# Mechanism Design

CS 886 Multiagent Systems University of Waterloo

#### Introduction

So far we have looked at

- Game Theory
  - Given a game we are able to analyze the strategies agents will follow
- Theory

   Given a set of agents'
  - Given a set of agents' preferences we can choose some outcome Ballot

Social Choice



X>Y>Z

#### Introduction

- Today, Mechanism Design
  - Game Theory + Social Choice
- Goal of Mechanism Design is to
  - Obtain some outcome
  - But agents are rational
- "Solution":
  - Define the rules of a game so that in equilibrium the agents do what we want
- CS Spin
  - Defining protocols for distributed systems

3

### Example: London Bus System

(as of April 2004)

- 5 million passengers each day
- 7500 buses
- 700 routes

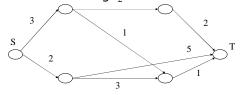


- The system has been privatized since 1997 by using competitive tendering
- Idea: Run an auction (mechanism) to allocate routes to companies

4

# Example

• Selfish Routing 2



Want to find the least-cost route from S to T.

You do not know costs.

You do know that each links wants to maximize revenue.

How do you use this information to extract information needed to find least-cost path?  $$^{5}$$ 

#### **Fundamentals**

- Set of possible outcomes, O
- Agents  $i \in N$ , |N| = n, each agent i has type  $\theta_i \in \Theta_i$  Type captures all private information that is relevant to agent's decision making
- Utility  $u_i(o, \theta_i)$ , over outcome  $o \in O$
- Recall: goal is to implement some system-wide solution
  - Captured by a social choice function

$$f{:}\Theta_1\times\ldots\times\Theta_n\to\mathcal{O}$$

 $\mathbf{f}(\boldsymbol{\theta}_1, \dots \boldsymbol{\theta}_n) = \mathbf{o}$  is a collective choice

# Examples of social choice functions

- Voting: choose a candidate among a group
- Public project: decide whether to build a swimming pool whose cost must be funded by the agents themselves
- Allocation: allocate a single, indivisible item to one agent in a group

#### Mechanisms

- Recall: We want to implement a social choice function
  - Need to know agents' preferences
  - They may not reveal them to us truthfully
- Example:
  - 1 item to allocate, and want to give it to the agent who values it the most
  - If we just ask agents to tell us their preferences, they may lie

I like th bear the most!







No, I do!

8

# Mechanism Design Problem

- By having agents interact through an institution we might be able to solve the problem
- Mechanism:

$$\begin{array}{c} M{=}(S_1,{\dots},S_n,g(\cdot))\\ \\ \nearrow \\ \text{Strategy spaces of agents} \end{array} \begin{array}{c} O\text{utcome function}\\ g{:}S_1{\times}{\dots}{\times}S_n{\to}O \end{array}$$

# Implementation

• A mechanism  $M=(S_1,...,S_n,g())$ implements social choice function  $f(\theta)$ if there is an equilibrium strategy profile  $s^*=(s_1^*,...,s_n^*)$  of the game induced by M such that

$$\begin{array}{l} -g(s_1^*(\theta_1),...,s_n^*(\theta_n)) = f(\theta_1,...,\theta_n) \\ \forall \ (\theta_1,...,\theta_n) \in \Theta_1 \mathbf{x}... \mathbf{x} \mathbf{\Theta}_n \end{array}$$

10

# Implementation

- We did not specify the type of equilibrium in the definition
- Nash

 $u_i(s_i^*(\boldsymbol{\theta}_i),s^*_{.i}(\boldsymbol{\theta}_{.i}),\boldsymbol{\theta}_i) \geq u_i(s_i^*(\boldsymbol{\theta}_i),s^*_{.i}(\boldsymbol{\theta}_{.i}),\boldsymbol{\theta}_i), \ \forall \ i, \ \forall \ \boldsymbol{\theta}, \ \forall \ s_i^* \neq s_i^*$ 

• Bayes-Nash

 $\mathrm{E}[\mathrm{u}_{i}(s_{i}^{*}(\theta_{i}),s^{*}._{i}(\theta_{.i}),\theta_{i})] \geq \mathrm{E}[\mathrm{u}_{i}(s_{i}^{*}(\theta_{i}),s^{*}._{i}(\theta_{.i}),\theta_{i})], \ \forall \ i, \ \forall \ \theta, \ \forall \ s_{i}^{*} \neq s_{i}^{*}$ 

• Dominant

 $u_i(s_i^*(\boldsymbol{\theta}_i),s_{\cdot i}(\boldsymbol{\theta}_i),\boldsymbol{\theta}_i) \geq u_i(s_i^*(\boldsymbol{\theta}_i),s_{\cdot i}(\boldsymbol{\theta}_{\cdot i}),\boldsymbol{\theta}_i), \ \forall \ i, \ \forall \ \boldsymbol{\theta}, \ \forall \ s_i^* \neq s_i^*, \ \forall \ s_{\cdot i}$ 

11

#### **Direct Mechanisms**

- Recall that a mechanism specifies the strategy sets of the agents
  - These sets can contain complex strategies
- Direct mechanisms:
  - Mechanism in which  $S_i=\Theta_i$  for all i, and  $g(\theta)=f(\theta)$  for all  $\theta\in\Theta_1\times...\times\Theta_n$
- Incentive compatible:
  - A direct mechanism is incentive compatible if it has an equilibrium  $s^*$  where  $s^*_i(\theta_i)=\theta_i$  for all  $\theta_i\in\Theta_i$  and all i
    - truth telling by all agents is an equilibrium
  - Strategy-proof if dominant-strategy equilibrium

#### **Dominant Strategy Implementation**

- Is a certain social choice function implementable in dominant strategies?
  - In principle we would need to consider all possible mechanisms
- Revelation Principle
  - Suppose there exists a mechanism  $M=(S_1,...,S_n,g(\cdot))$  that implements social choice function  $f(\cdot)$  in dominant strategies.
  - Then there is a direct strategy-proof mechanism, M', which also implements f().

13

# **Revelation Principle**

"the computations that go on within the mind of any bidder in the nondirect mechanism are shifted to become part of the mechanism in the direct mechanism" [McAfee&McMillian 87]

• Vickrey auction and English auction

14

### Revelation Principle: Proof

- M=(S<sub>1</sub>,...,S<sub>n</sub>,g()) implements SCF f() in dom str.
  - Construct direct mechanism  $M'=(\Theta^n,f(\theta))$
  - By contradiction, assume

 $\exists \theta_i^{'} \neq \theta_i \text{ s.t. } u_i(f(\theta_i^{'}, \theta_{-i}), \theta_i) > u_i(f(\theta_i, \theta_{-i}), \theta_i)$  for some  $\theta_i^{'} \neq \theta_i$ , some  $\theta_{-i}$ .

– But, because  $f(\theta)=g(s^*(\theta))$ , this implies  $u_i(g(s_i^*(\theta_i'),s_{-i}^*(\theta_{-i})),\theta_i)>u_i(g(s^*(\theta_i),s^*(\theta_{-i})),\theta_i)$ 

Which contradicts the strategy proofness of  $s^{\ast}$  in  $\mbox{\em M}$ 

15

#### **Revelation Principle:** Intuition Constructed "direct revelation" mechanism Agent 1's Strategy Original "complex preferences formulato "indirect" mechanism Outcome Strategy Agent IAI's preference 16

# Theoretical Implications

- Literal interpretation: Need only study direct mechanisms
  - This is a smaller space of mechanisms
  - Negative results
    - If no direct mechanism can implement SCF f() then no mechanism can do it
  - Analysis tool:
    - Best direct mechanism gives us an upper bound on what we can achieve with an indirect mechanism
    - Analyze all direct mechanisms and choose the best one

17

# **Practical Implications**

- Incentive-compatibility is "free" from an implementation perspective
- BUT!!!
  - A lot of mechanisms used in practice are not direct and incentive-compatible
    - Maybe there are some issues that are being ignored here

# Quick review

- · We now know
  - What a mechanism is
  - What is means for a SCF to be dominant strategy implementable
  - Implementable in dominant strategies ⇒ implementable by a direct incentivecompatible mechanism
- We do not know
  - What types of SCF are dominant-strategy implementable

#### Gibbard-Satterthwaite Thm

- Assume
  - $\mathcal{O}$  is finite and  $|\mathcal{O}|$  ≥ 3
  - Each  $o \in \mathcal{O}$  can be achieved by social choice function f() for some  $\theta$

#### Then:

f() is truthfully implementable in dominant strategies  $\leftrightarrow$ f() is dictatorial

# Circumventing G-S • Use a weaker equilibrium concept

- - Nash, Bayes-Nash
- Design mechanisms where computing a beneficial manipulation is hard
  - Many voting mechanisms are NP-hard to manipulate (or can be made NP-hard with small "tweaks) [Bartholdi, Tovey, Trick 89] [Conitzer, Sandholm 03]
- Randomization



Almost need this much

• Agents' preferences have special structure



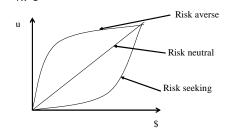
### Quasi-Linear Preferences

- Outcome: o=(x,t<sub>1</sub>,...,t<sub>n</sub>)
  - x is a "project choice"
  - t<sub>i</sub> is a "monetary" transfer
- Utility of agent i:
  - $-U_i(o,\theta_i)=u_i(x,\theta_i)-f(t_i)$
  - Preference of x is independent from the payment
  - Can choose to reward or punish by a monetary amount

22

# Quasi-linear preferences

- $U_i(o,\theta_i)=u_i(x,\theta_i)-f_i(t_i)$
- f<sub>i</sub>() gives i's risk attitude



23

#### SCF and quasi-linear settings

- $f:\Theta \rightarrow (x(\Theta),t(\Theta))$
- SCF f is efficient if for all types  $\theta = (\theta_1, ..., \theta_n)$ 
  - $\sum_{i=1}^{n} u_i(x(\theta), \theta_i) \ge \sum_{i=1}^{n} u_i(x'(\theta), \theta_i) \ \forall \ x'(\theta)$
  - Aka social welfare maximizing
- SCF f is budget-balanced if  $\sum_{i=1}^{n} t_i(\theta) = 0$
- SCF f is weakly budget-balanced if  $\sum_{i=1}^{n} t_i(\theta) \ge 0$

#### Mechanisms and quasi-linear utilities

- M=(S<sub>1</sub>,...,S<sub>n</sub>,(x(S),t(S))
- Valuation for choice x v<sub>i</sub>(x)=u<sub>i</sub>(x,θ<sub>i</sub>)
- Agents reveal their valuation functions in a direct mechanism
  - v'<sub>i</sub> denotes the valuation that agent i declares to the mechanism (may be different from true valuation  $v_i$ )
  - v=(v'<sub>1</sub>,...,v'<sub>n</sub>)

# Properties of mechanisms

- **Truthful**: ∀ i ∀ v<sub>i</sub>, the equilibrium strategy for agent i is to adopt  $v_i'=v_i$
- Efficient: Mechanism selects choice x such that  $\forall$  I  $\forall$   $v_i$   $\forall$  x'  $\sum_i v_i(x) \ge \sum_i v_i(x')$
- Budget balanced:  $\forall v' \sum_i t_i(v') = 0$
- Individually rational:  $v_i(s^*(v))-t_i(s(v)) \ge$ 0 where S\* is the equilibrium

#### **Groves Mechanisms**

[Groves 1973]

• A Groves mechanism,

 $M=(S_1,...,S_n, (x,t_1,...,t_n))$  is defined by

- Choice rule  $x^*(\theta')$  = argmax<sub>x</sub>  $\sum_i v_i(x,\theta_i')$
- Transfer rules

•  $t_i(\theta') = h_i(\theta_{-i}') - \sum_{i \neq j} v_i(x^*(\theta'), \theta'_i)$ 

where  $h_i(\cdot)$  is an (arbitrary) function that **does not depend** on the reported type  $\theta_i$ of agent i

#### **Groves Mechanisms**

- Thm: Groves mechanisms are strategyproof and efficient (We have gotten around Gibbard-Satterthwaite!)
- Proof: Agent i's utility for strategy  $\theta_i$ , given  $\theta_{-i}$ from agents j≠i is

 $U_i(\theta_i) = v_i(x^*(\theta), \theta_i) - t_i(\theta)$ 

 $= v_i(x^*(\theta'), \theta_i) + \sum_{j \neq i} v_j(x^*(\theta'), \theta'_j) - h_i(\theta'_{-i})$ 

Ignore  $h_i(\theta_{-i})$ . Notice that

 $x^*(\theta') = \operatorname{argmax} \sum_i v_i(x, \theta'_i)$ 

i.e. it maximizes the sum of reported values.

Therefore, agent i should announce  $\theta_{i}^{'}=\theta_{i}$  to maximize its own payoff

**Thm**: Groves mechanisms are unique (up to  $h_i(\theta_i)$ )

#### VCG Mechanism

(aka Clarke mechanism aka Pivotal mechanism)

• Def: Implement efficient outcome,

 $x^* = \max_{x} \sum_{i} v_i(x, \theta_i)$ 

Compute transfers

$$t_{i}(\theta') = \sum_{j \neq i} v_{j}(x^{-i}, \theta'_{j}) - \sum_{j \neq i} v_{j}(x^{*}, \theta'_{i})$$

Where  $x^{-i} = \max_{x} \sum_{i \neq j} v_i(x, \theta_i')$ 

VCG are efficient and strategy-proof

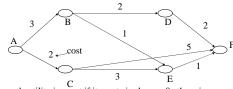
Agent's equilibrium utility is:

$$u_i(\boldsymbol{x}^{\star},\boldsymbol{t}_i,\boldsymbol{\theta}_i) \text{=} v_i(\boldsymbol{x}^{\star},\boldsymbol{\theta}_i) \text{-} [\boldsymbol{\Sigma}_{j\neq i} \ v_j(\boldsymbol{x}^{\text{-}i},\boldsymbol{\theta}_j) \ \text{-} \boldsymbol{\Sigma}_{j\neq i} v_j(\boldsymbol{x}^{\star},\boldsymbol{\theta}_j)]$$

= 
$$\sum_{j} v_{j}(x^{*}, \theta_{j}) - \sum_{j \neq i} v_{j}(x^{*}, \theta_{j})$$

= marginal contribution to the welfare of the system

# Example: Selfish Routing



Agent's utility is –cost if its route is chosen, 0 otherwise

 $x(v)=argmax \sum_{i} v_{i}(x) = ABEF$ Payments:

 $T_{AC} = 5-5=0$ Payments (Pivotal Agents):

T<sub>AB</sub>=2-6=-4 (paid 4 for its contribution)  $T_{CE} = 5-5=0$ 

 $T_{BD} = 5-5=0$ T<sub>BE</sub>=4-6=-2 (paid 2 for its contribution)

 $T_{DF} = 5-5=0$ T<sub>EF</sub>=4-7=-3 (paid 3 for its contribution)

 $T_{DF} = 5 - 5 = 0$ "Market Power"

# Example: Building a pool • The cost of building the pool is \$300

- If together all agents value the pool more than \$300 then it will be built
- Clarke Mechanism:
  - Each agent announces their value,  $\mathbf{v}_{i}$
  - If  $\sum v_i \ge 300$  then it is built and each pays 100
  - Payments  $t_i(\theta_i') = \sum_{j \neq i} v_j(x^{-i}, \theta_j') \sum_{j \neq i} v_j(x^*, \theta_i')$  if built, 0 otherwise

v1=50, v2=50, v3=250 Pool should be built

†<sub>1</sub>=(250+50)-(250+50)=0 †<sub>2</sub>=(250+50)-(250+50)=0 †<sub>3</sub>=(0)-(100)=-100

Not budget balanced

# **Example: Vickrey Auction**

- Highest bidder gets item, and pays second highest amount
- Also a VCG mechanism
  - Allocation rule: get item if b<sub>i</sub>=max<sub>i</sub>[b<sub>i</sub>]
  - Every agent pays

$$t_{i}(\theta_{i}^{'}) = \sum_{j \neq i} v_{j}(x^{-i}, \theta_{j}^{'}) - \sum_{j \neq i} v_{j}(x^{*}, \theta_{i}^{'})$$

$$\max_{j \neq i} [b_{j}] \text{ if } i \text{ is not the highest bidder, 0 if it is}$$