

Mechanism Design Powered by Social Interactions

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2009 DARPA Red Balloon Challenge

- The \$40,000 challenge award would be granted to the first team to submit the locations of 10 moored, 8-foot, red weather balloons at 10 previously undisclosed fixed locations in the continental United States.



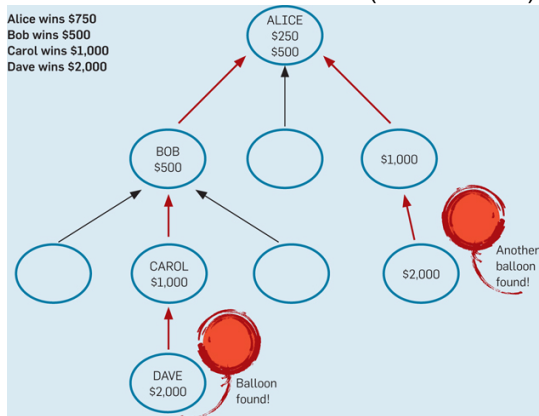
2009 DARPA Red Balloon Challenge

MIT Crowdsourced Solution (The Winner):

- *"We're giving **\$2000** per balloon to the first person to send us the correct coordinates, but that's not all – we're also giving **\$1000** to the person who invited them. Then we're giving **\$500** whoever invited the inviter, and **\$250** to whoever invited them, and so on ..."*
- got over **5,000** of participants, won the competition in under 9 hours.

2009 DARPA Red Balloon Challenge


MIT Crowdsourced Solution (The Winner):



- Pickard, G., et al., **Time-Critical Social Mobilization**. Science, 2011. 334(6055): p. 509-12.

PinDuoDuo (like Groupon)

人人都是产品经理
WWW.WOJIAOPIN.COM



The logo is a red heart shape composed of 12 hexagons. Each hexagon contains a white icon representing different product categories: a dress, a t-shirt, a wine glass, an apple, a handbag, a camera, a fish, a high-heeled shoe, a camera, a coffee cup, an umbrella, and a shopping bag. The center hexagon contains the large white Chinese character '拼' (Pīn), which means 'to join' or 'to share'.


3亿人都在“拼多多”

拼团价 ¥49.9

单买价 ¥99.9



The smartphone screen displays the PinDuoDuo app interface. At the top, it says '拼多多'. Below that, there's a banner for a group purchase of snacks, showing a pile of nuts and the group price '拼团价 ¥49.9'. Below the banner, it shows the individual price '单买价 ¥99.9'. At the bottom of the screen, there's a red button with the text '9.45.30'.



A woman in a red sweater is smiling and holding a wrapped gift box. She is standing next to the smartphone, which is displaying the group purchase offer.

What are the incentives?

More participants, higher chance to win!!!

- 2009 DARPA Red Balloon Challenge
 - Inviting more friends has **higher chance to win** (higher utility)
- PinDuoDuo
 - Inviting more friends has **higher chance to get cheap items** (higher utility)

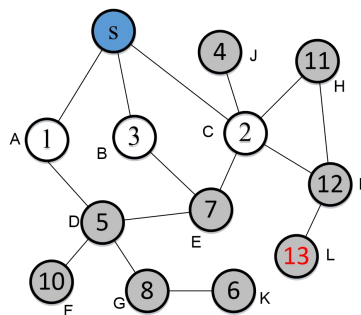
What if it is a competition?

- Resource allocation (auctions)
- Task allocation (crowdsourcing)
- Information propagation with budget
- Social choice (voting)

More participants means lower chance to win!!!

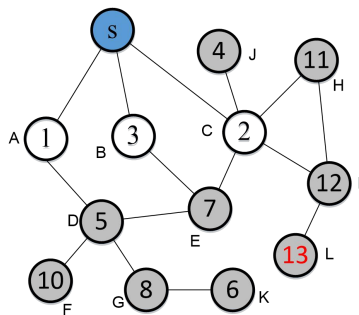
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Starter: Promote a Sale via Social Networks



- The seller (**blue node**) sells one item and has only three connections/neighbours in the network (A,B,C).
- Each node is a potential buyer and the value is her highest willing payment to buy the item (**valuation**).
- The seller's revenue of applying second price auction (VCG) without promotion is **2**.
- but the highest willing payment in the network is **13**.

Starter: Promote a Sale via Social Networks



Question

How the seller could do to increase her profit?

Traditional Sale Promotions

Traditional sale promotions:

- Promotions via agents
- Keywords based ads via **search engines** such as Google
- Ads via **social media** such as WeChat, Facebook, Twitter

Traditional Sale Promotions

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Challenge

- The return of these promotions are unpredictable.
- The seller may **LOSE** from the promotions.

Tackle the Challenge

Build promotion inside the market mechanism such that

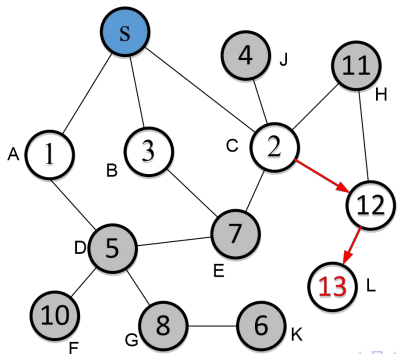
- 1 the promotion will **never bring negative utility/revenue** to the seller.
- 2 all **buyers** who are aware of the sale **are incentivized to diffuse** the **sale information** to all her neighbours.

"Diffusion Mechanism Design"

New Challenges

Why a buyer would bring more buyers to compete with her?

- only if their **diffusion** are **rewarded**, *but the seller doesn't want to lose!*
- we **cannot** just **pay** each node a **fixed amount** to incentivise them to diffuse the information.



Outline

1 Mechanism Design Review

- The History
- Second Price Auction (VCG)

2 Diffusion Mechanism Design

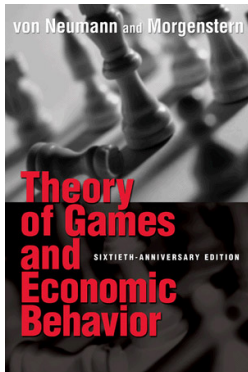
- Second Price Auction (VCG)

- Information Propagation

What is Mechanism/Market Design?

- it is known as Reverse Game Theory

- **Game theory** is the study of mathematical models of **conflict** and **cooperation** between intelligent rational decision-makers (wiki) [von Neumann and Morgenstern 1944].

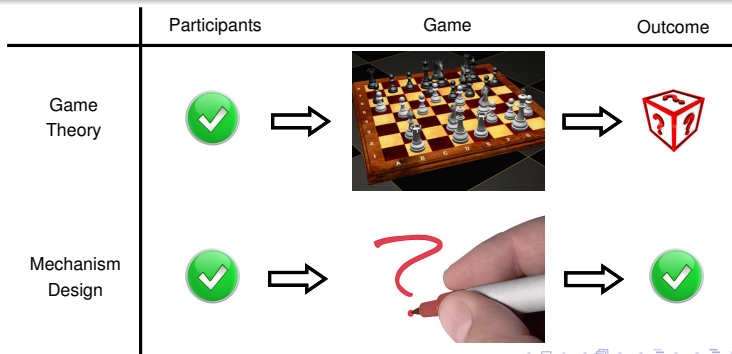


- **Non-cooperative games**: Go, poker, rock-paper-scissors
- **Cooperative games**: coordination games

Mechanism Design is to answer...

Question

How to **design** a mechanism/game, toward desired objectives, in strategic settings?



Mechanism Design (Reverse Game Theory)

Mechanism Design is to answer...

Question

How to **design** a mechanism/game, toward desired objectives, in strategic settings?

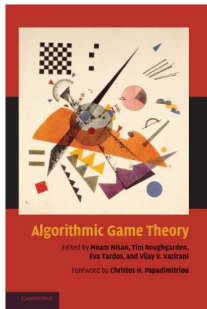


Roger B. Myerson (born March 29, 1951, University of Chicago, US)

- **Nobel Prize** for economics (2007), for "having laid the foundations of **mechanism design theory**."
- ***Eleven game-theorists** have won the economics Nobel Prize.*

Algorithmic Game Theory (AGT)

- **Algorithmic game theory** is an area in the intersection of **game theory** and **algorithm design**, whose objective is to design algorithms in strategic environments (wiki) [Nisan et al. 2007].



- *Computing in Games*: algorithms for computing equilibria
- *Algorithmic Mechanism Design*: design games that have both good game-theoretical and algorithmic properties
- ...

- Algorithmic game theory research in AI:
 - **Game Playing**: computation challenges, AlphaGo, poker
 - **Social Choice**: preferences aggregation, voting, prediction
 - **Mechanism Design**: the allocation of scarce resources, ad auctions
- Many IJCAI Computers and Thought Award (outstanding young scientists in artificial intelligence) winners had worked on AGT:
 - Sarit Kraus (1995), Nicholas Jennings (1999), Tuomas Sandholm (2003), Peter Stone (2007), Vincent Conitzer (2011), and Ariel Procaccia (2015)

- The History

- Second Price Auction (VCG)

- Resource Allocation

- Task Allocation

- Information Propagation

How can a house-seller sell her house with the "highest" revenue?



1.2 mil.



1.5 mil.

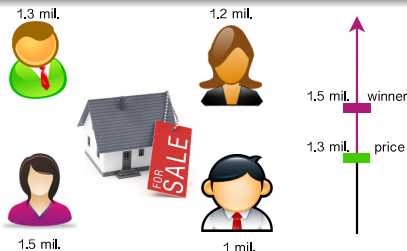
1 mil.

- **Challenge:** the seller **doesn't know** how much the buyers are willing to pay (**their valuations**).

A Mechanism Design Example

Design Goal

How can a house-seller sell her house with the "highest" revenue?



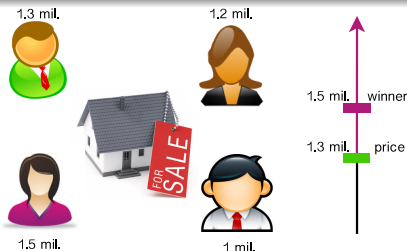
Solution: **Second Price Auction** (Vickrey Auction/VCG)

- **Input:** each buyer **reports a price/bid** to the seller
- **Output:** the seller decides
 - *allocation:* the agent with the **highest price wins**.
 - *payment:* the winner **pays the second highest price**.

A Mechanism Design Example

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Solution: **Second Price Auction** (Vickrey Auction/VCG)

Properties:

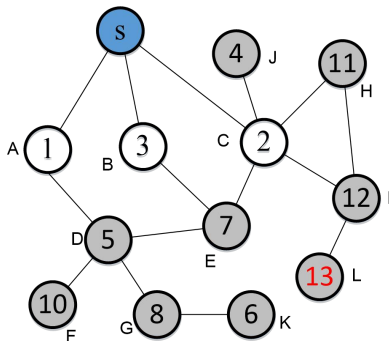
- **Efficient:** maximising **social welfare**
- **Truthful:** buyers **report** their valuations **truthfully**

What can the seller do to FURTHER increase her profit?

- estimate a good **reserve price** [Myerson 1981]
 - *requires a good estimation of buyers' valuations*
- **promotions**: let more people know/participate in the auction

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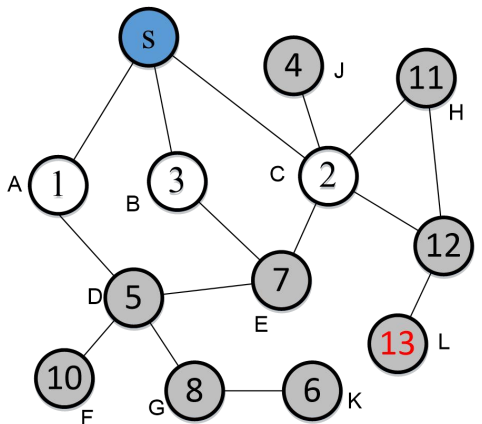
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Our Solutions: Information Diffusion Mechanisms

- Bin Li, Dong Hao, Dengji Zhao, Tao Zhou: **Mechanism Design in Social Networks**. AAAI'17.
- Dengji Zhao, Bin Li, Junping Xu, Dong Hao, Nick Jennings: **Selling Multiple Items via Social Networks**. AAMAS'18.
- Bin Li, Dong Hao, Dengji Zhao, Tao Zhou: *Customer Sharing in Economic Networks with Costs*. IJCAI-ECAI'18.
- Bin Li, Dong Hao, Dengji Zhao, Makoto Yokoo: *Diffusion and Auction on Graphs*. IJCAI'19.
- Wen Zhang, Dengji Zhao, Hanyu Chen: *Redistribution Mechanism on Networks*. AAMAS'20.
- Wen Zhang, Dengji Zhao, Yao Zhang: *Incentivize Diffusion with Fair Rewards*. ECAI'20.
- Bin Li, Dong Hao, Dengji Zhao: **Incentive-Compatible Diffusion Auctions**. IJCAI'20.

The First Diffusion Mechanism

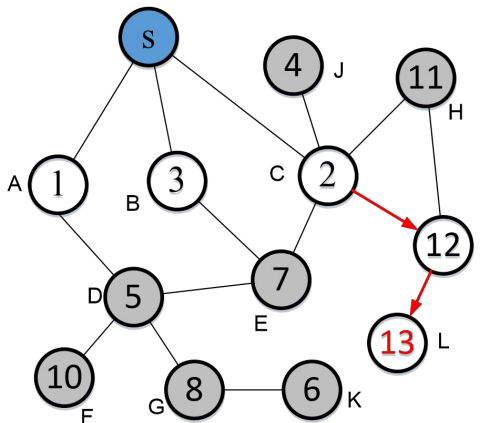
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Information Diffusion Paths

An information diffusion path from the seller to node L:

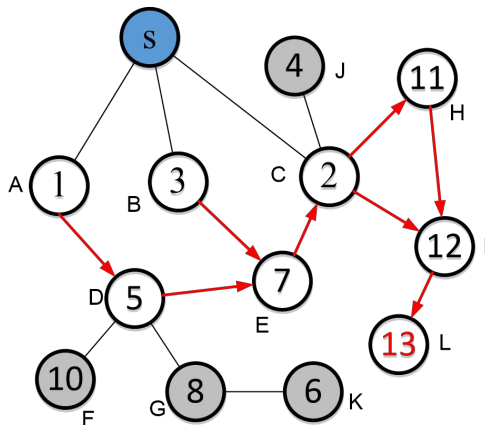
$s \rightarrow C \rightarrow I \rightarrow L$



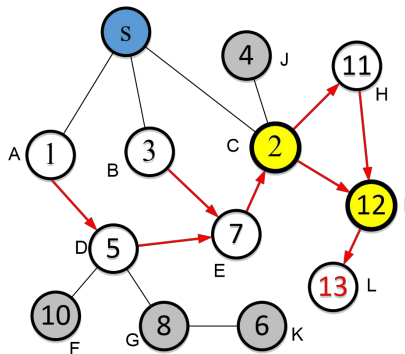
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An information diffusion path from the seller to node L:

$s \rightarrow C \rightarrow I \rightarrow L$



Diffusion Critical Nodes

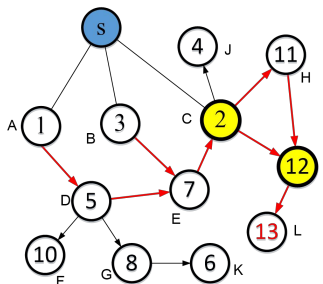


Definition

i is j 's **diffusion critical node** if **all** the information diffusion **paths** started from the seller s to j have to **pass** i .

- nodes **C** and **I** are L's only diffusion critical nodes.

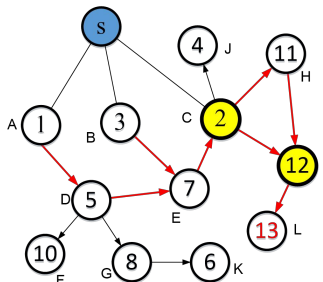
- If a buyer or one of her "*diffusion critical children*" gets the item, then the buyer pays **the highest bid of the others** (without the buyer's participation);
- otherwise, her payment is zero.



Information Diffusion Mechanism [Li et al. AAI'17]

The **payment** definition (second-price-like):

- If a buyer or one of her "*diffusion critical children*" gets the item, then the buyer pays **the highest bid of the others** (without the buyer's participation);
- otherwise, her payment is zero.



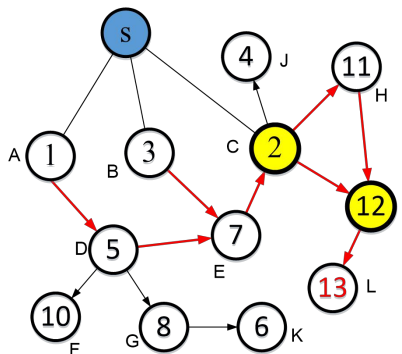
If the item is allocated to *L*, the payments of *C*, *I* and *L* are **10, 11, 12** respectively .

- Identify the node i with the highest bid and the node's **diffusion critical node path** $P_{c_i} = (c_i^1, c_i^2, \dots, i)$.
- Give the item to the first node of P_{c_i} , the node pays to the seller and then decides to whether keep the item or pass it to the next node in P_{c_i} :
 - If **the payment of the next node is greater than the bid of the current node**, passes it to the next node and receives the payment from the next node; the next node makes a similar decision;
 - otherwise, keep the item.

The Information Diffusion Mechanism

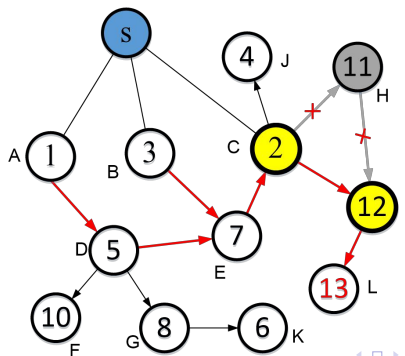
The **outcome** of the Information Diffusion Mechanism:

- the item is allocated to node I.
- node I pays 11 to C, C pays 10 to the seller.
- the **utilities** of I, C, the seller are 1, 1, 10.

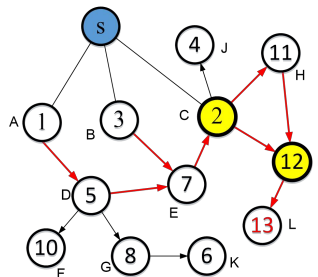


Why Buyers are Happy to Diffuse the Information?

- buyers receive the information earlier have higher priority to win the item (**C chooses before I and I chooses before L**).
- diffuse the information to more buyers will potentially increase their reward (**if C does not invite H, her utility is 0**).



Properties of the Information Diffusion Mechanism



- **Truthful**: report true valuation and diffuse the sale information to all her neighbours is a dominate strategy.
- **Individually Rational**: no buyer will receive a negative utility to join the mechanism.
- **Seller's Revenue Improved**: the seller's revenue is **non-negative** and is \geq that of the VCG without diffusion.

What Next?

- Diffusion mechanisms for combinatorial exchanges
- Diffusion with costs and delays
- Network structure based revenue analysis
- Applications/implementations in the existing social networks
- Other mechanisms to further improve the revenue and/or the efficiency

Diffusion Mechanisms for Combinatorial Exchanges

Challenge

How to generalise the mechanism to combinatorial settings?

Diffusion Mechanisms for Combinatorial Exchanges

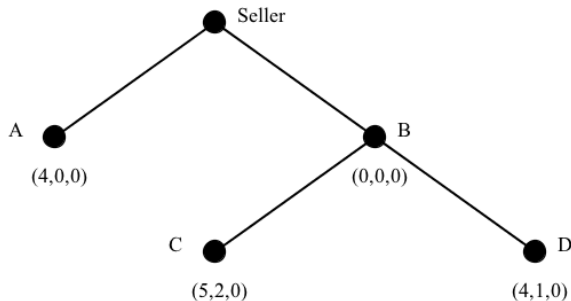
Consider the following simple setting:

- A seller sells three units of one commodity, e.g. MacBook computers.
- Each buyer has a **diminishing marginal utility** for consuming the goods.

Diffusion Mechanisms for Combinatorial Exchanges

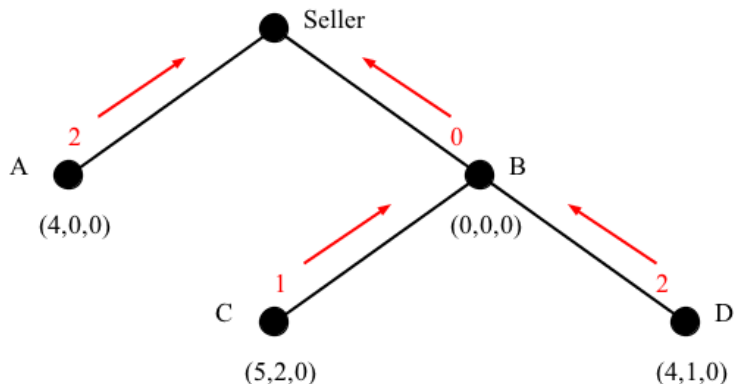
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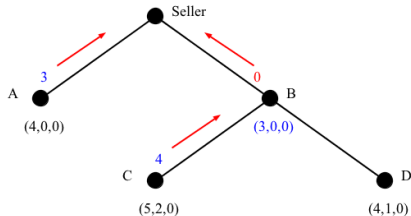
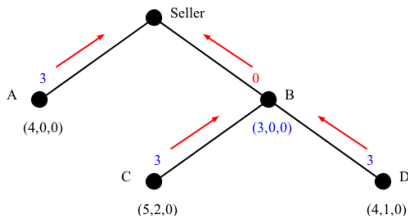
Diffusion Mechanisms for Combinatorial Exchanges

If we simply apply our information diffusion mechanism:



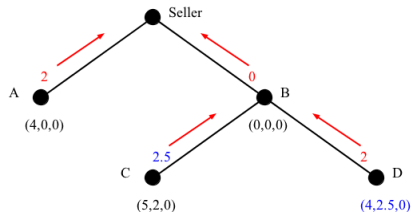
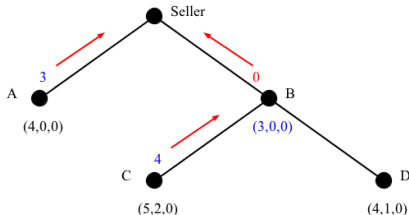
Diffusion Mechanisms for Combinatorial Exchanges

What if buyer B's valuation is $(3, 0, 0)$?



Diffusion Mechanisms for Combinatorial Exchanges

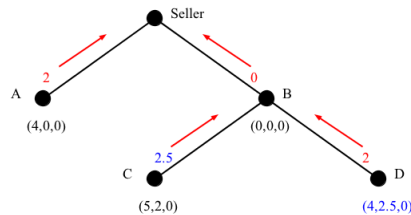
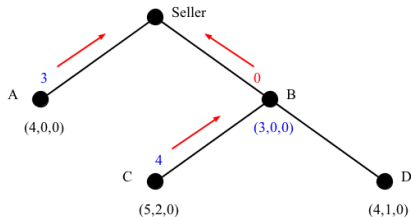
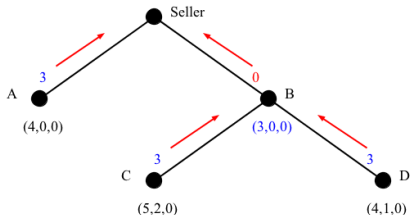
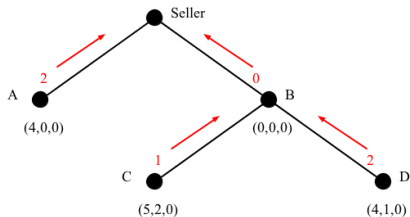
What if buyer D's valuation is $(4, 2.5, 0)$?



Diffusion Mechanisms for Combinatorial Exchanges

Challenge

There is a very complex **Decision Making** at each node!!!



Why is it so complex when there are multiple items?

To achieve truthfulness:

- The mechanism has to **maximise each node's utility** under truthful reporting.
- Each node's **payment** should **not depend on** her **valuation**.

The complexity we had:

- A node can **influence her payments** by controlling the items passed to her children.
- A node can **influence the payments of her peers**, without changing her own allocation and payments.
- **This leads to a decision loop (very complex optimization) and may not be able to maximise everyone's utility.**

Reduce the Complexity

The Main Idea

A node **CANNOT** influence the payments she receives by controlling the items passed to her children.

Simplify the decision complexity we had:

- A node can **influence her received payments** by controlling the items passed to her children.
- A node can **influence the payments of her peers**, without changing her own allocation and payments.
- This leads to a decision loop and may not be able to maximise everyone's utility.

Solution: Sells Multiple Homogeneous Items

Selling Multiple Items via Social Networks [Zhao et al. AAMAS'18]

- generalises the result from [Li et al. 2017];
- agent i 's reward/payment doesn't depends on how many of i 's children received items;
- agent pays to the seller directly rather than to their parent;

The setting:

- A seller sells $K \geq 1$ homogeneous items;
- each buyer requires at most one item (single-unit demand);
- the rest is the same as [Li et al. 2017].

Are the mechanisms fair?

- According to the theorem of small-world networks, the chance for a node to be a cut-point in a well-connected network is **very low**.
- We hope to give rewards to all the related buyers not only the cut-points on the paths to the winner.

Solution: Redistribute Rewards among Agents

Incentivize Diffusion with Fair Rewards [Zhang et al. ECAI'20]

- redistribute rewards among critical ancestors based on IDM
- all critical ancestors have positive expected utilities
- seller's revenue is not reduced

What if for non-profit purpose?

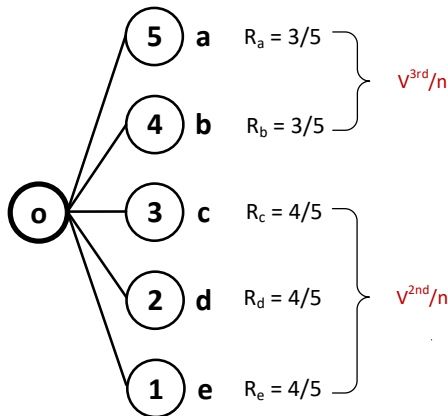
Redistribution mechanism is:

- to do the efficient resource allocation
- not for profit

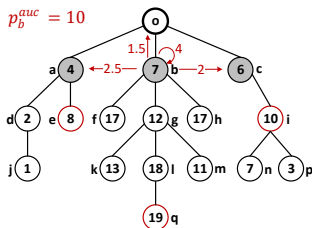
Challenges:

- how to achieve a more efficient allocation?
- how to maintain wealth among agents?

Cavallo Mechanism



- **Allocation:** agent with the highest valuation wins the item
- **Payment:** winner pays the second price
- **Redistribution:** each agent receives the owner's revenue without her participation divided by the total number of agents



- **Allocation:** keep the item if her valuation is greater than or equal to her payment
- **Payment:** the highest valuation without her participation
- **Redistribution:** a monotone increasing function to the number of descendants

A General Characterization of Diffusion Auction Mechanisms

Incentive-Compatible Diffusion Auctions [Li et al. IJCAI'20]

- characterize a **sufficient and necessary** condition for all incentive-compatible and individually rational diffusion auctions.
- propose a class of natural monotonic allocation policies with optimal payment policy that **maximizes the seller's revenue**.

The sufficient and necessary condition

Theorem:

- A diffusion auction (π, x) is incentive-compatible and individually rational if and only if for all type profile \mathbf{t} and all i , $P1 - P5$ are satisfied, where
 - $P1$: π is value-monotonic,
 - $P2$: \tilde{x}_i and \bar{x}_i are bid-independent,
 - $P3$: $\tilde{x}_i(r_i) - \bar{x}_i(r_i) = v_i^*(r_i)$,
 - $P4$: \tilde{x}_i and \bar{x}_i are diffusion-monotonic,
 - $P5$: $\bar{x}_i(\emptyset) \leq 0$.

Open Questions

- More general settings
 - characterize truthful diffusion mechanisms, **revenue monotonicity** is the key?
- When there is a diffusion cost
 - how to guarantee the seller will not lose?
- Privacy concern and the seller's strategies
 - the seller discovery the whole network and she may cheat as well!
- False-name manipulations
 - a node may create multiple ids as her neighbours to gain more payment?
- many more...

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 - **Task Allocation**
 - Information Propagation

Diffusion Mechanism Design for Task Allocation

- Wen Zhang, Yao Zhang, Dengji Zhao: *Collaborative Data Acquisition*. AAMAS'20.
- Yao Zhang, Xiuzhen Zhang, Dengji Zhao: *Sybil-proof Answer Querying Mechanism*. IJCAI'20.

Diffusion Mechanism Design for Task Allocation

Resource allocation vs task allocation:

- 1 task requires more participants' contribution (**collaboration**)
- 2 but participants' contribution may conflict with each other (**competition**)

Crowdsourcing Data Acquisition [Zhang et al. AAMAS'20]

- a requester is collecting data from the crowd
- more participants gives richer dataset
- participants' contribution depends on the quality of their provided data
- if two participants offer the same data, how to calculate their contribution?

Crowdsourcing Data Acquisition [Zhang et al. AAMAS'20]

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Shapley Value?

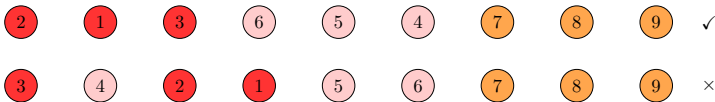
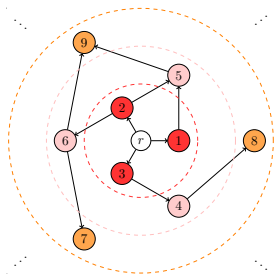
Shapley Value

The problem of the Shapley value:

- two participants offer the same data will share the same Shapley value
 - the Shapley value is doubled if one of them didn't participate

Solution: Layered Shapley Value

- participants are layered
- the Shapley value is calculated for each lower layer first
- the calculation for higher layer assumes that lower layers' participants are always in the coalition



Solution: Layered Shapley Value

$$\hat{\phi}_i = \sum_{S \subseteq L_i - \{i\}} \frac{|S|!(|L_i| - |S| - 1)!}{|L_i|!} \cdot \left(v \left(D'_{L_{i-1}^* \cup S \cup \{i\}} \right) - v \left(D'_{L_{i-1}^* \cup S} \right) \right)$$

Properties:

- participants are incentivized to invite more participants
(new participants do not compete with them)
- the requester does not need to pay for redundant data

Sybil-proof Answer Querying [Zhang et al. IJCAI'20]

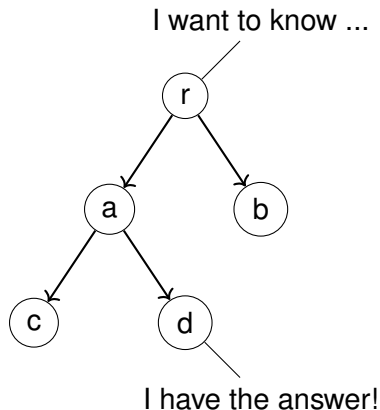


Figure: Query via Network

- Online networks has offered many opportunities for people to collaborate remotely in real time, e.g. P2P file sharing and Q&A platforms.
- Utilizing the social connections, we can enhance the power of answer querying via networks, e.g. DARPA Red Balloon Challenge.

Sybil-proof Answer Querying

- **Fact:** An SIR path mechanism cannot be both SP and CP.
- What if relax SIR to IR?
- **Theorem:** A path mechanism is IR, SP and CP if and only if it is a two-headed mechanism.

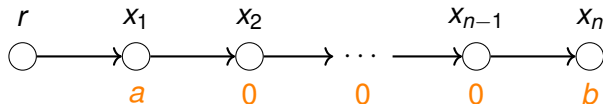


Figure: The rewards distributed by a two-headed mechanism

- What if relax CP to λ -CP?
- **New Idea:** Double Geometric Mechanism

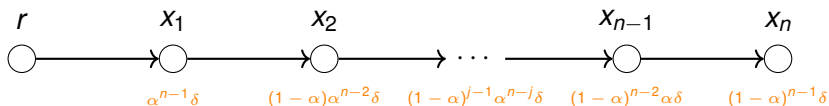


Figure: The rewards distributed by a double geometric mechanism

- Characterization:** Under mild condition, the properties of IC, SIR, BC, SP, 2-CP and ρ -SS determines a DGM.

- 1 Mechanism Design Review
 - The History
 - Second Price Auction (VCG)
- 2 Diffusion Mechanism Design
 - Resource Allocation
 - Task Allocation
 - Information Propagation

Mechanism Design for Information Propagation

- Haomin Shi, Yao Zhang, Zilin Si, Letong Wang, Dengji Zhao: *Maximal Information Propagation with Budgets*. ECAI'20.

Maximal Information Propagation with Budgets

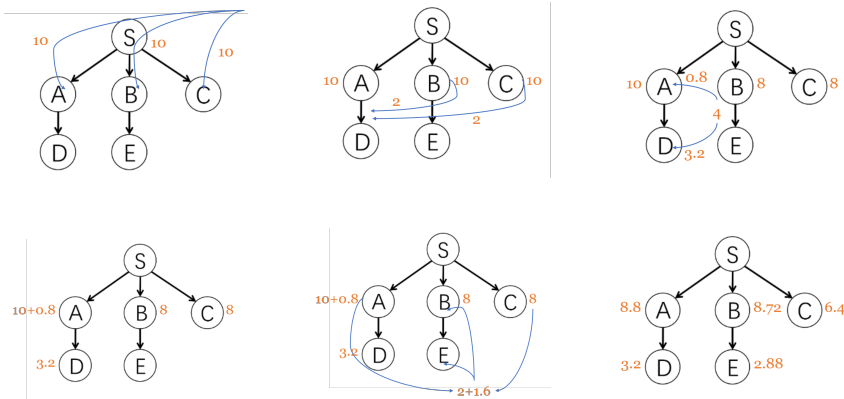
- The sponsor s wants to propagate some information to the social network modelled as a directed acyclic graph $G = (N, E)$.
- The sponsor holds a fixed budget B , which is prepared as agents' rewards.

Challenge

How to find a reward scheme that is propagation incentive compatible and budget balanced?

Maximal Information Propagation Example

● Example: Incentivies from Peer Pressure



What we have covered

Mechanism Design Powered by Social Interactions

- Diffusion Mechanism for Resource Allocation (competitive environment)
 - for selling single and multiple items
- Diffusion Mechanism for Task Allocation (both competitive and collaborative)
 - crowdsourcing, sybil-proof, execution uncertainty
- Diffusion Mechanism for Information Propagation
 - information propagation with budgets

<http://dengji-zhao.net/ijcai19.html>