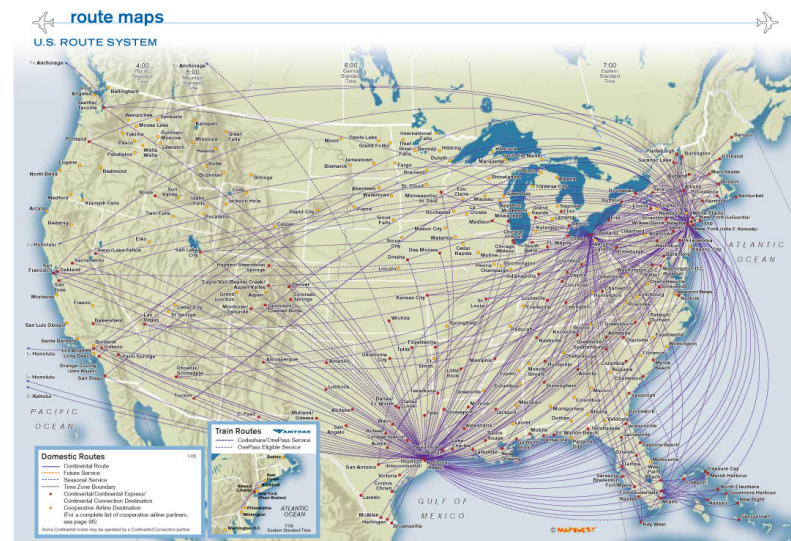


# May 5, 2016



# “Flows on spatial networks”

# Topics

- Last time:

Michael Gastner (SFI) and Mark Newman (U Mich)

I. Optimal allocation of facilities: Number of facilities within radius  $n(r)$  scales sublinear of density:  $n(r) \sim \rho(r)^{2/3}$ .

– Seems to hold true for distribution of public goods (hospitals, police stations, county seats, ...)

II. Optimal connection of facilities into a network:

– Linear tradeoffs between geometric and network metrics  
– From road networks to air transport

# Topics

- Today:

Network flows on road networks – Michael Zhang (UC Davis)  
(Details of demand, edge capacity, and feasible paths all extremely important)

- I. Optimization and network flow
- II. User vs System Optimal
- III. Braess' Paradox
- IV. Nash Equilibrium
- V. Price of anarchy

## **User optimal versus system optimal (In the traffic context)**

Act on self interests (User Equilibrium):

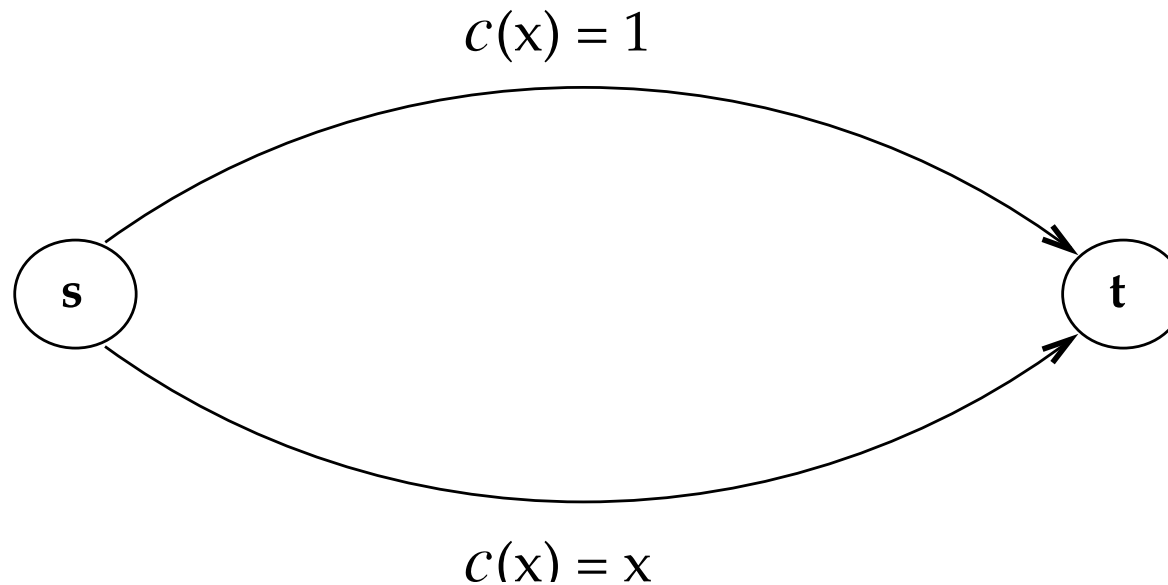
- Travelers have full knowledge of the network and its traffic conditions
- Each traveler minimizes his/her own travel cost (time)

Act on public interests (System Optimal):

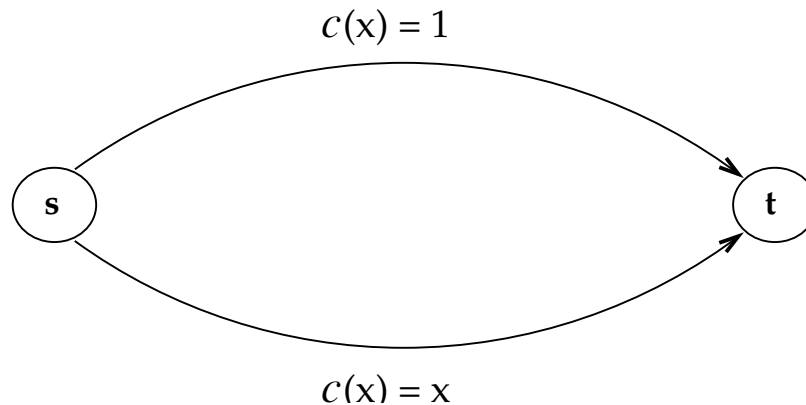
- Travelers choose routes to make the total travel time of all travelers minimal (which can be achieved through choosing the routes with minimal marginal travel cost)

$$\min \sum_{a \in A} t_a(v_a) v_a$$

## Pigou's example: User versus system optimal



- Two roads connecting source,  $s$ , and destination,  $t$
- Route 1, “infinite” capacity but circuitous; 1 hour travel time
- Route 2, direct but easily congested; travel time is 1 hour times the fraction of traffic on the route,  $x_2$ .
  - Route 1,  $c_1 = 1$  hour
  - Route 2,  $c_2 = x_2 \cdot 1$  hour.



- Everyone takes the bottom road!
  - It is never worse than the top road, and sometimes better
  - In general, an equilibrium exists when the travel times on all routes are equal. (See HW and later in lecture.)
- Average travel time
  - $\tau = x_1 \cdot c_1 + x_2 \cdot c_2$ .
  - Average travel time = 1 hour = 60 mins
- If could incentivize half the people to take the upper road, then lower road costs one-half hour.
  - Average travel time:  $0.5 \cdot 1 + 0.5 \cdot 0.5 = 0.75$  hour = 45 mins!

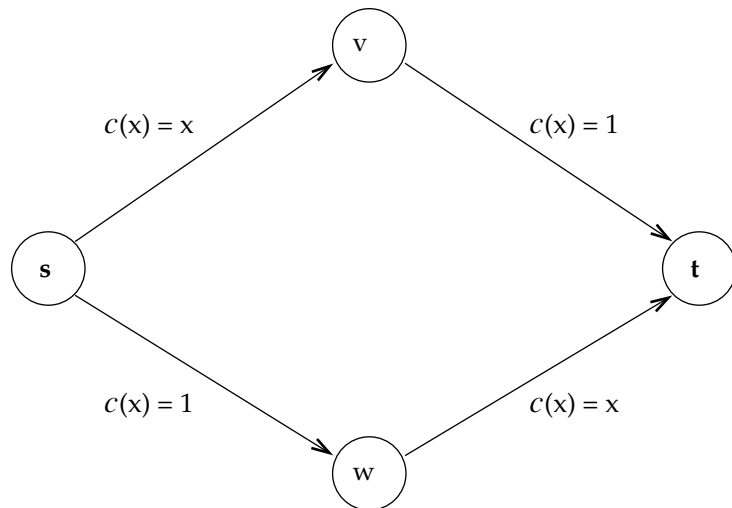
See Michael Zhang's slides ([zhang.pdf](#))

# Braess Paradox

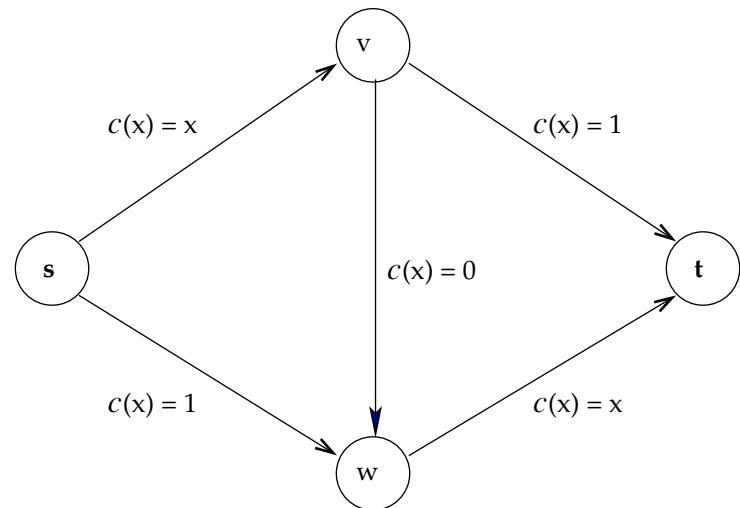
- Dietrich Braess, 1968

(Braess currently Prof of Math at Ruhr University Bochum, Germany)

- In a user-optimized network, when a new link is added, the change in equilibrium flows might result in a higher cost, implying that users were better off without that link.



(a) Initial network



(b) Augmented network



## Recall Zhang notation

# Flows in a Highway Network

$N$ : set of nodes

$A$ : set of links

$I$ : set of origins

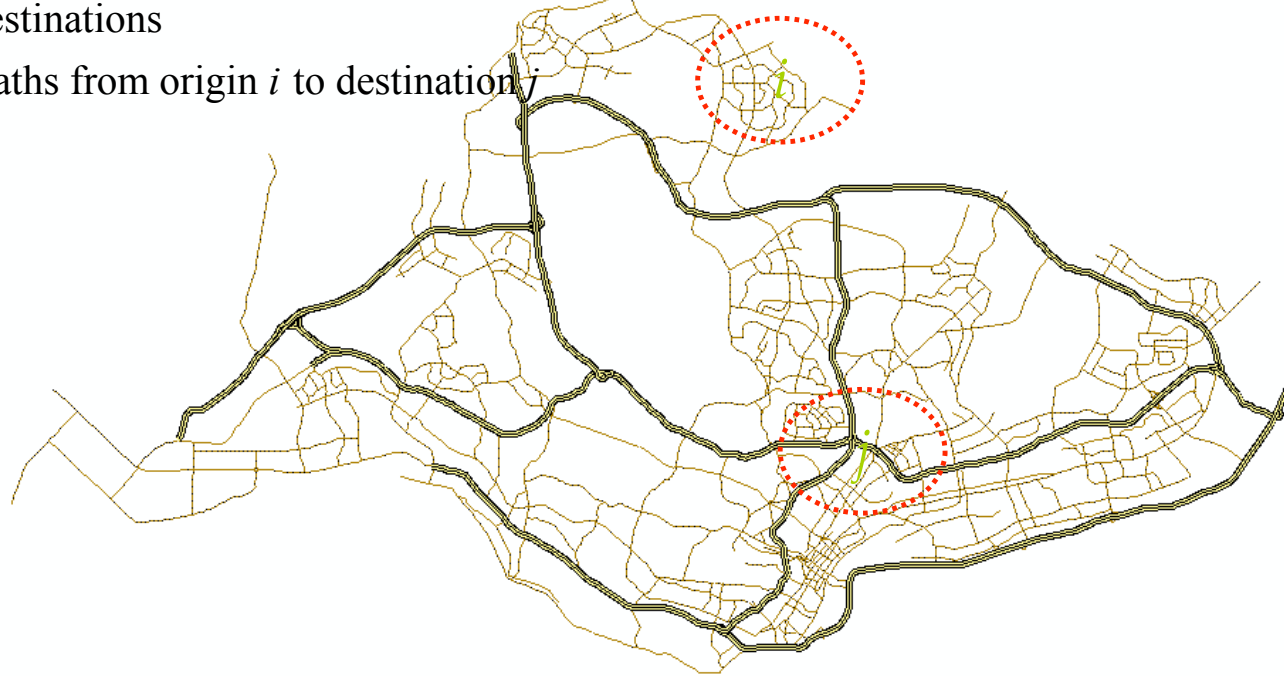
$J$ : set of destinations

$R_{ij}$ : set of paths from origin  $i$  to destination  $j$

$t_a(v_a, C_a)$ : link travel cost function

$q_{ij}$ : Traffic demand from origin  $i$  to destination  $j$

$C_a$ : Capacity on link  $a$



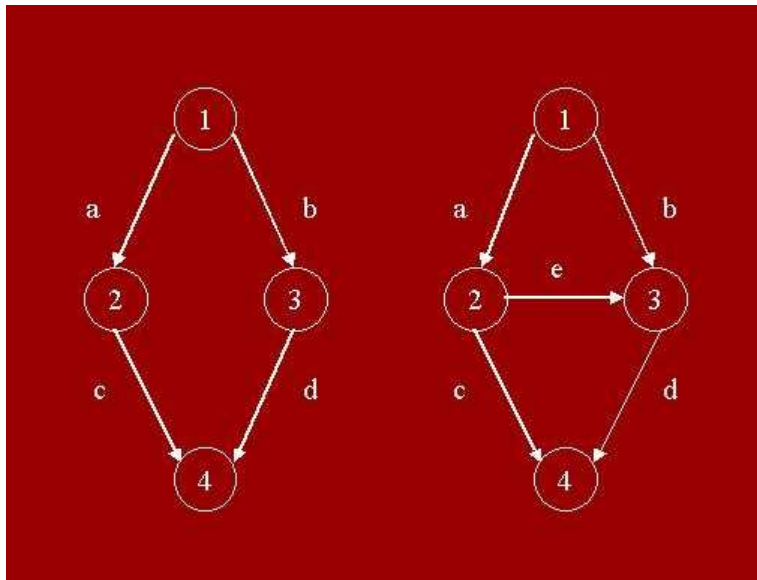
- Recall Zhang notation
  - $q_{ij}$  is overall traffic demand from node  $i$  to  $j$ .
  - $t_a(\nu_a)$  is travel cost along link  $a$ ,
  - which is a function of total flow that link  $\nu_a$ .
- Equilibrium is when the cost on all feasible paths is equal

# Getting from 1 to 4

Assume traffic demand  $q_{14} = 6$ . Originally 2 paths (a-c) and (b-d).

$$\begin{array}{ll} \bullet t_a(\nu_a) = 10\nu_a & \bullet t_c(\nu_c) = \nu_c + 50 \\ \bullet t_b(\nu_b) = \nu_b + 50 & \bullet t_d(\nu_d) = 10\nu_d \end{array} \implies \text{Eqm: } \nu = 3 \text{ on each link}$$

$$C_1 = C_2 = 83$$



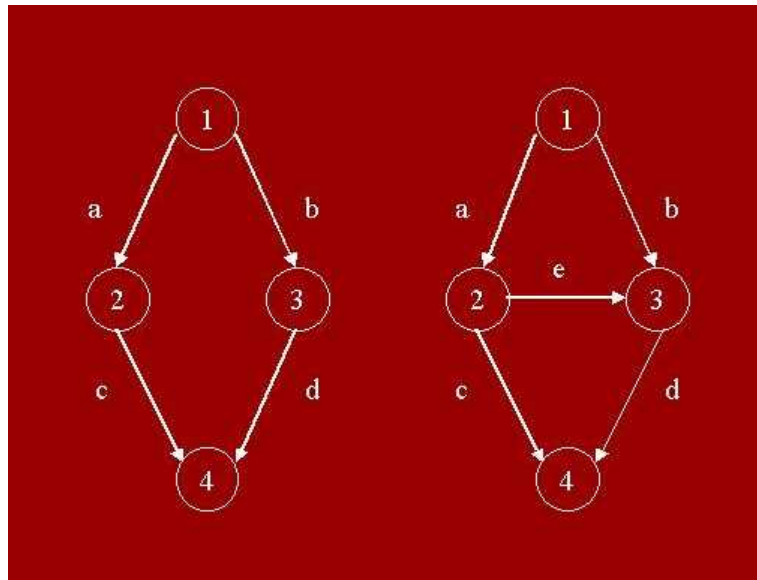
Add new link with  $t_e(\nu_e) = \nu_e + 10$

Now three paths:

Path 3 (a - e - d), with  $\nu_e = 0$  initially, so  $C_3 = 0 + 10 + 0 = 10$

$C_3 < C_2$  and  $C_1$  so new equilibrium needed.

- By inspection, shift one unit of flow from path 1 and from 2 respectively to path 3.
- Now all paths have flow  $f_1 = f_2 = f_3 = 2$ .
- Link flow  $\nu_a = 4, \nu_b = 2, \nu_c = 2, \nu_d = 4, \nu_e = 2$ .



$$t_a = 40, t_b = 52, t_c = 52, t_d = 40, t_e = 12.$$

$$C_1 = t_a + t_c = \mathbf{92}; C_2 = t_b + t_d = \mathbf{92}; C_3 = t_a + t_e + t_d = \mathbf{92}.$$

- $92 > 83$  so just increased the travel cost!

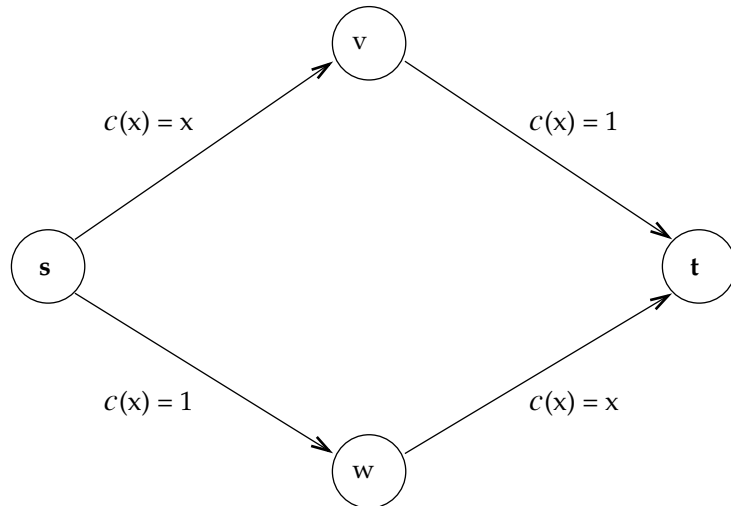
## Braess paradox – Real-world examples

- 42nd street closed in New York City. Instead of the predicted traffic gridlock, traffic flow actually improved.
- A new road was constructed in Stuttgart, Germany, traffic flow worsened and only improved after the road was torn up.

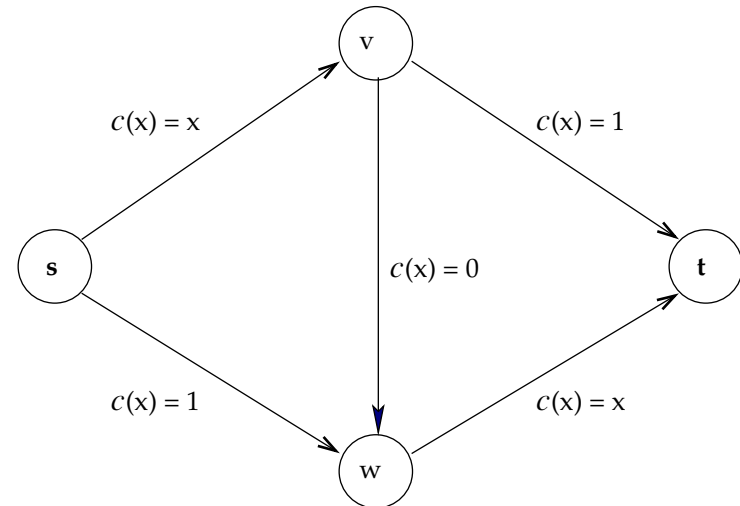
## Braess paradox depends on parameter choices

- “Classic” 4-node Braess construction relies on details of  $q_{14}$  and the link travel cost functions,  $t_i$ .
- The example works because for small overall demand ( $q_{14}$ ), links  $a$  and  $d$  are cheap. The new link  $e$  allows a path connecting them.
- If instead demand large, e.g.  $q_{14} = 60$ , now links  $a$  and  $d$  are costly! ( $t_a = t_d = 600$  while  $t_b = t_c = 110$ ). The new path a-e-d will always be more expensive so  $\nu_e = 0$ . No traffic will flow on that link. So Braess paradox does not arise for this choice of parameters.

## Another example of Braess



(a) Initial network



(b) Augmented network

## How to avoid Braess?

- Back to Zhang presentation .... typically solve for optimal flows numerically using computers. Can test for a range of choices of traffic demand and link costs.

## More flows and equilibrium

- David Aldous, “Spatial Transportation Networks with Transfer Costs: Asymptotic Optimality of Hub and Spoke Models”
- Marc Barthélemy, “Spatial networks” *Physics Reports* 499 (1), 2011.
- Flows of material goods, self-organization: Helbing et al.
- Jamming and flow (phase transitions):  
Nishinari, Liu, Chayes, Zechina.
- Algorithmic game theory: Multiplayer games for users connected in a network / interacting via a network.
  - Designing algorithms with desirable Nash equilibrium.
  - Computing equilibrium when agents connected in a network.



# User-centric behavior

- Utility functions
- Game theory
  - Normal form games & Nash equilibrium:
    - Prisoner's dilemma
    - Stag hunt

## Prisoner's Dilemma

	Cooperate	Defect
Cooperate	3, 3	0, 5
Defect	5, 0	1, 1

Blue has two strategies:

- Cooperates/Red Cooperates — Blue gets payout “3”
- Cooperates/Red Defects – Blue gets “0”
- Defects/Red Defects – Blue gets “1”
- Defects/Red Cooperates – Blue gets “5”

Ave payout: Cooperate = 1.5, Defect = 3

## Nash equilibrium

No player has anything to gain by changing only his or her own strategy.

- **Blue always chooses to Defect!** Likewise Red always chooses Defect.
- Both defect and get “1” (Nash), even though each would get a higher payout of “3” if they cooperated (Pareto efficient).

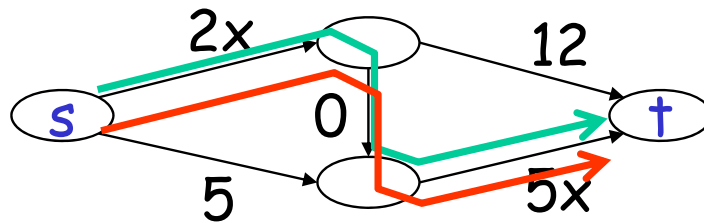
## **“The price of anarchy”**

E. Koutsoupas, C. H. Papadimitriou  
“Worst-case equilibria,” STACS 99.

Cost of worst case Nash equilibrium / cost of system optimal  
solution.

# The Price of Anarchy

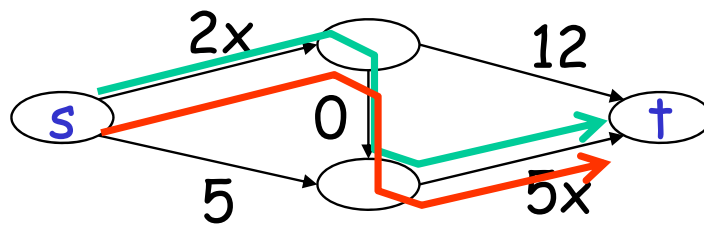
Nash Equilibrium:



$$\text{cost} = 14 + 14 = 28$$

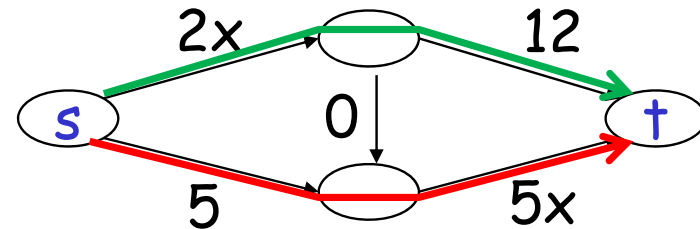
# The Price of Anarchy

Nash Equilibrium:



$$\text{cost} = 14 + 14 = 28$$

To Minimize Cost:



$$\text{cost} = 14 + 10 = 24$$

$$\text{Price of anarchy} = 28/24 = 7/6.$$

- if multiple equilibria exist, look at the *worst* one

## Selfish routing and the POA on the Internet

T. Roughgarden and E. Tardos, How Bad is Selfish Routing?,  
FOCS '00/JACM '02

- Routing in the Internet is *decentralized*: Each router makes a decision, so path dynamically decided as packet passed on.
- Cost of an edge  $c(e)$ , may be constant (infinite capacity) or depend on the load.
- “*Shortest path*” routing (really lowest  $\sum c(e)$  routing) typically implemented.
- This is equivalent to “selfish routing” (each router chooses best option available to it).
- **Resulting POA = 2!**

## **Braess and the POA for Internet traffic**

Greg Valiant, Tim Roughgarden, Eva Tardos

“Braess’s paradox in large random graphs”, Proceedings of the 7th ACM conference on Electronic commerce, 2006.

- Removing edges from a network with “selfish routing” can decrease the latency incurred by traffic in an equilibrium flow.
- With high probability, (as the number of vertices goes to infinity), there is a traffic rate and a set of edges whose removal improves the latency of traffic in an equilibrium flow by a constant factor.
- Braess paradox found in random networks often (not just “classic” 4-node construction).



## Algorithmic game theory

- Since we know users act according to Nash, can we design algorithms (mechanisms) that bring Nash and System Optimal as close together as possible?
- Typically we think of players who interact via a network, or who's connectivity is described by a network of interactions.
  - Multiplayer games for users connected in a network or interacting via a network.
  - Designing algorithms with desirable Nash equilibrium.
  - Computing equilibrium when agents connected in a network.

## mechanism design (or *inverse* game theory)

- agents have utilities – but these utilities are known *only to them*
- game designer prefers certain outcomes *depending on players' utilities*
- designed game (mechanism) has designer's goals as dominating strategies

# Some traditional games:

e.g.

matching pennies

1,-1	-1,1
-1,1	1,-1

prisoner's dilemma

3,3	0,4
4,0	1,1

chicken

0,0	0,1
1,0	-1,-1

auction

	1	...	$n$
1	$0, v - y$ $u - x, 0$		
·			
·			
$n$			

## Mechanism design example:

e.g., Vickrey auction

- sealed-highest-bid auction encourages gaming and speculation
- Vickrey auction: Highest bidder wins, pays second-highest bid

**Theorem:** Vickrey auction is a truthful mechanism.

**Theorem:** It maximizes social benefit *and* auctioneer expected revenue.

## **(Modified) Vickrey auctions in real life – Google AdWords, and Yahoo's ad sales**

- Bidding on a “keyword” so that your advertisement is displayed when a search user enters in this keyword
- You can safely bid the maximum price you think is fair, and if you win, you actually pay less!
- Mechanism design
  - Incentivizes users to bid what they think is fair (reveal their true utilities)
  - Keeps more people in the bidding
  - Does not necessarily maximize profits for seller

## Summary of spatial flows and games

- Optimal location of facilities to maximize access for all.
- Designing “optimal” spatial networks (collection/distribution networks – subways, power lines, road networks, airline networks).
- Details of flows on actual networks make all the difference!
  - Users act according to Nash
  - Braess paradox (removing edges may improve a network’s performance!)
  - The “Price of Anarchy” ( $\text{cost of worst Nash eqm} / \text{cost of system optimal}$ )