

# Social and Information Networks

University of Toronto CSC303  
Winter/Spring 2023

Week 11: March 27-31

# This week's agenda

- Congestion Networks
  - ▶ Traffic networks (Ch 8.1)
  - ▶ Nash Equilibrium
  - ▶ Braess' Paradox (Ch 8.2)
  - ▶ Social cost
    - ★ Tragedy of the commons
    - ★ Price of anarchy
- Kidney Exchange
- Recap

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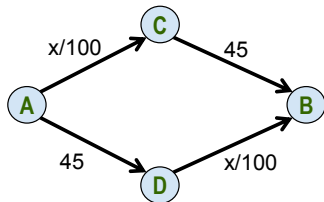
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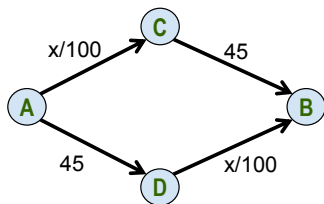
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  - ▶ ... however, our model uses linear congestion, therefore roads make more sense

## A simple but interesting example



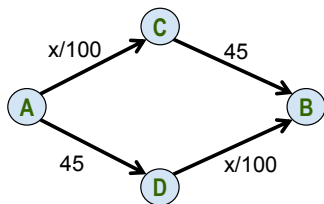
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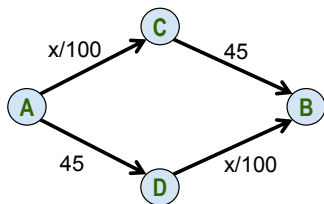
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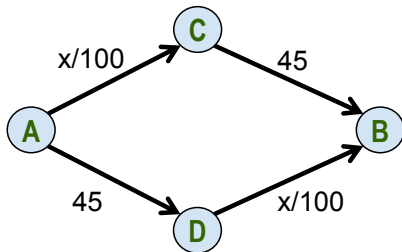
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- **Question** What route should they decide to take?

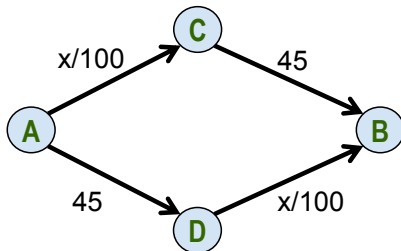
## The traffic network example continued



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- Formally, there are  $2^{4000}$  possible outcomes
- However, we view commuters as equivalent and thus there are 4001 outcomes
- All outcomes with  $x$  people using the path via  $C$  (and  $4000 - x$  using the path via  $D$ ) are all equivalent and we will just view them as one outcome

# What is a Nash Equilibrium for this traffic network game?

- We are interested in a Nash Equilibrium (NE)
  - ▶ an “outcome  $x$ ” (i.e., with  $x$  using the path via  $C$ ) such that no individual will want to change routes in order to save time (under the assumption that no one else changes their route)



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Proof of Claim: In the outcome with  $x = 2000$  commuters using the path via  $C$  (and hence also 2000 commuters using the path via  $D$ ), if any individual changes their route, then their commute time increases from  $t = 45 + 2000/100 = 65$  to  $t' = 45 + 2001/100 > 65$ .

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While this would unlikely be noticed by a single individual, what happens when more and more decide to switch?

## Wed. Mar 29: Announcements and Corrections

- Course evaluations are now open on Quercus!
  - ▶ Good or bad, please do fill them out if you can; it's both formative feedback for myself and the university
- Individual peer feedback for the critical reviews is due this Friday (March 31st)
- You will then have just under a week to review the feedback, before submitting the final draft that will be graded (April 6th)
- Last tutorial is this Friday
- Last class is Wednesday April 5th; we'll be doing a fun review exercise! :)
- The final is open book, take home
  - ▶ starts Tuesday April 11th at 9:00AM Toronto time, and ends April 13th at 9:00AM Toronto time

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  - ▶ 2001 commuters will increase their commute time by .01 minutes while only 1999 will save .01 minutes
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  - ▶ A similar observation applies for the outcome when 1999 go via  $C$  and 2001 go via  $D$
- While any individual commuter is unlikely to notice this, larger deviations are apparent
  - ▶ Suppose now that 3000 go via  $C$ , then the total commute time will now increase by 20,000 minutes  $\approx$  2 weeks worth of time
  - ▶ If everyone takes the same route, then the total commute time will increase by 80,000 minutes  $\approx$  2 months of time



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Essentially we would expect random uncoordinated decisions will gradually lead individuals to work towards solutions that come close to an equilibrium. The study of the Braess paradox comes, of course, before the use of GPS systems. With GPS people can change routes dynamically based on real-time information.

## Braess' Paradox

Suppose the premier decides to build a new superhighway (or super fast rail line) and add this to the existing traffic network.

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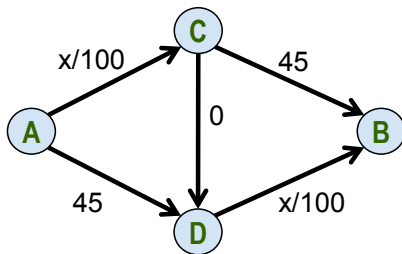
Suppose the premier decides to build a new superhighway (or super fast rail line) and add this to the existing traffic network.

Lets even imagine that the time to traverse this new additional link is negligible (and hence approximated by 0 time). It seems that this can only improve the life of commuters. So lets add a directed link from  $C$  to  $D$  in our example traffic network.

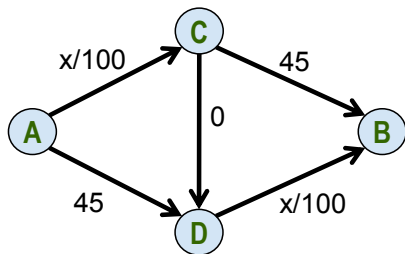
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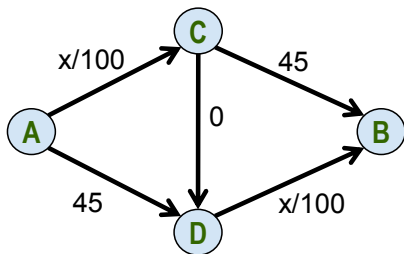
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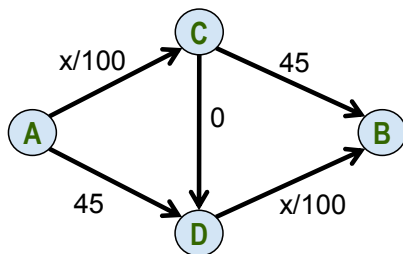


What will happen? Demo time!

**Claim:** There is a new unique NE. Everyone now will want to take the route  $A \rightarrow C \rightarrow D \rightarrow B$ . And the individual commute time of this NE is 80 minutes! That is, by building the new superhighway (rail link) everyone has an additional 15 minutes of commuting.

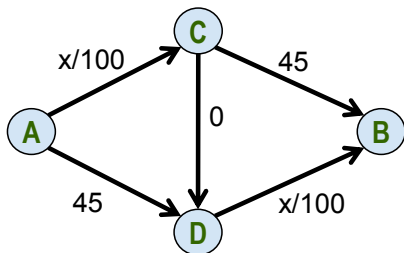
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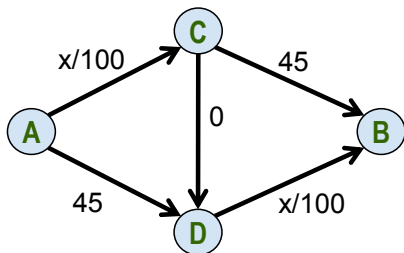
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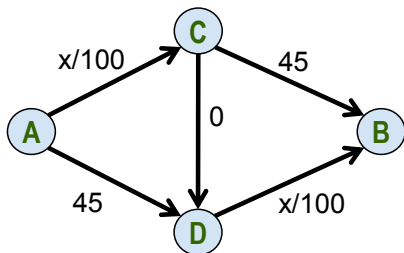
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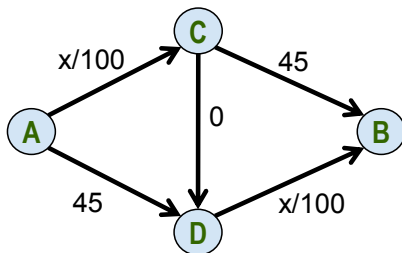
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  - ▶ From  $C$ , it is better to take the indirect path (via  $D$ ) to  $B$  than taking the direct  $(C, B)$  link

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  - ▶ This has been observed in some cases

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**Claim:** The following is a socially optimal solution:

- 1750 take  $A \rightarrow C \rightarrow B$  route
- 500 take  $A \rightarrow C \rightarrow D \rightarrow B$  route
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- On the other hand, the  $1750 + 1750 = 3500$  commuters taking the more direct  $A \rightarrow C \rightarrow B$  or the  $A \rightarrow D \rightarrow B$  routes will each have travel time 67.50 minutes incurring an additional 2.5 minutes of commute time.

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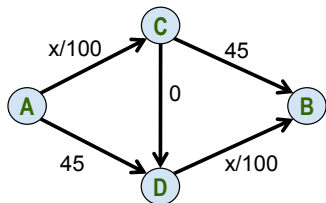
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So the *total time* saved is  $(500 \times 20 - 3500 \times 2.5) = 1250$  minutes each way, each day. **On average** (over the 4000 commuters), it is a saving of  $1250/4000 = .3125$  minutes per commuter. If this doesn't sound sufficiently impressive, suppose time was being measure in hours; that is, we can scale the edge costs by any fixed factor.

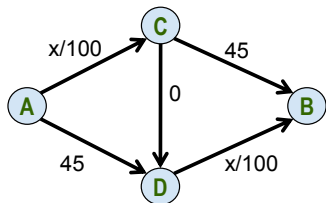
And beyond time lost, a social optimum reduces pollution.

## Understanding the partition into 3 routes



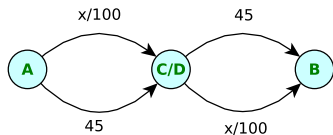
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We can lower bound the optimal solution by solving a simpler case. By introducing the edge  $(D, C)$  all previous traffic patterns are valid.



## Understanding the partition into 3 routes

We can prove it is optimal by solving a quadratic equation to determine the  $x$  commuters who will directly go to  $C$  (i.e., take the  $x/100$  route) and the  $4000 - x$  that will directly go to  $D$  (i.e., take the 45 route).

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With this, we're practically done!

- Taking the derivative and setting it to 0, we get:

$$T'(x) = .02x - 45 = 0$$

- ▶ Solving, we get  $x = 2250$ 
  - ★ 2250 will take the  $A - C$  route
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- We can now pretend that all vehicles start at node  $C$  (since we merged nodes  $C$  and  $D$ ), and repeat the above process
  - ▶ 2250 will take the  $D - B$  route
  - ▶ 1750 will take the  $C - B$  route



## Understanding the partition into 3 routes

This is an optimal solution to a relaxation where we merged  $C$  and  $D$ ; as the solution is valid in the original, optimality is preserved.

Therefore in the optimal solution we have 2250 going to  $C$  (with 1750 going on directly to  $B$  and 500 taking the  $C - D$  road) and 1750 going to  $B$  via the  $A - D - B$  route.

**How could the government obtain the socially optimum solution?**

## How could the government obtain the socially optimum solution?

- If the government selects some number (say 500) of commuters (e.g. those involved in essential services) then we can achieve the social optimum (e.g. HOV lanes). Or it can allow commuters to buy a special license for the road (e.g. HOT lanes) and hopefully let self interest lead to the social optimum
- Another implicit way to hopefully influence drivers to converge towards the socially better equilibrium is to place a toll on the new link. By adjusting the pricing on the new link, the idea would be that commuters who have the money and value their time more would start taking the new route
- They could alternatively limit the number of commuters taking the  $C - D$  road by telling commuters (by say signs at the entrance to the highway system) when the road is open or closed for the commute

# The Tragedy of the Commons and the Price of Anarchy

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- Note: In general, there can be many pure and mixed NE



## The Price of Anarchy continued

The Price of Anarchy was introduced by Papadimitriou.

For a more optimistic perspective there is also a **Price of Stability** defined as:  $\frac{\text{cost}(S)}{\text{cost}(OPT)}$  where now  $S$  is a NE solution having the least cost.

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- 1 The POA is no more than  $\frac{4}{3}$
- 2 This result is tight in the sense that if we change the fixed cost in the simple 4 node network from 45 to 40, the POA would be  $\frac{4}{3}$ .

# Recap

- Congestion Networks
  - ▶ Traffic networks (Ch 8.1)
  - ▶ Nash Equilibrium
  - ▶ Braess' Paradox (Ch 8.2)
  - ▶ Social cost
    - ★ Tragedy of the commons
    - ★ Price of anarchy

## Mon. April 3: Announcements and Corrections

- Course evaluations are still open on Quercus!
  - ▶ A big thank you to those who've filled them out already :)
- The final draft of the critical review is due April 6th on MarkUs
- Four new practice questions are on the course website
- Additional details about the final are forthcoming
- Last class is Wednesday April 5th; be ready for fun review, and *prizes!*

## **New topic: Kidney exchanges**

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Some (admittedly outdated) statistics:

- In the US, each year there are 50,000 new cases of potentially lethal kidney disease
- There are two possible treatments: dialysis or transplant
- Transplants can come from live donations or from transplants for someone who has just died (e.g., in car accident). All else being equal, live donations are much more successful
- Each year there are  $\approx 10,000$  transplants from someone deceased and  $\approx 6,500$  from live donations
- The waiting list for a transplant in the US is  $\approx 75,000$  people who usually wait between 2 and 5 years. During this waiting time,  $\approx 4000$  people die each year



## More facts concerning kidney exchanges

Live donations are possible since everyone has two kidneys and only one is needed. Moreover, when people incur kidney diseases, usually both kidneys are affected so the “additional kidney” is rarely needed.

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However, people are reluctant to donate kidneys and live donations usually come from close relatives and friends.

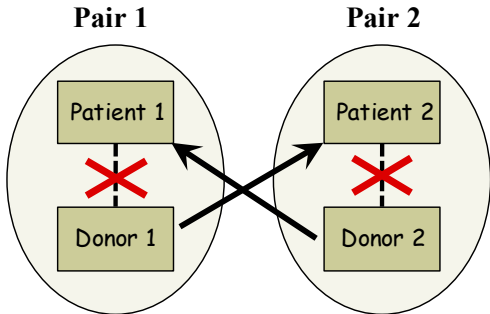
There are many biological compatibility requirements in order to do a transplant so there is often no one available and willing to do a donation.

- Blood compatibility
- Tissue compatibility

Even if possible, some donor-recipient transplants are better than others.

## Pairing up transplants

So if a willing donor for a recipient is not compatible (or if the match is not that great), there may be another recipient-donor pair that are having the same issue and are willing to do a ‘swap’. Consider the following possibility for a pair swapping:

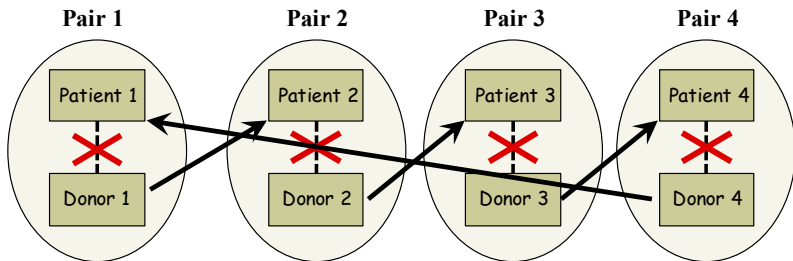


Here an edge means that the Patient (i.e. the recipient) and Donor are compatible. Edges can be weighted to reflect some objective as to how good is the match. The weight could also reflect geographic distance.

## Extending to bigger cycles

The idea of pairs swapping as just illustrated was first proposed in 1986 and only realized in 2003.

This idea has been extended to bigger cycles as in the next illustration:



## How practical are such swaps and cycles?

The are “logistical” issues that impact the practicality of such swaps and cycles, and the bigger the cycle the more problematic logistically.

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What if a potential donor, say Donor  $i$  reneges (or dies, or gets ill) once his/her paired recipient Patient  $i$  has already received their (from Donor  $i - 1$ ) kidney from the person with whom they are compatible? Now Patient  $i + 1$  has lost a valuable resource his/her (i.e., the intended Donor they brought to the exchange) if Donor  $i + 1$  has already given their kidney to Patient  $i + 2$ .

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This requires that the donation and transplant must all basically be done *simultaneously*. For cycles of length  $k$ , this requires  $2k$  simultaneous operations, where each transplantation requires both a donation and transplant operation.

Furthermore, live kidneys from donors travel best inside the donor, so need these operations to be geographically close (i.e. same or nearby hospital).  
Note: Some hospitals will not accept organs transplanted by air.

The net effect is that this severely limits the length of cycles in practice.



## Altruistic donors

What happens if we have one altruistic donor who is willing to donate a kidney without having someone with whom he/she wishes to be paired?

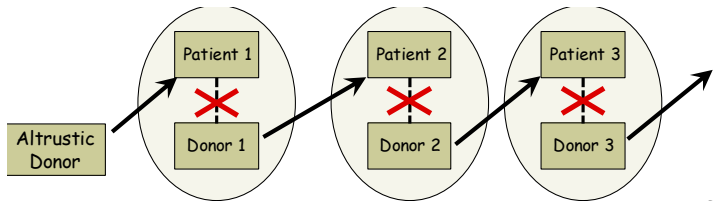
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After we have an altruistic donor, we can proceed in what potentially can be an arbitrarily long chain as below. Here each Patient must still be willing to bring a willing Donor to the exchange. But now if some donor reneges, etc, the next recipient has not lost their paired donor.



There has been at least one chain of length 30 (ending in February 2012) and some chains may be still be ongoing.

## Some final comments

- Due to biological, logistical, and incentive issues, kidney exchange requires efficient algorithmic solutions
  - ▶ Fairly large scale networks; i.e., say tens of thousands of nodes when considered nationwide.
- When restricted to pairs, this is a (possibly weighted) matching problem in a non-bipartite graph
- Introducing cycles and chains makes the problem becomes much harder
  - ▶ Need to find “practically feasible” cycles and chains
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  - ▶ When an altruistic donor arrives, do you use up that valuable resource now or wait for a better match that might lead to a longer chain?
  - ▶ Are there incentive issues for say hospitals to want to do more of the transplants by themselves than join in a broader exchange?

## Reflect :)

- Discuss with the people around you!
  - ▶ What were any recurring themes or concepts?
  - ▶ What do you think were important ideas?
  - ▶ What was your favourite material or part of the course! Why? :)



# A recap of the course

- I would say that the central theme of the course is the attempt to more precisely model sociological phenomena
  - ▶ I'd argue this includes "information networks" (e.g., the web) as it is humans that create this network
  - ▶ How we link and rank documents, and "navigate" within this network of documents fits into social networks

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  - ▶ How we link and rank documents, and "navigate" within this network of documents fits into social networks
- The main mathematical framework (and hence the course name) centres around networks
  - ▶ Modeling social networks presents significant challenges
  - ▶ Often, we have only initial insights and we are far from realistic models and analysis of social phenomena

## Recap continued

- As current social networks are often extremely large, it is necessary to be able to “think algorithmically”
- It’s also necessary to appreciate the fundamental insights and studies that have evolved and continue to evolve from sociology, economics, biology, physics, and other fields
  - ▶ While we can often abstract problems to the same framework, understanding the area, and the constraints and details lost in abstraction, is crucial for correctness!
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  - ▶ These insights may also be directed towards concrete applications!



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Needless to say however, the more we know about the content relating to the nodes and edges in a network, the more we should be able to make informative findings

# Some of the major topics in the text and the course

Here are some of the major topics in course:

- The concept of strong and weak ties and their relative role in obtaining “social capital”
- Different types of closing of triangles: triadic closure, focal closure, membership closure
- Homophily and influence
  - ▶ Are our friendships derived from similar interests and behaviour, or are our friendships influencing our interests and behaviour?
  - ▶ A central issue in social relations and one where any findings can be controversial!
    - ★ e.g., whether or not “obesity is contagious”
- A number of topics relate to different equilibrium concepts
  - ▶ Schelling’s segregation model, structural balance in friend/enemy networks, balanced outcomes in bargaining networks, stable matchings, and Nash equilibria in a congestion network, page rank, etc...

## Some major topics continued

- A number of topics relate to navigation in a social network and in particular to the small world phenomena based on geographic or social distance
  - ▶ Also was related to power law distributions in social and information networks
- Influence spread in social networks and disease spread in contact networks
- Am I missing any major themes that we discussed?



# Thank you!

You've been a wonderful class, and it's been a pleasure teaching you. I hope to see you in the last lecture, but if not, then best of luck on the final and have a great summer!