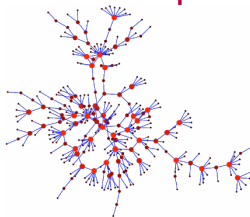


01RMHNG-03RMHPF-01RMING

# Network Dynamics

## Week 4

### Network Flow Optimization



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# This week

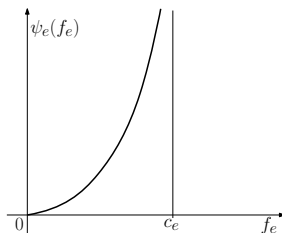
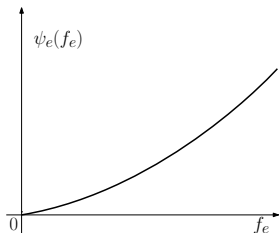
- ▶ Network Flow Optimization
- ▶ Lagrangian Techniques
- ▶ User Equilibrium vs System Optimum in Traffic Networks
- ▶ Price of Anarchy and Optimal Tolling

# Network flows

- ▶ multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ , node-link incidence matrix  $B$
- ▶ exogenous net flow vector  $\nu \in \mathbb{R}^{\mathcal{V}}$  such that  $\sum_i \nu_i = 0$
- ▶ network flow is a vector  $f \in \mathbb{R}_+^{\mathcal{E}}$  such that  $Bf = \nu$
- ▶ typically several feasible solutions when problem feasible
- ▶ network flow optimization: selection of “best” flow

# Network flow optimization

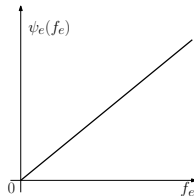
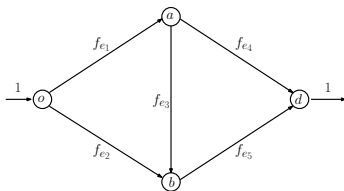
- convex nondecreasing cost functions  $\psi_e(f_e)$  on every link  $e \in \mathcal{E}$



- capacity  $c_e = \inf\{f_e \geq 0 : \psi_e(f_e) = +\infty\}$  either finite or infinite
- **Network flow optimization problem:** given exogenous net-flow  $\nu$

$$M(\nu) := \min_{\substack{f \in \mathbb{R}_+^{\mathcal{E}} \\ Bf = \nu}} \sum_{e \in \mathcal{E}} \psi_e(f_e)$$

## Example 1: Shortest path and optimal transport



►  $b_e = \text{length of link } e > 0$

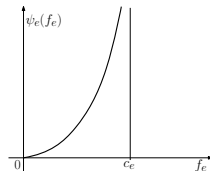
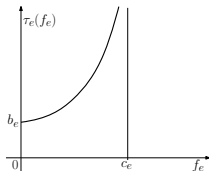
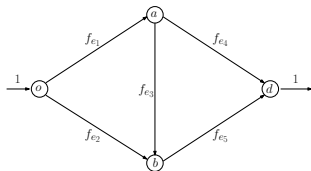
► link cost functions

$$\psi_e(f_e) = b_e f_e$$

►  $\nu = \delta(o) - \delta(d)$

► if  $\nu = \nu^+ - \nu^-$  with  $\nu^+, \nu^- \in \mathbb{R}_+^{\mathcal{V}}$ , optimal transport

## Ex. 2: System-Optimum Traffic Assignment (SO-TAP)



- convex nondecreasing delay functions  $\tau_e(f_e)$ . E.g.,

$$\tau_e(f_e) = \frac{b_e}{1 - f_e/c_e}, \quad e \in \mathcal{E}$$

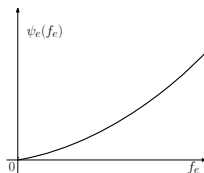
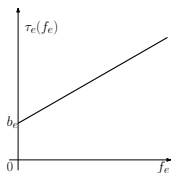
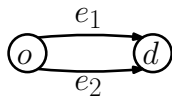
- corresponding convex nondecreasing cost

$$\psi_e(f_e) = f_e \cdot \tau_e(f_e)$$

- SO-TAP:

$$\begin{aligned} \min_{f \in \mathbb{R}_+^{\mathcal{E}}} \quad & \sum_{e \in \mathcal{E}} f_e \tau_e(f_e) \\ \text{s.t.} \quad & Bf = \nu \end{aligned}$$

## Ex. 2a: SO-TAP on two parallel links, affine delays



- $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  with  $\mathcal{V} = \{o, d\}$ ,  $\mathcal{E} = \{e_1, e_2\}$ , where

$$\theta(e_1) = \theta(e_2) = o \quad \kappa(e_1) = \kappa(e_2) = d$$

- let  $\tau_e(f_e) = a_e f_e + b_e$ , where  $a_e > 0$  and  $b_e > 0$ , so that

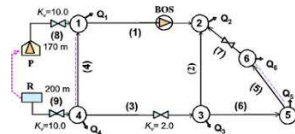
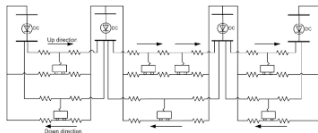
$$\psi_e(f_e) = f_e \cdot \tau_e(f_e) = a_e f_e^2 + b_e f_e$$

- for  $v > 0$  solution of  $\min_{\substack{f_{e_1} \geq 0, f_{e_2} \geq 0 \\ f_{e_1} + f_{e_2} = v}} \{\psi_{e_1}(f_{e_1}) + \psi_{e_2}(f_{e_2})\}$  is

$$f_{e_1} = v - f_{e_2} = \begin{cases} 0 & \text{if } \frac{b_2 - b_1}{2v} < -a_2 \\ \frac{2a_2 v + b_2 - b_1}{2(a_1 + a_2)} & \text{if } -a_2 \leq \frac{b_2 - b_1}{2v} \leq a_1 \\ v & \text{if } \frac{b_2 - b_1}{2v} > a_1 \end{cases}$$

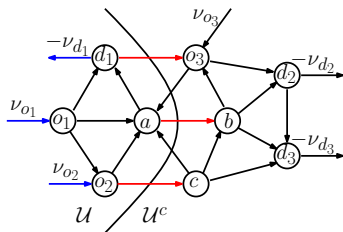
## Example 3: Power dissipation

- ▶ Undirected  $\mathcal{G} = (\mathcal{V}, \mathcal{E}, h)$ , link conductances  $h_e = h_{\bar{e}} > 0$
- ▶ Power dissipation on link  $e$ :  $\psi_e(f_e) = \frac{f_e^{\alpha+1}}{(\alpha+1)h_e^\alpha}$
- ▶ Total power dissipation:  $\sum_{e \in \mathcal{E}} \psi_e(f_e) = \sum_{e \in \mathcal{E}} \frac{f_e^{\alpha+1}}{(\alpha+1)h_e^\alpha}$
- ▶  $\alpha = 1 \Rightarrow$  direct current (DC) power networks
- ▶  $\alpha = 2 \Rightarrow$  gas networks
- ▶  $\alpha = 1/1.85 \simeq 0.54 \Rightarrow$  hydraulic networks





# Feasibility



**Proposition:** On  $\mathcal{G}$  with capacities  $c_e$ , network flow optimization

$$M(\nu) := \min_{\substack{f \in \mathbb{R}_+^{\mathcal{E}} \\ Bf = \nu}} \sum_{e \in \mathcal{E}} \psi_e(f_e)$$

is **feasible** if and only if

$$\sum_{i \in \mathcal{U}} \nu_i < c_{\mathcal{U}} \quad \text{for every } \mathcal{U} \subseteq \mathcal{V} \text{ s.t. } \sum_{i \in \mathcal{U}} \nu_i > 0$$

## Lagrangian techniques and duality

$$M(\nu) := \min_{\substack{f \in \mathbb{R}_+^{\mathcal{E}} \\ Bf = \nu}} \sum_{e \in \mathcal{E}} \psi_e(f_e)$$

► Lagrange multipliers  $\lambda_i$  for every node  $i \in \mathcal{V}$

► Lagrangian function

$$L(f, \lambda, \nu) = \sum_{e \in \mathcal{E}} \psi_e(f_e) + \sum_{i \in \mathcal{V}} \lambda_i \left( \sum_{e \in \mathcal{E}: \kappa(e)=i} f_e - \sum_{e \in \mathcal{E}: \theta(e)=i} f_e + \nu_i \right)$$

## Lagrangian techniques and duality

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► Lagrange multipliers  $\lambda_i$  for every node  $i \in \mathcal{V}$

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$$\begin{aligned} L(f, \lambda, \nu) &= \sum_{e \in \mathcal{E}} \psi_e(f_e) + \sum_{i \in \mathcal{V}} \lambda_i \left( \sum_{e \in \mathcal{E}: \kappa(e)=i} f_e - \sum_{e \in \mathcal{E}: \theta(e)=i} f_e + \nu_i \right) \\ &= \sum_{e \in \mathcal{E}} (\psi_e(f_e) - f_e(\lambda_{\theta(e)} - \lambda_{\kappa(e)})) + \sum_{i \in \mathcal{V}} \lambda_i \cdot \nu_i \end{aligned}$$

# Lagrangian techniques and duality

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► **Dual function** (always **concave** in  $\lambda$ ):

$$\begin{aligned} D(\lambda, \nu) &:= \inf_{f \in \mathbb{R}_+^{\mathcal{E}}} L(f, \lambda, \nu) \\ &= \sum_{e \in \mathcal{E}} \inf_{f_e \geq 0} \{ \psi_e(f_e) - f_e(\lambda_{\theta(e)} - \lambda_{\kappa(e)}) \} + \sum_{i \in \mathcal{V}} \lambda_i \cdot \nu_i \end{aligned}$$

## Complementary slackness conditions

► **Lemma:** For convex nondecreasing cost  $\psi_e : [0, +\infty) \rightarrow [0, +\infty]$  and pair of Lagrange multipliers  $(\lambda_{\theta(e)}, \lambda_{\kappa(e)})$  in  $\mathbb{R}^2$

$$f_e^* \in \operatorname{argmin}_{f_e \geq 0} \{ \psi_e(f_e) - (\lambda_{\theta(e)} - \lambda_{\kappa(e)}) f_e \}$$

if and only if the following **CS** conditions hold true

$$f_e^* \geq 0 \quad \psi_e'(f_e^*) \geq \lambda_{\theta(e)} - \lambda_{\kappa(e)} \quad f_e^* (\psi_e'(f_e^*) - (\lambda_{\theta(e)} - \lambda_{\kappa(e)})) = 0$$

## Complementary slackness conditions

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► **Remark:** if  $\psi_e(x)$  is strictly convex on the interval  $[0, c_e)$ , and

$$\lambda_{\theta(e)} - \lambda_{\kappa(e)} < \sup\{\psi_e'(x) : x \in [0, c_e)\},$$

then the solution of **CS** is unique and given by

$$f_e^* = \begin{cases} 0 & \text{if } \lambda_{\theta(e)} - \lambda_{\kappa(e)} \leq \psi_e'(0) \\ (\psi_e')^{-1}(\lambda_{\theta(e)} - \lambda_{\kappa(e)}) & \text{if } \lambda_{\theta(e)} - \lambda_{\kappa(e)} > \psi_e'(0). \end{cases}$$

## Lagrangian techniques and duality

**Proposition:**  $f^*$  in  $\mathbb{R}_+^{\mathcal{E}}$  s.t.  $Bf^* = \nu$  and  $\lambda^*$  in  $\mathbb{R}^{\mathcal{V}}$  satisfy CS conditions on every  $e$  in  $\mathcal{E}$  if and only if

(i)  $f^*$  is optimal solution of

$$M(\nu) := \min_{\substack{f \in \mathbb{R}_+^{\mathcal{E}} \\ Bf = \nu}} \sum_{e \in \mathcal{E}} \psi_e(f_e)$$

(ii)  $\lambda^*$  is an optimal solution of

$$M^*(\nu) = \max_{\lambda \in \mathbb{R}^{\mathcal{E}}} D(\lambda, \nu)$$

(iii)  $M(\nu) = M^*(\nu)$

## Lagrangian techniques and duality

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(iii)  $M(\nu) = M^*(\nu)$

► **Application:** to find  $f^*$ , solve CS as a function of  $\lambda$ , finding  $f^*(\lambda)$ , then impose conservation constraints

$$Bf^*(\lambda) = \nu.$$

Solution  $\lambda^*$  to this nonlinear system is such that  $f^*(\lambda^*)$  is an optimal flow and  $\lambda^*$  is an optimal solution of dual problem.



# Lagrangian techniques and duality

► note that

$$D(\lambda, \nu) = \sum_i \lambda_i \cdot \nu_i - \sum_e \psi_e^*(\lambda_{\theta(e)} - \lambda_{\kappa(e)})$$

where the dual cost on link  $e \in \mathcal{E}$

$$\psi_e^*(y_e) = \sup_{f_e \geq 0} \{y_e f_e - \psi_e(f_e)\}$$

is the **Fenchel transform** of the cost  $\psi_e(f_e)$  and represents the maximum profit that a link operator can make if it charges  $y_e$  per unit of flow and pays  $\psi_e(f_e)$  to transport  $f_e$  units

# Lagrangian techniques and duality

**Theorem:**  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ , nonincreasing costs  $\psi_e(f_e)$ , capacities  $c_e$

- (i) optimal network flow problem feasible, i.e.,  $M(\nu) < +\infty$ , if and only if  $\sum_{i \in \mathcal{U}} \nu_i < c_{\mathcal{U}}$ , for every  $\mathcal{U} \subset \mathcal{V}$  s.t.  $\sum_{i \in \mathcal{U}} \nu_i > 0$ .

If  $\psi_e(f_e)$  convex differentiable on  $[0, +\infty)$ , then for all  $\nu$  as above

- (ii) flow vector  $f^*$  optimal  $\Leftrightarrow$  satisfies CS conditions for Lagrange multipliers  $\lambda = \lambda^*$  that solve dual optimization problem

$$M^*(\nu) = \max_{\lambda \in \mathbb{R}^{\mathcal{V}}} D(\lambda, \nu)$$

- (iii) if optimal cost  $M(\nu)$  is differentiable in  $\nu$ , then

$$\frac{\partial}{\partial \nu_i} M(\nu) - \frac{\partial}{\partial \nu_j} M(\nu) = \lambda_i^* - \lambda_j^*$$

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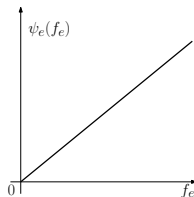
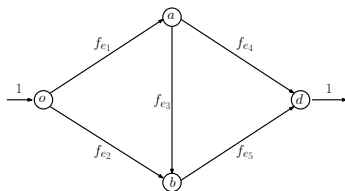
$$M^*(\nu) = \max_{\lambda \in \mathbb{R}^{\mathcal{V}}} D(\lambda, \nu)$$

- (iii) if optimal cost  $M(\nu)$  is differentiable in  $\nu$ , then

$$\frac{\partial}{\partial \nu_i} M(\nu) - \frac{\partial}{\partial \nu_j} M(\nu) = \lambda_i^* - \lambda_j^*$$

► **Application** (of (ii)): find  $\lambda^*$  solution of dual optimization problem (which is convex and unconstrained), then solve CS to find  $f^*(\lambda^*)$ . Such  $f^*(\lambda^*)$  is an optimal flow.

## Example – complementary slackness for shortest path



► cost functions  $\psi_e(f_e) = b_e f_e$  where  $b_e = \text{length of link } e > 0$

►  $\nu = \delta(o) - \delta(d)$

► CS conditions yield

$$b_e \begin{cases} = \lambda_{\theta(e)} - \lambda_{\kappa(e)} & \text{if } f_e^* > 0 \\ \geq \lambda_{\theta(e)} - \lambda_{\kappa(e)} & \text{if } f_e^* = 0 \end{cases} \quad \forall e \in \mathcal{E}$$

► For every  $o - d$  path  $\gamma = (e_1, \dots, e_l)$

$$\sum_{j=1}^l b_{e_j} \begin{cases} = \lambda_o - \lambda_d & \text{if } f_{e_j}^* > 0 & \text{for all } 1 \leq j \leq l \\ \geq \lambda_o - \lambda_d & \text{if } f_{e_j}^* = 0 & \text{for some } 1 \leq j \leq l \end{cases}$$

## Example – complementary slackness for power dissipation

$$\psi_e(f_e) = \frac{f_e^{\alpha+1}}{(\alpha+1)h_e^\alpha} \quad e \in \mathcal{E}$$

- ▶ For  $f_e \geq 0$ ,  $\psi'_e(f_e) = f_e^\alpha / h_e^\alpha$
- ▶ complementary slackness conditions yield

$$f_e^* = h_e [\lambda_{\theta(e)} - \lambda_{\kappa(e)}]_+$$

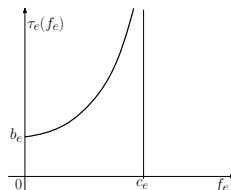
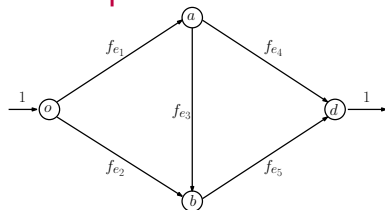
- ▶ for DC power networks ( $\alpha = 1$ ), let

$$W_{ij} = \sum_{e \in \mathcal{E}: \theta(e)=i, \kappa(e)=j} h_e$$

then the net-flow  $z_{ij}$  from node  $i$  to node  $j$  is

$$z_{ij} = W_{ij}(\lambda_i - \lambda_j) \quad \Longleftrightarrow \quad \text{Ohm's law}$$

# System Optimum Traffic Assignment (SO-TAP)



- convex nondecreasing delay functions  $\tau_e(f_e)$ . E.g.,

$$\tau_e(f_e) = \frac{b_e}{1 - f_e/c_e}, \quad e \in \mathcal{E}$$

- corresponding convex nondecreasing cost

$$\psi_e(f_e) = f_e \cdot \tau_e(f_e)$$

- SO-TAP:

$$\begin{aligned} \min_{f \in \mathbb{R}_+^{\mathcal{E}}} \quad & \sum_{e \in \mathcal{E}} f_e \cdot \tau_e(f_e) \\ \text{s.t.} \quad & Bf = \nu \end{aligned}$$

## When drivers choose their route

- ▶ Drivers, total amount  $v$ , can choose different  $o$ - $d$  paths
- ▶  $\Gamma_{od}$  the set of all  $o$ - $d$  paths
- ▶ Link-path incidence matrix  $A \in \{0, 1\}^{\mathcal{E} \times \Gamma_{od}}$

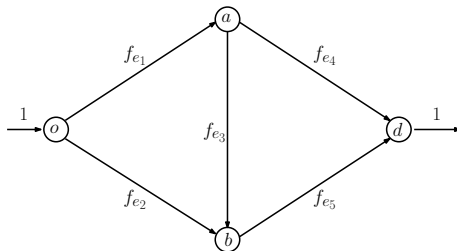
$$A_{e\gamma} = \begin{cases} 1 & \text{if link } e \text{ is along path } \gamma \\ 0 & \text{if link } e \text{ is not along path } \gamma \end{cases}$$

- ▶ Path flow  $z \in \mathbb{R}^{\Gamma_{od}}$ ,  $\mathbf{1}'z = v$ ,  $z \geq 0$
- ▶ Recall that

$$(BA)_{i\gamma} = \begin{cases} +1 & \text{if } i = o \\ -1 & \text{if } i = d \\ 0 & \text{if } i \neq o, d \end{cases}$$

- ▶ Link flow  $f = Az$

## Example – Link-path incidence matrix



$$A = \begin{array}{ccc|c} \gamma_1 & \gamma_2 & \gamma_3 & \\ \left[ \begin{array}{ccc} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{array} \right] & \begin{array}{l} e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \end{array} \end{array}$$

$$B = \begin{array}{ccccc|c} e_1 & e_2 & e_3 & e_4 & e_5 & \\ \left[ \begin{array}{ccccc} +1 & +1 & 0 & 0 & 0 \\ -1 & 0 & +1 & +1 & 0 \\ 0 & -1 & -1 & 0 & +1 \\ 0 & 0 & 0 & -1 & -1 \end{array} \right] & \begin{array}{l} o \\ a \\ b \\ d \end{array} \end{array}$$



# Wardrop equilibrium

► Wardrop equilibrium  $f^{(0)}$ : The flow vector

$$f^{(0)} = Az$$

where  $z \in \mathbb{R}^{\Gamma_{od}}$  is such that  $z \geq 0$ ,  $\mathbb{1}'z = v$ , and for  $\gamma \in \Gamma_{od}$

$$z_\gamma > 0 \quad \implies \quad \underbrace{\sum_{e \in \mathcal{E}} A_{e\gamma} \tau_e(f_e^{(0)})}_{\text{total delay on path } \gamma} \leq \underbrace{\sum_{e \in \mathcal{E}} A_{e\tilde{\gamma}} \tau_e(f_e^{(0)})}_{\text{total delay on path } \tilde{\gamma}} \quad \forall \tilde{\gamma} \in \Gamma_{od}$$

► Interpretation: drivers choose their fastest path

# Wardrop equilibrium

- Wardrop equilibrium  $f^{(0)}$ : The flow vector

$$f^{(0)} = Az$$

where  $z \in \mathbb{R}^{\Gamma_{od}}$  is such that  $z \geq 0$ ,  $\mathbb{1}'z = v$ , and for  $\gamma \in \Gamma_{od}$

$$z_\gamma > 0 \implies \underbrace{\sum_{e \in \mathcal{E}} A_{e\gamma} \tau_e(f_e^{(0)})}_{\text{total travel time on path } \gamma} \leq \underbrace{\sum_{e \in \mathcal{E}} A_{e\tilde{\gamma}} \tau_e(f_e^{(0)})}_{\text{total travel time on path } \tilde{\gamma}} \quad \forall \tilde{\gamma} \in \Gamma_{od}$$

- Interpretation: drivers choose their fastest path
- Proposition: Wardrop equilibrium = solution of UO-TAP

$$\min_{f \in \mathbb{R}_+^{\mathcal{E}}} \sum_{e \in \mathcal{E}} \int_0^{f_e} \tau_e(s) ds$$
$$Bf = v(\delta^{(o)} - \delta^{(d)})$$

# Wardrop equilibrium

► **Proposition:** Wardrop equilibrium = solution of UO-TAP

$$\begin{aligned} \min_{f \in \mathbb{R}_+^{\mathcal{E}}} \quad & \sum_{e \in \mathcal{E}} \int_0^{f_e} \tau_e(s) ds \\ & Bf = v(\delta^{(o)} - \delta^{(d)}) \end{aligned}$$

► **Proof:**

# Price of Anarchy

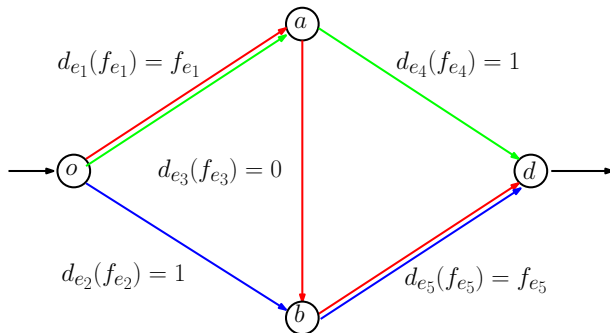
► price of anarchy of Wardrop equilibrium  $f^{(0)}$  is

$$\text{PoA}(0) = \frac{\sum_{e \in \mathcal{E}} f_e^{(0)} \tau_e(f_e^{(0)})}{\min_{\substack{f \in \mathbb{R}_+^{\mathcal{E}} \\ Bf = v(\delta^{(o)} - \delta^{(d)})}} \sum_{e \in \mathcal{E}} f_e \tau_e(f_e)},$$

total delay at the Wardrop equilibrium / total delay at system optimum

► Observe:  $\text{PoA}(0) \geq 1$

## Example - Braess paradox



- ▶ Three paths  $\gamma_1$ ,  $\gamma_2$ ,  $\gamma_3$
- ▶ Wardrop equilibrium:  $f^{(0)} = (1, 0, 1, 0, 1)$
- ▶ Social optimum:  $f^* = (1/2, 1/2, 0, 1/2, 1/2)$

# Toll design

- ▶ Vector of **tolls**:  $\omega = (\omega_e)_{e \in \mathcal{E}}$
- ▶ **Perceived cost** =  $\omega_e + \tau_e(f_e)$  on link  $e$
- ▶ Wardrop equilibrium with tolls

$$f^{(\omega)} = Az$$

where  $z \in \mathbb{R}^{\Gamma_{od}}$  is such that  $z \geq 0$ ,  $\mathbb{1}'z = v$ , and for  $\gamma \in \Gamma_{od}$

$$z_\gamma > 0 \implies \underbrace{\sum_{e \in \mathcal{E}} A_{e\gamma} \left( \tau_e(f_e^{(\omega)}) + \omega_e \right)}_{\substack{\text{total perceived} \\ \text{cost on path } \gamma}} \leq \underbrace{\sum_{e \in \mathcal{E}} A_{e\tilde{\gamma}} \left( \tau_e(f_e^{(\omega)}) + \omega_e \right)}_{\substack{\text{total perceived} \\ \text{cost on path } \tilde{\gamma}}} \quad \forall \tilde{\gamma} \in \Gamma_{od}$$

- ▶ Can we find  $\omega$  s.t.  $\text{PoA}(\omega) = 1$ ?

# Toll design

► **Theorem:** For nondecreasing  $\tau_e(f_e)$ , convex  $f_e \cdot \tau_e(f_e)$

Wardrop equilibrium with tolls  $\omega$  satisfies

$$f^{(\omega)} = \underset{\substack{f \in \mathbb{R}_+^{\mathcal{E}} \\ Bf = v(\delta^{(o)} - \delta^{(d)})}}{\operatorname{argmin}} \sum_{e \in \mathcal{E}} \left( \int_0^{f_e} \tau_e(s) ds + \omega_e f_e \right)$$

► **Corollary:** With **marginal cost tolls**

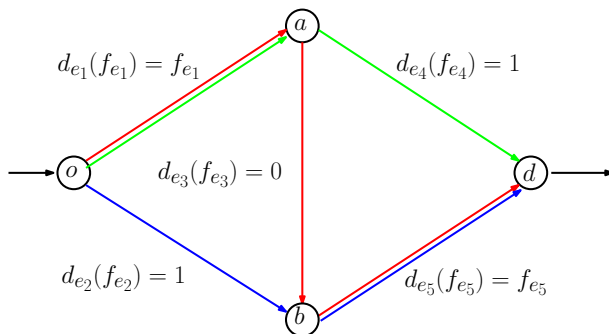
$$\omega_e^* = f_e^* \cdot \tau_e'(f_e^*)$$

computed at social optimum  $f^* = \underset{Bf = v(\delta^{(o)} - \delta^{(d)})}{\operatorname{argmin}} \sum_{e \in \mathcal{E}} f_e \cdot \tau_e(f_e)$

$$f^{(\omega^*)} = f^*$$

► marginal cost tolls  $\Leftrightarrow$  “**internalize negative externality**”

## Example – Toll design



- Tolls  $\omega_1^* = \omega_5^* = 1/2$ ,  $\omega_2^* = \omega_3^* = \omega_4^* = 0$
- Wardrop equilibrium with tolls:  $f^{(\omega^*)} = (0.5, 0.5, 0, 0.5, 0.5)$
- Social optimum:  $f^* = (0.5, 0.5, 0, 0.5, 0.5)$
- $\exists$  other optimal toll choices, e.g.,  $\omega = (0, 0, \beta, 0, 0)$ ,  $\beta \geq 1/2$