### CMSC 474, Game Theory

#### 2. Analyzing Normal-Form Games

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Chapter 2 of the textbook, plus several related topics

# How to reason about games?

- In single-agent decision theory, look at an **optimal** strategy
  - Maximize the agent's expected payoff in its environment
- What is your optimal strategy if you're interacting with other agents?
  - Depends on both your choices and theirs
- Identify certain subsets of outcomes called **solution concepts**
- Chapter 2 discusses two solution concepts:
  - Pareto optimality
  - Nash equilibrium
- Chapter 3 will discuss several others

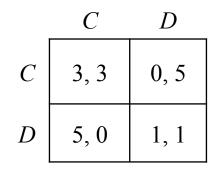
### **Pareto Optimality**

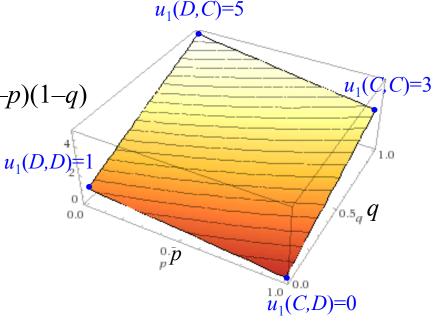
- Vilfredo Pareto (1848–1923)
- A strategy profile **s Pareto dominates** a strategy profile **s'** if
  - rile.,  $u_i(\mathbf{s}) \ge u_i(\mathbf{s}')$  for all i,
  - > at least one agent does better with s than with s', i.e.,  $u_i(s) > u_i(s')$  for at least one i
- s is Pareto optimal (or Pareto efficient) if there's no other strategy profile that Pareto dominates s
  - > Every game has at least one Pareto optimal profile
  - Always at least one Pareto optimal profile in which the strategies are pure

# **Examples**

#### Prisoner's Dilemma

- (D,D) isn't Pareto optimal
  - $\triangleright$  (C,C) Pareto dominates it
- $\bullet$  (*D*, *C*) is Pareto optimal
  - $\triangleright$  For all other strategy profiles,  $u_1$  is lower
  - Let  $s_1 = \{(p, C), (1-p, D)\}$ and  $s_2 = \{(q, C), (1-q, D)\}$
  - $u_1(s_1, s_2) = 3pq + 5(1-p)q + 0 + (1-p)(1-q)$
- (*C*,*D*) is Pareto optimal
  - For other strategy profiles,  $u_2$  is smaller
- (C,C) is Pareto optimal
  - In all other strategy profiles, either  $u_1$  is smaller or  $u_2$  is smaller
- $(s_1, s_2)$  is Pareto optimal for every  $(s_1, s_2)$  except (D,D)





# **Examples**

- Which Side of the Road
  - Pareto optimal: (Left,Left) and (Right,Right)

	Left	Right
Left	1, 1	0, 0
Right	0, 0	1, 1

- In common-payoff games, all Pareto optimal strategy profiles have the same payoffs
  - ➤ If not, the one with lower payoffs wouldn't be Pareto optimal

• **Poll 2.1**: Here's a common-payoff game. Which strategy profiles are Pareto optimal?

	Left	Right
Left	1, 1	2,2
Right	0, 0	2,2

### **Best Response**

- Suppose agent i knows how the others are going to play
- Then *i* has an ordinary optimization problem:
  - Maximize i's expected utility
- We'll use  $\mathbf{s}_{-i}$  to mean a strategy profile for all of the agents except i

$$\mathbf{s}_{-i} = (s_1, ..., s_{i-1}, s_{i+1}, ..., s_n)$$

Notation: if  $s_i$  is a strategy for agent i, then

$$(s_i, \mathbf{s}_{-i}) = (s_1, ..., s_{i-1}, s_i, s_{i+1}, ..., s_n)$$

•  $s_i$  is a **best response** to  $s_{-i}$  if for every strategy  $s_i'$  available to agent i,

$$u_i(s_i, \mathbf{s}_{-i}) \geq u_i(s_i', \mathbf{s}_{-i})$$

There is always at least one best response

# **Examples**

- Poll 2.2: Suppose 1's strategy is Left
  - ➤ What are 2's best responses?

	Left	Right
Left	1, 1	0, 0
Right	0, 0	1, 1

- **Poll 2.3**: Suppose 1's strategy is  $\{(\frac{1}{2}, \text{Left}), (\frac{1}{2}, \text{Right})\}$ 
  - ➤ What are 2's best responses?
- **Poll 2.4**: Suppose 1's strategy is  $\{(^2/_3, \text{Left}), (^1/_3, \text{Right})\}$ 
  - > What are 2's best responses?

# **Announcements (Sept 6)**

- I'm recovering from an illness
- Ali Shafahi's office hours: Wednesdays & Thursdays, 11am–12pm
- What we've covered so far:
  - Pareto dominance, Pareto optimality
    - Prisoner's Dilemma: all strategy profiles are Pareto optimal except (D,D)
    - Common-payoff games: all Pareto optimal strategy profiles have same payoffs

$\triangleright$ Notation: $\mathbf{s}_{-i}$		Left	Right
• $(S_i, \mathbf{S}_{-i})$	Left	1, 1	0, 0
Best response:	Right	0, 0	1, 1

- Given  $\mathbf{s}_{-i}$ , a strategy that maximizes  $u(s_i, \mathbf{s}_{-i})$
- How does *i* know what <sub>-i</sub> is?

### **Best Response**

**Theorem.** Given  $\mathbf{s}_{-i}$ , there are only two possibilities:

- (1) *i* has a pure strategy  $s_i$  that is the *only* best response to  $\mathbf{s}_{-i}$
- (2) *i* has *infinitely many* best responses to  $\mathbf{s}_{-i}$

*Proof strategy*: Show that if (1) is false then (2) must be true.

**Proof.** Let  $s_i$  be a best response to  $\mathbf{s}_{-i}$ , and suppose (1) is false.

- Either  $s_i$  isn't the only best response to  $\mathbf{s}_{-i}$  or  $s_i$  isn't pure
- Case 1:  $s_i$  isn't the only best response to  $\mathbf{s}_{-i}$ 
  - $\triangleright$  The others must have the same expected utility as  $s_i$
  - > Thus every mixture of them is a best response, so (2) holds.
- Case 2:  $s_i$  isn't pure. It's a mixture of at least two pure strategies
  - $\triangleright$  Each of them must have the same expected utility as  $s_i$
  - ➤ They're both best responses, so this reduces to Case 1.

# Nash Equilibrium

- $\mathbf{s} = (s_1, ..., s_n)$  is a **Nash equilibrium** if for every  $i, s_i$  is a best response to  $\mathbf{s}_{-i}$ 
  - > Every agent's strategy is a best response to the other agents' strategies
  - ➤ No agent can do better by *unilaterally* changing his/her strategy
- Theorem (Nash, 1951): Every game with finitely many agents and action profiles has at least one Nash equilibrium
- Prisoner's Dilemma: (*D*,*D*)

	C	D
C	3, 3	0, 5
D	5, 0	1, 1

### Nash Equilibrium

- $\mathbf{s} = (s_1, ..., s_n)$  is a **Nash equilibrium** if for every  $i, s_i$  is a best response to  $\mathbf{s}_{-i}$ 
  - > Every agent's strategy is a best response to the other agents' strategies
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- Theorem (Nash, 1951): Every game with finitely many agents and action profiles has at least one Nash equilibrium
- Prisoner's Dilemma: (*D*,*D*)

	C	D
C	3, 3	0, 5
D	5, 0	1, 1

- Modified Prisoner's Dilemma:
  - (s,D), s is any mixture of D and E

	C	D
C	3, 3	0, 5
D	5, 0	1, 1
E	5, 0	1, 1

# Strict and Weak Nash Equilibria

- Let  $\mathbf{s} = (s_1, \dots, s_n)$  be a Nash equilibrium
  - > s is strict if each  $s_i$  in **s** is the *only* best response to  $\mathbf{s}_{-i}$ 
    - any agent who unilaterally changes strategy will do worse
  - > Otherwise s is weak
- If s includes a mixed\* strategy  $s_i$ , then s is weak
  - $\triangleright$  e.g., (s,D) where  $s = \{(\frac{1}{2}, D), (\frac{1}{2}, E)\}$

• For s to be strict, all strategies in s must be pure
--

	C	D
C	3, 3	0, 5
D	5, 0	1, 1

	C	D
C	3, 3	0, 5
D	5, 0	1, 1
E	5, 0	1, 1

**Poll 2.5**: if all strategies in **s** are pure, is **s** guaranteed to be strict?

# Strict and Weak Nash Equilibria

- Weak Nash equilibria often are less stable than strict Nash equilibria
  - ➤ If s is weak, at least one agent has infinitely many best responses and only one of them is in s
- Example:  $s_1 = s_2 = \{(\frac{1}{2}, \text{Left}), (\frac{1}{2}, \text{Right})\}$ 
  - $\triangleright$  For 2, Left is also a best response to  $s_1$
  - ➤ If 2 unilaterally switches to Left, 1's strategy is no longer a best response

	Left	Right
Left	1, 1	0, 0
Right	0, 0	1, 1

- Difficult in general
  - > Easier if we can identify the support of the equilibrium strategies
- In 2x2 games, it's easy
  - > Each agent has two actions, support must include both of them
- If there's a mixed\*-strategy Nash equilibrium  $(s_1, s_2)$  then
  - $\triangleright$   $s_1$  is a mixture of actions a and a' that have same expected utility given  $s_2$
  - $\triangleright$   $s_2$  is a mixture of actions b and b' that have same expected utility given  $s_1$
- Look for  $s_1$  and  $s_2$  that make those things true
  - > Solve linear equations

#### • Example: Battle of the Sexes

$$> s_1 = \{(p, A), (1-p, B)\}$$

$$> s_2 = \{(q, A), (1-q, B)\}$$

• If 2's strategy isn't pure, 2's actions must have same expected utility

$$u_2(s_1,A) = u_2(s_1,B)$$
  
 $p = 2/3$ 

$$> s_1 = \{(2/3, A), (1/3, B)\}$$

	A	B
A	2,1	0,0
В	0,0	1,2

$$1p + 0(1-p) 0p + 2(1-p)$$
$$p = 2(1-p)$$

#### • Example: Battle of the Sexes

$$> s_1 = \{(p, A), (1-p, B)\}$$

$$> s_2 = \{(q, A), (1-q, B)\}$$

• If 2's strategy isn't pure, 2's actions must have same expected utility

$$u_2(s_1,A) = u_2(s_1,B)$$
  
 $p = 2/3$ 

$$> s_1 = \{(2/3, A), (1/3, B)\}$$

 $\triangleright$  **Poll 2.6**: what is q?

$$\begin{array}{c|cccc}
 A & B \\
 A & 2,1 & 0,0 \\
 B & 0,0 & 1,2
\end{array}$$

$$1p + 0(1-p) 0p + 2(1-p)$$
$$p = 2(1-p)$$

**Example: Battle of the Sexes** 

$$> s_1 = \{(p, A), (1-p, B)\}$$

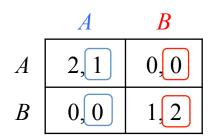
$$> s_2 = \{(q, A), (1-q, B)\}$$

• If 2's strategy isn't pure, 2's actions must have same expected utility

$$u_2(s_1,A) = u_2(s_1,B)$$
  
 $p = 2/3$   
 $s_1 = \{(2/3, A), (1/3, B)\}$ 

• If 1's strategy isn't pure, 1's actions must have same expected utility

$$u_1(A, s_2) = u_1(B, s_2)$$
  
 $y = 1/3$   
 $v = \{(1/3, A), (2/3, B)\}$ 



$$1p + 0(1-p) 0p + 2(1-p)$$
$$p = 2(1-p)$$

• Example: Battle of the Sexes

$$> s_1 = \{(p, A), (1-p, B)\}$$

$$> s_2 = \{(q, A), (1-q, B)\}$$

• If 2's strategy isn't pure, 2's actions must have same expected utility

$$u_2(s_1,A) = u_2(s_1,B)$$
  
 $p = 2/3$   
 $s_1 = \{(2/3, A), (1/3, B)\}$ 

• If 1's strategy isn't pure, 1's actions must have same expected utility

$$u_1(A, s_2) = u_1(B, s_2)$$
  
 $y_1(A, s_2) = u_1(B, s_2)$   
 $y_2(B, s_2) = \{(1/3, A), (2/3, B)\}$ 

$$\begin{array}{c|cccc}
 A & B \\
 A & 2,1 & 0,0 \\
 B & 0,0 & 1,2
\end{array}$$

$$1p + 0(1-p) 0p + 2(1-p)$$
$$p = 2(1-p)$$

• What will happen if there's no mixed\*-strategy equilibrium?

#### **Matching Pennies**

- No pure-strategy Nash equilibrium
  - ➤ In every case, one of the agents can do better by changing strategy

	Heads	Tails
Heads	1, -1	-1, 1
Tails	-1, 1	1, -1

- There's a mixed-strategy equilibrium
  - Get it the same way as in the Battle of the Sexes
    - Result is (s,s), where  $s = \{(\frac{1}{2}, \text{Heads}), (\frac{1}{2}, \text{Tails})\}$
  - More about this in Chapter 3

# **Another Interpretation of Mixed Strategies**

- Suppose agent *i* has a deterministic method for picking a strategy, but it depends on things that aren't part of the game itself
  - $\triangleright$  If *i* plays a game several times, *i* may pick different strategies
- If the other players don't know how *i* picks a strategy, they'll be uncertain what *i*'s strategy will be
  - Agent *i*'s mixed strategy is **everyone else's assessment** of how likely *i* is to play each pure strategy
- Example:

In a series of soccer penalty kicks, use a pseudo-random number generator to decide whether to kick left or right

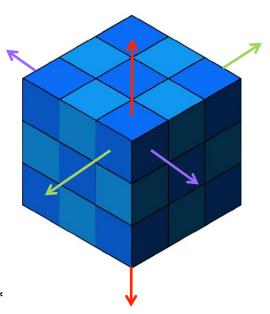
# Finding Nash Equilibria

- For 2x2 games:
  - Pure-strategy equilibria
    - Look for cells where neither player can do better by switching to the other action

- Mixed-strategy equilibria
  - Write 1's strategy as  $\{(p, a), (1-p, a')\}$ 
    - $\rightarrow$  Look for p such that 2 gets same expected utility for b and b'
  - Write 2's strategy as  $\{(q, b), (1-q, b')\}$ 
    - $\rightarrow$  Look for q such that 1 gets same expected utility for a and a'
- > Equilibria where one strategy is pure and the other is mixed\*
  - If there is a weak pure-strategy equilibrium, then look for mixed-strategy best-responses
- What about the general case?

# Finding Nash Equilibria

- General case (not in the book):
  - $\triangleright$  *n* players,  $m_i$  actions for player *i*
  - $\triangleright$  size of payoff matrix:  $m_1 m_2 \dots m_n$
- Brute-force approach:
  - > Pure-strategy equilibria
    - Look for cells where no player can do better by unilaterally choosing a different action
    - Time is polynomial in the size of the matrix
  - > Equilibria in which one or more strategies are mixed\*
    - For every possible combination of supports for  $s_1, ..., s_n$ 
      - > Solve sets of simultaneous equations
    - Exponentially many combinations of supports → exponential time
- Can it be done more quickly?



### **Complexity of Finding Nash Equilibria**

- 2 players:
  - Lemke & Howson (1964): solve a set of simultaneous equations that includes all possible sets of supports
    - Some of the equations are quadratic => worst-case exponential time
  - Porter, Nudelman, & Shoham (2004)
    - AI methods (constraint programming)
  - Sandholm, Gilpin, & Conitzer (2005)
    - Mixed Integer Programming (MIP) problem
- *n*-player games
  - > van der Laan, Talma, & van der Heyden (1987)
  - ➤ Govindan, Wilson (2004)
  - > Porter, Nudelman, & Shoham (2004)
- Worst-case running time still is exponential in the size of the payoff matrix

# **Complexity of Finding Nash Equilibria**

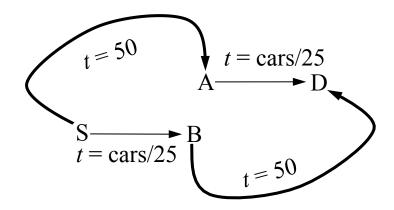
- For the general case,
  - > It's unknown whether there are polynomial-time algorithms to do it
  - ➤ It's unknown whether there are polynomial-time algorithms to compute approximations
- One of the most important open problems in computational complexity theory
- Some special cases can be done in polynomial time
  - > Finding pure-strategy Nash equilibria
    - Check each square of the payoff matrix
  - Finding Nash equilibria in zero-sum games
    - Linear programming

#### **Problem Structure**

- In some classes of problems, can reason about problem structure
- Example (not in the book):
  - > Suppose 1,000 drivers want to go from *S* (start) to *D* (destination)
- $t = \frac{\text{cars}}{25}$
- $\triangleright$  Two routes:  $S \rightarrow A \rightarrow D$  and  $S \rightarrow B \rightarrow D$
- $\triangleright$   $S \rightarrow A$  and  $B \rightarrow D$  are long and wide
  - t = 50 minutes, no matter how many cars
- $\rightarrow$  A $\rightarrow$ D and S $\rightarrow$ B are short, but narrow
  - t = (number of cars)/25
- Assume each driver's utility is -t
- Huge payoff matrix:
  - $\gt$  2<sup>1000</sup> action profiles
- But we don't need to write the matrix

#### **Problem Structure**

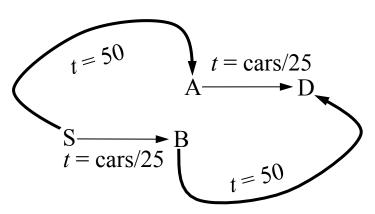
- Nash equilibrium:
  - > 500 cars go through A
  - > 500 cars through B
- Everyone's expected time: 50 + 20 = 70 minutes



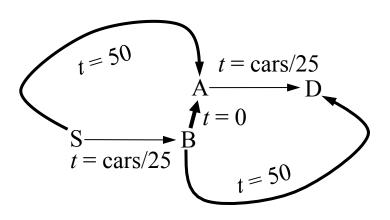
- Consider a driver whose strategy is  $s = S \rightarrow A \rightarrow D$ 
  - $\triangleright$  Suppose the driver changes unilaterally to  $S \rightarrow B \rightarrow D$ 
    - 501 cars on  $S \rightarrow B \rightarrow D$
    - Expected travel time: 50 + 501/25 = 70.04
- Can generalize to the case where  $s = \{(p, S \rightarrow A \rightarrow D), (1-p, S \rightarrow B \rightarrow D)\}\$

#### **Problem Structure**

- Nash equilibrium:
  - ➤ On average, 500 cars go  $S \rightarrow A \rightarrow D$ and 500 cars go  $S \rightarrow B \rightarrow D$
- Everyone's expected travel time: 50 + 20 = 70 minutes

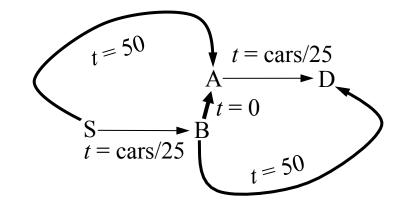


- Modify the network
  - Add a road from B to A that's very short and very wide
  - > 0 minutes, regardless of how many cars
- Three possible routes:
  - $\triangleright S \rightarrow A \rightarrow D$
  - $\triangleright S \rightarrow B \rightarrow D$
  - $\triangleright$  S $\rightarrow$ B $\rightarrow$ A $\rightarrow$ D
- **Poll 2.7**: which one would you take?



#### **Braess's Paradox**

- Nash equilibrium: all cars go  $S \rightarrow B \rightarrow A \rightarrow D$ 
  - $\rightarrow$  time for S  $\rightarrow$  B is 1000/25 = 40
  - $\triangleright$  same for  $A \rightarrow D$
- Total time is 40 + 40 = 80 minutes
- To see that it's a Nash equilibrium:
  - ➤ If a driver unilaterally switches to  $S \rightarrow A \rightarrow D$  or  $S \rightarrow B \rightarrow D$



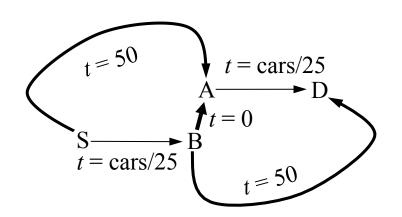
- driving time is 50 + 40 = 90 minutes
- > If a driver unilaterally switches to
  - $s = \{(p, S \rightarrow B \rightarrow A \rightarrow D), (q, S \rightarrow A \rightarrow D), (1-p-q, S \rightarrow B \rightarrow D)\}$ with p < 1
  - What happens?

#### **Braess's Paradox**

- To see that  $S \rightarrow B \rightarrow A \rightarrow D$  is the *only* Nash equilibrium:
- Let
  - $\rightarrow a = \text{expected } \# \text{ of cars } S \rightarrow A \rightarrow D$
  - $\rightarrow b = \text{expected } \# \text{ of cars } S \rightarrow B \rightarrow D$
  - $\rightarrow$  0 < a + b < 1000
- Times:
  - $\rightarrow$  time to go S $\rightarrow$ A $\rightarrow$ D:

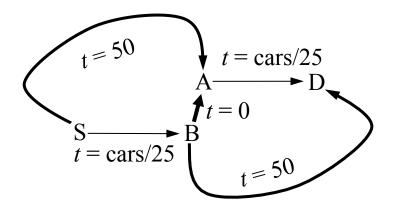
• 
$$50 + (1000-b)/25 = 90 - b/25$$

- $\triangleright$  time to go S $\rightarrow$ B $\rightarrow$ D:
  - 50 + (1000-a)/25 = 90 a/25
- $\rightarrow$  time to go S $\rightarrow$ B $\rightarrow$ A $\rightarrow$ D:
  - (1000-a)/25 + (1000-b)/25 = 80 a/25 b/25
- $\rightarrow$  Any driver that goes  $S \rightarrow A \rightarrow D$  or  $S \rightarrow B \rightarrow D$ can get lower travel time by switching to  $S \rightarrow B \rightarrow A \rightarrow D$



#### **Discussion**

- Travel time
  - > 70 minutes before adding the road
  - > 80 minutes after
- Suggests that sometimes adding road capacity can hurt
- We assumed
  - $\rightarrow$  t = 0 regardless of how many cars
  - > t = 50 regardless of how many cars
  - $\rightarrow t = cars/25$
- Is that realistic?
- Can this really happen in practice?

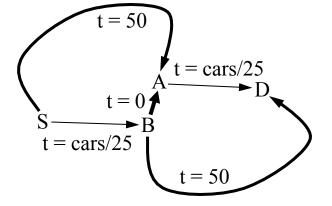


#### **Braess's Paradox in Practice**

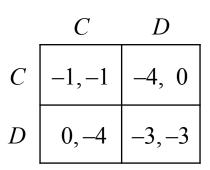
- 1969, Stuttgart, Germany when a new road to the city center was opened, traffic got worse; didn't improve until the road was closed
- 1990, Earth day, New York closing 42nd street improved traffic flow
- 1999, Seoul, South Korea closing a tunnel improved traffic flow
- 2003, Seoul, South Korea traffic flow was improved by closing a 6-lane motorway and replacing it with a 5-mile-long park
- 2010, New York closing parts of Broadway has improved traffic flow
- Sources
  - http://www.umassmag.com/transportationandenergy.htm
  - http://www.cs.caltech.edu/~adamw/courses/241/lectures/brayes-j.pdf
  - http://www.guardian.co.uk/environment/2006/nov/01/society.travelsenvironmentalimpact
  - ► <a href="http://www.scientificamerican.com/article.cfm?id=removing-roads-and-traffic-lights">http://www.scientificamerican.com/article.cfm?id=removing-roads-and-traffic-lights</a>
  - http://www.lionhrtpub.com/orms/orms-6-00/nagurney.html

#### **Discussion**

- Nash equilibrium:
  - $\rightarrow$  All 1000 cars go S $\rightarrow$ B $\rightarrow$ A $\rightarrow$ D
  - > Total time is 80 minutes



Compare with the Prisoner's Dilemma



#### **Comments**

- Braess's paradox can also occur in other kinds of networks
  - Queueing networks
  - Communication networks
- In principle, it can occur in Internet traffic
  - ➤ I don't know enough about this topic to know how much of a problem it is

#### Here's Another Game

- All of you can play!
  - $\triangleright$  Choose a number in the range  $0 \le x \le 100$
  - Write your choice on a piece of paper
  - > Fold the paper so nobody else can see your number
  - > Pass the paper to the front of the room
- The winner(s) will be those whose number is closest to 2/3 of the average of all the numbers

• I'll present the results next time

# **Announcements (Sept 8)**

- What we covered last time:
  - Best response
  - Nash equilibrium
    - strict, weak
    - finding mixed-strategy Nash equilibria
    - problem structure (Braess's paradox)
  - > Game:
    - choose a number between 0 and 100
    - winner(s): whoever chooses a number that's closest to 2/3 of the average

(not in the book)

- In the Prisoner's dilemma, recall that
  - (C,C) is the action profile that provides the best outcome for everyone
  - > If we assume each payer acts to maximize his/her utility without regard to the other, we get (D,D)
  - > By choosing (C,C), each player could have gotten 3 times as much

	C	D
C	3, 3	0, 5
D	5, 0	1, 1

	C	D
C	3, 3	0, 5
D	5, 0	1, 1

Let's generalize "best outcome for everyone"

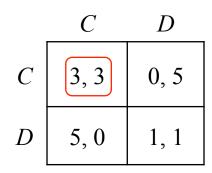
- Social welfare function w(s)
  - measures the players' welfare given s
  - ▶ **Utilitarian** welfare function: w(s) = average expected utility
  - **Egalitarian** welfare function: w(s) = minimum expected utility
- Social optimum: benevolent dictator chooses  $s^*$  that optimizes w
  - $> s* = arg max_s w(s)$
  - > Is this Pareto optimal?
- **Anarchy**: no dictator; every player selfishly tries to optimize his/her own expected utility, disregarding the welfare of the other players
  - Get a strategy profile s (e.g., a Nash equilibrium)
  - ➤ In general,  $w(\mathbf{s}) \le w(\mathbf{s}^*)$
- Price of anarchy =  $w(s^*)/w(s)$

- Example: the Prisoner's Dilemma
  - > Utilitarian welfare function:  $w(\mathbf{s})$  = average expected utility
- Social optimum:  $s^* = (C,C)$

$$> w(s^*) = 3$$

- Anarchy:  $\mathbf{s} = (D,D)$ 
  - > w(s) = 1
- Price of anarchy

$$= w(s^*)/w(s) = 3/1 = 3$$

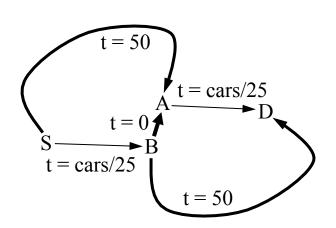


	C	D
C	3, 3	0, 5
D	5, 0	1, 1

What would the answer be if we used the egalitarian welfare function?

- Sometimes instead of maximizing a welfare function w, we want to minimize a cost function c
  - ightharpoonup Utilitarian function:  $c(\mathbf{s}) = \text{avg. expected cost}$
  - $\triangleright$  Egalitarian function:  $c(\mathbf{s}) = \max$ . expected cost
- Need to adjust the definitions
  - > Social optimum:  $s^* = \arg\min_{s} c(s)$
  - > Anarchy: every player selfishly tries to minimize his/her own cost, disregarding the costs of the other players
    - Get a strategy profile s (e.g., a Nash equilibrium)
    - In general,  $c(\mathbf{s}) \ge c(\mathbf{s}^*)$
  - > Price of anarchy =  $c(s) / c(s^*)$ 
    - reciprocal of what we had before

- Example: Braess's Paradox
  - $\triangleright$  Utilitarian cost function:  $c(\mathbf{s})$  = average expected cost
- Social optimum:
  - $> s^* = [500 \text{ go S} \rightarrow A \rightarrow D; 500 \text{ go S} \rightarrow B \rightarrow D]$
  - $c(s^*) = 70$
- Anarchy:  $\mathbf{s} = [1000 \text{ drivers go S} \rightarrow \mathbf{B} \rightarrow \mathbf{A} \rightarrow \mathbf{D}]$ 
  - c(s) = 80
- Price of anarchy =  $c(\mathbf{s}) / c(\mathbf{s}^*) = 8/7$



- What would the answer be if we used the egalitarian cost function?
- This can be generalized

### **Summary**

- Pareto optimality
  - > Prisoner's Dilemma, Which Side of the Road
- Best responses and Nash equilibria
  - Battle of the Sexes, Matching Pennies
- Finding pure-strategy and mixed-strategy Nash equilibria
  - Methods for special cases
- Not in the book:
  - Brief discussion of computational complexity
- Road-network example (not in the book)
  - Braess's paradox
- Price of anarchy (not in the book)
  - Prisoner's dilemma, road networks