Dynamic Marginal Contribution Mechanism

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Intertemporal Efficiency with Private Information

- random arrival of buyers, sellers and/or objects
 - selling seats for an airplane with random arrival of buyers
 - bidding on ebay
 - bidding for construction projects with uncertain arrival of new projects
- bidding for links in sponsored search (Google, Yahoo, etc.)
 - uncertainty about click-through probability
 - uncertainty about conversion probability
- leasing resource over time
 - auction of renewable license, right, capacity over time
 - web serving, computational resource (bandwidth, CPU)

private value environment

- Vickrey (1961): single or multiple unit discriminatory auctions implement socially efficient allocation
 - in private value environments
 - in (weakly) dominant strategies
- Clarke (1971) and Groves (1973) extend to general allocation problems in private value environments
 - agent *i* internalizes the social objective and is led to report her type truthfully

Pivot Mechanism

- Green & Laffont (1977) analyze specific VCG mechanism
- *i* internalizes social objective if *i* pays her externality cost
- externality cost:

utility of $I \setminus i$ given *i* is present - utility of $I \setminus i$ given *i* is absent

- marginal contribution of i = utility of i externality cost of i
- In Pivot mechanism:
- payoff of *i* is her marginal contribution to social value
- 2 participation constraint holds ex post and no budget deficit

Dynamic Marginal Contribution Mechanism

- marginal contribution = payoff in Pivot mechanism
- develop marginal contribution mechanism in intertemporal environments with new arrival of information regarding:
 - preferences
 - agents
 - allocations
- design sequence of payments so that each agent receives flow marginal contribution in every period

- solve intertemporal problem as a completely contigent plan
- embed intertemporal problem in a static problem (as in an Arrow Debreu economy) ...
- ... and then appeal to the classic VCG results.
- but the contingent view fails to account for strategic possibilities of the agents in the sequential model

- information arrives over time
- report of agent *i* in period *t* responds to private information of agent *i*, but may also respond to past reports of other agents (possibly inferred from allocative decisions)
- truthtelling (generally) fails to be a weakly dominant strategy
- with forward looking agents, participation constraint is required to be satisfied at every point in time (and not only in the initial period)

Results

- marginal contribution mechanism is dynamically efficient
- periodic ex post: with respect to information available at period t
- satisfies (periodic) ex post incentive constraints
- satisfies (periodic) ex post participation constraints
- adding efficient exit condition (weak "online" condition): if agent *i* does not impact future decisions, then agent *i* does not receive future payments, uniquely identifies marginal contribution mechanism

- Dolan (RAND 1978): priority queuing
- Parkes et al. (2003): delayed VCG without participation or budget balance constraints
- Bergemann & Valimaki (JET 2006): complete information, repeated allocation of single object over time, first price bidding
- Athey & Segal (2007): balanced budget rather than participation constraints

- scheduling tasks
- discrete time, infinite horizon: t = 0, 1, ...
- common discount factor δ
- finite number of agents: $i \in \{0, 1, ..., I\}$
- each agent *i* has a single task
- value of task for i is:

 $v_i > 0$

• quasilinear utility: $v_i - p_i$

• values are given wlog in descending order:

 $v_0 > v_1 > \cdots > v_l > 0$

- marginal contribution of task i : difference in welfare with i and without i
- efficient task assignment policy:

policy without i = 0 1 \cdots i = 1 i + 1 i + 2 \cdots 1 policy with i = 0 1 \cdots i = 1 i i + 1 i + 2 \cdots

Marginal Contribution

policy without i = 0 1 \cdots i - 1 i+1 i+2 \cdots I policy with i = 0 1 \cdots i - 1 i i+1 i+2 \cdots I

- insert valuable task *i*:
- raise the value of all future tasks: t > i
- marginal contribution *M_i*:

$$M_i = \sum_{t=0}^{l} \delta^t \mathbf{v}_t - \left(\sum_{t=0}^{i-1} \delta^t \mathbf{v}_t + \sum_{t=i+1}^{l} \delta^{t-1} \mathbf{v}_t\right)$$

or

$$M_i = \sum_{t=i}^{\prime} \delta^t \left(v_t - v_{t+1} \right) \ge 0$$

• from marginal contribution to externality pricing:

 $M_i = v_i - p_i$

externality cost of task i is:

$$p_i = v_{i+1} - \sum_{t=i+1}^{l} \delta^{t-i} \underbrace{(v_t - v_{t+1})}^{>0}$$

- task *i* directly replaces task i + 1, but also:
- task i raises the value of all future tasks

Incomplete Information

- suppose v_i is private information to agent i at t = 0
- incentive compatibility and efficient sorting
- when would agent *i* like to win against *j* versus j + 1:

$$(\mathbf{v}_{i} - \mathbf{v}_{j}) - \sum_{t=j}^{l} \delta^{t-(j-1)} (\mathbf{v}_{t} - \mathbf{v}_{t+1}) \ge \delta (\mathbf{v}_{i} - \mathbf{v}_{j+1}) - \sum_{t=j+1}^{l} \delta^{t-j} (\mathbf{v}_{t} - \mathbf{v}_{t+1})$$

reduces to cost of delay:

$$(1-\delta)$$
 $\mathbf{v}_{\mathbf{i}} \geq (1-\delta)$ $\mathbf{v}_{\mathbf{j}}$.

report thruthfully if others report truthfully: ex post equilibrium

Bidding vs Direct Revelation Mechanism

- an ascending (English) auction in every period
- winning bidder i pays bid of second highest bidder
- bid by agent *i* in period *t*:

bįt

- bid should reflect value of task but ...
- value of task today versus value of task tomorrow
- value = utility option value

Option Value

- bidding strategy b^t_i determined recursively in i and t
- option value is value of realizing task tomorrow

$$\delta\left(\boldsymbol{v}_{i}-\boldsymbol{p}_{i}^{t+1}\right)$$

and the price tomorrow is

$$\boldsymbol{p}_{i}^{t+1} \triangleq \max_{j \neq i} \left\{ ..., \boldsymbol{b}_{j}^{t+1}, ... \right\}$$

net value of realizing task today is

$$\mathbf{v}_i - \delta\left(\mathbf{v}_i - \mathbf{p}_i^{t+1}\right)$$

Dynamic Bidding

• bidding strategy of agent *i* is given

$$\boldsymbol{b}_{i}^{t} = \boldsymbol{v}_{i} - \delta \left(\boldsymbol{v}_{i} - \boldsymbol{p}_{i+1}^{t} \right) = (1 - \delta) \, \boldsymbol{v}_{i} + \delta \boldsymbol{b}_{i+1}^{t+1}$$

- ascending auction gives efficient assignment in all periods
- Bergemann and Valimaki (JET 2006): dynamic price competition, complete information, first price bidding
- Edelmann, Ostrovsky and Schwarz (AER 2007): static price competition, incomplete information, second price bidding

- sequential allocation of a single indivisible object with initially uncertain value to the bidders
- bidder *i* receives additional information only in periods in which *i* is assigned the object
- license to use facility or to explore resource for a limited time

Single Unit Auction

- single unit auction repeated over time
- discrete time, infinite horizon: t = 0, 1, ...
- finite number of bidders: $i \in \{1, ..., I\}$
- realized value of object for winning bidder in period t is

 $\mathbf{V}_{i,t} = \omega_i + \varepsilon_{i,t}$

- $\varepsilon_{i,t}$ is i.i.d. over time with $\mathbb{E}\left[\varepsilon_{i,t}\right] = 0$
- ω_i is true value of object
- ε_{i,t} is random noise

Information Flow

- at t = 0: common prior distribution $F_i(\omega_i)$ for each agent *i*
- at $t \ge 0$: winning bidder receives informative signal $s_{i,t+1}$:

$$\mathbf{S}_{i,t+1} = \mathbf{V}_{i,t} = \omega_i + \varepsilon_{i,t}$$

- realized value in period t constitutes private information for period t + 1
- at t ≥ 0: loosing bidders don't receive additional information:

$$\mathbf{s}_{i,t+1} = \mathbf{s}_{i,t}$$

• private history of bidder *i*:

$$\mathbf{s}_{i}^{t} = \left(\mathbf{s}_{i,0}, ..., \mathbf{s}_{i,t}\right)$$

• expected value for bidder *i* in perid *t* :

 $\mathbf{v}_{i,t}\left(\mathbf{s}_{i}^{t}\right) \triangleq \mathbb{E}\left[\omega_{i} \left| \mathbf{s}_{i}^{t} \right. \right]$

renting store space in mall

- current winner (lessee) gets traffic data, purchase behavior
- current looser does not get traffic data, purchase behavior
- bidding for keywords
 - current winner gets information about click-through rate, sales conversion rate
 - current looser doesn't get information about click-through rate, sales conversion rate

Dynamic Direct Mechanism

bidder *i* is asked to report her signal in every period *t*initial reports:

$$r_0 = (r_{1,0}, ..., r_{l,0})$$

inductively, a history of reports:

$$r^t = \left(r^{t-1}, r_{1,t}, ..., r_{l,t}\right) \in \mathbf{R}^t$$

allocation rule:

$$x_t: \mathbf{R}^{t-1} \times \mathbf{R}_t \rightarrow [0, 1]^{t}$$

• transfer (or pricing) rule is given by:

$$p_t: \mathbf{R}^{t-1} \times \mathbf{R}_t \to \mathbb{R}^t$$

• reporting strategy for agent *i*:

$$r_{i,t}: R^{t-1} \times S_i \to S_i.$$

• expected payoff for bidder i :

$$\mathbb{E}\sum_{t=0}^{\infty}\delta^{t}\left[\mathbf{x}_{i,t}\left(r^{t}\right)\mathbf{v}_{i}\left(\mathbf{s}_{i}^{t}\right)-\mathbf{p}_{i,t}\left(r^{t}\right)\right].$$

reporting strategy of *i* solves sequential optimization problem V_i(s^t_i, r^{t-1}):

$$\max_{r_{i,t}\in S_{i}} \mathbb{E}\left\{ x_{i,t}\left(r^{t}\right) v_{i,t}\left(s_{i}^{t}\right) - p_{i,t}\left(r^{t}\right) + \delta V_{i}\left(s_{i}^{t+1}, r^{t}\right) \right\}$$

• taking expectation \mathbb{E} wrt $(s_{-i,t}, r_{-i,t})$

Equilibrium

- denote by $\mathbf{s}_{-(i,t)}^t \triangleq \mathbf{s}^t \setminus \mathbf{s}_{i,t}$
- Bayesian incentive compatible if $r_{i,t} = s_{i,t}$ solves

$$\max_{r_{i,t}\in\mathcal{S}_{i}}\mathbb{E}\left\{x_{i,t}\left(r_{i,t},\mathbf{s}_{-(i,t)}^{t}\right)\mathbf{v}_{i}\left(\mathbf{s}_{i}^{t}\right)-\mathbf{p}_{i,t}\left(r_{i,t},\mathbf{s}_{-(i,t)}^{t}\right)+\delta V_{i}\left(r_{i,t},\mathbf{s}_{-(i,t)}^{t}\right)\right\}$$

- periodic ex post: with respect to all the information available at period t
- (periodic) ex post incentive compatible if $r_{i,t} = s_{i,t}$ solves

$$\max_{r_{i,t}\in\mathcal{S}_{i}}\left\{\mathbf{x}_{i,t}\left(r_{i,t},\mathbf{s}_{-(i,t)}^{t}\right)\mathbf{v}_{i}\left(\mathbf{s}_{i}^{t}\right)-\mathbf{p}_{i,t}\left(\mathbf{s}_{i,t},\mathbf{s}_{-(i,t)}^{t}\right)+\delta \mathbf{V}_{i}\left(r_{i,t},\mathbf{s}_{-(i,t)}^{t}\right)\right\}$$

for all $s_{-i,t} \in S_{-i}$

Social Efficiency

socially efficient assignment policy

$$W(s^{u}) = \max_{\{x_{t}(s^{t})\}_{t=u}^{\infty}} \mathbb{E} \sum_{t=u}^{\infty} \sum_{i=1}^{N} \delta^{t-u} x_{i,t}(s^{t}) v_{i}(s_{i}^{t})$$

optimal assignment is a multi–armed bandit problem
optimal policy is an index policy:

$$\gamma_{i}(\mathbf{s}_{i}^{u}) = \max_{\tau} \mathbb{E} \left\{ \frac{\sum_{t=0}^{\tau} \delta^{t} \mathbf{v}_{i}\left(\mathbf{s}_{i}^{u+t}\right)}{\sum_{t=0}^{\tau} \delta^{t}} \right\}$$

• socially efficient allocation policy $\mathbf{x}^* = \{\mathbf{x}_t^*\}_{t=0}^\infty$:

$$\mathbf{x}_{i,t}^* > \mathbf{0} \text{ if } \gamma_i \left(\mathbf{s}_i^t \right) \geq \gamma_j \left(\mathbf{s}_j^t \right) \text{ for all } j.$$

Marginal Contribution

• value of social program after removing bidder i

$$W_{-i}\left(\boldsymbol{s}^{u}\right) = \max_{\left\{\boldsymbol{x}_{-i,t}\left(\boldsymbol{s}^{t}\right)\right\}_{t=u}^{\infty}} \mathbb{E}\sum_{t=u}^{\infty}\sum_{j\neq i}\delta^{t-u}\boldsymbol{x}_{j}^{t}\left(\boldsymbol{s}^{t}\right)\boldsymbol{v}_{j}\left(\boldsymbol{s}_{j}^{t}\right)$$

- marginal contribution $M_i(s^t)$ of bidder *i* at history s^t is: $M_i(s^t) = W(s^t) - W_{-i}(s^t)$
- value *M* conditional on history s^u and allocation x_u : $M(s^u, x_u)$
- flow marginal contribution $m_i(s^t)$: $M_i(s^t) = m_i(s^t) + \delta M_i(s^t, x_t^*)$

Flow Marginal Contribution

flow marginal contribution:

$$\boldsymbol{m}_{i}\left(\boldsymbol{s}^{t}\right) = \boldsymbol{M}_{i}\left(\boldsymbol{s}^{t}\right) - \delta \boldsymbol{M}_{i}\left(\boldsymbol{s}^{t}, \boldsymbol{x}_{t}^{*}\right)$$

expanding flow expression with respect to time

$$m_{i}\left(\mathbf{s}^{t}\right) = \overbrace{\left(W\left(\mathbf{s}^{t}\right) - W_{-i}\left(\mathbf{s}^{t}\right)\right)}^{\mathsf{M}_{i} \text{ starting at } t+1 \text{ and } \mathbf{x}_{t}^{*}} \overbrace{\delta\left(W\left(\mathbf{s}^{t}, \mathbf{x}_{t}^{*}\right) - W_{-i}\left(\mathbf{s}^{t}, \mathbf{x}_{t}^{*}\right)\right)}^{\mathsf{M}_{i} \text{ starting at } t+1 \text{ and } \mathbf{x}_{t}^{*}}$$

expanding flow expression with respect to identity

$$m_{i}\left(s^{t}\right) = \underbrace{\left(W\left(s^{t}\right) - \delta W\left(s^{t}, x_{t}^{*}\right)\right)}_{\text{current value without i but } x_{t}^{*}} \underbrace{\left(W_{-i}\left(s^{t}\right) - \delta W_{-i}\left(s^{t}, x_{t}^{*}\right)\right)}_{\text{current value without i but } x_{t}^{*}}$$

• note
$$W_{-i}\left(\mathbf{s}^{t}\right) \geq W_{i}\left(\mathbf{s}^{t}, \mathbf{x}_{t}^{*}\right)$$

 $\boldsymbol{m}_{i}\left(\boldsymbol{s}^{t}\right) = \left(\boldsymbol{W}\left(\boldsymbol{s}^{t}\right) - \delta\boldsymbol{W}\left(\boldsymbol{s}^{t},\boldsymbol{x}_{t}^{*}\right)\right) - \left(\boldsymbol{W}_{-i}\left(\boldsymbol{s}^{t}\right) - \delta\boldsymbol{W}_{-i}\left(\boldsymbol{s}^{t},\boldsymbol{x}_{t}^{*}\right)\right)$

- consider efficient assignment $x_t^* = i$:
- information about $x_t^* = i$ is worthless without *i*:

$$W_{-i}\left(\mathbf{s}^{t},i\right)=W_{-i}\left(\mathbf{s}^{t}\right)$$

leads to

$$m_i(\mathbf{s}^t) = \mathbf{v}_i(\mathbf{s}_i^t) - (1 - \delta) W_{-i}(\mathbf{s}^t)$$

• consider inefficient bidder: $x_t^* \neq j$:

$$\mathbf{x}_{-j,t}^* = \mathbf{x}_t^*$$

leads to

$$m_{j}\left(s^{t}
ight)=0$$

- match net payoff to flow marginal contribution
- for winner *i*:

$$\boldsymbol{m}_{i}\left(\boldsymbol{s}^{t}\right)=\boldsymbol{v}_{i}\left(\boldsymbol{s}^{t}\right)-\boldsymbol{p}_{i}\left(\boldsymbol{s}^{t}\right)$$

• for losers, $j \neq i$: $m_j(s^t) = -p_j(s^t)$

Theorem (Dynamic Second Price Auction)

The socially efficient allocation rule **x**^{*} satisfies ex post incentive and participation constraints with payment **p**^{*}:

$$p_{j}^{*}(s^{t}) = \begin{cases} (1-\delta) W_{-j}(s^{t}) & \text{if } x_{j,t}^{*} = 1, \\ 0 & \text{if } x_{j,t}^{t*} = 0. \end{cases}$$

- price equals intertemporal opportunity cost
- delay (1δ) of the optimal program for all but *j*

- with private values, static mechanism satisfies incentive compatibility in weakly dominant strategies
- in dynamic mechanism, dominant incentive compatibility fails to hold in private value environment
- truthtelling after all histories fails to be a weakly dominant strategy as it removes the ability to respond to past announcements
- yet ex post incentive compatibility can be satisfied in dynamic mechanism

General Allocation Problems

- description of a dynamic Vickrey-Clarke-Groves mechanism
- general specification of utility of each agent and arrival of private information over time
- dynamic VCG mechanism is time consistent
 - social choice function can be implemented by a sequential mechanism without ex ante commitment by the designer
 - thruthtelling strategy in the dynamic setting forms an ex-post equilibrium rather than an equilibrium in weakly dominant strategies

General Allocation Problem

- extend single unit auction to general allocation model
- net expected flow utility of agent *i* in period *t* :

 $\mathbf{v}_i\left(\mathbf{x}_t, \mathbf{s}_i^t\right) - \mathbf{p}_{i,t}$

 private signal of agent *i* in period *t* + 1 is generated by conditional distribution function:

 $\mathbf{s}_{i,t+1} \sim \mathbf{G}_i\left(\cdot \left| \mathbf{x}_t, \mathbf{s}_i^t \right) \right)$

 generalize information flow by allowing signal s_{i,t+1} of agent *i* in period t + 1 to depend on current decision x_t and entire past history of private signals of *i*

Dynamic VCG Mechanism

- efficiency
- marginal contribution pricing

Theorem (Dynamic VCG Mechanism)

The socially efficient allocation rule $\{x^*\}$ satisfies ex post incentive and ex post participation constraint with payment p^* :

$$p_{i,t}^{*}\left(x^{*}\left(s^{t}\right),s_{-i}^{t}\right)=v_{i}\left(x^{*}\left(s^{t}\right),s_{i}^{t}\right)-m_{i}\left(s^{t}\right).$$

- characterization of transfer prices via marginal contribution
- in specific environments additional insights from observing how social policies are affected by removal of agent *i*

- many transfer rules support ex post incentive and ex post participation constraints in dynamic setting
- temporal separation between allocative influence and monetary payments may be undesirable for may reasons:
 - agent *i* could be tempted to leave the mechanisms and break her commitment *after* she ceases to have a pivotal role but *before* her payments come due
 - if arrival and departure of agents is random, then an agent could falsely claim to depart to avoid future payments
- in intertemporal environment if agent *i* ceases to influence current or future allocative decisions in period *t*, then she also ceases to have monetary obligations

given state s^τ the presence of *i* is immaterial for the efficient decision x^{*}_u if

$$\tau = \min \left\{ t \left| \begin{array}{c} x_u^* \left(s^u \right) = x_{-i,u}^* \left(s^u \right), \forall u \ge t, \\ \forall s^u = \left(s^\tau, s_{\tau+1}, ..., s_t, ..., s_u \right) \end{array} \right\}$$

Definition (Efficient Exit)

A mechanism satisfies the efficient exit condition if for all *i*, s^{τ} and τ :

 $p_{i,u}(s^u) = 0$, for all $u \ge \tau$.

 weak online requirement: decisions regarding agent *i* have to be made in the presence of agent *i*

Theorem (Uniqueness)

If a dynamic direct mechanism is efficient, satisfies ex post incentive and participation constraints and the efficient exit condition, then it is the dynamic marginal contribution mechanism.

Conclusion

- direct dynamic mechanism in private value environments with transferable utility
- design of monetary transfers relies on notions of marginal contribution and flow marginal contribution
- transfer the insights of VCG mechanism from static to dynamic settings
- many interesting questions are left open
 - current contribution is silent on issue of revenue maximizing mechanisms
 - characterization of implementable allocations in dynamic setting will first be necessary
 - restriction to private value environments