

# ***Structure and stability of equilibria in a queue-based traffic model***

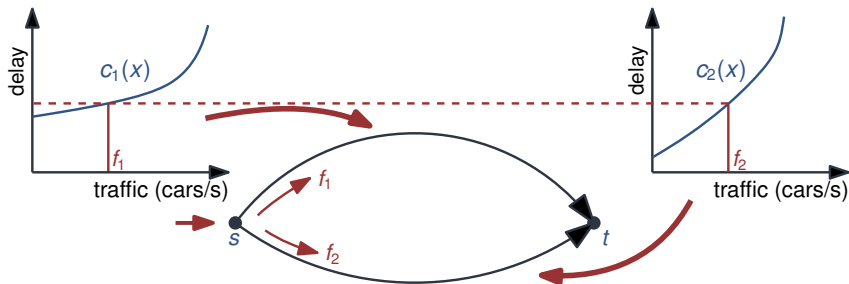
**Neil Olver**



Dutch Seminar on Optimization, September 2024

# Static models of traffic

Static models well-studied from the algorithmic game theory perspective



- Equilibria computable via a convex program *Beckmann et al. '56*
- Price of anarchy bounds *Roughgarden & Tardos, ...*
- Braess's paradox
- ...

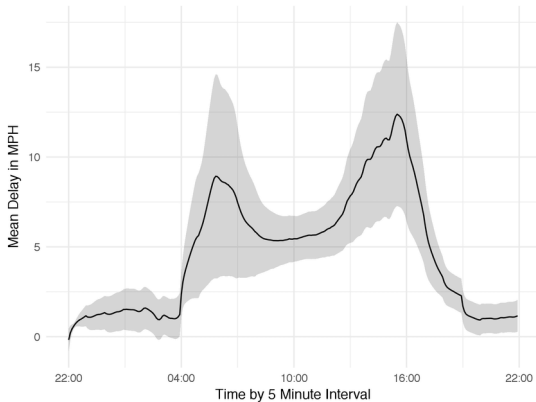
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Static models can be useful, but they do miss something...



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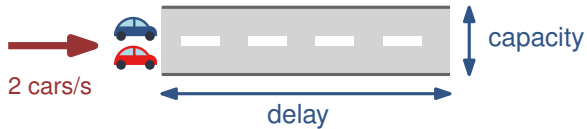
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*Credit: Brent, Beland (2020). Traffic congestion, transportation policies, and the performance of first responders, J. Environmental Economics and Management.*

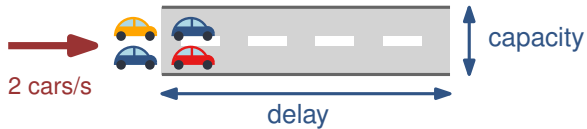
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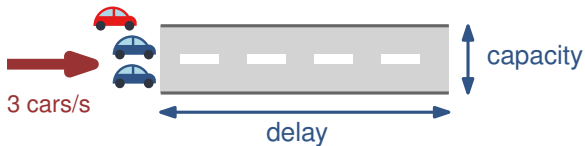
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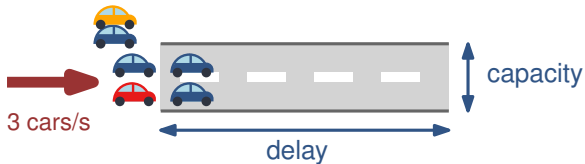
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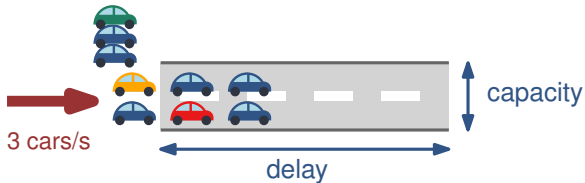
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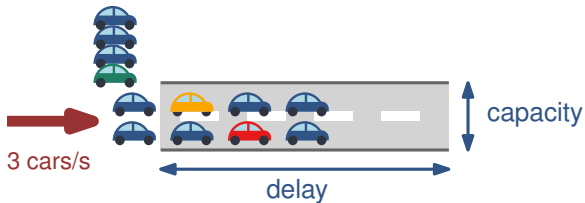
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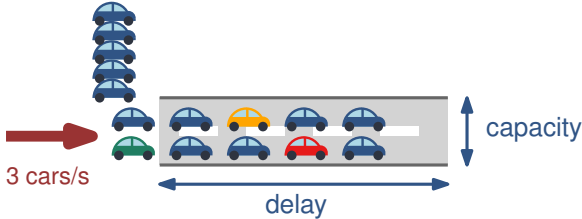
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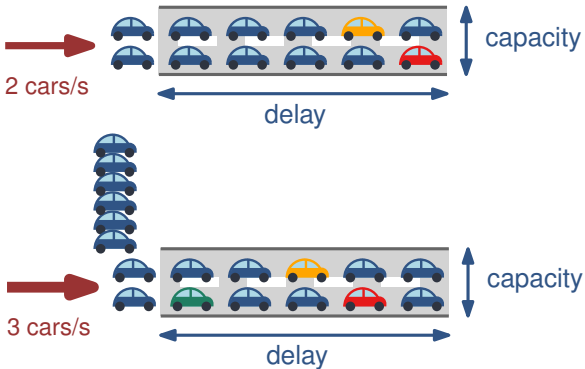
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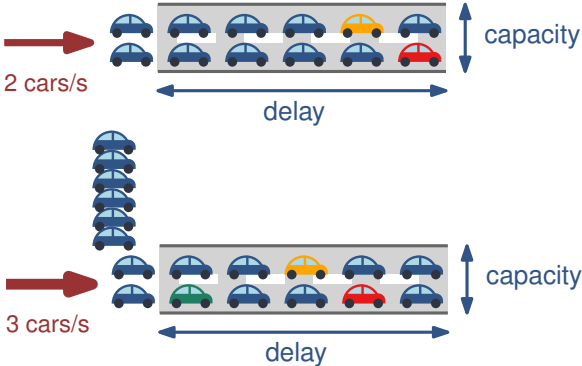
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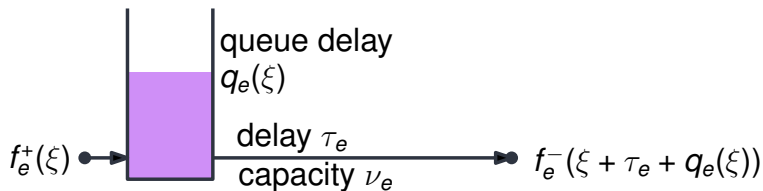
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- Continuous, nonatomic limit



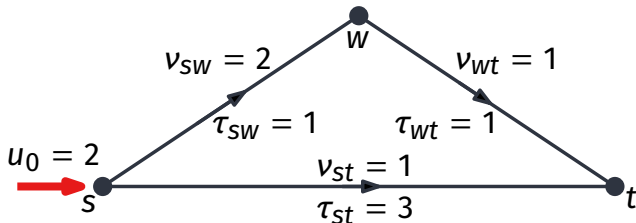
# The Vickrey bottleneck model



$$\frac{dq_e(\xi)}{d\xi} = \begin{cases} \frac{1}{\nu_e} [f_e^+(\xi) - \nu_e] & \text{if } q_e(\xi) > 0 \\ \frac{1}{\nu_e} [f_e^+(\xi) - \nu_e]^+ & \text{if } q_e(\xi) = 0 \end{cases}$$

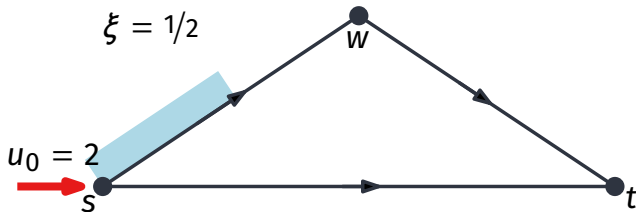
# Networks of Vickrey bottlenecks

- Each link behaves as per the Vickrey bottleneck model:  $(f_e^+, f_e^-, z_e)$ .
- Flow conservation: except for  $s, t$ , flow in = flow out at all times.
- All traffic from  $s$  to  $t$ ; constant inflow rate  $u_0$  from time 0.



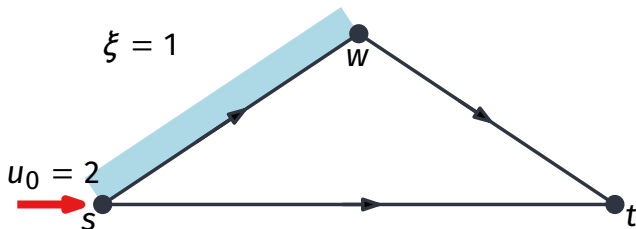
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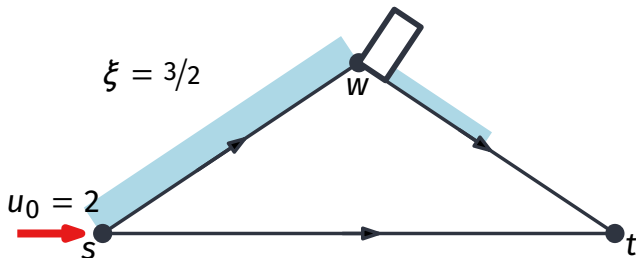
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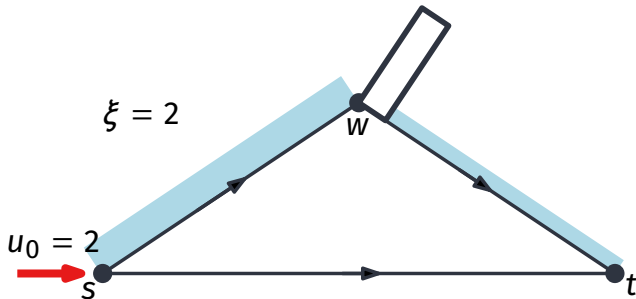
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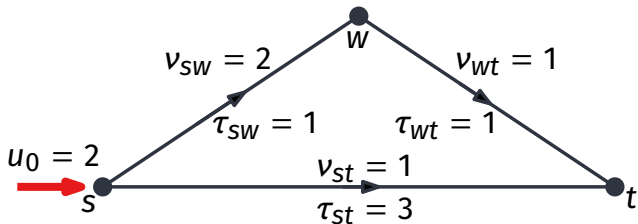


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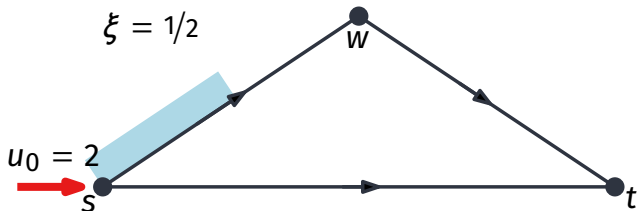
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- **Dynamic equilibria:** users choose routes to arrive as early as possible, given congestion (queues) induced by other users.
- In the sense of Nash: no deviating improvement possible in hindsight.

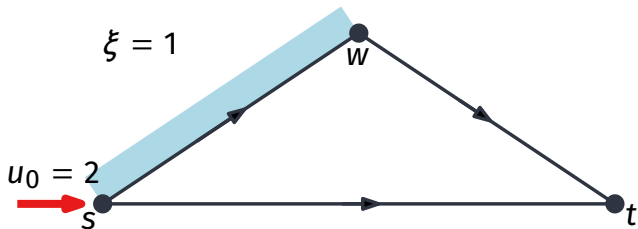


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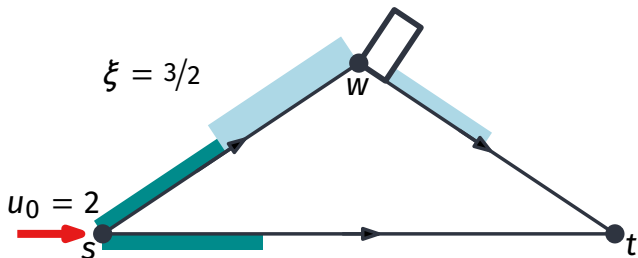
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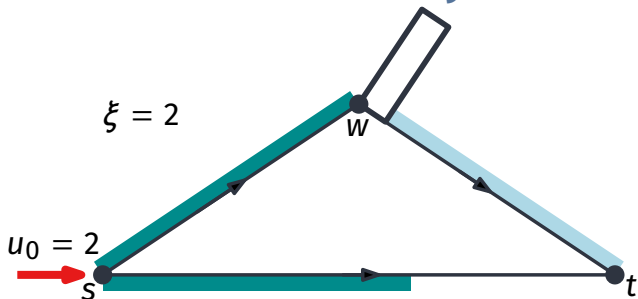
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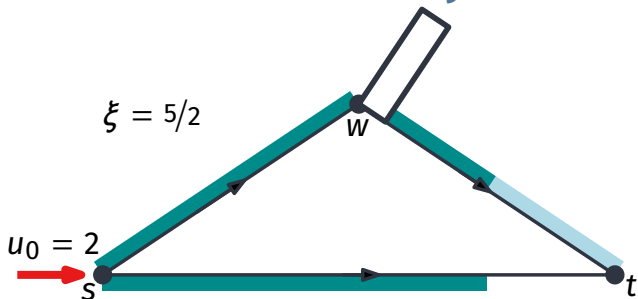
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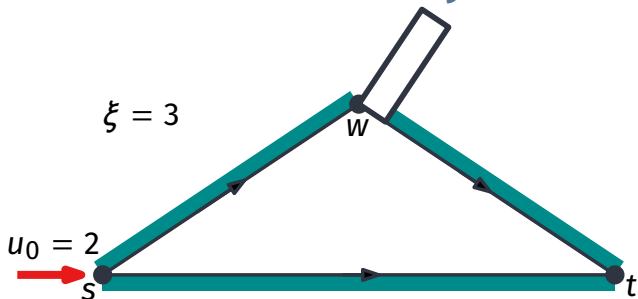
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## Relevance

- Model is a continuous approximation of a discrete reality
- Travel times on roads are likely to be noisy
- Users may take only approximately shortest routes, not precisely shortest routes
- ...

Are dynamic equilibria “stable” under perturbations?  
Do they have anything to do with equilibria in models that are “almost” the same?

## A more precise question (1)

One can define discrete (packet) versions of the deterministic queueing model.

*Hoefler et al. '11, Werth et al. '14, ...*

**Equilibrium:** no packet can arrive to the sink earlier using a different route.

Suppose we fix an instance, but divide up the flow into packets of smaller and smaller size.

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## A more precise question (2)

An  $\epsilon$ -equilibrium is a joint strategy choice of all users in which each user arrives at the sink at most  $\epsilon$  later than the earliest possible, given the delays caused by other users.

Fix an instance. Let  $\epsilon_1, \epsilon_2, \dots$  be a sequence converging to 0. Let  $\phi_i$  be an  $\epsilon_i$ -equilibrium for each  $i$ .

**The hope:**  $\phi_i$  converges to the exact dynamic equilibrium as  $i \rightarrow \infty$ .

(Informal.) In both of these situations (among others), convergence to the dynamic equilibrium is guaranteed.

- Point in favour of “meaningfulness” of the equilibrium concept
- Allows for results in the deterministic queueing model to be ported to other models
  - If the network capacity is at least as large as the inflow, queues stay bounded in dynamic equilibria
- So the same holds for packet models, for sufficiently small packet sizes
- Shows that discretization can be used to compute approximate equilibria in the nonatomic model

*Correa-Cominetti-O. '17, '22*

## Labels and equilibrium conditions

- **Agent set**  $\mathcal{A}$ . Agent  $a \in \mathcal{A}$  departs source at time  $\vartheta_a$ .
- **Strategy profile**  $\varphi : \mathcal{A} \rightarrow \mathcal{P} = \{\text{s-t paths}\}$ .  
 $\Rightarrow$  a flow  $x'(\theta)$  of value  $u_0$  describing what particles departing at time  $\theta$  do.



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From this, we can determine:

- **Departure times:** for  $v \in \varphi(a)$ ,  $d_v(a)$  is the time that agent  $a$  departs node  $v$ .
- **Earliest arrival labels:**  $\ell_v(\theta)$  is the earliest time a particle leaving  $s$  at time  $\theta$  can arrive at  $v$ , taking into account queues caused by earlier particles.

## Dynamic equilibrium conditions

A strategy profile  $\varphi$  is an equilibrium if

$$d_v(a) = \ell_v(\partial_a) \quad \text{for all } a \in \mathcal{A}, v \in \varphi(a).$$

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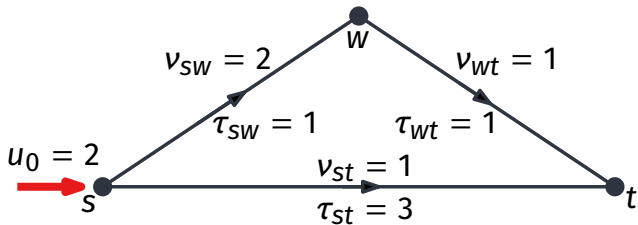
### Active arcs:

$$E'_\theta := \{e = vw \in E : \ell_w(\theta) = \ell_v(\theta) + \tau_e + \underbrace{z_e(\ell_v(\theta))}_{q_e(\theta)} / v_e\}$$

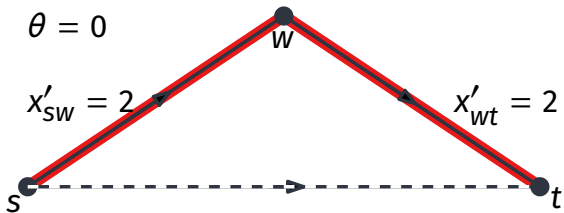
## Dynamic equilibrium conditions (alternative)

$$x'_e(\theta) > 0 \quad \Rightarrow \quad e \in E'_\theta \quad \text{for all } \theta, e.$$

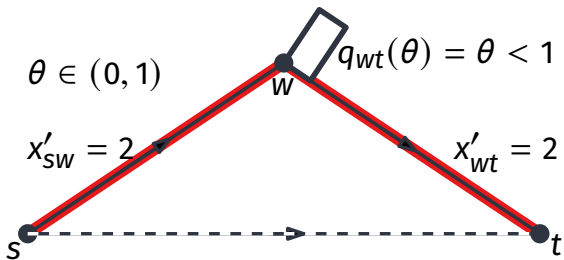
## Example redux



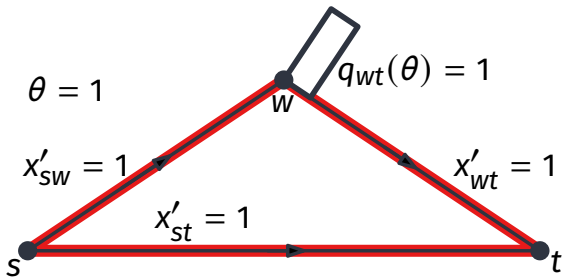
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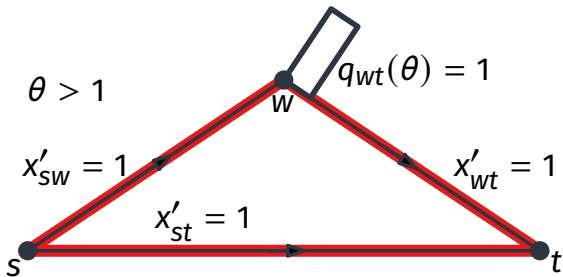


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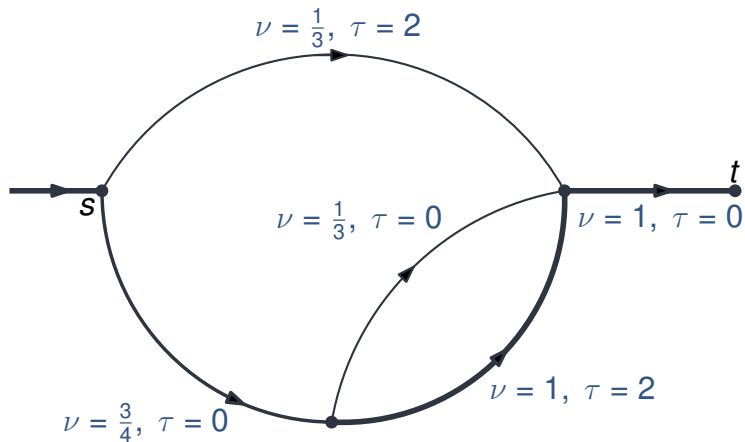




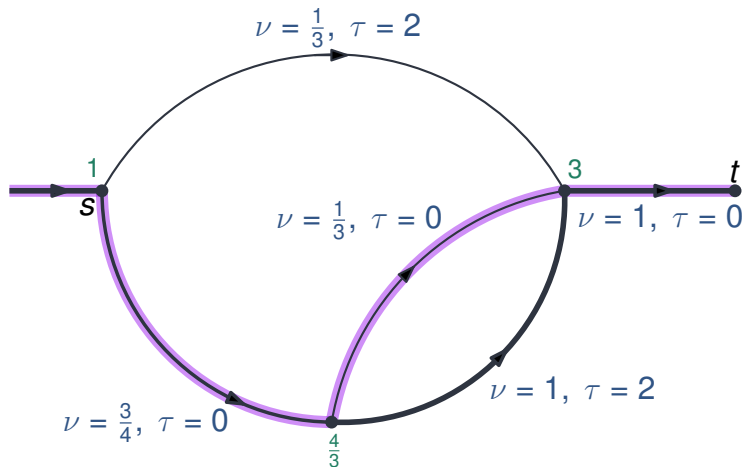
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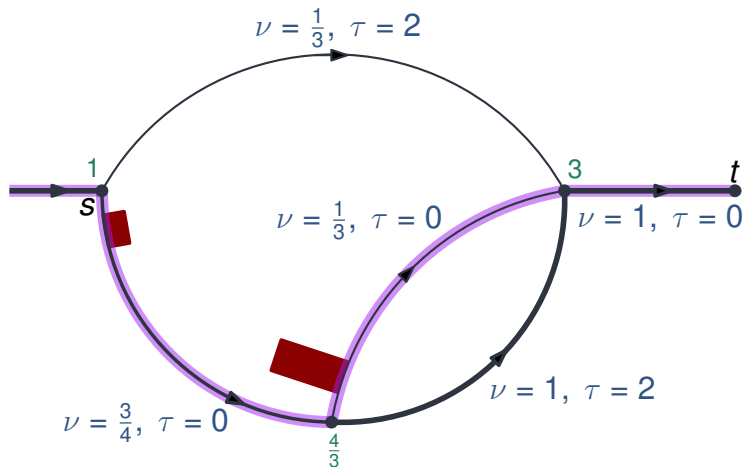
## A more surprising example



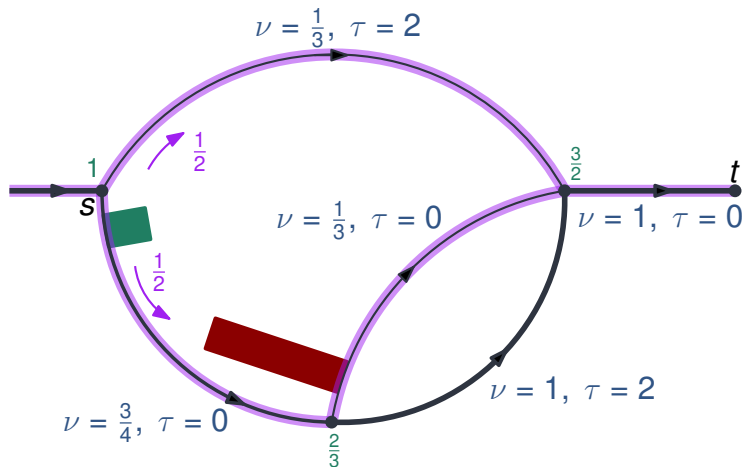
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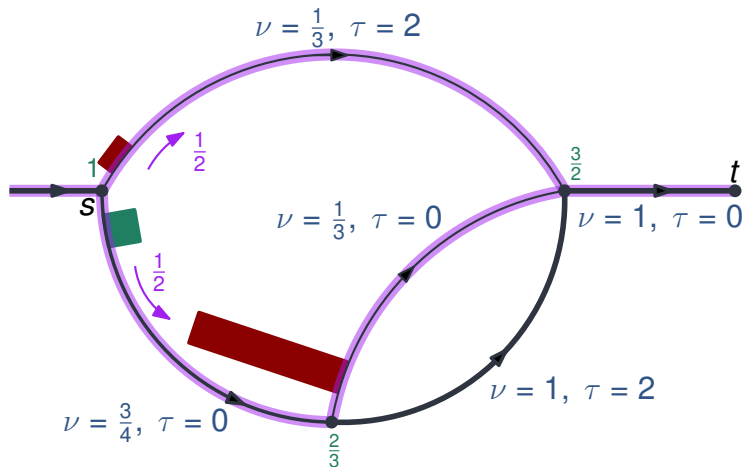
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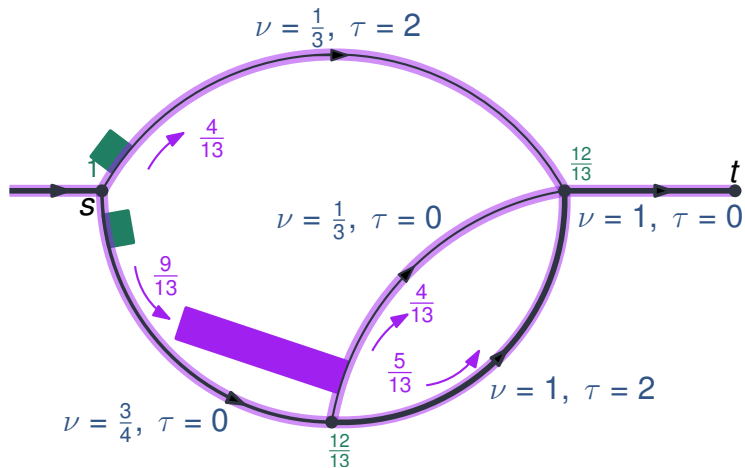
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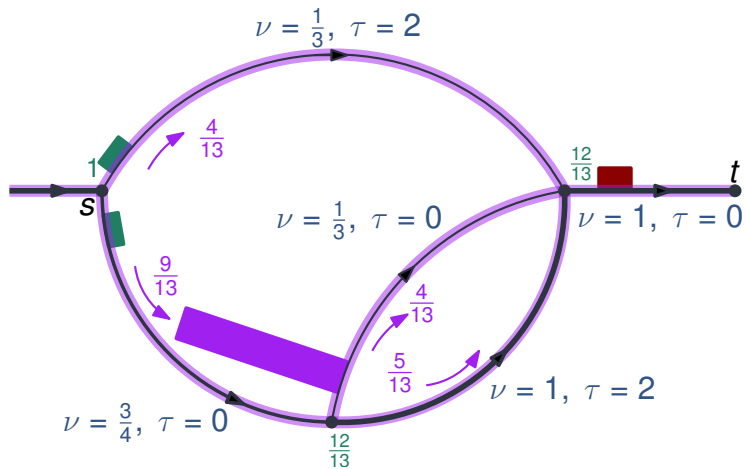
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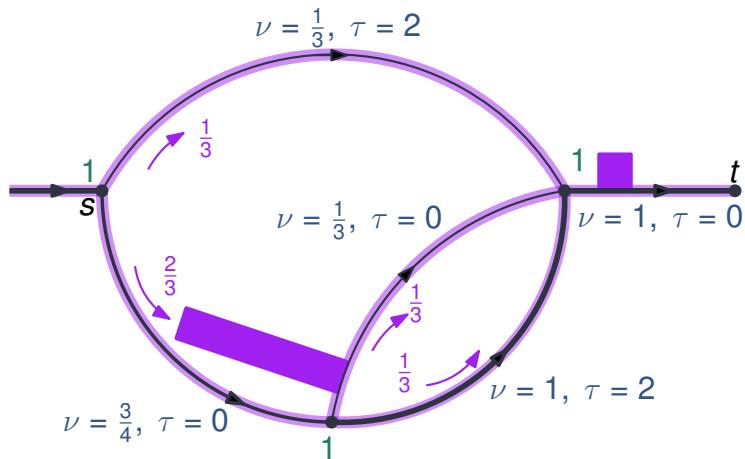


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- Labels suffice to determine the set of active arcs and the (relevant) queue lengths: for  $e = vw$ ,

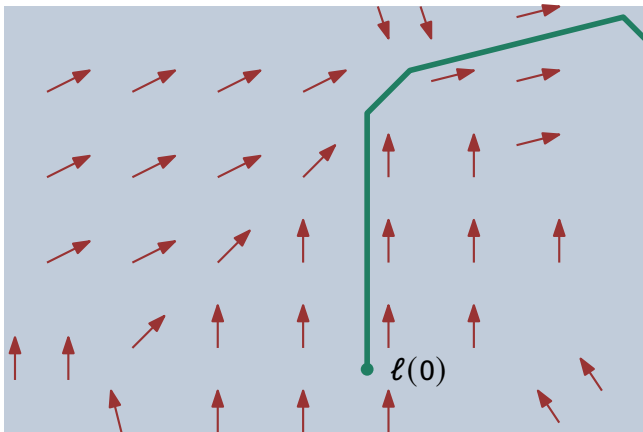
$$e \in E'_\theta \quad \text{if and only if} \quad \ell_w(\theta) \geq \ell_v(\theta) + \tau_e$$

$$q_e(\theta) = [\ell_w(\theta) - \ell_v(\theta) - \tau_e]^+$$

# Equilibrium structure

There is a vector field  $Z : \mathbb{R}^V \rightarrow \mathbb{R}^V$  s.t. for any equilibrium,

$$\ell'(\theta) = Z(\ell(\theta)) \quad \text{for almost every } \theta.$$

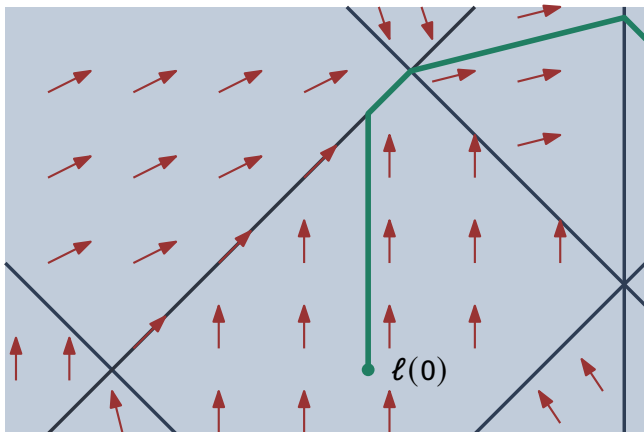


# Equilibrium structure

Further, we can write  $Z(\ell(\theta)) := Z(E'_{\ell(\theta)}, E^*_{\ell(\theta)})$ , where

$$E'_l := \{e = vw \in E : l_w - l_v \geq \tau_e\}$$

$$E^*_l := \{e = vw \in E : l_w - l_v > \tau_e\}$$



# Equilibrium structure

- $Z(l)$  is defined as a solution to a certain nonlinear system of equations (the “thin flow equations”) *Koch-Skutella '11*
- This system always has a **unique** solution (so  $Z$  is well-defined) *Cominetti-Correa-Larré '16*

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- $\ell(\theta)$  is piecewise-linear; we call each linear segment a **phase**.

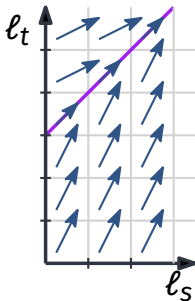
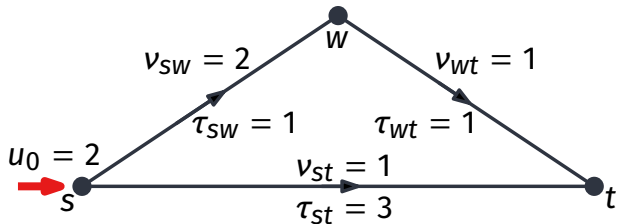
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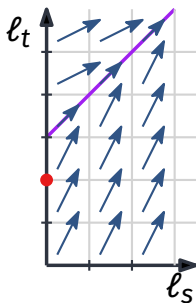
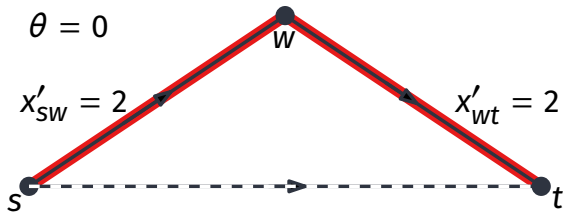
We don't know if  $Z(\cdot)$  can be efficiently computed.



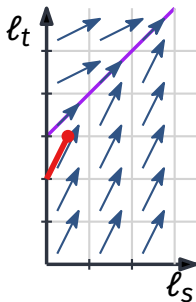
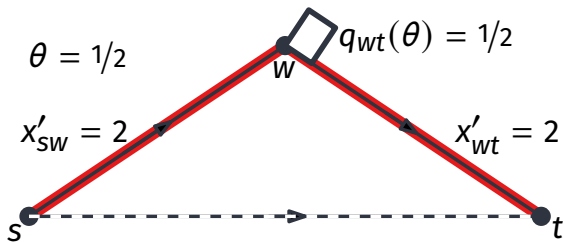
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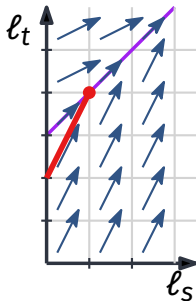
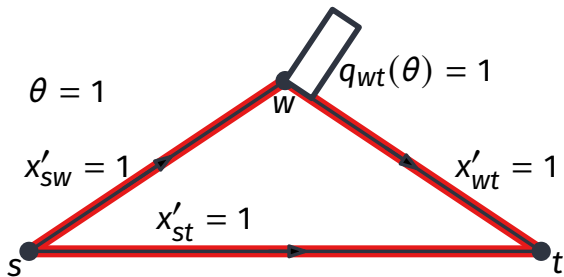
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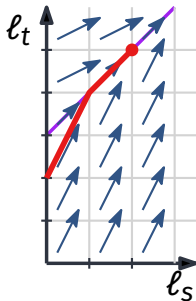
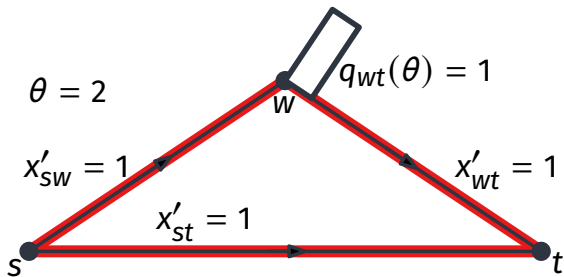
## Example again



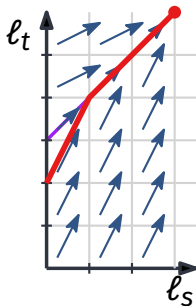
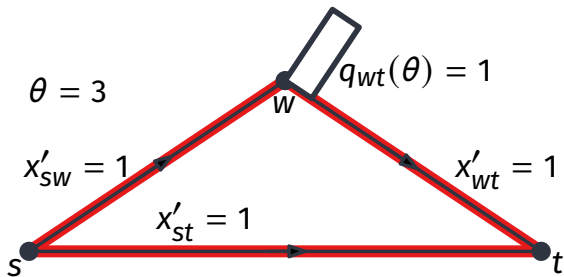
## Example again



## Example again

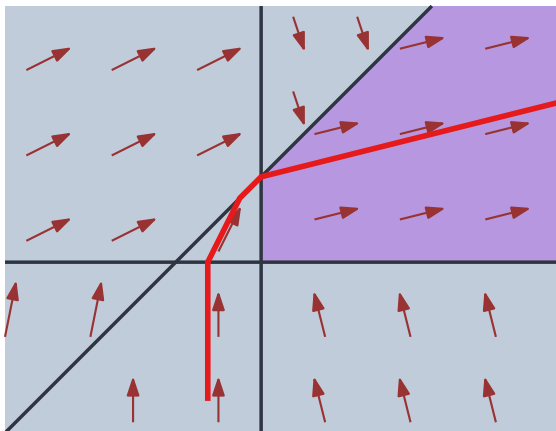


## Example again



# Long-term behaviour

**Q:** Does an equilibrium always reach a **steady state**, after which  $\ell$  is linear?



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## Theorem

*O.-Serinig-Vargas Koch '21*

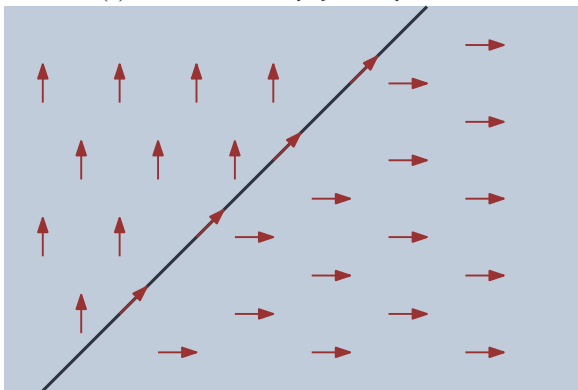
A steady state is always reached in finite time.

- Builds on [Cominetti-Correa-O. \(2017, 2021\)](#), which shows this under the condition that the capacity of the network is at least  $u_0$ .  
Implies bounded queues in this case.
- Key is the construction of a (rather non-obvious) non-decreasing potential.



# Uniqueness and continuity

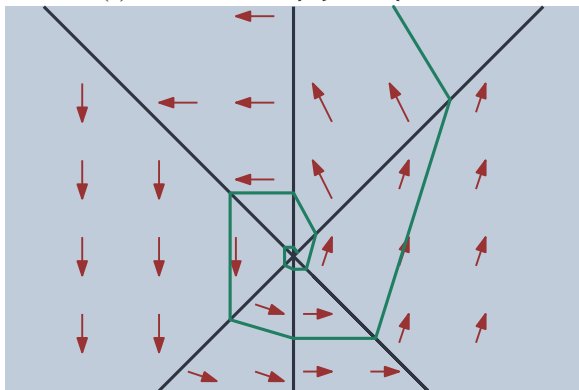
Uniqueness of  $Z(\cdot)$  does not imply uniqueness of  $\ell$ .





# Uniqueness and continuity

Uniqueness of  $Z(\cdot)$  does not imply uniqueness of  $\ell$ .



## Theorem

*O.-Sering-Vargas Koch '21*

Equilibrium trajectories are unique and depend continuously on initial conditions  $\ell(0)$ .

The vector fields that describe equilibria dynamics are very special!

# Back to stability

## Main theorem

*O.-Sering-Vargas Koch FOCS '23*

(Informal.) For packet-based models (as packet size goes to zero) and for  $\epsilon$ -equilibria (as  $\epsilon \rightarrow 0$ ), convergence to the dynamic equilibrium is guaranteed.

# Strict $\delta$ -equilibria

## Exact equilibria

A strategy profile  $\varphi$  is an exact equilibrium if

$$d_v(a) = \ell_v(\vartheta_a) \quad \text{for all } a \in \mathcal{A}, v \in \varphi(a);$$

the labels  $\ell_v(\theta)$  then define an **equilibrium trajectory**.

## Strict $\delta$ -equilibria

A strategy profile  $\tilde{\varphi}$  is a **strict  $\delta$ -equilibrium** if

$$\tilde{d}_v(a) \leq \tilde{\ell}_v(\vartheta_a) + \delta \quad \text{for all } a \in \mathcal{A}, v \in \tilde{\varphi}(a);$$

the labels  $\tilde{\ell}_v(\theta)$  then define a  **$\delta$ -trajectory**.

- An  $\epsilon$ -approximate equilibrium is a strict  $\epsilon$ -equilibrium (but not conversely).

# Formal theorem statement

## Theorem

- $\epsilon$ -equilibria are strict  $O(\epsilon)$ -equilibria.
- Packet equilibria with packets of size  $\epsilon$  are strict  $O(\epsilon)$ -equilibria

## Main theorem

*O.-Serjng-Vargas Koch FOCS '23*

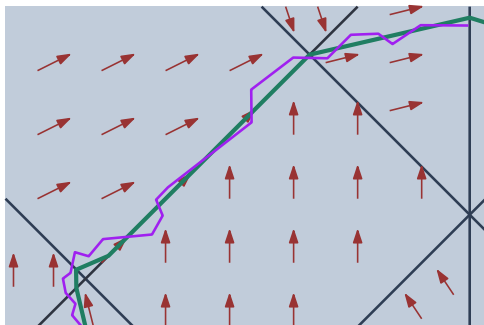
Strict  $\delta$ -equilibria converge to exact dynamic equilibria as  $\delta \rightarrow 0$ .

More precisely: if  $\ell(\theta)$  is an equilibrium trajectory, and  $\tilde{\ell}^{(i)}(\theta)$  a  $\delta^{(i)}$ -trajectory for each  $i$ , with  $\delta^{(i)} \rightarrow 0$ , then

$$\sup_{\theta \geq 0} \|\ell(\theta) - \tilde{\ell}^{(i)}(\theta)\| \rightarrow 0.$$

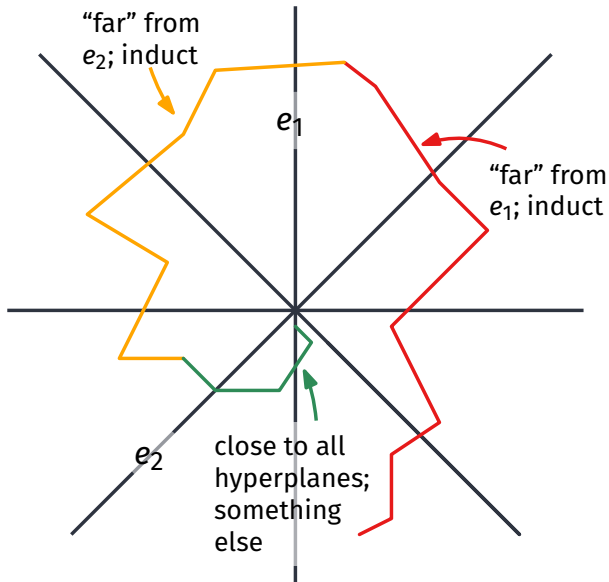
# Continuity vs stability

- Our continuity result shows that equilibria are stable under a single perturbation (or finite number of perturbations).
- Clearly necessary, but not nearly enough:
  - $\tilde{\ell}$  need not follow the vector field anywhere.
  - A slow drift away is not acceptable; something must “pull  $\tilde{\ell}$  back”.



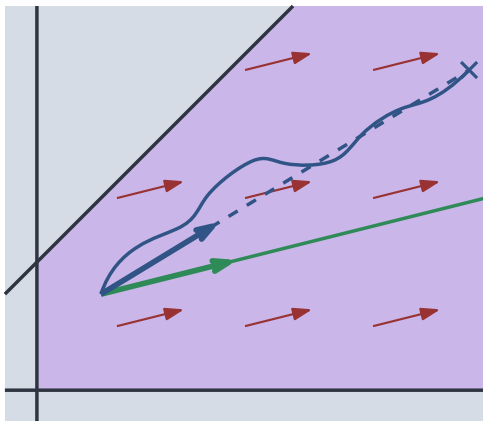


Proof heavily exploits induction on the number of hyperplanes.



**Base case:** equilibrium trajectory is in “steady state”: all labels and queues change linearly forever.

We give a “robust” version of proof by Cominetti, Correa and Larré that  $Z(\cdot)$  is unique.



# Conclusion

## Theorem

*O.-Sering-Vargas Koch '23*

Strict  $\delta$ -equilibria converge to exact dynamic equilibria as  $\delta \rightarrow 0$ .

- Dependence on  $\delta$  is horrible...  
Can it be shown that  $\sup_{\theta} \|\ell(\theta) - \tilde{\ell}(\theta)\| = o(\delta)$ ?

Many basic open questions about equilibria remain:

- Computational complexity of computing  $X(\cdot)$
- Price of anarchy
- Structure of equilibria with multiple origin-destination pairs
- ...

# Conclusion

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**Thank you!**