

# A.P. SHAH INSTITUTE OF TECHNOLOGY

# Department of Computer Science and Engineering Data Science



# Vickrey and VCG mechanisms

a Vickrey-Clarke-Groves (VCG) mechanism is a generic truthful mechanism for achieving a socially optimal solution. It is a generalization of a Vickrey-Clarke-Groves auction. A VCG auction performs a specific task: dividing items among people. A VCG mechanism is more general: it can be used to select any outcome out of a set of possible outcomes.

### **Notation**

There is a set X of possible outcomes.

There are n agents, each of which has a set of outcome valuations. The valuation of agent i is represented as a function:

$$v_i:X o \mathbb{R}_+$$

which expresses the value it has for each alternative, in monetary terms.

It is assumed that the agents have quasilinear utility functions; this means that, if the outcome is x and in addition the agent receives a payment  $p_i$  (positive or negative), then the total utility of agent i is:

$$u_i := v_i(x) + p_i$$

Our goal is to select an outcome that maximizes the sum of values, i.e.:

$$x^{opt}(v) = rg \max_{x \in X} \sum_{i=1}^n v_i(x)$$

In other words, our social-choice function is utilitarian.

The VCG family is a family of mechanisms that implements the utilitarian welfare function. A typical mechanism in the VCG family works in the following way:

- 1. It asks the agents to report their value function. I.e, each agent i should report  $v_i(x)$  for each option x.
- 2. Based on the agents' report-vector v, it calculates  $x^*=x^{opt}(v)$  as above.
- 3. It pays, to each agent i, a sum of money equal to the total values of the *other* agents:

$$p_i := \sum_{j \neq i} v_j(x^*)$$

4. It pays, to each agent i, an additional sum, based on an arbitrary function of the values of the other agents:

$$p_i + h_i(v_{-i})$$

where  $v_{-i} = (v_1, \dots, v_{i-1}, v_{i+1}, \dots, v_n)$ , that is,  $h_i$  is a function that depends only on the valuations of the other agents.



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Every mechanism in the VCG family is a truthful mechanism, that is, a mechanism where bidding the true valuation is a dominant strategy.

The agent is paid the total value of the other agents; hence, together with its own value, the total welfare of the agent is exactly equal to the total welfare of society. Hence, the incentives of the agent are aligned with those of the society and the agent is incentivized to be truthful in order to help the mechanism achieve its goal.

The function hi, in step 4, does not affect the agent's incentives, since it depends only on the declarations of the other agents.

### The Clarke pivot rule

The function  $h_i$  is a parameter of the mechanism. Every selection of  $h_i$  yields a different mechanism in the VCG family.

We could take, for example:

$$h_i(v_{-i}) = 0,$$

but then we would have to actually pay the players to participate in the auction. We would rather prefer that players give money to the mechanism.

An alternative function is:

$$h_i(v_{-i}) = -\max_{x \in X} \sum_{j 
eq i} v_j(x)$$

It is called the *Clarke pivot rule*. With the Clarke pivot rule, the total amount paid by the player is:

(social welfare of others if i were absent) - (social welfare of others when i is present).

This is exactly the externality of player i.

When the valuations of all agents are weakly-positive, the Clarke pivot rule has two important properties:

- Individual rationality: for every player i,  $v_i(x) + p_i \ge 0$ . It means that all the players are getting positive utility by participating in the auction. No one is forced to bid.
- No positive transfers: for every player  $i, p_i \leq 0$ . The mechanism does not need to pay anything to the bidders.

This makes the VCG mechanism a win-win game: the players receive the outcomes they desire, and pay an amount which is less than their gain. So the players remain with a net positive gain, and the mechanism gains a net positive payment.

### Weighted VCG mechanism



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Instead of maximizing the sum of values, we may want to maximize a weighted sum:

$$x^{opt}(v) = rg \max_{x \in X} \sum_{i=1}^n w_i v_i(x)$$

where wi is a weight assigned to agent

The VCG mechanism from above can easily be generalized by changing the price function in step 3 to:

$$p_i := rac{1}{w_i} \sum_{j 
eq i} w_j v_j(x^*)$$

### **Cost minimization**

The VCG mechanism can be adapted to situations in which the goal is to minimize the sum of costs (instead of maximizing the sum of gains). Costs can be represented as negative values, so that minimization of cost is equivalent to maximization of values.

The payments in step 3 are negative: each agent has to pay the total cost incurred by all other agents. If agents are free to choose whether to participate or not, then we must make sure that their net payment is non-negative (this requirement is called individual rationality). The Clarke pivot rule can be used for this purpose: in step 4, each agent i is paid the total cost that would have been incurred by other agents, if the agent i would not participate. The net payment to agent i is its marginal contribution to reducing the total cost.

### **Applications**

# Auctions

Vickrey–Clarke–Groves auction is an application of VCG mechanism for welfare maximization. Here, X is the set of all possible allocations of items to the agents. Each agent assigns a personal monetary value to each bundle of items, and the goal is to maximize the sum of the values of all agents.

A well-known special case is the Vickrey Auction. Here, there is only a single item, and the set X contains n+1 possible outcomes: either sell the item to one of the n agents, or not sell it at all. In step 3, the winner agent is paid 0 (since the total value of the others is 0) and the losers receive a payment equal to the declared value of the winner. In step 4, the winner pays the second-highest bid (the total value of the others had he not participated) and the losers pay the declared value of the winner (the total value of the others had they not participated). All in all, the winner pays the second-highest bid and the losers pay 0.

A VCG mechanism can also be used in a double auction. It is the most general form of incentive-compatible double-auction since it can handle a combinatorial auction with arbitrary value functions on bundles. Unfortunately, it is not budget-balanced: the total value paid by the



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buyers is smaller than the total value received by the sellers. Hence, in order to make it work, the auctioneer has to subsidize the trade.

### **Quickest paths**

The **quickest path** problem is a cost-minimization problem.<sup>[3]</sup>The goal is to send a message between two points in a communication network, which is modeled as a graph. Each computer in the network is modeled as an edge in the graph. Different computers have different transmission speeds, so every edge in the network has a numeric cost equal to the number of milliseconds it takes to transmit the message. Our goal is to send the message as quickly as possible, so we want to find the path with the smallest total cost.

If we know the transmission-time of each computer (-the cost of each edge), then we can use a standard algorithm for solving the shortest path problem. If we do not know the transmission times, then we have to ask each computer to tell us its transmission-time. But, the computers have their own selfish interests so they might not tell us the truth. For example, a computer might tell us that its transmission time is very large, so that we will not bother it with our messages.

The VCG mechanism can be used to solve this problem. Here, X is the set of all possible paths; the goal is to select a path  $x \in X$  with minimal total cost.

The value of an agent,  $v_i(x)$ , is minus the time it spent on the message: it is negative if  $i \in x$  and it is zero if  $i \notin x$ . The payment in step 3 is negative: each agent should pay to us the total time that the other agents spent on the message (note that the value is measured in units of time. We assume that it is possible to pay computers in units of time, or that it there is a standard way to translate time to money). This means that, together with its own spent time, each agent actually loses the total time it took the message to arrive its destination, so the agent is incentivized to help the mechanism achieve the shortest transmission time. computers in units of time, or that it there is a standard way to translate time to money). This means that, together with its own spent time, each agent actually loses the total time it took the message to arrive its destination, so the agent is incentivized to help the mechanism achieve the shortest transmission time.

The Clarke pivot rule can be used to make the mechanism individually-rational: after paying us the cost, each agent receives from us a positive payment, which is equal to the time it would have taken the message to arrive at its destination if the agent would not have been present. Obviously, this time is weakly larger than the time required when the agent is present, so the net gain of every agent is weakly positive. Intuitively, each agent is paid according to its marginal contribution to the transmission.

### **Computational issues**

A VCG mechanism has to calculate the optimal outcome, based on the agents' reports (step 2 above). In some cases, this calculation is computationally difficult. For example, in combinatorial auctions, calculating the optimal assignment is NP-hard.