

Task Routing for Prediction Tasks

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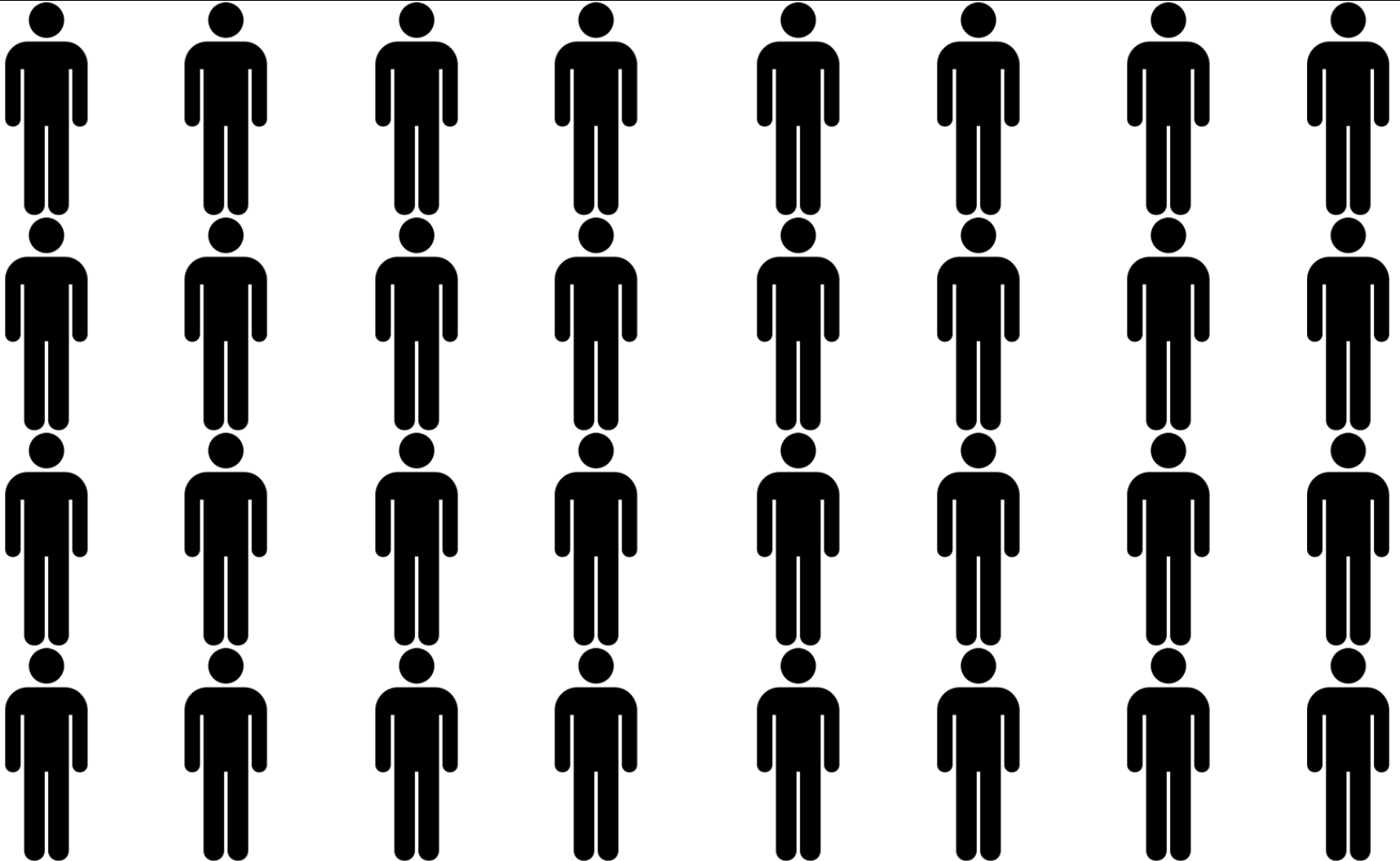
Eric Horvitz

Yiling Chen

David Parkes

Let's do a poll.

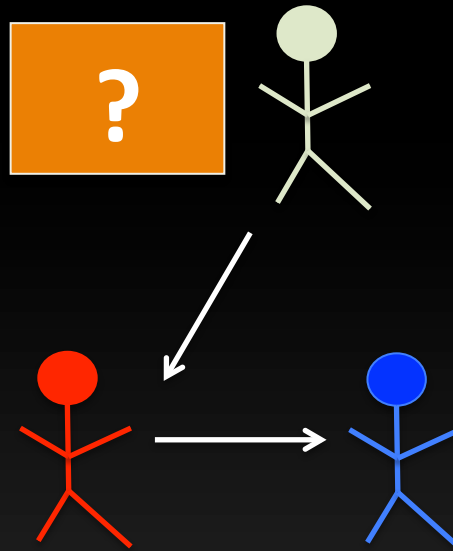
Expertise is among us



We know each other's expertise



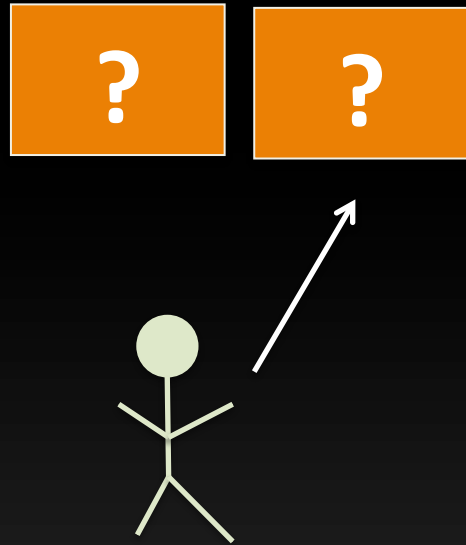
Leveraging expertise and knowledge of others' expertise



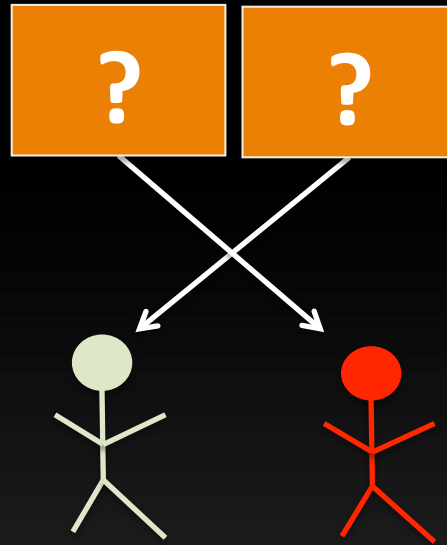
Task Routing

Leverage people's abilities to jointly solve problems and to route problems to others who can further contribute.

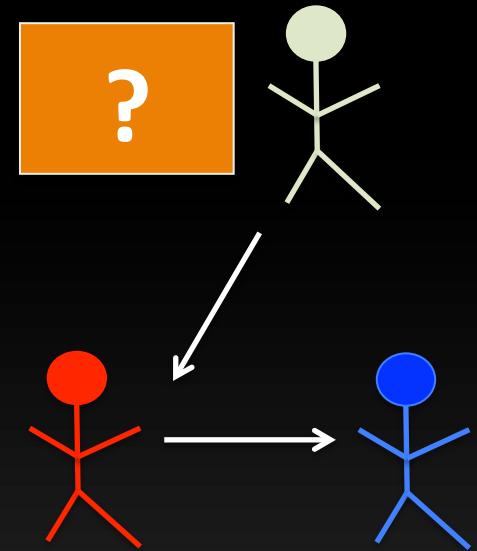
Pull



Push



Route



Our work

Our work

- Task routing for prediction tasks
 - Incentives for solving and routing

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- Two cases:
 -

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 - Common knowledge

Our work

- Task routing for prediction tasks
 - Incentives for solving and routing
- Two cases:
 - Common knowledge
 - Local common knowledge
 - Local routing rules (**tractable and good**)

MIT's red balloon mechanism

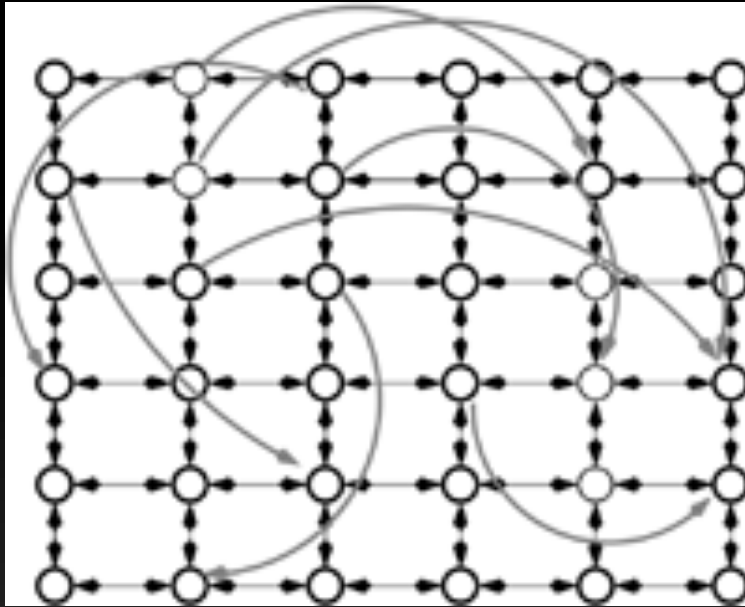
[Pickard et al.]



Provides incentives for
“spreading the word.”

Decentralized Search

[Kleinberg]



Route message towards destination using only **local information**

Model

Model

- Task: predict answer to a **yes/no question** whose answer is revealed in the future

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- Discrete time $t = 1, 2, 3, \dots, R$

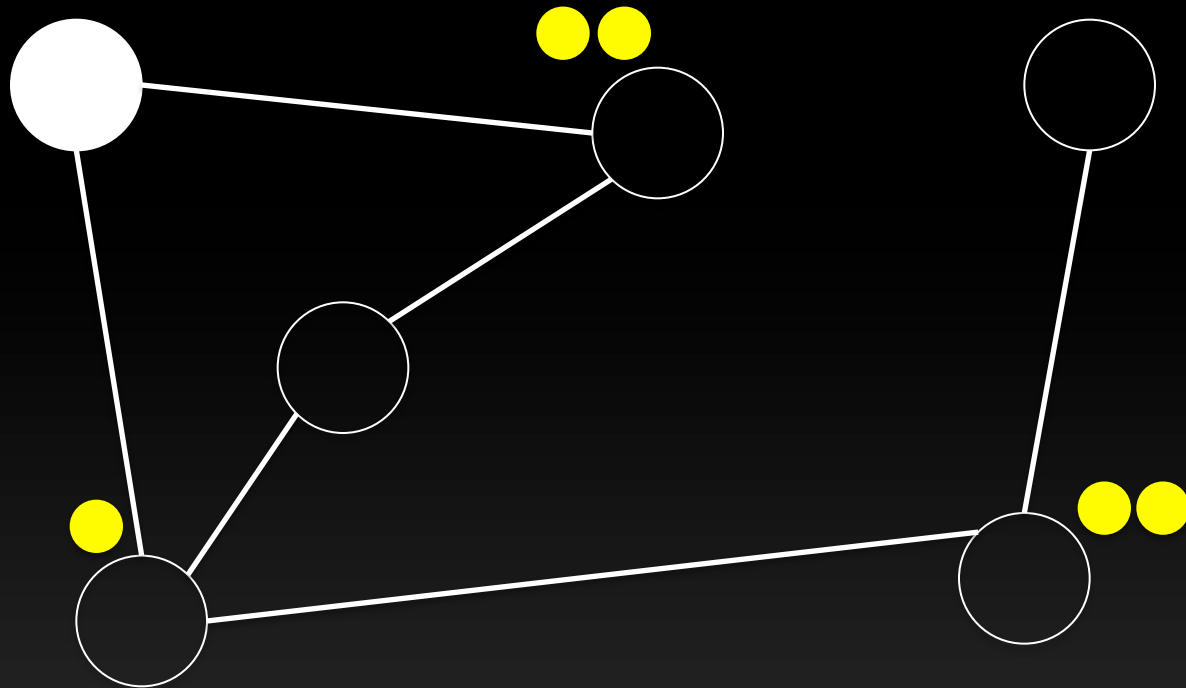
Model

- Task: predict answer to a **yes/no question** whose answer is revealed in the future
- Discrete time $t = 1, 2, 3, \dots, R$
- A person receiving the task can:
 - update the current probability estimate
 - route the task to a neighbor on the network

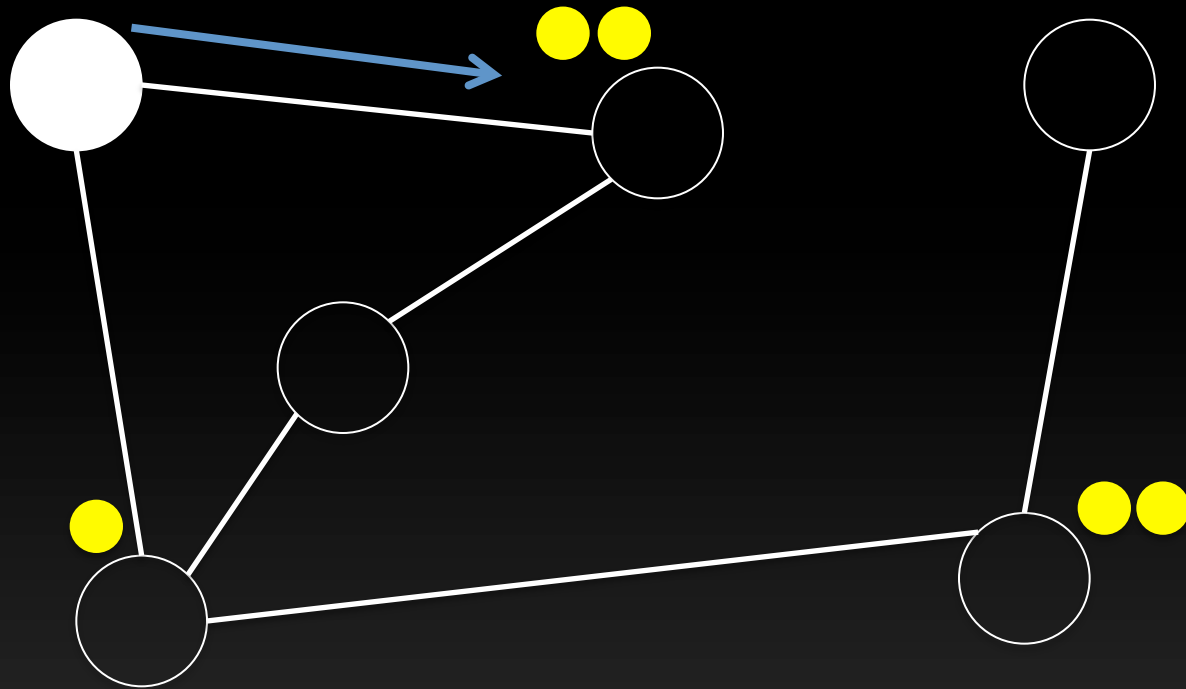
Information structure

- Players observe **coin flips**
 - Probabilistic signal of what the answer is
 - More coin flips == more information
- Assume coin flips are **conditionally independent**
- Players may know how many coin flips others observe

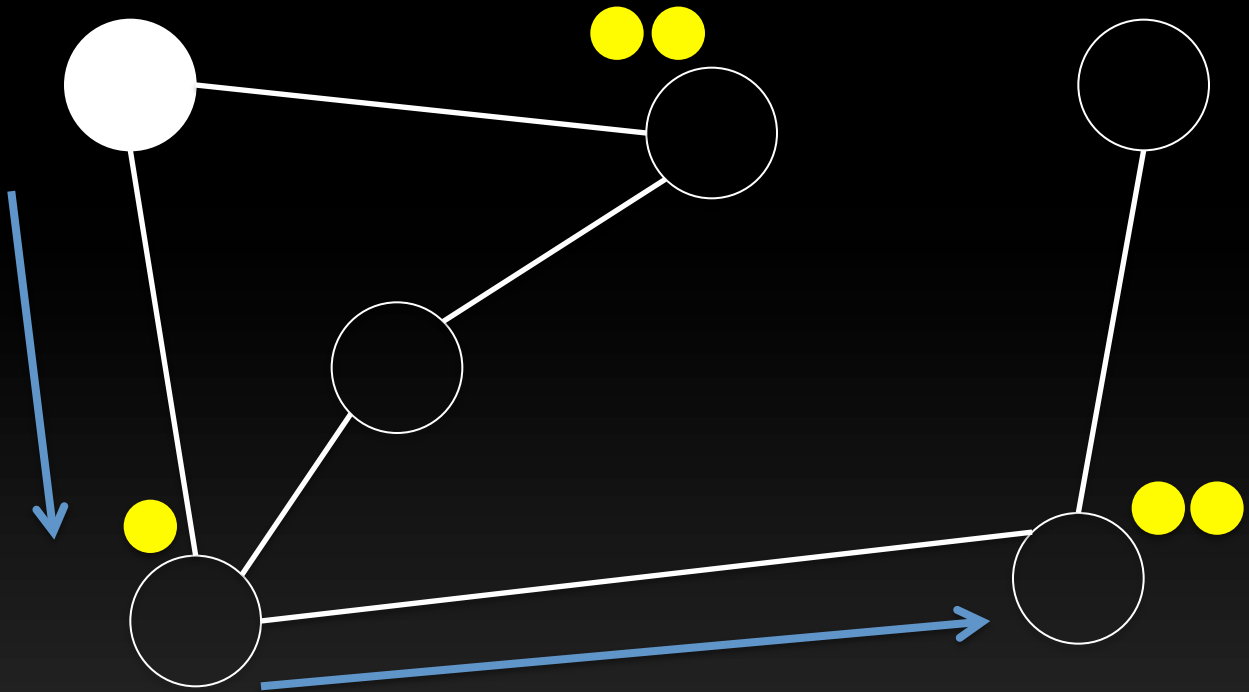
Example



Example with routing



Example with routing



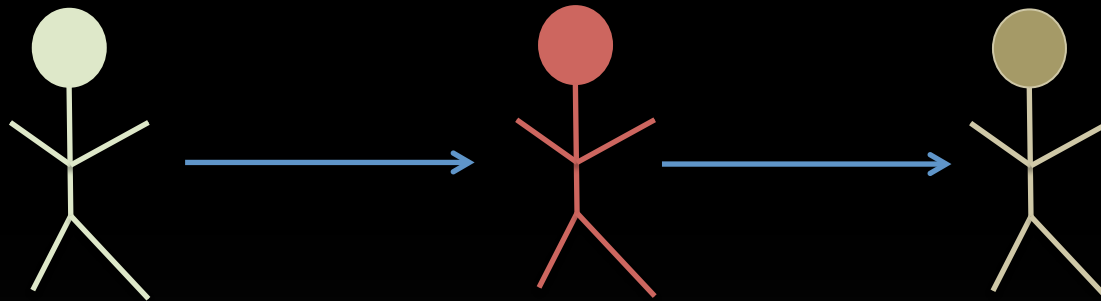
Goal

- Want each player to:
 - Properly update probabilities
 - Route task to someone with many coin flips
- Idea:
 - Use **proper scoring rules** to incentivize update
 - ‘Properly score’ routing decisions?

Common knowledge

Assumption: everyone knows how many coin flips everyone else has, and this is common knowledge.

k-step routing scoring rule



$$\alpha S_i - S_{i-1} + (1-\alpha)S_k$$

Score for update

Score for
routing decision

Theorem

Under the R -step routing scoring rule, it is a Perfect Bayesian Equilibrium for people to update probabilities truthfully and to route to the next person in the optimal path of length R .

Implications

- Can incentivize local routing decisions that follow the optimal coin-collecting path.
- But...
 - Computing the optimal path is in general NP-hard
 - Common knowledge assumption is unreasonable for large networks

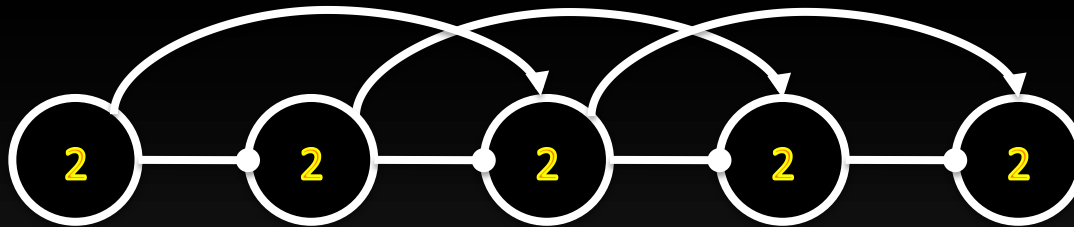
Without common knowledge

- People may have different information about others' expertise on the network
- To route optimally, have to reason about what other's know and don't know
 - “optimal routing” is **highly impractical**

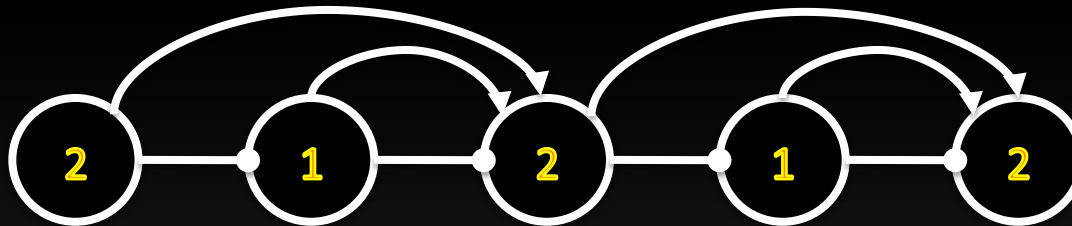
Local common knowledge

Assumption: everyone that is within some distance of a person has the same information about that person.

2-2-2-2



2-1-2-1



Breaks the chain of reasoning

m-hop local routing rules

1-1-1-1 (myopic)

2-1-2-1

3-2-1-3-2-1

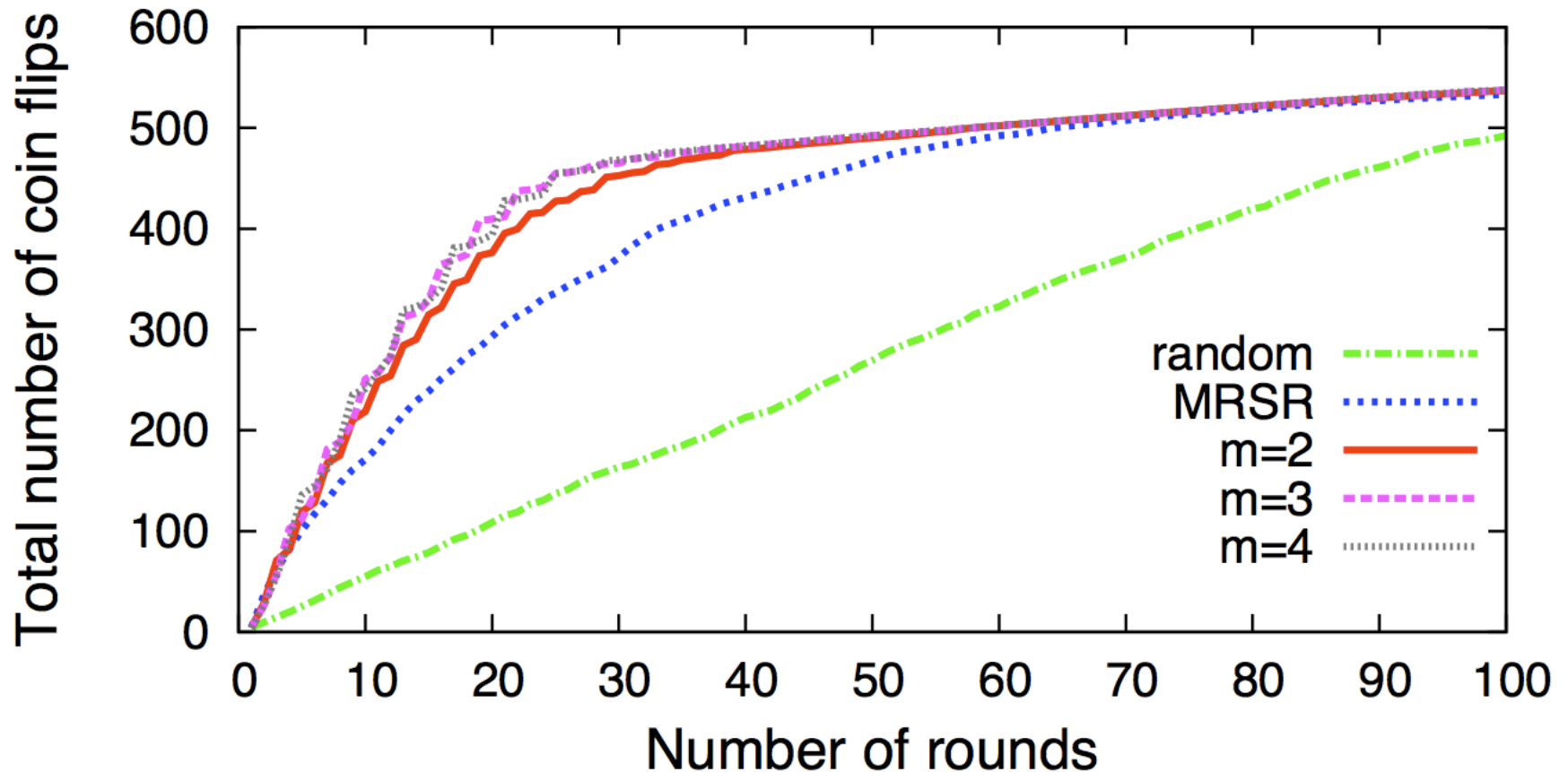
4-3-2-1-4-3-2-1

....

Theorem

Assume m -hop local knowledge holds. Under a m -hop local routing rule, it is a Perfect Bayesian Equilibrium for everyone to update probabilities truthfully and to route to the next person in the optimal path of length m .

Simulation results

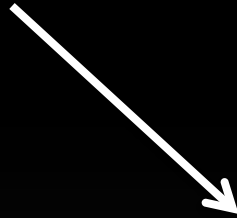


Conclusion

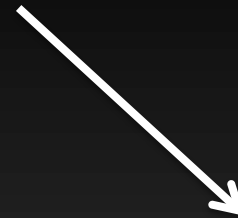
- Task routing for prediction tasks explores connection among incentives, solving, routing, and knowledge
- Intriguing class of local routing rules that induce effective routing behaviors while promoting a simple equilibrium where people only reason about what they know
- Rich space of practical and theoretical questions

Thank you

Comments



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my co-authors