this is fine.)

So  $x^1 = (\frac{3}{4}, 0, \frac{1}{4})$  and  $x^2 = (0, \frac{1}{3}, \frac{2}{3})$  are in the best responses for each other, and  $(x^1, x^2)$  is a NE.

**Note:**

One “algorithm” for finding NE is by looking at possible combinations of the supports for each player. In example above, if we ask “suppose support for p1 is  $\{A, C\}$  and support for p2 is  $\{E, F\}$ ” then we can use support characterization to find a NE or prove that none exist for these supports.

Problem: There are exponentially many support sets each player ( $\sim 2^k$  if there are  $k$  pure strategies). Not practical.

**Exercise:**

Show that in the game of rock paper scissors, both players playing  $(\frac{1}{3}, \frac{1}{3}, \frac{1}{3})$  is the only Nash equilibrium.

## 2.6 Voting game

**Downs paradox** Voting has costs. The probability that one vote is a decisive vote is very small. Costs outweigh benefits.

Expectation: People don’t vote. Reality: people do vote.

### Model for voter participation

Suppose there are two candidates  $A, B$ , and the number of supporters are  $a, b$ , respectively.

WLOG, assume  $a \geq b$ . Each person can choose to “vote” or “abstain”. If they vote, then they incur a cost of  $c$  where  $0 < c < 1$ . Regardless voting or abstaining, each person gets a payoff of 2 if their supporting candidate wins, 1 for a tie, 0 for a loss.

### Pure NE

Suppose  $a = b = 1$ .

		PII (B)	
		A	V
PI (A)	A	1, 1	0, 2-c
	V	2-c, 0	1-c, 1-c

It’s like prisoner’s dilemma: both players vote, get lower utility than both players abstain.

Now suppose  $a = b \geq 2$ . 4 cases:

1. Everyone votes. There is a tie, everyone has utility  $1 - c$ , switching gives 0. NE
2. Not everyone votes, and there is a tie. One who abstains can vote,  $1 \rightarrow 2 - c > 1$ . Not NE
3. One candidate wins by 1 vote. One who abstains for the losing candidate can vote,  $0 \rightarrow 1 - c > 0$ . Not NE
4. One candidate wins by at least 2 votes. One who votes for the winning candidate can abstain,  $2 - c \rightarrow 2$ . Not NE

In a close election, we expect more people to vote.

**Exercise:**

Show that when  $a > b$ , there is no pure Nash equilibrium.

### Mixed NE

Then we consider mixed Nash equilibrium: one possible scenario for a mixed NE.

Suppose  $a > b$ . Among all  $A$  supporters,  $b$  of them will vote and  $a - b$  of them will abstain. Suppose

every  $B$  supporter will vote with the same probability  $p$ . So the best that  $B$  can do is a tie. It is easy to check that  $p = 0$  or  $p = 1$  is not a NE. Assume  $p \in (0, 1)$ .

Consider a  $B$  supporter. If they abstain, then  $B$  cannot win. So utility of “abstain” as pure strategy is 0. If they vote, then  $B$  ties only if all other  $B$  supporters vote (utility  $1 - c$ ), otherwise  $B$  loses (utility  $-c$ ). Expected utility of “vote” as pure strategy is

$$\underbrace{p^{b-1}}_{b-1 \text{ vote}} \underbrace{(1-c)}_{\text{utility of a tie}} + \underbrace{(1-p^{b-1})}_{\text{not all } b-1 \text{ vote}} \underbrace{(-c)}_{\text{utility of a loss}} = p^{b-1} - c$$

When is it possible that this is in a NE?  $p \in (0, 1)$ , so both strategies have positive probabilities. To be in the best response, support characterization implies the two utilities are equal. So  $0 = p^{b-1} - c$ , or  $p = c^{\frac{1}{b-1}}$ .

Given this  $p$ , are  $A$  supporters incentivized to change their mixed strategies? Currently, all of them are playing pure strategies. In order to switch, the utility of switching to the other pure strategy must be greater.

1. Consider an  $A$  who abstained. Expected utility is  $\underbrace{p^b}_{b \text{ vote}} \cdot \underbrace{1}_{\text{utility of a tie}} + \underbrace{(1-p^b)}_{< b \text{ vote}} \cdot \underbrace{2}_{\text{utility of a win}} = 2 - p^b$

Expected utility of voting is  $2 - c$  ( $A$  guaranteed to win).  $2 - c = 2 - p^{b-1} \leq 2 - p^b$  ( $0 < p < 1$ )

Switching to a pure strategy does not increase utility. So switching to any mixed strategy does not increase utility. No reason to switch.

2. Consider an  $A$  supporter who voted. Expected utility is  $\underbrace{p^b(1-c)}_{\text{tie}} + \underbrace{(1-p^b)(2-c)}_{\text{win}} = 2 - p^b - c$

If they abstain...

- $A$  loses if all  $B$  supporters vote;
- $A$  ties if  $b - 1$   $B$  supporters vote, 1 abstain;
- $A$  wins otherwise.

Utility of abstaining is

$$\underbrace{p^b \cdot 0}_{\text{choices of who abstains}} + \underbrace{b \cdot p^{b-1}}_{b-1 \text{ votes}} \cdot \underbrace{(1-p)}_{1 \text{ abstain}} \cdot 1 + \underbrace{(1-p^b - b \cdot p^{b-1} \cdot (1-p))}_{\text{remaining probability}} \cdot 2 = 2 - 2p^b - bp^{b-1}(1-p)$$

and we know:  $2 - p^b - c \geq 2 - 2p^b - bp^{b-1}(1-p)$ . No reason to switch.

When  $p = c^{\frac{1}{b-1}}$ , this is a mixed NE.

**Q** What happens to voter participation as cost increase?

If  $c$  increase, then  $p$  increase, so more voters will vote.

## 2.7 Two-player zero-sum game

### zero-sum

A strategic game is a **zero-sum** game if for all strategy profiles  $s \in S$ ,  $\sum_{i \in N} u_i(s) = 0$ .

Examples: Matching pennies and rock paper scissors.



For a two-player zero-sum game, let  $s_1 = \{1, \dots, m\}$  and  $s_2 = \{1, \dots, n\}$ . Define such a game with a payoff matrix  $A \in \mathbb{R}^{m \times n}$  where  $u_1(i, j) = A_{ij}$  and  $u_2(i, j) = -A_{ij}$ .

Example:

$$\begin{array}{c} \text{PI} \end{array} \begin{array}{c} \begin{array}{c} \text{PII} \\ \begin{array}{ccc} 1 & 2 & 3 \end{array} \end{array} \begin{array}{c} \begin{array}{ccc} 3 & 5 & -2 \\ -5 & 7 & 1 \end{array} \end{array} = A$$

payoff for PI

$$\begin{array}{c} \text{PI} \end{array} \begin{array}{c} \begin{array}{c} \text{PII} \\ \begin{array}{ccc} 1 & 2 & 3 \end{array} \end{array} \begin{array}{c} \begin{array}{ccc} -3 & -5 & 2 \\ 5 & -7 & -1 \end{array} \end{array} = -A$$

payoff for PII

Note:

For a mixed strategy profile  $x = (x^1, x^2)$ ,  $u_1(x^1, x^2) = -u_2(x^1, x^2)$ .

We use min-max argument for finding a NE: Given a strategy that we play, the opposing player will maximize their utility, which maximizes our utility. Knowing how they would play, what can we do to maximize our own utility?

Player I's perspective: Suppose player I plays  $x^1$ . They expect player II to play from their best response.

PII's expected utility for playing pure strategy  $j$  is  $-(x^1)^T A_{\cdot j}$  ( $A_{\cdot j}$  is the  $j$ -th column of  $A$ )

Utility of PII's best response is equal to the maximum of these values,

$$\max_{j \in \{1, \dots, n\}} -(x^1)^T A_{\cdot j} = - \min_{j \in \{1, \dots, n\}} (x^1)^T A_{\cdot j}$$

So utility for PI is  $\min_{j \in \{1, \dots, n\}} (x^1)^T A_{\cdot j}$

PI wants to maximize this:

$$\begin{aligned} \max \quad & \min_{j \in \{1, \dots, n\}} (x^1)^T A_{\cdot j} \\ \text{s.t.} \quad & \sum_{i=1}^m x_i^1 = 1 \\ & x^1 \geq 0 \end{aligned}$$

which is not an LP. So we turn it into

$$\begin{aligned} \max \quad & u_1 \\ \text{s.t.} \quad & u_1 \leq (x^1)^T A_{\cdot j} \quad \forall j \in \{1, \dots, n\} \\ & \sum_{i=1}^m x_i^1 = 1 \\ & x^1 \geq 0 \end{aligned}$$

Example:

Expected utilities for PII's 3 strategies are

$$\begin{aligned} u_2(1, x^1) &= -3x_1^1 + 5x_2^1, \\ u_2(2, x^1) &= -5x_1^1 - 7x_2^1, \\ u_2(3, x^1) &= 2x_1^1 - x_2^1 \end{aligned}$$

Look for

$$\begin{aligned} & \max \{-3x_1^1 + 5x_2^1, -5x_1^1 - 7x_2^1, 2x_1^1 - x_2^1\} \\ & = \min \{3x_1^1 - 5x_2^1, 5x_1^1 + 7x_2^1, -2x_1^1 + x_2^1\} \end{aligned}$$

$$\begin{aligned} \max \quad & u_1 \\ \text{s.t.} \quad & u_1 \leq 3x_1^1 - 5x_2^1 \\ & u_1 \leq 5x_1^1 + 7x_2^1 \\ & u_1 \leq -2x_1^1 + x_2^1 \\ & x_1^1 + x_2^1 = 1 \\ & x^1 \geq 0 \end{aligned}$$

Player II's perspective: Suppose PII plays  $x^2$ . Then PI will play from their best response.

Utility of PI's best response is  $\max_{i \in \{1, \dots, m\}} -(x^2)^T A_i$ ,  
where  $A_i$  is the  $i$ -th row of  $A$ .

PII's utility is  $-\max_{i \in \{1, \dots, m\}} (x^2)^T A_i$ .

Maximizing this is equivalent to minimizing  
 $\max_{i \in \{1, \dots, m\}} (x^2)^T A_i$ .

PI wants to maximize this:

$$\begin{aligned} \min \quad & \max_{i \in \{1, \dots, m\}} (x^2)^T A_i \\ \text{s.t.} \quad & \sum_{j=1}^n x_j^2 = 1 \\ & x^2 \geq 0 \end{aligned}$$

which is not an LP. So we turn it into

$$\begin{aligned} \min \quad & u_2 \\ \text{s.t.} \quad & (x^2)^T A_i \leq u_2 \quad \forall i \in \{1, \dots, m\} \\ & \sum_{j=1}^n x_j^2 = 1 \\ & x^2 \geq 0 \end{aligned}$$

**Example:**

PI's best response has utility

$$\max\{3x_1^2 + 5x_2^2 - 2x_3^2, -5x_1^2 + 7x_2^2 + x_3^2\}$$

Thus

$$\begin{aligned} \min \quad & u_2 \\ \text{s.t.} \quad & 3x_1^2 + 5x_2^2 - 2x_3^2 \leq u_2 \\ & -5x_1^2 + 7x_2^2 + x_3^2 \leq u_2 \\ & x_1^2 + x_2^2 + x_3^2 = 1 \\ & x^2 \geq 0 \end{aligned}$$

**Exercise:**

The LPs for player I and player II are duals of each other.

Both LPs are feasible (take  $x^1, x^2$  to be any probability distribution,  $u_1, u_2$  as max/min values).

So both have optimal solutions with the same objective value. (Note: obj value of PI's LP is the utility of PI, so the obj value of PII's LP is the negative of the utility of PII.) The optimal solutions are best responses to each other, so they form a NE. Solve this using simplex (a modified version of simplex is provably polynomial time).

### Theorem 2.11

Assume finite pure strategies, any two-player zero-sum game has a mixed Nash equilibrium, and this can be efficiently computed.

**Example:**

For our 2 LPs above, an optimal solution is

$$\begin{aligned} \text{PI:} \quad & x_1^1 = \frac{6}{11}, x_2^1 = \frac{5}{11}, u_1 = -\frac{7}{11} \quad (u_1 \text{ is the utility of PI}) \\ \text{PII:} \quad & x_1^2 = \frac{3}{11}, x_2^2 = 0, x_3^2 = \frac{8}{11}, u_2 = -\frac{7}{11} \quad (-u_2 \text{ is the utility of PII}) \end{aligned}$$

**Note:**

Computing NE in general is difficult. Even in the 3-player zero-sum game or 2-player general-sum game, no polynomial time algorithm is known.

## 2.8 Nash's theorem

### Theorem 2.12: Nash

Every strategies game with finitely many players and pure strategies has a Nash equilibrium.

### 2.8.1 Brouwer's fixed point theorem

#### Brouwer

Let  $X$  be a convex and compact set in a finite-dimensional Euclidean space, and let  $f : X \rightarrow X$  be a continuous function. Then there exists  $x_0 \in S$  such that  $f(x_0) = x_0$  ("fixed point")

#### Example:

Let  $X = [0, 1]$ . Consider any continuous function  $f : [0, 1] \rightarrow [0, 1]$ .



The graph of  $f$  will always intersect  $f(x) = x$ , producing a fixed point. This is a consequence of the intermediate value theorem (apply to  $f(x) - x$ )

Terminology from the theorem:

- We will think of an Euclidean space as  $\mathbb{R}^n$  with the standard dot product, which defines how we measure distance and angle.
- A set is convex if for any two points in the set, the line segment joining them is also in the set.

Precise definition:  $S$  is convex if for all  $u, v \in S$ ,  $\lambda u + (1 - \lambda)v \in S$  for all  $\lambda \in [0, 1]$ .

Note: The convex combination of any set of points is convex.

$$S = \{\lambda_1 v_1 + \dots + \lambda_n v_n : \lambda_1, \dots, \lambda_n \geq 0, \lambda_1 + \dots + \lambda_n = 1\}$$

- A set is compact if it is closed and bounded<sup>1</sup>.

#### Note:

This is a deep theorem from analysis. We will not prove it here, though there are many fascinating proofs of it (suggestion: look into the combinatorial proof using Sperner's Lemma). None of the proofs are constructive: we know that a fixed point exists, but the proofs do not tell us how to find one.

#### Illustrations

1. Print a world map and place it on your desk. This is a continuous mapping from the surface of Earth to the part of the surface occupied by the map on your desk. The theorem implies there is a fixed point: some point on the map is directly on top of the point it represents on your desk.
2. Take a cup of tea and stir it. Let it settle. Then some part of the liquid is in the same spot before the stir.

<sup>1</sup>This is not true in general, but it works for a subset of  $\mathbb{R}^n$  by The Heine-Borel Theorem. See more in lec 13 and 20 in <https://notes.sibeliusp.com/pdfs/1201/amath331.pdf>

## Relation to strategic games

We want to use Brouwer's fixed point theorem when  $X$  is the set of all mixed strategy profiles of a finite strategic game. Need to verify that  $\Delta$  is convex and compact.

Start with just one player  $i$  and their set of mixed strategies  $\Delta^i$ . If the set of pure strategies is  $\{1, \dots, k\}$ , then  $\Delta^i = \{(x_1^i, \dots, x_k^i) : x_j^i \geq 0, x_1^i + \dots + x_k^i = 1\}$

$$k = 2 : \Delta^i = \{(p, 1-p) : p \in [0, 1]\}$$

$$k = 3 : \Delta^i = \{(p, q, r) : p + q + r = 1, 0 \leq p, q, r \leq 1\}$$

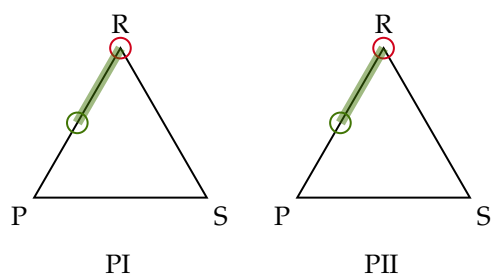


In the case of  $k = 3$ , it is a triangle, that's why we call it  $\Delta$ . We can see (without proof) that  $\Delta^i$  is compact: it is closed and any 2 points have distance at most 1.  $\Delta^i$  is convex: it is the convex combination of the standard basis vectors  $e_1, \dots, e_k$ . (An element of  $\Delta^i$  has the form  $x_1^i e_1 + \dots + x_k^i e_k$  where  $x_1^i + \dots + x_k^i = 1, x_j^i \geq 0$ .) These  $e_1, \dots, e_k$  are the pure strategies of player  $i$ .

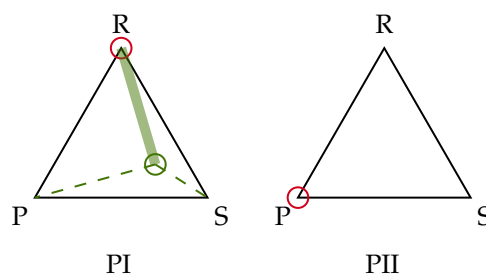
The set of all strategy profiles is  $\Delta = \Delta^1 \times \dots \times \Delta^n$ . We can "pretend" that this is a set in  $\mathbb{R}^{|S_1| + \dots + |S_n|}$ . It is still compact (a result, Tychonoff's Theorem, from analysis is that the cartesian product of compact sets is compact). It is also convex. So we can use  $\Delta$  as the set in Brouwer's fixed point theorem. Now we need to find a continuous function  $f : \Delta \rightarrow \Delta$  that relates fixed points to mixed Nash equilibria.

Given a strategy profile  $x = (x^1, \dots, x^n)$ , a player  $i$  will look at possibly switching to a pure strategy to gain utility against  $x^{-i}$ . If pure strategy  $s$  improves utility, then player  $i$  wants to shift the probability distribution so that  $s$  receives higher probability. The function will take  $x$ , and map it to another strategy profile where each player improves their utility.

### Example: Rock paper scissors



Suppose both play rock as a pure strategy. They can increase utility by moving toward paper.



Suppose PI plays rock, PII plays paper. PII cannot improve utility by moving to paper or scissors. PI will move more towards scissors than paper.

What is the meaning of a fixed point? No player can improve their utility. So it must be a Nash equilibrium.

## Defining the function

First define  $\Phi$  which records the improvement of a player in switching to a pure strategy. Given strategy profile  $x \in \Delta$ , a player  $i$ , and a pure strategy  $s \in S_i$ , define  $\Phi_s^i(x) = \max\{0, u_i(s, x^{-i}) - u_i(x)\}$ . If playing  $s$  increases utility for player  $i$ , then  $\Phi_s^i(x)$  represents this increase. Otherwise  $\Phi_s^i(x) = 0$ .

For player  $i$  and strategies  $s$  where  $\Phi_s^i(x) > 0$ , we want to increase probability on  $s$ . We want to replace  $x_s^i$  by  $x_s^i + \Phi_s^i(x)$ . But the sum of probabilities is greater than 1. We can normalize this by dividing by  $\sum_{s' \in S_i} (x_{s'}^i + \Phi_{s'}^i(x)) = 1 + \sum_{s' \in S_i} \Phi_{s'}^i(x)$

We define  $f : \Delta \rightarrow \Delta$  by  $f(x) = \bar{x}$  where for each player  $i$  and strategy  $s \in S_i$ ,  $\bar{x}_s^i = \frac{x_s^i + \Phi_s^i(x)}{1 + \sum_{s' \in S_i} \Phi_{s'}^i(x)}$

We can verify that  $f(x) \in \Delta$ .

### Example:

In rock paper scissors where PI plays rock and PII plays paper, the strategy profile is  $x = ((1, 0, 0), (0, 1, 0))$ . For PII,  $\Phi_s^2(x) = 0$  for each  $s \in \{R, P, S\}$ . For PI,  $\Phi_R^1(x) = 0, \Phi_P^1(x) = 1, \Phi_S^1(x) = 2$ . So the new strategy for PI is

$$\bar{x}_R^1 = \frac{1+0}{1+3} = \frac{1}{4}, \quad \bar{x}_P^1 = \frac{0+1}{1+3} = \frac{1}{4}, \quad \bar{x}_S^1 = \frac{0+2}{1+3} = \frac{1}{2}.$$

Thus,  $f(x) = ((\frac{1}{4}, \frac{1}{4}, \frac{1}{2}), (0, 1, 0))$ .

## 2.8.2 Completing the proof of Nash's theorem

Given  $x \in \Delta$ , consider  $\Phi$  and  $f : \Delta \rightarrow \Delta$  defined above. We see that  $f$  is continuous since  $\Phi$  is continuous. By Brouwer's fixed point theorem, there exists  $\hat{x} \in \Delta$  such that  $f(\hat{x}) = \hat{x}$ . We prove that  $\hat{x}$  is a NE by showing  $\hat{x}^i \in B_i(\hat{x}^{-i})$ .

For player  $i$ , let  $s \in S_i$  be a pure strategy such that  $\hat{x}_s^i > 0$  and  $u_i(s, \hat{x}^{-i}) \leq u_i(\hat{x})$ . (Exercise: show such  $s$  exists.) Then  $\Phi_s^i(\hat{x}) = 0$ . Since  $\hat{x}$  is a fixed point,  $\hat{x}_s^i = (f(\hat{x}))_s^i = \hat{x}_s^i / (1 + \sum_{s' \in S_i} \Phi_{s'}^i(\hat{x}))$ . Since  $\hat{x}_s^i > 0$ , the denominator must be 1. So  $\sum_{s' \in S_i} \Phi_{s'}^i(\hat{x}) = 0$ . But  $\Phi$  is non-negative, so  $\Phi_{s'}^i(\hat{x}) = 0$  for all  $s' \in S_i$ . This means that  $u_i(s', \hat{x}^{-i}) \leq u_i(\hat{x})$  for all  $s' \in S_i$ . So playing  $\hat{x}^i$  gives the highest utility against  $\hat{x}^{-i}$ , so  $\hat{x}^i \in B_i(\hat{x}^{-i})$ . Since this holds for all players,  $\hat{x}$  is a Nash equilibrium.  $\square$

### Note:

This proves that a NE always exists, but the proof does not show us how to find such a NE, as it depends on Brouwer's fixed point theorem.

## Cooperative games

### 3.1 Introduction

There are games where group of players can work together to obtain higher utility.

#### Example: Ice cream

Alice, Bob, Carol want to buy ice cream. Three sizes: 1L, 1.5L, 2L with costs \$6, \$9, \$11 respectively. A has \$3, B has \$4, C has \$5. On their own, they cannot buy any. But if they pool money together, they can get some ice cream. (e.g. B + C can buy 1.5L)

#### cooperative game

A **cooperative game** is given by a set of players  $N$  and a characteristic function  $v : 2^N \rightarrow \mathbb{R}$  that assigns a value  $v(S)$  to each coalition  $S \subseteq N$  of players. We use  $(N, V)$  to represent this game. The set  $N$  is the **grand coalition**.

#### Example:

In the ice cream game,  $N = \{A, B, C\}$ , and  $v$  is defined by

$S$	$\emptyset, \{A\}, \{B\}, \{C\}$	$\{A, B\}$	$\{A, C\}$	$\{B, C\}$	$\{A, B, C\}$
$v(S)$	0	1	1	1.5	2

General assumptions:  $v(\emptyset) = 0$ ,  $v(S) \geq 0$  for all  $S \subseteq N$ .

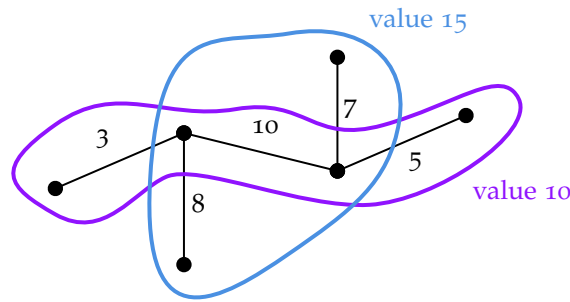
#### Example: 101-member parliament

A country has a 101-member parliament. There are 4 parties  $A, B, C, D$  with 40, 22, 30, 9 members, respectively. They need to decide how to spend a \$1 billion windfall, they need to form a majority to spend it. Thus  $N = \{A, B, C, D\}$  and

$$v(S) = \begin{cases} 10^9 & \text{parties in } S \text{ have } \geq 51 \text{ members} \\ 0 & \text{otherwise} \end{cases}$$

#### Example: matching game

In a matching game, we are given a graph  $G = (V, E)$  and edge weights  $\omega : E \rightarrow \mathbb{R}$ . The players are the vertices,  $V = N$ . The weight of an edge represents the benefits if two vertex players work together. For any subset  $S \subseteq N$ , the value is the maximum weight of a matching using vertices in  $S$ .



## Outcomes of cooperative games

Outcomes of Strategic games: Strategy profiles (pure or mixed). Which strategy is played by each player?

Outcomes of Cooperative games:

1. Divide the players into groups, we call them coalitions. “coalition structure”  
Each coalition will generate their assigned value.
2. Distribute the value that each coalition generates among its members. “payoff vector”

### coalition structure

Given a cooperative  $(N, v)$ , a **coalition structure** is a partition  $\pi$  of  $N$ , i.e.,  $\pi = (C^1, \dots, C^k)$  where each  $C^i \subseteq N$ ,  $C^i \cap C^j = \emptyset$  whenever  $i \neq j$ , and  $C^1 \cup \dots \cup C^k = N$ .

### payoff vector

A **payoff vector** is a vector  $x \in \mathbb{R}^n$  such that  $x \geq \mathbf{0}$  and

$$\sum_{i \in C^j} x_i \leq v(C^j)$$

for all  $j = 1, \dots, k$ .

Notation: For any  $T \subseteq N$ ,  $x(T) = \sum_{i \in T} x_i$ . So the inequality here is  $x(C^j) \leq v(C^j)$ .

### efficient outcome

An **outcome** consists of  $(\pi, x)$ . Such an outcome is **efficient** if  $x(C^j) = v(C^j)$  for all  $j$ .

### Example:

An outcome of the ice cream game is  $(\pi, x)$  where  $\pi = (\{A, B\}, \{C\})$ , and  $x_A = x_B = 0.5$ ,  $x_C = 0$ . This outcome is efficient:  $v(\{A, B\}) = 1 = x_A + x_B$ ,  $v(\{C\}) = 0 = x_C$ .

## Some classes of games

1. **Monotone games**:  $S \subseteq T \implies v(S) \leq v(T)$ . “more players produce more value”
2. **Superadditive games**: for disjoint  $S, T$ ,  $v(S) + v(T) \leq v(S \cup T)$ .

“forming coalitions is always better”

Superadditive  $\implies$  monotone, converse is not true.

We usually only consider the grand coalition:  $\pi = (N)$ .

3. **Convex games:** for any  $S, T$ ,  $v(S) + v(T) \leq v(S \cup T) + v(S \cap T)$ . (supermodularity inequality)

Convexity  $\implies$  superadditivity, converse is not true.

#### Proposition 3.1

A game  $(N, v)$  is convex if and only if for every  $S, T$  where  $S \subseteq T \subseteq N$  and for every player  $i \in N \setminus T$ ,

$$v(T \cup \{i\}) - v(T) \geq v(S \cup \{i\}) - v(S)$$

“A player is more useful in larger coalitions”

## 3.2 Shapley values

Two desirable properties of an outcome in cooperative games:

1. Fairness. The payoff vector should reflect the contribution of the players to their coalitions.
2. Stability. We want to incentivize the players to stay in their assigned coalition in the coalition structure.

Shapley values deal with the fairness of the payoff vector. Assume players form the grand coalition. (If not, look at individual coalitions separately.)

*How to quantify a player's contribution?*

**Idea 1:** Compare the value of the coalition before and after joins it.

Example: Ice cream game. The contribution of  $A$  is  $v(\{A, B, C\}) - v(\{B, C\}) = 0.5$

Problem: The sum of the payoffs may exceed the value of coalition  $x(N) > v(N)$ .

**Idea 2:** Fix a sequence of players, and see their contribution sequentially.

Example: Use sequence  $A, B, C$ .  $v(\{A\}) = 0$ , so  $A$  contributes 0.  $v(\{A, B\}) = 1$ , so  $B$  contributes 1.  $v(\{A, B, C\}) = 2$ , so  $C$  contributes 1. This is efficient,  $x(N) = v(N)$ .

Problem: Different orderings produce different results.

*Shapley's idea:* Look at all possible orderings of players, average a player's contributions.

$S_N$

A permutation of  $N$  has the form  $\sigma = (\sigma_1, \dots, \sigma_n)$  where each  $\sigma_i$  is a distinct element of  $N$ . The element  $\sigma_i$  is at the  $i$ -th position of  $\sigma$ . The set of all permutations of  $N$  is denoted  $S_N$ .

#### marginal contribution

Given a permutation  $\sigma \in S_N$ , the **marginal contribution** of player  $\sigma_i$  is

$$\Delta_\sigma(\sigma_i) = v(\{\sigma_1, \dots, \sigma_i\}) - v(\{\sigma_1, \dots, \sigma_{i-1}\})$$



### Shapley value

The **Shapley value** of player  $i$  is

$$\varphi_i = \frac{1}{n!} \sum_{\sigma \in S_N} \Delta_{\sigma}(i)$$

**Example:**

In the ice cream game,  $N = \{A, B, C\}$ , and there are 6 permutations:

$$\begin{aligned} \sigma^1 &= (A, B, C), & \sigma^2 &= (A, C, B), & \sigma^3 &= (B, A, C), \\ \sigma^4 &= (B, C, A), & \sigma^5 &= (C, A, B), & \sigma^6 &= (C, B, A) \end{aligned}$$

We calculate the marginal contribution of  $A$  in each of the permutations:

$$\begin{aligned} \Delta_{\sigma^1}(A) &= v(\{A\}) - v(\emptyset) = 0 \\ \Delta_{\sigma^2}(A) &= v(\{A\}) - v(\emptyset) = 0 \\ \Delta_{\sigma^3}(A) &= v(\{B, A\}) - v(\{B\}) = 1 \\ \Delta_{\sigma^4}(A) &= v(\{B, C, A\}) - v(\{B, C\}) = 0.5 \\ \Delta_{\sigma^5}(A) &= v(\{C, A\}) - v(\{C\}) = 1 \\ \Delta_{\sigma^6}(A) &= v(\{C, B, A\}) - v(\{C, B\}) = 0.5 \end{aligned}$$

So the Shapley value for  $A$  is  $\varphi_A = \frac{1}{6}(0 + 0 + 1 + 0.5 + 1 + 0.5) = \frac{1}{2}$

Other Shapley values;  $\varphi_B = \varphi_C = \frac{3}{4}$

## 4 good properties of Shapley values

1. **Efficiency:** it distributes  $v(N)$  to all players.

### Proposition 3.2

$$\sum_{i \in N} \varphi_i = v(N)$$

**Proof:**

For any  $\sigma \in S_N$ , the sum of all marginal contributions is

$$\begin{aligned} \sum_{i \in N} \Delta_{\sigma}(i) &= \sum_{i=1}^n \Delta_{\sigma}(\sigma_i) \quad \text{since permutation is a bijection} \\ &= [v(\{\sigma_1\}) - v(\emptyset)] + [v(\{\sigma_1, \sigma_2\}) - v(\{\sigma_1\})] + \cdots + [v(\{\sigma_1, \dots, \sigma_n\}) - v(\{\sigma_1, \dots, \sigma_{n-1}\})] \\ &= v(\{\sigma_1, \dots, \sigma_n\}) - v(\emptyset) \\ &= v(N) \end{aligned}$$

So the sum of Shapley values is

$$\sum_{i \in N} \varphi_i = \sum_{i \in N} \frac{1}{n!} \sum_{\sigma \in S_N} \Delta_{\sigma}(i) = \frac{1}{n!} \sum_{\sigma \in S_N} \sum_{i \in N} \Delta_{\sigma}(i) = \frac{1}{n!} \sum_{\sigma \in S_N} v(N) \underset{|S_N| = n!}{=} \frac{1}{n!} (n!) v(N) = v(N)$$

□

2. **Symmetric.**

### symmetric players

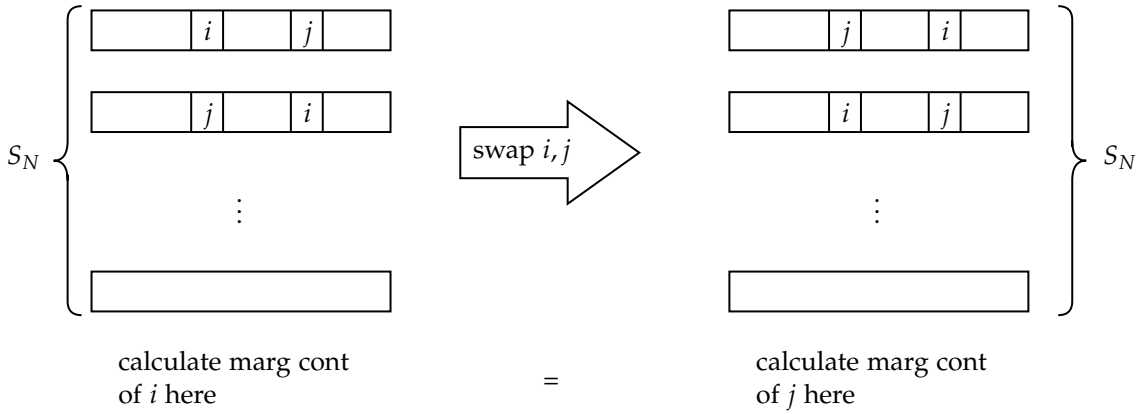
Two players  $i, j$  are **symmetric** if  $v(C \cup \{i\}) = v(C \cup \{j\}) \forall C \subseteq N \setminus \{i, j\}$ . (they contribute to coalitions equally)

#### Example:

In the ice cream game,  $B, C$  are symmetric.  $v(\emptyset \cup \{B\}) = v(\emptyset \cup \{C\}) = 0$ ,  $v(\{A\} \cup \{B\}) = v(\{A\} \cup \{C\}) = 1$ .

### Proposition 3.3

If  $i, j$  are symmetric players, then  $\varphi_i = \varphi_j$ .



#### Proof:

Define  $f : S_N \rightarrow S_N$  where  $f(\sigma)$  is obtained from  $\sigma$  by swapping  $i$  and  $j$ . This is a bijection  $f^{-1} = f$ . We claim  $\Delta_\sigma(i) = \Delta_{f(\sigma)}(j)$ . Two cases:

- Suppose  $i$  precedes  $j$  in  $\sigma$ . Let  $C$  be all elements preceding  $i$ . In  $f(\sigma)$ ,  $C$  is also the elements preceding  $j$ . So

$$\Delta_\sigma(i) = v(C \cup \{i\}) - v(C) \text{ and } \Delta_{f(\sigma)}(j) = v(C \cup \{j\}) - v(C)$$

Since  $C \subseteq N \setminus \{i, j\}$  and  $i, j$  are symmetric,  $v(C \cup \{i\}) = v(C \cup \{j\})$ . So  $\Delta_\sigma(i) = \Delta_{f(\sigma)}(j)$ .

- Suppose  $j$  precedes  $i$  in  $\sigma$ . Let  $C$  be all elements preceding  $i$  except  $j$ . In  $f(\sigma)$ ,  $C$  is also the elements that precedes  $j$  except  $i$ . So

$$\Delta_\sigma(i) = v(C \cup \{j\} \cup \{i\}) - v(C \cup \{j\}) \text{ and } \Delta_{f(\sigma)}(j) = v(C \cup \{i\} \cup \{j\}) - v(C \cup \{i\})$$

Since  $C \subseteq N \setminus \{i, j\}$  and  $i, j$  are symmetric,  $v(C \cup \{j\}) = v(C \cup \{i\})$ , so  $\Delta_\sigma(i) = \Delta_{f(\sigma)}(j)$ .

Therefore,

$$\varphi_i = \frac{1}{n!} \sum_{\sigma \in S_N} \Delta_\sigma(i) \underset{\text{from above}}{=} \frac{1}{n!} \sum_{\sigma \in S_N} \Delta_{f(\sigma)}(j) \underset{\text{since } f \text{ is a bijection}}{=} \frac{1}{n!} \sum_{\sigma \in S_N} \Delta_\sigma(j) = \varphi(j)$$

□

#### Example: Unanimity Game

Suppose  $|N| = n$  and  $v(S) = \begin{cases} 1 & S = N \\ 0 & \text{otherwise} \end{cases}$

Any pair of players is symmetric, so  $\varphi_i = \varphi_j$  for any  $i, j$ . Since  $\varphi$  is efficient, the sum is  $v(N) = 1$ . So  $\varphi_i = \frac{1}{n}$  for each  $i \in N$ .

## 3. Dummy player.

**dummy player**

$i$  is a **dummy player** if  $v(S \cup \{i\}) = v(S)$ ,  $\forall S \subseteq N \setminus \{i\}$ . The player does not contribute to any coalition.

**Example: 101-seat parliament**

$A, B, C, D$  with 40, 22, 30, 9 seats. Party D is a dummy player: no combination of  $A, B, C$  exists where it is not a majority, but adding 9 gives a majority.

**Proposition 3.4**

If  $i$  is a dummy player, then  $\varphi_i = 0$ .

**Proof:**

For any  $\sigma \in S_N$ , say  $i$  is at the  $j$ -th position ( $\sigma_j = i$ ), the marginal contribution of  $i$  is  $\Delta_\sigma(i) = v(\{\sigma_1, \dots, \sigma_{j-1}, i\}) - v(\{\sigma_1, \dots, \sigma_{j-1}\}) = 0$  by definition of a dummy player. So

$$\varphi_i = \frac{1}{n!} \sum_{\sigma \in S_N} \Delta_\sigma(i) = 0$$

□

**Note:**

The converse is not true. If a game is monotone, then the converse is true.

4. **Additivity:** Suppose there are multiple games with the same set of players. We add the values together to get a new game. Then the Shapley values are also added together.

**Proposition 3.5**

Let  $(N, v^1), (N, v^2)$  be two cooperative games. Define  $v^3(S) = v^1(S) + v^2(S)$ ,  $\forall S \subseteq N$ . Let  $\varphi_i^j$  be the Shapley values of player  $i$  in  $(N, v^j)$ ,  $j = 1, 2, 3$ . Then  $\varphi_i^3 = \varphi_i^1 + \varphi_i^2$  for all  $i$ .

**Summary** The Shapley values satisfy 4 good properties: efficiency, symmetric, dummy player, additivity. *Deep result:* The Shapley value function is the only one that satisfies all 4 properties. (If  $f$  is a function that maps  $(N, v)$  to a real vector  $\mathbb{R}^n$  and all properties hold, then  $f$  gives the Shapley values.)

## 3.3 The core

**Stability:** Given an outcome, what would be a reason that players want to deviate from it? A group of players could generate more value than what they are receiving.  $x(C) < v(C)$

**core**

The **core** of a cooperative game  $(N, v)$  consists of all outcomes  $(\pi, x)$  such that  $x(C) \geq v(C)$  for all  $C \subseteq N$ .

**Example: Ice cream game**

Consider  $(\pi, x)$  with  $\pi = (\{A, B\}, \{C\})$  and  $x_A = 0.5, x_B = 0.5, x_C = 0$ . If  $C$  joins with  $\{A, B\}$ , then they produce value 2, while currently their combined payoffs is 1. Better if they form  $\{A, B, C\}$ , not in the core.

Same reasoning gives  $\pi = (N)$  if  $(\pi, x)$  is in the core. If  $x_A = 2, x_B = x_C = 0$ , then  $\{B, C\}$  can get more value. If  $x_A = 0, x_B = x_C = 1$ , then  $(\pi, x)$  is in the core. This satisfies these inequalities:

$$x_A + x_B + x_C \geq 2, \quad x_A + x_B \geq 1, \quad x_A + x_C \geq 1, \quad x_B + x_C \geq 1.5, \quad x_A \geq 0, \quad x_B \geq 0, \quad x_C \geq 0$$

## Properties of the outcomes in the core

1. They are efficient within the coalition structure.

### Proposition 3.6

If  $(\pi, x)$  is in the core, then  $x(C) = v(C)$  for each  $C \in \pi$ .

**Proof:**

Let  $C \in \pi$ . By the definition of the core,  $x(C) \geq v(C)$ . Since  $(\pi, x)$  is a valid outcome,  $x(C) \leq v(C)$ . Therefore,  $x(C) = v(C)$  for all  $C \in \pi$ .  $\square$

2. The coalition structure generates the maximum amount of total value among all outcomes. “social welfare”

$v(\pi)$

$$v(\pi) = \sum_{C \in \pi} v(C)$$

### Proposition 3.7

If  $(\pi, x)$  is in the core, then  $v(\pi) \geq v(\pi')$  for all partitions  $\pi'$  of  $N$ .

**Proof:**

$$v(\pi) = \sum_{C \in \pi} v(C) \underset{\text{By Proposition 3.6}}{=} \sum_{C \in \pi} x(C) \underset{\text{Since } \pi \text{ is a partition of } N}{=} \sum_{i \in N} x_i \underset{\text{Since } \pi' \text{ is a partition of } N}{=} \sum_{C' \in \pi'} x(C') \underset{\text{Since } (\pi, x) \text{ is in the core}}{\geq} \sum_{C' \in \pi'} v(C') = v(\pi') \quad \square$$

**Note:**

This proposition only says that coalition structures that maximize total value are eligible to be in the core. It does not mean that there exists an outcome in the core with this structure.

## Games with empty cores

**Example: 3-player majority game**

$$N = \{1, 2, 3\}, \quad v(S) = \begin{cases} 1 & |S| \geq 2 \\ 0 & \text{otherwise} \end{cases}$$

We claim that no outcome is in the core. Suppose  $(\pi, x)$  is in the core. Then  $x_1 + x_2 + x_3 \geq 1$ ,  $x_1, x_2, x_3 \geq 0$ . So  $x_i \geq \frac{1}{3}$  for some  $i$ . The value of any coalition structure is at most 1, so  $x_1 + x_2 + x_3 \leq 1$ . This means  $x(N \setminus \{i\}) \leq \frac{2}{3}$ . However,  $v(N \setminus \{i\}) = 1 > x(N \setminus \{i\})$ . This contradicts the assumption that  $(\pi, x)$  is in the core, which implies core is empty.

Main question: which games have non-empty cores?

### 3.3.1 Cores of superadditive games

**Goal:** Determine when a superadditive game has a non-empty core. We can narrow the search: we only need to consider outcomes that form the grand coalition.

**Proposition 3.8**

Let  $(N, v)$  be a superadditive game. If  $(\pi, x)$  is in the core, then  $((N), x)$  is in the core.

**Proof:**

We need to prove: the core conditions holds, and  $((N), x)$  is a valid outcome.

Since  $(\pi, x)$  is in the core,  $x(C) \geq v(C)$  for all  $C \subseteq N$ . This still holds for  $((N), x)$ .

To show that  $((N), x)$  is a valid outcome, we need to show that  $x(N) \leq v(N)$ .

$$x(N) = \sum_{C \in \pi} x(C) \leq \sum_{C \in \pi} v(C) \leq v(N)$$

$\uparrow$  Since  $\pi$  is a partition of  $N$        $\uparrow$  Since  $(\pi, x)$  is a valid outcome       $\uparrow$  Superadditivity

□

**Example: Unanimity game**

$$v(S) = \begin{cases} 1 & S = N \\ 0 & \text{otherwise} \end{cases}$$

To determine if the core is non-empty, we only need to consider  $((N), x)$ .  $x$  satisfies  $\sum_{i \in N} x_i = 1$  (Proposition 3.6) and  $\sum_{i \in S} x_i \geq 0$  for all  $S \subseteq N$ . e.g.,  $x_i = \frac{1}{n}, \forall i$ , or  $x_1 = 1, x_i = 0$  if  $i \neq 1$ .

**Characterizing superadditive games with non-empty cores**

Given an outcome  $((N), x)$ , what must  $x$  satisfy to be in the core? We claim that  $x$  must be in the set

$$\mathcal{C} = \{x \in \mathbb{R}^N : x(N) = v(N), \quad x(C) \geq v(C) \quad \forall C \subseteq N\}$$

$\uparrow$  Proposition 3.6       $\uparrow$  Definition of the core

**Example: Ice cream game**

$$\mathcal{C} = \left\{ x \in \mathbb{R}^N : \begin{array}{l} x_A + x_B + x_C = 2, \\ x_A \geq 0, \\ x_B \geq 0, \\ x_C \geq 0, \\ x_A + x_B \geq 1, \\ x_A + x_C \geq 1, \\ x_B + x_C \geq 1.5, \\ x_A + x_B + x_C \geq 2 \end{array} \right\}$$

Now  $\mathcal{C}$  is the intersection of halfspaces, so it is a polyhedron. *Mini-result:*  $(N, v)$  has a non-empty core if and only if  $\mathcal{C}$  is non-empty.

We can solve the problem of “is  $\mathcal{C}$  non-empty” using a linear program.

Let (P) be the following LP:

$$\begin{array}{ll} \min & x(N) \\ \text{s.t.} & x(C) \geq v(C) \quad \forall C \subseteq N \end{array}$$

Take the dual (D):

$$\begin{array}{ll} \max & \sum_{C \subseteq N} y_C v(C) \\ \text{s.t.} & \sum_{C \subseteq N, i \in C} y_C = 1 \quad \forall i \in N \\ & y \geq 0 \end{array}$$

Example:

$$\begin{array}{llll}
 \min & x_A & +x_B & +x_C \\
 \text{s.t.} & x_A & & \geq 0 \\
 & & x_B & \geq 0 \\
 & & & x_C \geq 0 \\
 & x_A & +x_B & \geq 1 \\
 & x_A & & +x_C \geq 1 \\
 & & x_B & +x_C \geq 1.5 \\
 & x_A & +x_B & +x_C \geq 2
 \end{array}$$

$$\begin{array}{llllll}
 \max & & y_{AB} & +y_{AC} & +1.5y_{BC} & +2y_{ABC} \\
 \text{s.t.} & y_A & & +y_{AB} & +y_{AC} & & +y_{ABC} = 1 \\
 & & y_B & & +y_{AB} & +y_{AC} & & +y_{ABC} = 1 \\
 & & & y_C & & +y_{AC} & +y_{BC} & +y_{ABC} = 1 \\
 & & & & & & & y \geq 0
 \end{array}$$

(P) has an optimal value  $v(N) \Leftrightarrow$  (D) has optimal value  $v(N) \Leftrightarrow \sum_{C \subseteq N} y_C v(C) \leq v(N)$  for all feasible  $y$ .

Rationale: (P) is feasible (take large  $x$ ), and  $x(N) \geq v(N)$  is a constraint, so (P) is bounded.  $\Rightarrow$  (P) has an optimal solution. If optimal value is  $v(N)$ , then we have optimal solution  $x$  with  $x(N) = v(N)$  and  $x(C) \geq v(C)$ ,  $\forall C \subseteq B$ , so  $x \in \mathcal{C}$ .

(Subtle pt: Is it possible that  $\sum_{C \subseteq N} y_C v(C) < v(N)$  for all  $y$ ?)

What is the meaning of the dual?

1. Feasible solution:  $\sum_{C \subseteq N, i \in C} y_C = 1$ .

Example:

$$\begin{array}{ccc}
 y_{AB} = y_C = 1 & \text{or} & y_{ABC} = 1 & \text{or} & y_{AB} = y_{BC} = y_{AC} = \frac{1}{2} \\
 \uparrow & & \uparrow & & \uparrow \\
 (\{A, B\}, \{C\}) & & (\{A, B, C\}) & & \text{"fractional" partition} \\
 & & & & \begin{array}{l} A \text{ works for} \\ \{A, B\} \text{ } 1/2 \text{ the time,} \\ \{A, C\} \text{ } 1/2 \text{ the time,} \\ \Rightarrow \text{Total } 1 \text{ for } A \end{array}
 \end{array}$$

A feasible solution  $y$  is a generalized coalition structure.  $y_C$  represents the fraction of time each member will contribute to  $C$ , with a total time of 1 from each player. Any such feasible  $y$  is called a **balancing weight**.

2. Objective.  $\sum_{C \subseteq N} y_C v(C) \leq v(N)$ .

Example:

$$\begin{array}{c}
 y_{AB} v(\{A, B\}) + y_C v(\{C\}) \leq 2 \\
 \uparrow \qquad \qquad \qquad \uparrow \\
 \text{value of the coalition} \quad \text{value of the grand} \\
 \text{structure } (\{A, B\}, \{C\}) \quad \text{coalition } (N)
 \end{array}$$

By Proposition 3.7, maximize social welfare.

Then  $\sum_{C \subseteq N} y_C v(C)$  is the total value of the fractional partition represented by  $y$ . Then inequality  $\sum_{C \subseteq N} y_C v(C) \leq v(N)$  means the value of the grand coalition is maximum over the values of any fractional partition. This generates Proposition 3.7.

A game that satisfies this inequality for all balancing weight  $y$  is called a **balanced game**.

#### Theorem 3.9: Bondareva-Shapley

A superadditive game has a non-empty core if and only if it is balanced.

### Game with non-empty cores

In superadditive games with non-empty cores, there is always an outcome in the core with the grand coalition. This is not necessarily the case for cooperative games in general.

**Example:**

$$\text{Let } N = \{1, 2, 3, 4\}, v(S) = \begin{cases} 2 & |S| \geq 2 \\ 0 & \text{otherwise} \end{cases}$$

By Proposition 3.7, coalition structure in the core has highest value.  $v(N) = 2$ . But  $v(\{1, 2\}, \{3, 4\}) = 4$ . So the grand coalition cannot be in any outcome of the core. The core is non-empty:  $\pi = (\{1, 2\}, \{3, 4\})$  with  $x = (1, 1, 1, 1)$  is in the core.

Checking if the core is non-empty cannot be reduced to checking only the grand coalition. But we can relate this to superadditive games.

#### superadditive cover

For any cooperative game  $(N, v)$ , its **superadditive cover** is  $(N, v^*)$  where, for each  $S \subseteq N$ ,

$$v^*(S) = \max\{v(\pi) : \pi \text{ is a partition of } S\}$$

**Example:**

The superadditive cover for example above is  $(N, v^*)$  where  $v^*(S) = \begin{cases} 4 & |S| = 4 \\ 2 & |S| = 2, 3 \\ 0 & |S| = 0, 1 \end{cases}$

For example,

$$v^*(\{1, 2, 3\}) = \max\{v(\{1, 2, 3\}), v(\{1\}, \{2, 3\}), v(\{2\}, \{1, 3\}), v(\{3\}, \{1, 2\}), v(\{1\}, \{2\}, \{3\})\} = 2$$

Then we can prove that superadditive cover is superadditive.

#### Proposition 3.10

A cooperative game  $(N, v)$  has a non-empty core if and only if its superadditive cover  $(N, v^*)$  has a non-empty core.

**Example:**

Check that  $((N), (1, 1, 1, 1))$  is in the core of  $(N, v^*)$  above.

**Proof:**

$(\Rightarrow)$  Let  $(\pi, x)$  be in the core of  $(N, v)$ .

Note that by Proposition 3.7,  $v(\pi)$  has the maximum value among all partitions of  $N$ . So by the definition of superadditive cover,  $v^*(N) = v(\pi)$ .

1.

$$v^*(N) = v(\pi) = \sum_{C \in \pi} v(C) \underset{\substack{\uparrow \\ \text{Proposition 3.6} \\ (\pi, x) \text{ is in the core}}}{=} \sum_{C \in \pi} x(C) \underset{\substack{\uparrow \\ \text{Since } \pi \text{ is a} \\ \text{partition of } N}}{=} \sum_{i \in N} x_i = x(N)$$

2. Let  $C \subseteq N$ . Suppose  $v^*(C) = v(\pi')$  for some partition  $\pi'$  of  $C$ . Then

$$v^*(C) = v(\pi') = \sum_{C' \in \pi'} v(C') \underset{\substack{\uparrow \\ (\pi, x) \text{ is} \\ \text{in the core}}}{\leq} \sum_{C' \in \pi'} x(C') \underset{\substack{\uparrow \\ \text{Since } \pi' \text{ is a} \\ \text{partition of } C}}{=} x(C)$$

So  $((N), x)$  is in the core of  $(N, v^*)$ .

( $\Leftarrow$ ) Since  $(N, v^*)$  is superadditive, there exists a payoff vector  $x$  such that  $((N), x)$  is in the core (Proposition 3.8). By the definition of the superadditive cover,  $v^*(N) = v(\pi)$  for some partition  $\pi$  of  $N$ .

1. It is a valid outcome,  $v(C) \geq x(C)$  for all  $C \in \pi$ ; and
2. core condition is satisfied,  $v(C) \leq x(C)$  for all  $C \subseteq N$ . We leave the proof of them as exercises.

□

### Corollary 3.11

A cooperative game has a non-empty core if and only if its superadditive cover is balanced.

**Proof:**

Combine the Bondareva-Shapley theorem with Proposition 3.10.

□

### 3.3.2 Cores of convex games

**Example: Bankruptcy game**

Bob the Banker is bankrupt. He owes three people money, \$100, \$200, \$300 each. Bob only has \$200. How should Bob's money be divided? \$50, \$50, \$100? \$0, \$0, \$200? which ones are stable?

*General model:* Bob has \$ $M$ , he owes money to  $n$  people  $N$ , amounts owed are  $d \in \mathbb{R}^N$ . Assume  $0 \leq M \leq \sum_{i \in N} d_i$ .

*Cooperative game:*  $(N, v)$  where  $v(S) = \max\{0, M - \sum_{i \in N \setminus S} d_i\}$  for each  $S \subseteq N$ .

*Meaning:* Players are taking a pessimistic view,  $v(S)$  is the amount left if the remaining players take what they owed.

With numbers above,  $M = 200, d = (100, 200, 300)$ . Examples of values:  $v(\{2, 3\}) = 200 - d_1 = 100, v(\{1, 2\}) = 0$ .

**Exercise:**

Show that is a convex game.

### Proposition 3.12

Convex games have non-empty cores.

*Idea:* The marginal contributions of the players in any permutation form the payoff vector in the core.

**Proof:**

Since convex games are superadditive, it suffices to find  $x$  such that  $((N), x)$  is in the core (Proposition 3.8). Let  $\sigma \in S_N$ . WLOG, assume  $\sigma = (1, 2, \dots, n)$ . Define  $x$  by  $x_i = \Delta_\sigma(i)$ .

1. In the proof of Proposition 3.2,  $x(N) = \sum_{i=1}^n \Delta_\sigma(i) = v(N)$ .
2. Let  $C \subseteq N$ . Suppose  $C = \{i_1, \dots, i_k\}$  where  $i_1 < \dots < i_k$ .

For any  $i_j$ , the “equivalent convex condition” implies

$$v(\{1, \dots, i_j\}) - v(\{1, \dots, i_j - 1\}) \geq v(\{i_1, \dots, i_j\}) - v(\{i_1, \dots, i_{j-1}\})$$



So

$$\begin{aligned}
 x(C) &= \sum_{j=1}^k \Delta_{\sigma}(i_j) \\
 &= \sum_{j=1}^k (v(\{1, \dots, i_j\}) - v(\{1, \dots, i_j - 1\})) \\
 &\geq \sum_{j=1}^k (v(\{i_1, \dots, i_j\}) - v(\{i_1, \dots, i_j - 1\})) \\
 &= v(\{i_1, \dots, i_k\}) - v(\emptyset) \\
 &= v(C)
 \end{aligned}$$

□

**Note:**

Any vector of marginal contributions is in the core with the grand coalition. The set of all vectors in the core is  $\mathcal{C}$ , which is a polyhedron, which is convex. The vector of Shapley values  $\varphi$  is a convex combinations of the marginal contributions, so it is also in the core.

*Stronger result:*  $\mathcal{C}$  is precisely equal to the convex combinations of all marginal contribution vectors. (We will not prove this)

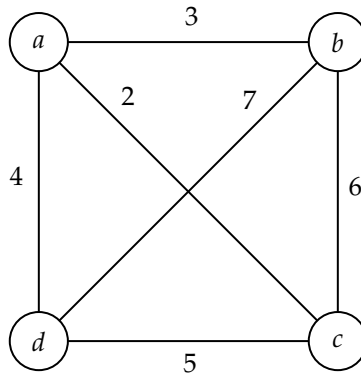
**Example: Bankruptcy game**

Shapley values:  $(33 + \frac{1}{3}, 83 + \frac{1}{3}, 83 + \frac{1}{3})$ . This is in the core with the grand coalition. Order the players  $\sigma = (3, 2, 1)$ , the marginal contributions  $(100, 100, 0)$  is also in the core.

### 3.4 Matching games

Recall that matching game consists of a graph  $G$ , the players are the vertices  $N = V(G)$ , and non-negative weights  $w$ . The value  $v(S)$  of  $S \subseteq N$  is the maximum weight of a matching using vertices in  $S$ . Notation: For a matching  $M$ ,  $w(M)$  is the total edge weight of the matching.

**Example:**



$$v(\{a, b\}) = 3, \quad v(\{a, b, c\}) = 6, \quad v(\{a, b, c, d\}) = 10$$

We interpret the weight of an edge as the value generated by the two players on both sides when they work together. This game is superadditive (check), so we only need to consider the grand coalition when determining if it has a non-empty core. Recall:  $((N), x)$  is in the core if and only if  $x \in \mathcal{C}$  where

$$\mathcal{C} = \{x \in \mathbb{R}^N : x(N) = v(N), x(C) \geq v(C), \quad \forall C \subseteq N\}$$

There are exponentially many inequalities. We can make a simplification: Define

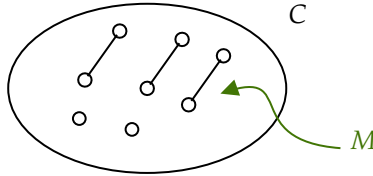
$$\mathcal{C}' = \{x \in \mathbb{R}^N : x(N) = v(N), x_u + x_v \geq w_{uv} \quad \forall uv \in E(G), x \geq \mathbf{0}\}$$

**Proposition 3.13**

$$\mathcal{C} = \mathcal{C}'$$

**Proof:**

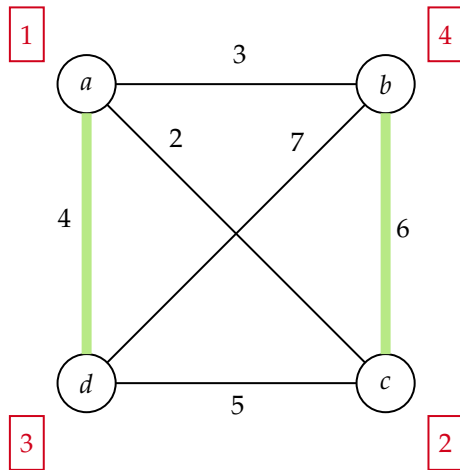
If  $x \in \mathcal{C}$ , then  $x$  satisfies all constraints in  $\mathcal{C}'$  (take  $C = \{u, v\}$  for each  $uv \in E(G)$ ; take  $C = \{v\}$  to get  $x \geq 0$ ). So  $\mathcal{C} \subseteq \mathcal{C}'$ .



Let  $x \in \mathcal{C}'$ . Then  $x(N) = v(N)$ . Let  $C \subseteq N$ , and let  $M$  be a maximum weight matching in  $C$ . Then

$$\begin{aligned} x(C) &\geq \sum_{uv \in M} x_u + x_v \quad \text{since } M \text{ is a matching, and let } M \text{ be a maximum weight matching in } C \\ &\geq \sum_{uv \in M} w_{uv} \quad \text{since } x \in \mathcal{C}' \\ &= w(M) = v(C) \end{aligned}$$

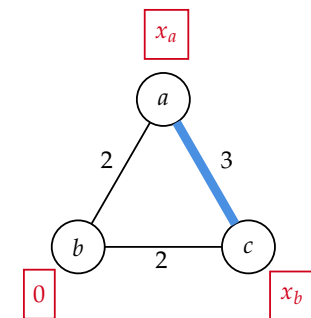
□

**Example:**

□ in the core with grand coalition  
 — maximum matching

Check:  $x_a + x_b = 5 \geq 3 = w_{ab}$   
 $x_b + x_c = 6 \geq 6 = w_{bc}$   
 $\vdots$

**Observations:** The edges in the maximum matching are “efficient” ( $x_u + x_v = w_{uv}$ ) for all  $uv \in M$ .

**Example:**

$$v(\{a, b, c\}) = 3$$

If  $((N), x)$  is in the core, then  $x_a + x_c = 3$ . This implies that  $x_b = 0$  because  $x(N) = v(N)$ . At least one of  $x_a, x_c$  is at most 1.5. Say it is  $x_a$ .

Then  $x_a + x_b \leq 1.5$ , contradicting  $x_a + x_b \geq 2$ ,  $\implies$  the core is empty.

**LP formulation**

We can determine if  $\mathcal{C}'$  is non-empty by solving the following LP:

$$\begin{array}{ll} \min & x(N) \\ \text{s.t.} & x_u + x_v \geq w_{uv} \quad \forall uv \in E(G) \\ & x \geq \mathbf{0} \end{array} \quad (\text{P})$$

$\mathcal{C}'$  is non-empty if and only if (P) has optimal value  $v(N)$ .

$$\begin{array}{ll} \max & \sum_{uv \in E(G)} w_{uv} y_{uv} \\ \text{s.t.} & \sum_{\{uv: uv \in E(G)\}} y_{uv} \leq 1 \quad \forall u \in N \\ & y \geq \mathbf{0} \end{array} \quad (\text{D})$$

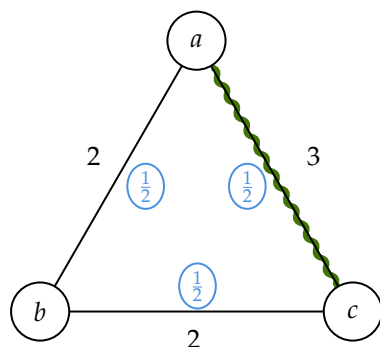
$\mathcal{C}'$  is non-empty if and only if (D) has optimal value  $v(N)$ .

Example:

$$\begin{array}{ll} \min & x_a + x_b + x_c + x_d \\ \text{s.t.} & x_a + x_b \geq 3 \\ & x_a + x_c \geq 2 \\ & x_a + x_d \geq 4 \\ & \vdots \\ & x \geq \mathbf{0} \end{array}$$

$$\begin{array}{ll} \max & 3y_{ab} + 2y_{ac} + 4y_{ad} + \dots \\ \text{s.t.} & y_{ab} + y_{ac} + y_{ad} \leq 1 \\ & \vdots \\ & y \geq \mathbf{0} \end{array}$$

Meaning of  $y$ : If  $y$  is integral, then its values are 0 or 1. For each vertex, the number of incident edges with value 1 is at most 1. So the edges with value 1 form a matching. Maximizing integral solutions will give us the weight of a max matching, which is  $v(N)$ . But  $y$  could be fractional. So feasible solutions to (D) are generalized matchings.



○ fractional “matching”

~ max matching,  $v(N) = 3$

sum of weights of the fractional soln is  $\frac{1}{2} \cdot 2 + \frac{1}{2} \cdot 2 + \frac{1}{2} \cdot 3 = \frac{7}{2} > 3$   
 $\Rightarrow$  optimal value of (D)  $\neq v(N) \Rightarrow$  empty core

As long as no fractional matching has weight higher than  $v(N)$ , then  $v(N)$  is the optimal value achieved by a maximum matching.

#### Proposition 3.14

The core of the matching game is non-empty if and only if (D) has an integral optimal solution.

**Proof:**

It suffices to prove that (P) has optimal value  $v(N)$  if and only if (D) has an integral optimal solution.

( $\Rightarrow$ ) Suppose (P) has optimal value  $v(N)$ . Suppose  $v(N) = w(M)$  for some maximum matching  $M$ . Define  $y \in \mathbb{R}^{E(G)}$  by

$$y_e = \begin{cases} 1 & e \in M \\ 0 & e \notin M \end{cases} \quad \forall e \in E(G)$$

Then  $y$  is feasible for (D) (since  $M$  is matching) with objective value  $w(M) = v(N)$ . By weak duality,  $y$  is optimal for (D), and it is an integral optimal solution.

( $\Leftarrow$ ) Suppose  $y$  is an integral optimal solution for (D). Let  $M' = \{e \in E(G) : y_e = 1\}$ . Then  $M'$  is a matching with optimal value  $w(M')$ . Since  $y$  is optimal,  $M'$  must be a maximum matching. So  $w(M') = v(N)$ . So (P) has optimal value  $v(N)$ .  $\square$

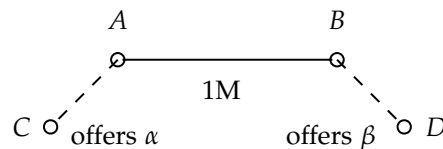
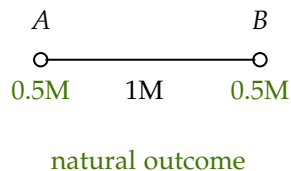
**Note:**

When the graph is bipartite, (D) always has an integral optimal solution. Hence the core is non-empty. For general graphs, there is an efficient algorithm to determine if (D) has an integral optimal solution.

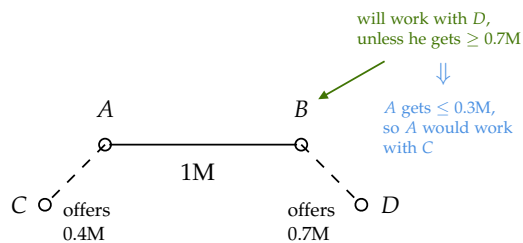
### 3.5 Network bargaining game

#### Two-player bargaining

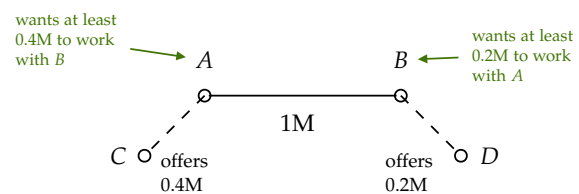
Alice and Bob are negotiating over how to split \$1 M. They can only take the money if both agree on how to split.



Suppose there are outside options: Carol offers  $\alpha$  to Alice if they work together, Dan offers  $\beta$  to Bob if they work together.



Negotiation between A, B  
breaks down. This happens  
if  $\alpha + \beta > 1M$



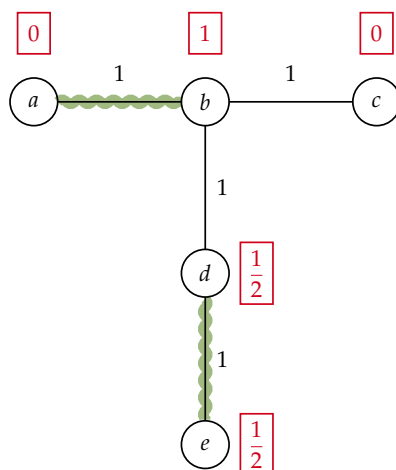
There are 0.4M left. Split then equally.  
 $\Rightarrow$  A gets 0.6M, B gets 0.4M

*Nash's bargaining solution:* Players A, B try to split  $w$ , each has outside options  $\alpha, \beta$  respectively. If  $\alpha + \beta > w$ , then no split is possible. Otherwise,  $x_A = \alpha + \frac{w - \alpha - \beta}{2}$ ,  $x_B = \beta + \frac{w - \alpha - \beta}{2}$ .

#### Network bargaining

Given a graph  $G$ , the players are the vertices  $N = V(G)$ . Each edge  $e$  has weight  $w_e \geq 0$ . Which parism form? How do they split their value?

Example:



5 players negotiating. Who has more negotiating power?  
One possible result with partners  $\sim$  and payoffs  $\square$   
Player b has the most powerful position.

An outcome of the network bargaining game consists of a matching  $M$  and a payoff vector  $x \in \mathbb{R}^N$  such that

$$x_u + x_v = w_{uv} \text{ for all } uv \in M, \quad x_v = 0 \text{ if } v \text{ is not matched by } M, \quad x \geq 0$$

The stability of an outcome depends on the outside options of the players.

#### outside option

For an outcome  $(M, x)$  and a player  $u$ , the **outside option** of  $u$  is

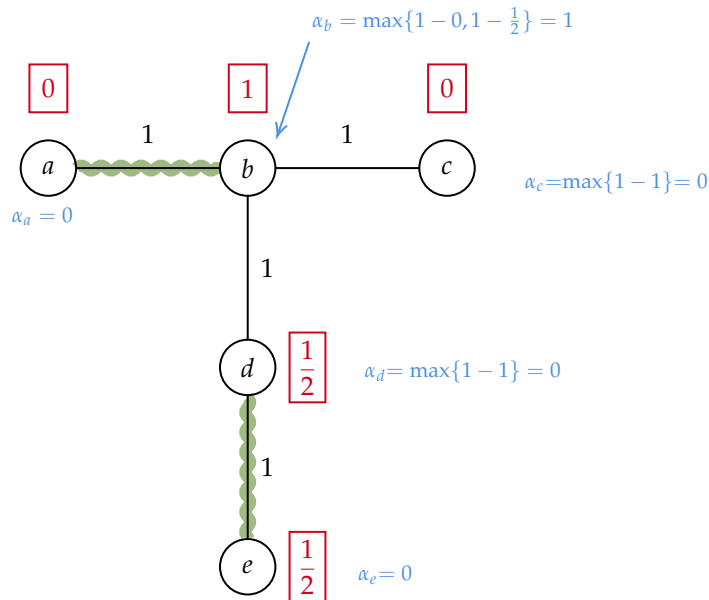
$$\alpha_u = \max(\{0\} \cup \{w_{uv} - x_v : uv \in E \setminus M\})$$

This is the maximum value that  $u$  can get by defecting from their current partner in  $M$ . No need to defect if  $\alpha_u \leq x_u$ .

#### stable outcome

An outcome  $(M, x)$  is **stable** if  $\alpha_u \leq x_u$  for all  $u \in N$ .

Example:



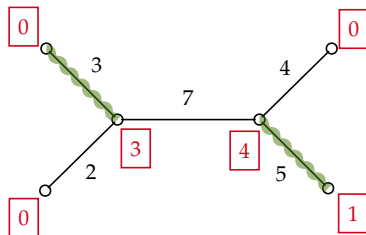
This is a stable outcome

To be stable, for each edge  $uv$ ,  $x_u \geq \alpha_u \geq w_{uv} - x_v$ . So  $x_u + x_v \geq w_{uv}$ . This is in  $\mathcal{C}'$  from the matching game.

#### Proposition 3.15

A stable outcome exists in a networking bargaining game if and only if the corresponding matching game has a non-empty core.

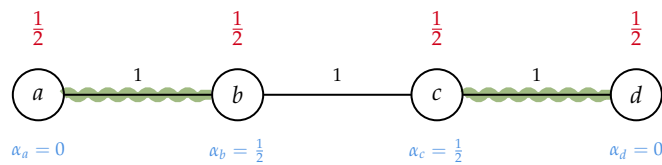
Example:



Find  $x$  in the core of the matching game. Need maximum matching. Use LP to solve for  $x$  (or try values).  $x_u + x_v \geq w_{uv} \quad \forall uv \in E(G)$   
Check: This is stable.

Recall in a stable outcome,  $x_u \geq \alpha_u$ . A stable outcome might not reflect real life experimental results.

Example:



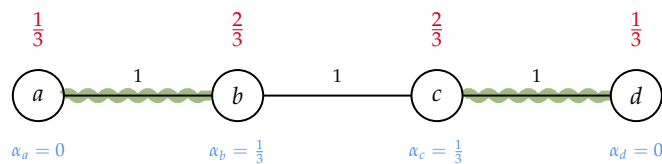
This is a stable outcome. In experiments,  $b$  and  $c$  tend to get more than  $a$  and  $d$ .

*Idea from Nash's bargaining solution:* In a matched pair, each player take at least their outside option and split the rest. In other words, after deducting their outside options, they are equal.

### balanced outcome

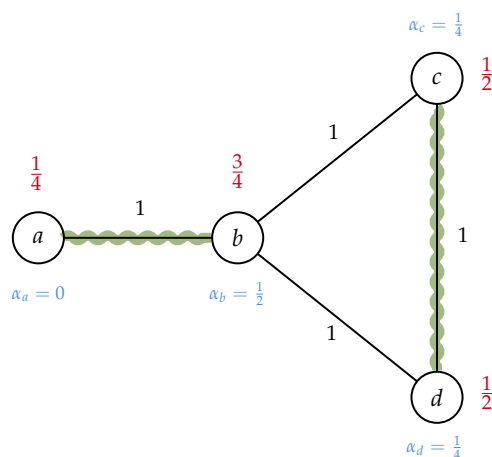
An outcome  $(M, x)$  is **balanced** if  $x_u - \alpha_u = x_v - \alpha_v \geq 0$  for all  $uv \in M$ .

Example:



This is a balanced outcome.

Example:



Balanced.  
Player  $b$  is in a more powerful position.

## Summary of results

### Theorem 3.16

A network bargaining game has a balanced outcome if and only if it has a stable outcome.

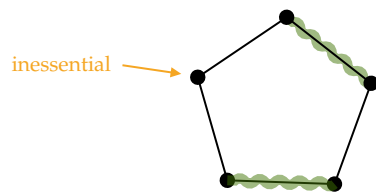
Note:

It is not hard to prove that balanced outcomes are stable.

### inessential vertex

A vertex  $v$  in a graph  $G$  is **inessential** if there exists a maximum matching in  $G$  that does not use  $v$ .

Example:



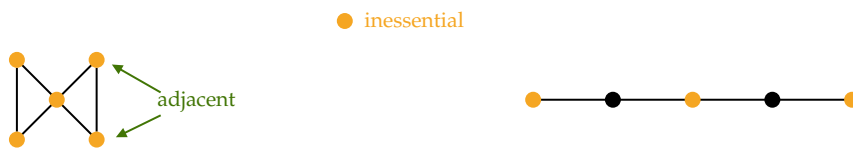
Maximum matching has size 2.  
By symmetry, every vertex is inessential.

### Theorem 3.17: Kleinberg, Tardos

In a graph  $G$  with weight 1 for each edge, the following are equivalent:

1. There is a stable outcome.
2. There is a balanced outcome.
3. No two inessential vertices are adjacent.

Example:



No balanced outcomes exist

Balanced outcomes exist

*Aside:* The set of inessential vertices is the set  $D$  in the Gallai-Edmonds structure theorem. (3) means  $D$  is independent.

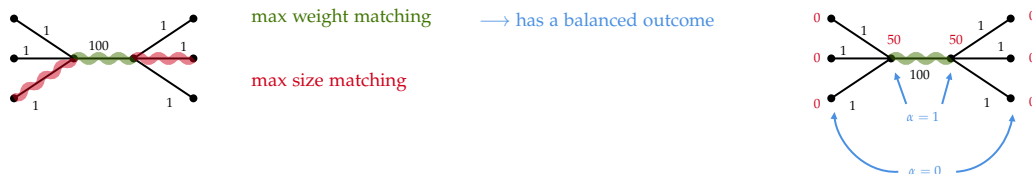
For general weights...

### Theorem 3.18: Kleinberg, Tardos

If a network bargaining game has a balanced outcome, then the balanced outcome must use the maximum weight matching, and there exists a polynomial time algorithm for finding a balanced outcome.

Note:

Maximum weight matching might not be a matching with maximum number of edges.



# Mechanism Design

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*Mechanism design:* We want to design games so that players are incentivized to achieve certain overall goals.

## Example:

1. Elections: If more people prefer  $A$  over  $B$ , then  $A$  should win.
2. Auctions: The player who values the item the most should win. Or, maximize revenue for the seller.
3. Sports tournaments: best team should win, teams play the best they can.

*Problem:* Players look after themselves, will find loop holes and advantages to avoid playing as the game designer intended.

*Main question:* How to set rules so that players who play for themselves will also achieve these global goals?

## 4.1 Ideal auctions

In an auction, the auctioneer is selling something. players bid on the item. Auctioneer need to decide the rules for two things: 1. who wins what item; 2. who pays what amount.

Second price auction: 1. Player who bids most wins; 2. Winner pays 2nd highest bid.

Nice properties of second-price auctions:

- Recall: A player bidding their valuation is a dominant strategy. “Truthful bidding.” This is easy to play: even without knowing everyone else’s valuation, bidding one’s own valuation will maximize utility. (As an exercise, truthful bidding is not a dominant strategy for the first-price auction.)
- If players give truthful bids. then their utility is never negative. (This is not true for “all-pay” auction, where every player pays regardless if they win or lose.)

### dominant-strategy incentive-compatible

An auction is **dominant-strategy incentive-compatible**(DSIC) if truthful bidding is a dominant strategy, and yields non-negative utility.



**Note:**

Dominant strategy here means “weakly dominant” except we don’t require a case that decrease utility.

Having DSIC is not good enough. For example, an auction where we give the item for free to player 1 is DSIC. We need an overall goal: the item should go to the player with the highest valuation. Recall the “social welfare” is the sum total of values received by all players.

For one-item auction, only one player receives the item. If player  $i$  wins the auction, then they receive value  $v_i$  (their valuation). Remaining players receive nothing. So the social welfare is  $v_i$ .

To maximize social welfare in this case, we make sure that the player with the highest valuation wins. This is easy if players bid truthfully.

**welfare-maximizing**

An auction is **welfare-maximizing** if truthful bidding results in maximum social welfare.

*Last good property:* second-price auction is easy to run. We can quickly determine the highest bidder and the 2nd highest bid.

**ideal auction**

An auction is **ideal** if

1. It is DSIC;
2. It is welfare-maximizing; and
3. It can be implemented efficiently.

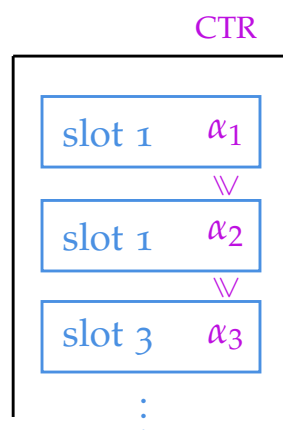
**Theorem 4.1**

The second-price auction is ideal.

*Result from the future:* This is the only single-item auction that is ideal.

## 4.2 Sponsored search auctions

Search engines may show ads as the first entries of a search result. Selecting which ads to show their order of placement is done through auctions.



**Model**

Suppose there are  $k \geq 1$  slots for sponsored links on a search results page. There are advertisers  $N$  bidding for these slots with related keywords. The slots have different “values”: top ones tend to get clicked more often. “click through rate” CTR. For each slot  $j$ , let  $\alpha_j \in [0, 1]$  be the probability that a user clicks on an ad at slot  $j$ .

*Assumption 1:*  $\alpha_1 \geq \alpha_2 \geq \alpha_3 \geq \dots$

*Assumption 2:* CTR is independent of the quality of the ad.

We are looking for ideal auctions: DSIC, welfare-maximizing, efficient. Need to determine who wins what, and who pays what. General approach:

1. Assume truthful bidding, how can we assign the items that is welfare-maximizing and efficient?
2. Given the way we assign items, how can we set prices so that it is DSIC?

**Social welfare**

We want to maximize the overall value for the players. What is the value to a player? Each player  $i$  has a valuation  $v_i$  of how much on their ad is worth. If they are assigned a slot with CTR  $\alpha_i$ , then their expected value is  $v_i \alpha_i$ . Social welfare is  $\sum_{i \in N} v_i \alpha_i$ .

**Example:**

2 slots, 3 players. CTR:  $\alpha_1 = 0.7, \alpha_2 = 0.5$ . Player valuations:  $v_1 = 10, v_2 = 9, v_3 = 2$ . If we assign slot 1 to player 3 and slot 2 to player 1, then the social welfare is  $10 \cdot 0.5 + 0 + 2 \cdot 0.7 = 6.4$

What is a rule that maximizes social welfare? Assign slot 1 to the player with the highest valuation, slot 2 to the 2nd highest, etc. This can be efficiently implemented. This resolves (1) above.

**Note:**

Efficiency is critical. Lots of sponsored search auctions run at the same time. Has to be almost instantaneous.

**Payment rule**

We want a payment rule that is DSIC, so players are incentivized to bid truthfully. This requires Myerson's Lemma.

*Aside:* Consider the generalized second-price auction. The player who wins the  $j$ -th slot will pay the  $(j + 1)$ -st highest bid times their CTR  $\alpha_j$ . This is not DSIC.

In our example above, player 1 gets CTR  $\alpha_1$ , and gets value  $v_1 \alpha_1 = 7$ . Their payment is  $v_2 \alpha_1 = 6.3$ . Utility = 0.7. If player 1 bids 8 instead, then they get CTR  $\alpha_2$ , value  $v_1 \alpha_2 = 5$ . Their payment is  $v_3 \alpha_2 = 1$ . Utility = 4. Truthful bidding is not a dominant strategy.

**4.3 Myerson's Lemma**

Myerson's Lemma characterizes auctions that have DSIC payment rule, and gives a formula for the payment rule. First need an abstraction of the auction model.

### single-parameter environment

In a **single-parameter environment**, we are given

- a set of players  $N$ ;
- a private valuation  $v_i \geq 0$  for each  $i \in N$ , which is player  $i$ 's valuation for one unit of goods; and
- a set  $X \subseteq \mathbb{R}^N$  of vectors  $(x_1, \dots, x_n)$  that describe feasible allocations, i.e.,  $x_i$  is the amount of goods given to player  $i$ .

#### Note:

This is “single-parameter” since each player has only one piece of private information.

#### Example:

In a single-item auction,  $X$  is the set of all standard basis vectors in  $\mathbb{R}^n$  ( $n$ -tuples with one 1 and  $n - 1$  0's). In the sponsored search auction,  $X$  is the set of vectors where for each slot  $j$ , at most one  $i \in N$  satisfies  $x_i = \alpha_j$ , 0 otherwise.

We are interested in mechanisms in this environment, i.e., rules of the auction. Let  $B \subseteq \mathbb{R}^N$  be the set of all possible player bids.

### direct revelation mechanism

Given a single-parameter environment, a **direct revelation mechanism**...

- collects bids  $b \in B$  from all players;
- choose a feasible allocation  $x(b) \in X$  based on the bids; and
- choose a payment  $p(b) \in \mathbb{R}^N$  where player  $i$  pays  $p_i(b)$ .

We call  $x : B \rightarrow X$  an **allocation rule**, and  $p : B \rightarrow \mathbb{R}^N$  a **payment rule**. The utility of player  $i$  is  $u_i(b) = v_i x_i(b) - p_i(b)$ .

As part of the DSIC condition, we assume  $0 \leq p_i(b) \leq v_i x_i(b)$ , i.e., truthful bidding will result in non-negative utility.

#### Example:

For the second-price auction,

$$x_i(b) = \begin{cases} 1 & b_i \text{ is max in } b \\ 0 & \text{otherwise} \end{cases} \quad p_i(b) = \begin{cases} \max_{j \neq i} b_j & b_i \text{ is max in } b \\ 0 & \text{otherwise} \end{cases}$$

### Two terms for an allocation rule

1. An allocation rule  $x : B \rightarrow X$  is **implementable** if there exists a payment rule  $p : B \rightarrow \mathbb{R}^N$  such that  $(x, p)$  is DSIC. The single-item auction where we give the item to the highest bidder is implementable, by using the second-price rule as payment.
2. An allocation rule  $x : B \rightarrow X$  is **monotone** if  $\forall i \in N$  and  $\forall b_{-i} \in B_{-i}$ ,  $x_i(z, b_{-i}) \geq x_i(y, b_{-i})$  whenever  $z \geq y$ . Translation: The higher you bid, the more you get. For single-item auctions, highest bid wins is monotone, but second highest bid wins is not monotone (you can lose things by bidding higher).

**Theorem 4.2: Myerson's Lemma**

For a single parameter environment, an allocation rule  $x : B \rightarrow X$  is implementable if and only if  $x$  is monotone. Moreover, given a monotone allocation rule  $x : B \rightarrow X$ , there exists a unique payment rule  $p : B \rightarrow \mathbb{R}^N$  such that  $(x, p)$  is DSIC and  $p_i(b) = 0$  whenever  $b_i = 0$ .

**Lemma 4.3**

If an allocation rule  $x : B \rightarrow X$  is implementable, then  $x$  is monotone.

**Proof:**

Suppose  $x$  is implementable. Then there exists a payment rule  $p : B \rightarrow \mathbb{R}^n$  such that  $(x, p)$  is DSIC.

Fix a player  $i \in N$ , and bids  $b_{-i} \in B_{-i}$ . We use  $x(y), p(y)$  to represent  $x_i(y, b_{-i})$  and  $p_i(y, b_{-i})$ . Want to prove: if  $y \geq z$ , then  $x(y) \geq x(z)$ . Assume  $y \geq z$ .

Key observation: the rules  $(x, p)$  are defined independent of the players' valuations.

1.  $(x, p)$  is DSIC if player  $i$ 's valuation is  $y$ . So utility of bidding  $y \geq$  utility of bidding  $z$ , which implies

$$y \cdot x(y) - p(y) \geq y \cdot x(z) - p(z)$$

2.  $(x, p)$  is DSIC if player  $i$ 's valuation is  $z$ , which implies

$$z \cdot x(z) - p(z) \geq z \cdot x(y) - p(y)$$

Rearranging the inequalities to get

$$z(x(y) - x(z)) \leq p(y) - p(z) \leq y(x(y) - x(z))$$

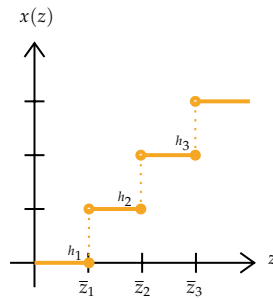
which is called the **payment sandwich**.

Since  $y \geq z \geq 0$ , we get  $x(y) - x(z) \geq 0$ . So  $x(y) \geq x(z)$ . □

For the converse, we will only prove the case where  $x$  is piecewise constant.

*Approach:* Assume  $x$  is monotone and  $(x, p)$  is DSIC. Derive that  $p$  is unique with  $p_i(b) = 0$  whenever  $b_i = 0$ . Then prove that using this particular  $p$ ,  $(x, p)$  is indeed DSIC.

We will now describe a generic  $x$  that is monotone and piecewise constant. There are jump points  $\bar{x}_1 < \bar{x}_2 < \dots < \bar{x}_q$  such that  $x$  is constant in the intervals  $[0, \bar{x}_1], (\bar{x}_1, \bar{x}_2], \dots, (\bar{x}_{q-1}, \bar{x}_q], (\bar{x}_q, \infty)$ . Define  $h_j$  to be the jump at  $\bar{x}_j$ . (i.e.,  $h_j = \lim_{w \rightarrow \bar{x}_j^+} x(w) - x(\bar{x}_j)$ )



**Lemma 4.4**

If  $x$  is monotone and piecewise constant, and  $(x, p)$  is DSIC with  $p_i(b) = 0$  whenever  $b_i = 0$ , then  $p$  is unique.

**Proof:**

We first show that  $p$  is also piecewise constant. (If  $z$  is inside an interval, then  $p'(z) = 0$ .)

Assume  $y > z$ . Suppose  $z \in (\bar{z}_j, \bar{z}_{j+1})$  for some  $j$ . We assume DSIC, so the payment sandwich applies. Divide by  $y - z$  to get

$$z \cdot \frac{x(y) - x(z)}{y - z} \leq \frac{p(y) - p(z)}{y - z} \leq y \cdot \frac{x(y) - x(z)}{y - z}$$

Take the limit as  $y \rightarrow z^+$ . Then

$$\lim_{y \rightarrow z^+} \frac{x(y) - x(z)}{y - z} = x'(z) = 0$$

since  $x$  is constant at  $z$ . So both ends of inequalities approach 0. By Squeeze Theorem,

$$\lim_{y \rightarrow z^+} \frac{p(y) - p(z)}{y - z} = p'(z) = 0$$

So  $p$  is piecewise constant with jumps at  $\bar{z}_1, \dots, \bar{z}_q$ . Let  $z = \bar{z}_j$  for some  $j$ . Look at the payment sandwich:

$$\lim_{y \rightarrow z^+} y(x(y) - x(z)) = z \cdot h_j, \quad \lim_{y \rightarrow z^+} z(x(y) - x(z)) = z \cdot h_j$$

By Squeeze Theorem,

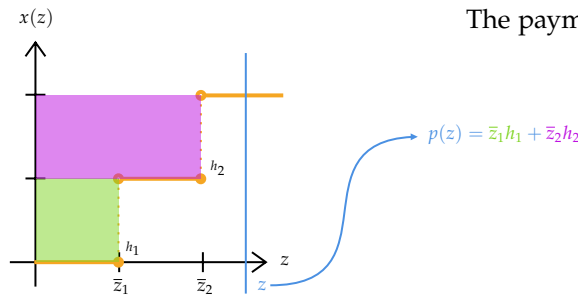
$$\lim_{y \rightarrow z^+} p(y) - p(z) = z \cdot h_j = \bar{z}_j h_j$$

So the payment jump at  $\bar{z}_j$  is  $\bar{z}_j h_j$ .

By assumption,  $p(0) = 0$ , so this uniquely defines the function  $p$ : If  $j$  is then largest index where

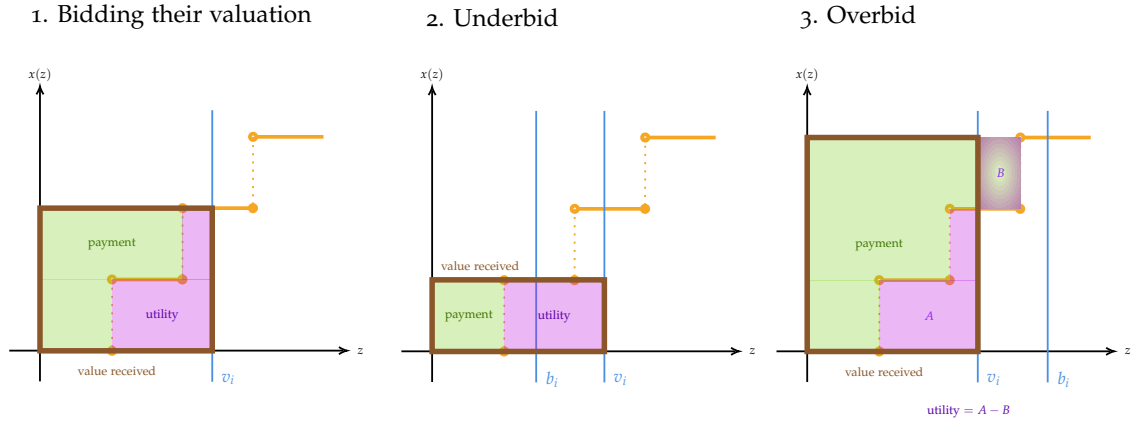
$$z \geq \bar{z}_j, \text{ the } p(z) = p_i(z, b_{-i}) = \sum_{k=1}^j \bar{z}_k h_k. \quad \square$$

Visualization:

**Lemma 4.5**

The payment rule  $p$  from Lemma 4.4 is DSIC.

**Proof by picture:**



The utilities for (2) and (3) are at most the utility of (1). Utility of (1) is non-negative. Hence DSIC.  $\square$

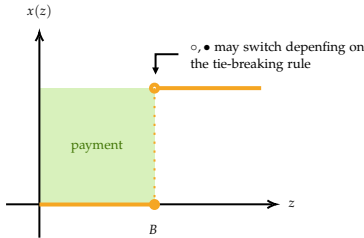
## Applying Myerson's Lemma

Recall: Payment rule:  $p_i(z, b_{-i}) = \sum_{k=1}^j \bar{z}_k h_k$ .

### 1. Single item auction

Let  $x$  be the allocation rule that gives the item to the higher bidder (with a consistent tie-breaking rule). This is monotone. Consider a player  $i$  and bids  $b_{-i} \in B_{-i}$ . Let  $B = \max_{j \neq i} b_j$ .

The allocation function for player  $i$ :



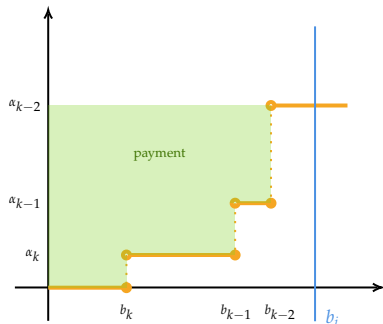
There is one jump at  $z = B$ . When  $b_i > B$ , player  $i$  wins, and the payment is  $B \cdot 1 = B$ . This is the second-price rule, and it is unique assuming  $b_i = 0$  implies  $p_i = 0$ .

### 2. Sponsored search auction

$k$  slots with CTR  $\alpha_1 \geq \alpha_2 \geq \dots \geq \alpha_k$ . Our allocation function  $x$  assigns  $\alpha_j$  to the  $j$ -th highest bidder. This is monotone. Apply Myerson's lemma for the payment rule.

Consider a player  $i$  and bids  $b_{-i} \in B_{-i}$ . Assume  $b_1 \geq b_2 \geq \dots \geq b_k$  and we ignore the remaining bid.

Player  $i$  gets  $\alpha_k$  if  $b_k < b_i \leq b_{k-1}$ ,  $\alpha_{k-1}$  if  $b_{k-1} < b_i \leq b_{k-2}$ ,  $\dots$



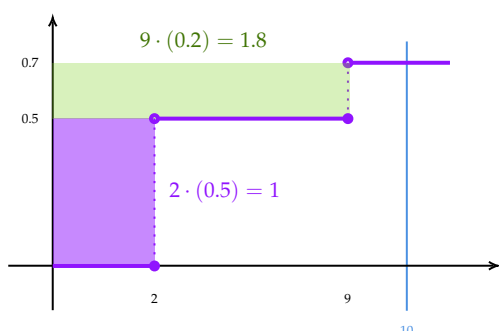
The height of each jump is  $\alpha_{j-1} - \alpha_j$ . Using the formula,  $p_i(z) = \sum_{\{j: b_j < z\}} b_j(\alpha_j - \alpha_{j+1})$ , where we are using  $\alpha_{k+1} = 0$

This is the unique DSIC payment rule.

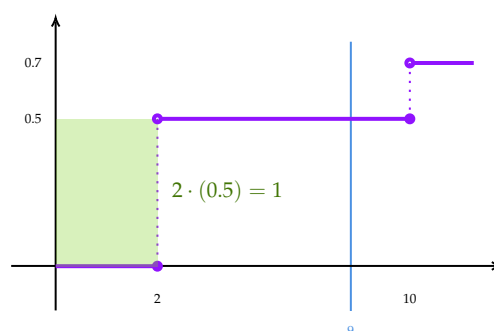
**Example:**

2 slots  $\alpha_1 = 0.7, \alpha_2 = 0.5$ . 3 bidders:  $v_1 = 10, v_2 = 9, v_3 = 2$ .

Allocation for player 1: Payment = 2.8



Allocation for player 2: Payment = 1

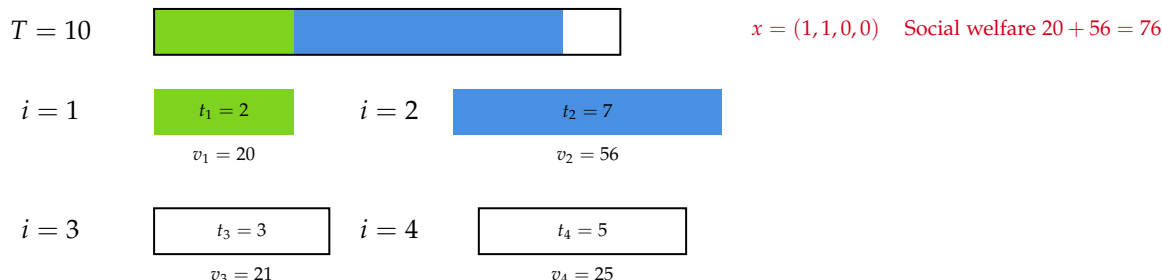


## 4.4 Knapsack auction

Recall that an ideal auction is (1) DSIC (2) welfare-maximizing (3) efficient. We consider an auction that cannot be ideal.

**Example:**

We are managing advertising slots for a TV station. We have  $T$  seconds to fill and there are potential advisers  $N$ . Advertiser  $i$  has an ad of length  $t_i$  and gets value of  $v_i$ . We run an auction to determine whose ads to air.



The set of feasible allocations:  $X = \left\{ x \in \{0,1\}^N : \sum_{i \in N} t_i x_i \leq T \right\}$ . Social welfare is  $\sum_{i \in N} v_i x_i$

To maximize social welfare, we need  $\max\{v^T x : x \in X\}$ .

Recall the first stage of designing this auction is to assume players bid truthfully and find an allocation rule that is welfare-maximizing. Then Myerson's Lemma gives us a payment rule that is DSIC.

The problem is that maximizing social welfare here is equivalent to the knapsack problem, which is NP-hard, so it is not possible to find an optimal solution efficiently (unless  $P = NP$ ). Thus we cannot fulfill (2) and (3) of ideal auction simultaneously.

The solution is to relax (2) or (3). We should not give up (3) because we want this to be practical and run in real time. So we need to give up on (2). Instead of finding a welfare-maximizing allocation exactly, we will find an approximation that is "close" to the maximum while keeping efficiency. Myerson's Lemma does not require social welfare maximization so as long as our approximation is good enough and monotone we can apply Myerson's Lemma to get a DSIC payment rule with respect to our approximation.

**Goal:** Approximate the welfare-maximizing function for the knapsack auction.

**Aside:** Typically with approximation algorithms, we want some performance guarantee. "The solution

produced by the algorithm must be at least this close to the theoretical optimal."

**Example:**

$i$	1	2	3	4
$t_i$	2	7	3	5
$v_i$	20	56	21	25
$v_i/t_i$	10	8	7	5

$T = 10$

*Idea:* we have limited space, so we want to fill in each unit with as much value as possible  $\Rightarrow$  Calculate value density. Higher density should give better value. We sort items in non-increasing value density, i.e.,  $\frac{v_1}{t_1} \geq \frac{v_2}{t_2} \geq \dots \frac{v_n}{t_n}$

**First try:** Pick items 1, 2, ... until we cannot put the next one in the knapsack. Consider the example above, we pick 1 and 2.  $t_1 + t_2 = 9 \leq 10$ , no room for 3. Total value 76. Is this a good approximation? No, there are cases where the solution produced is very bad...

**Example:**

2 players,  $v_1 = t_1 = 1$ ,  $v_2 = T - 1$ ,  $t_2 = T$ .

1 has higher density than 2. If we pick 1, then we do not have room for 2. Algorithm value 1. However, obviously picking 2 is optimal with optimal value  $T - 1$ . Our solution is  $\frac{1}{T-1}$  of the optimal.

**Better idea:** Suppose we stop at item  $i$  above. Check the value of item  $i + 1$ . If  $v_{i+1} > v_1 + \dots + v_i$ , then we will say that  $i + 1$  is the only winner. (Assumption:  $t_i \leq T$  for all  $i$ )

**Example:**

Consider 2-player example above. Stop at  $i = 1$ . We see  $v_2 > v_1$ , so we pick  $v_2$  instead.

---

**Algorithm 1:** Full approximation algorithm

---

**Assumption:**  $N = \{1, \dots, n\}$  with  $\frac{v_1}{t_1} \geq \frac{v_2}{t_2} \geq \dots \frac{v_n}{t_n}$

- 1 Find  $i$  such that it is the largest index with  $t_1 + \dots + t_i \leq T$
  - 2 If  $v_1 + \dots + v_i \geq v_{i+1}$ , then  $\{1, \dots, i\}$  are the winners. Otherwise,  $\{i + 1\}$  is the only winner.
- 

Suppose APX is the value of the winners produced by the approximation algorithm, OPT is the actual optimal value.

**Theorem 4.6**

$$\text{APX} \geq \frac{1}{2} \text{OPT}$$

The solution produced by the approximation algorithm is guaranteed to be at least halfway to the optimal.

**Proof:**

A welfare-maximizing solution is optimal for this integer program:

$$\max\{v^T x : t^T x \leq T, x_i \in \{0, 1\} \ \forall i\} \quad (\text{IP})$$

Its LP relaxation:

$$\max\{v^T x : t^T x \leq T, x_i \in [0, 1] \ \forall i\} \quad (\text{LP})$$

Solutions of it can be fractional.

Let  $i$  be the largest index with  $t_1 + \dots + t_i \leq T$ . Consider the solution  $x$  where

$$x_1 = \dots = x_i = 1, \quad x_{i+1} = \frac{T - t_1 - \dots - t_i}{t_{i+1}}, \quad x_{i+1} = \dots = x_n = 0$$



Idea: Fill in the remaining space with a fraction of item  $i + 1$ . In example above, we have  $x_1 = 1, x_2 = 1, x_3 = \frac{1}{3}, x_4 = 0$

As we have completely filled in the knapsack with the highest-density items,  $x$  is optimal, which can be proved using duality.

Suppose optimal value of (LP) is  $v^*$  (obtained by  $x$ ). Since (LP) is relaxation of (IP),  $\text{OPT} \leq v^*$ . We want to prove  $\text{APX} \geq \frac{v^*}{2}$ .

Let  $S = \{1, \dots, i\}$ . We see that  $v^* \leq v_1 + \dots + v_i + v_{i+1} = v(S) + v_{i+1}$ . Two cases:

1. If  $v(S) \geq v_{i+1}$ , then the algorithm picks  $S$ , so  $\text{APX} = v(S)$ . Then

$$v^* \leq v(S) + v_{i+1} \leq v(S) + v(S) = 2 \text{APX}$$

2. If  $v(S) < v_{i+1}$ , then the algorithm picks  $\{i + 1\}$ , so  $\text{APX} = v_{i+1}$ . Then

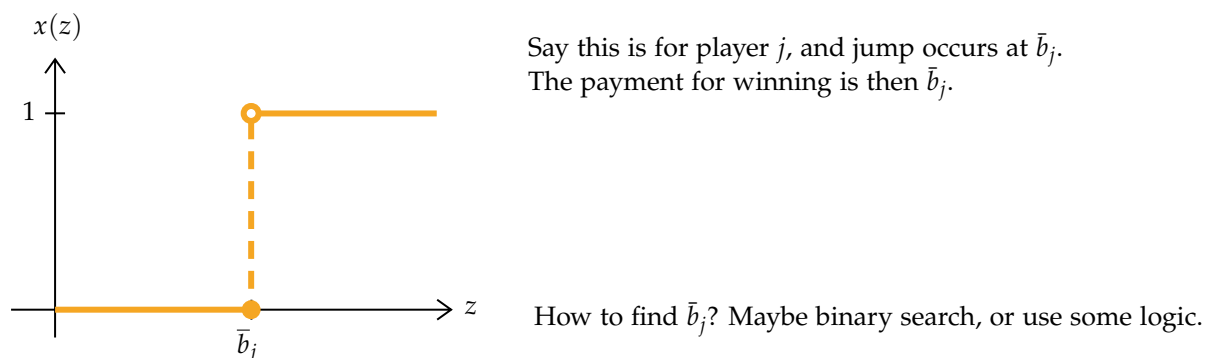
$$v^* \leq v(S) + v_{i+1} < 2v_{i+1} = 2 \text{APX}$$

So  $\text{APX} \geq \frac{v^*}{2} \geq \frac{1}{2} \text{OPT}$ . □

#### Exercise:

Prove that the allocation rule derived from the approximation algorithm is monotone.

Myerson's lemma gives a payment rule for our allocation. A player either wins or not, so the allocation functions looks like:



#### Example:

$i$	1	2	3	4
$t_i$	2	7	3	5
$v_i$	20	56	21	25
$v_i/t_i$	10	8	7	5

$$T = 10$$

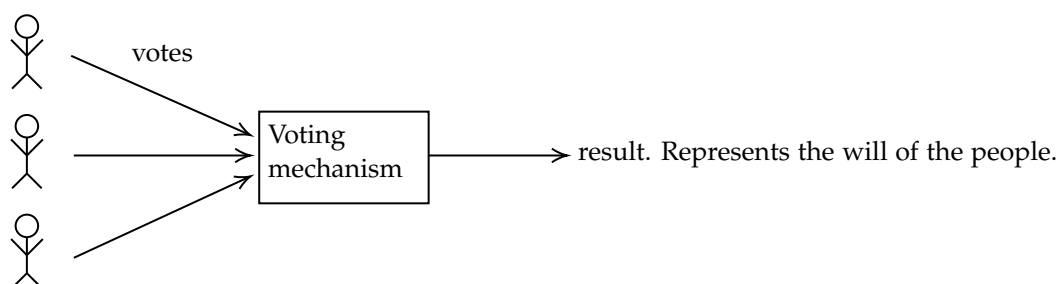
Approximation algorithm picks 1 and 2 as winners.

What is the payment for player 2? The allocation is monotone, so to find  $\bar{b}_2$ , we need to lower  $v_2$  until the approximation algorithm does not pick player 2.

As long as  $\frac{v_2}{t_2} \geq 7$ , the algorithm will pick 2.  
So player 2 can go as low as 49.

What happens if  $v_2$  goes below 49? Say it has value density  $\geq 5$ . The approximation algorithm will pick 1 and 3 with combined value 41, then look at  $v_2$ . If  $v_2 > 41$ , then the algorithm will still pick 2. If  $v_2 \leq 41$ , then the algorithm will stick with 1, 3. So  $\bar{b}_2 = 41$ , and the payment is 41.

## 4.5 Voting mechanism



If there are 2 candidates, this is easy: decide on the majority. Reflects the preferences of the people.

### Condorcet's paradox

3 candidates  $a, b, c$ .

3 voters: voter 1:  $a >_1 b >_1 c$

voter 2:  $b >_2 c >_2 a$

voter 3:  $c >_3 a >_3 b$

A majority of voters prefer  $a$  over  $b$ ,  $b$  over  $c$ ,  $c$  over  $a$ , which implies  $a > b > c > a$ , not possible.

If we only choose 1 candidate, then a majority wants a different one.

This could lead to "strategic voting": a voter may be incentivized to not vote for their preferred candidate.

*Example:*  $a, b$  are front runners, more prefer  $a$  over  $b$ . A group of voters prefer  $c$ , who has no chance of winning. But they would rather  $b$  win than  $a$ , so they vote for  $b$  instead.

*Problem:* The usual voting mechanism does not incentivize voters to vote truthfully. So the votes that are submitted do not represent the will of the people.

*Question:* Is there a voting mechanism where truthful voting is a dominant strategy?

*Answer:* No, if we want our voting mechanism to have obvious good properties.

### Social welfare and social choice

#### Set up

Candidates  $A$ , voters  $\{1, \dots, n\}$ . Let  $L$  be the set of all total orders<sup>a</sup> on  $A$ . Each voter  $ii$  has a total order of preferences  $<_i \in L$ .  $a <_i b$  means voter  $i$  prefers  $b$  over  $a$ .

A collection of all voter preferences  $(<_1, \dots, <_n) \in L^n$ .

<sup>a</sup>antisymmetric:  $a \neq b \Rightarrow a > b \vee b > a$ . transitive:  $a > b \wedge b > c \Rightarrow a > c$

Given a collection of voter preferences (how voters would vote), there are 2 outcomes that may be generated by a voting mechanism:

1. A **social welfare function** will output another total order on  $A$  that represents "society's preference."  $F : L^n \rightarrow L$ .
2. A **social choice function** will output one candidate from  $A$  that represents "society's choice."  $F : L^n \rightarrow A$ .

## Properties of a social welfare function

Let  $F : L^n \rightarrow L, (\prec_1, \dots, \prec_n) \in L^n$ . Suppose  $F(\prec_1, \dots, \prec_n) = \prec$ .

**Good property 1: Unanimity.** If every voter prefers  $a$  over  $b$ , then the society prefers  $a$  over  $b$ . (If  $b \prec_i a \ \forall i \in N$ , then  $b \prec a$  in  $F$ .)

**Good property 2: Independence of irrelevant alternatives (IIA).** The society's preference between two candidates should only depend on the individual votes' preference of these two candidates, not on others.

Suppose  $(\prec_1, \dots, \prec_n), (\prec'_1, \dots, \prec'_n) \in L^n, F(\prec_1, \dots, \prec_n) = \prec, F(\prec'_1, \dots, \prec'_n) = \prec'$ . For two candidates  $a, b \in A, a \prec_i b \Leftrightarrow a \prec'_i b$  for each  $i \in N$  (the relative rank of  $a, b$  doesn't change between  $\prec_i$  and  $\prec'_i$ ) implies  $a \prec b \Leftrightarrow a \prec' b$  (the relative rank of  $a \prec b$  for  $F$  does not change from  $\prec$  to  $\prec'$ )

*Flowed illustration of IIA:* Suppose you can choose between studying game theory and cryptography. Of course you would choose game theory. Now you have the additional option of studying Galois theory, and you decide to study cryptography instead. Should not happen.

**Bad property: Dictatorship.** The society's preferences completely agrees with one particular voter's preferences in all cases. For all  $(\prec_1, \dots, \prec_n) \in L^n, F(\prec_1, \dots, \prec_n) = \prec_i$ .

## 4.6 Arrow's impossibility theorem

### Theorem 4.7: Arrow's impossibility theorem

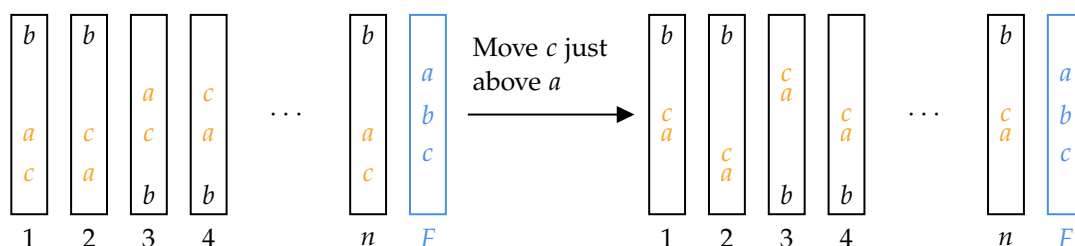
Every social welfare function over a set of at least 3 candidates that satisfies unanimity and IIA is a dictatorship.

If we want the good properties, we must also have the bad property. No way to design a good voting mechanism.

#### Proof sketch:

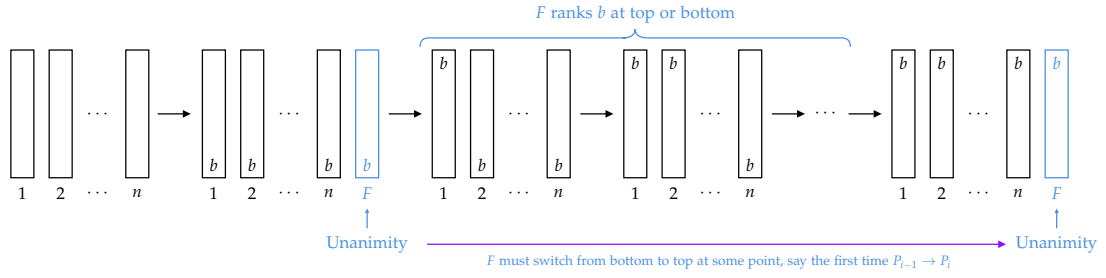
Suppose  $F : L^n \rightarrow L$  satisfies unanimity and IIA.

1. We prove that given a candidate  $b$ , if each voter ranks  $b$  at the top or bottom of their list, then  $F$  also ranks  $b$  at the top or the bottom. Suppose otherwise, and there exist  $a, c \in A$  such that  $F$  ranks  $c \prec b \prec a$ .

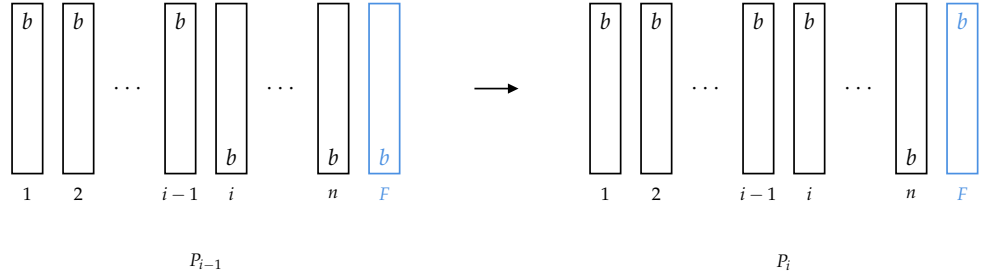


Relative rankings of  $a, b$  do not change. By IIA, relative rankings of  $a, b$  do not change in  $F$ . Similar for the relative rankings of  $b, c$ . So  $F$  keeps  $a \succ b \succ c$  in the new preferences. But  $c$  is higher than  $a$  in all voters' lists, so by unanimity,  $c \succ a$  in  $F$ . Contradicts transitivity.

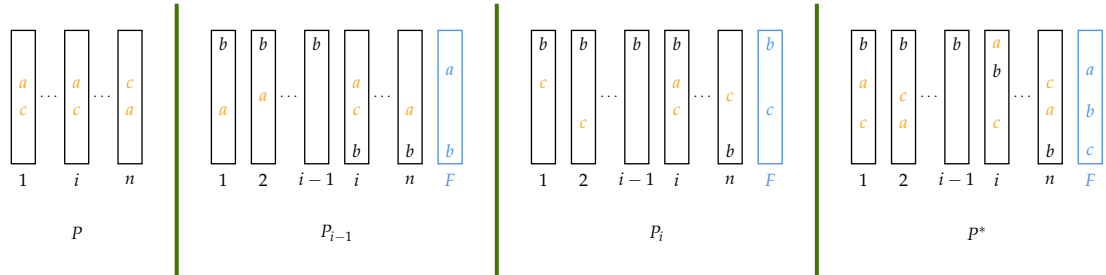
2. Find a potential voter that can be a dictator. Take an arbitrary voter preference  $P \in L^n$ . Create a sequence of preferences  $P_0, P_1, \dots, P_n$  by first moving  $b$  to the bottom, then move  $b$  to the top one voter one at a time.



Possible dictator voter  $i$ :



3. Let  $a, c$  be two candidates other than  $b$ . (They exist since there are at least 3 candidates.) We will show that  $F$  agrees with voter  $i$  on the relative ranking of  $a, c$ . WLOG voter  $i$  ranks  $a$  above  $c$  in the original  $P$ . Create a new voter reference  $P^*$  from  $P_i$  by moving  $a$  to the top of the list for voter  $i$ .



Note: Between  $P$  and  $P^*$ , the relative rankings of  $a, c$  do not change. By IIA,  $F$  ranks  $a, c$  the same in  $P$  and  $P^*$ .

Goal:  $F$  ranks  $a$  above  $c$  in  $P^*$  (agreeing with voter  $i$ ).

Compare  $P_{i-1}$  and  $P^*$ . The relative rankings of  $a, b$  do not change. By IIA,  $F$  ranks  $a, b$  the same in  $P_{i-1}$  and  $P^*$ ,  $F$  ranks  $b$  at the bottom in  $P_{i-1}$ , so  $a > b$  for both  $P_{i-1}$  and  $P^*$ .

Compare  $P_i$  and  $P^*$ . The relative rankings of  $b, c$  do not change. By IIA,  $F$  ranks  $b, c$  the same in  $P_i$  and  $P^*$ .  $F$  ranks  $b$  at the top in  $P_i$ , so  $b > c$  for both  $P_i$  and  $P^*$ .

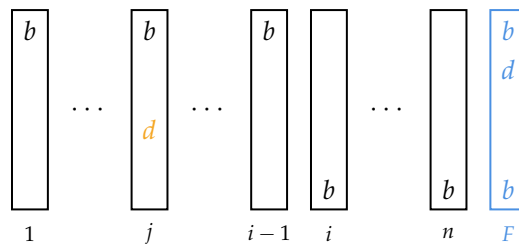
By transitivity,  $F$  ranks  $a > c$ , which matches the ranking for voter  $i$ .

So  $F$  agrees with voter  $i$  on the relative rankings of any two candidates other than  $b$ .

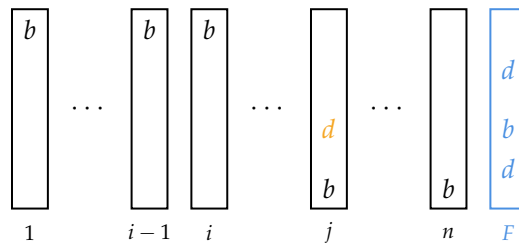
4. Let  $d$  be any candidate other than  $b$ . We will show that  $F$  agrees with voter  $i$  on  $b, d$ . There exists candidate  $e$  that is not  $b, d$  (recall we have  $\geq 3$  candidates).

Run the arguments from (2) and (3) with  $e$  instead of  $b$ . This gives us a voter  $j$  such that  $F$  agrees with voter  $j$  for any two candidates that are not  $e$ . We show that  $i = j$ .

Suppose otherwise. Say  $j < i$ . Consider  $P_{i-1}$ .  $F$  ranks  $b$  at the bottom, so  $d > b$ . For voter  $j$ ,  $d$  is lower than  $b$ . And  $F$  must agree with voter  $j$  in all cases not involving  $e$ . So  $F$  ranks  $b > d$ . This contradicts antisymmetric property of  $F$ .



Suppose  $j > i$ . Consider  $P_i$ .  $F$  ranks  $b > d$ . But voter  $j$  ranks  $d$  over  $b$ . And  $F$  agrees with voter  $j$ . So  $d > b$  for  $F$ , contradiction.



So  $i = j$ .

$F$  agrees with voter  $i$  on the relative rankings of any two candidates. Hence voter  $i$  is a dictator.  $\square$

## 4.7 Gibbard–Satterthwaite theorem

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