

Networks, Potential Functions, and the Price of Anarchy

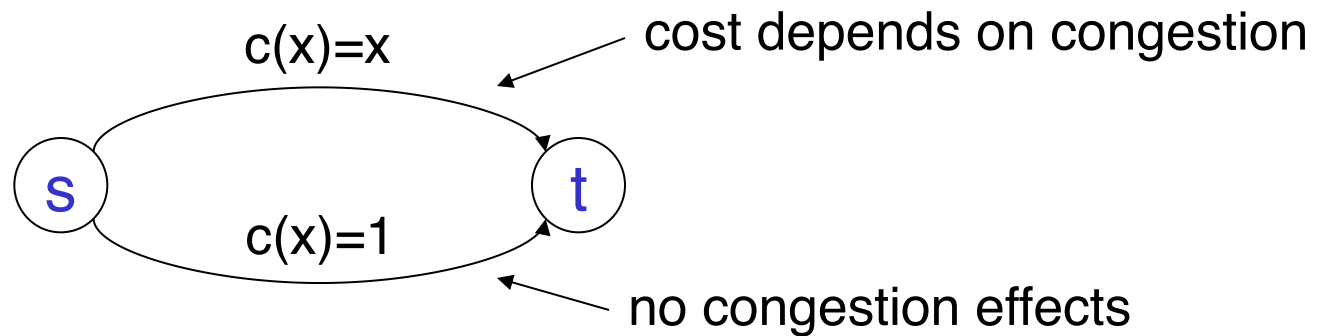
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PRICE OF ANARCHY (1G)

Pigou's Example

Example: one unit of traffic wants to go from **s** to **t**

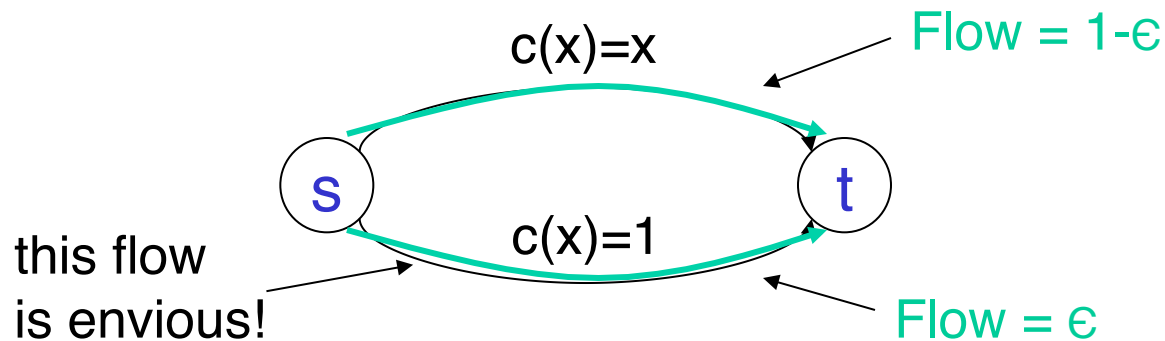


Question: what will selfish network users do?



- assume everyone wants smallest-possible cost
- [Pigou 1920]

Motivating Example

Claim: all traffic will take the top link.

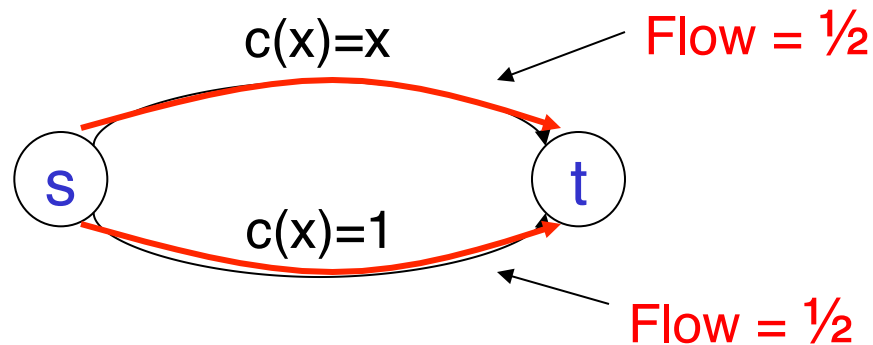


Reason:

- $\epsilon > 0$  traffic on bottom is envious
- $\epsilon = 0$  equilibrium
 - all traffic incurs one unit of cost

Can We Do Better?

Consider instead: traffic split equally

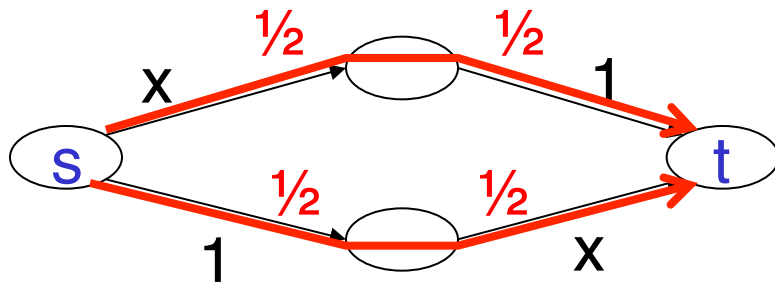


Improvement:

- half of traffic has cost 1 (same as before)
- half of traffic has cost $\frac{1}{2}$ (much improved!)

Braess's Paradox

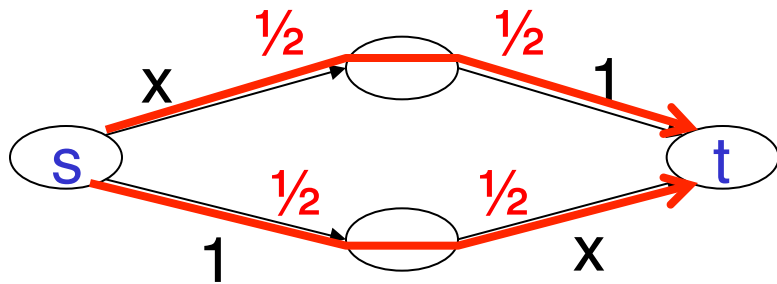
Initial Network:



Cost = 1.5

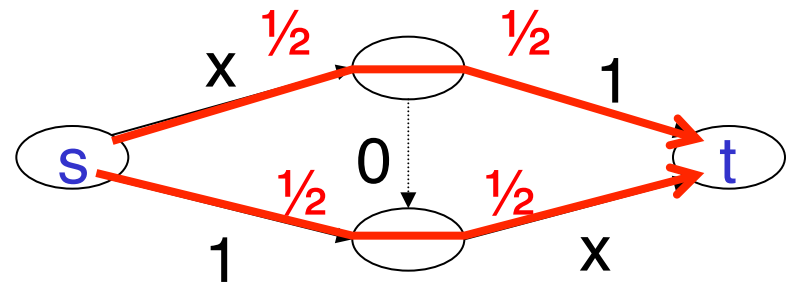
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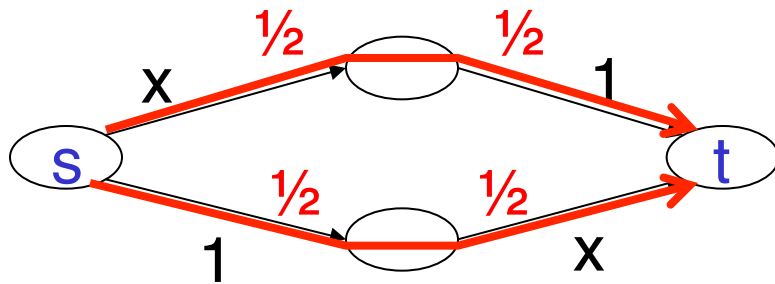
Augmented Network:



Now what?

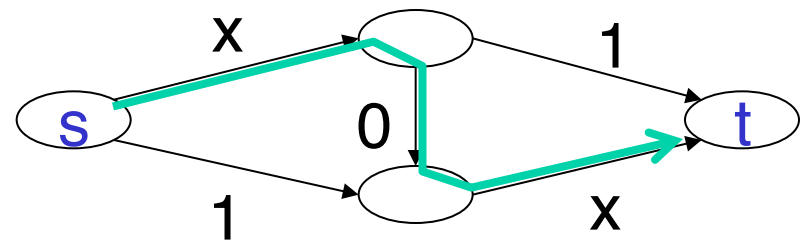
Braess's Paradox

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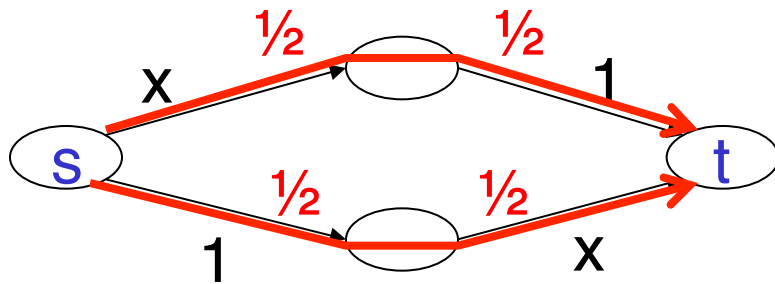
Augmented Network:



Cost = 2

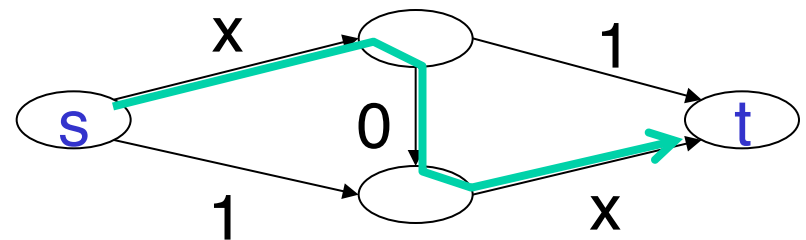
Braess's Paradox

Initial Network:



Cost = 1.5

Augmented Network:



Cost = 2

All traffic incurs more cost! [Braess 68]

- also has physical analogs [Cohen/Horowitz 91]

High-Level Overview

Motivation: equilibria of noncooperative network games typically **inefficient**

- e.g., Pigou's example + Braess's Paradox
- don't optimize natural objective functions

Price of anarchy: **quantify** inefficiency with respect to an objective function

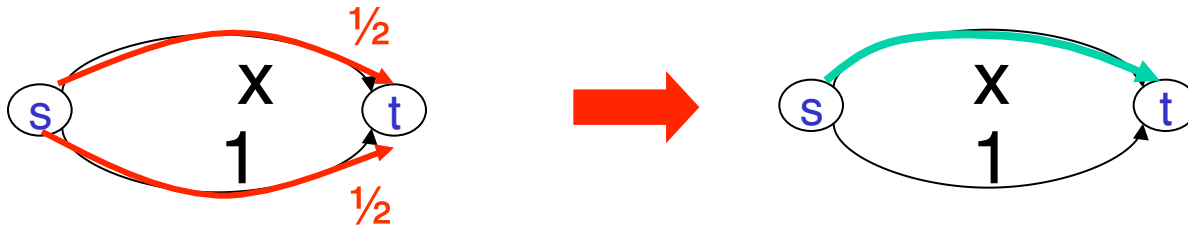
Our goal: when is the price of anarchy small?

- when does competition approximate cooperation?
- benefit of centralized control is small

Nonatomic Selfish Routing

- directed graph $G = (V, E)$
- source-destination pairs $(s_1, t_1), \dots, (s_k, t_k)$
- r_i = amount of traffic going from s_i to t_i
- for each edge e , a cost function $c_e(\cdot)$
 - assumed continuous and nondecreasing

Defn: a multicommodity flow is an *equilibrium* if all traffic routed on shortest paths.

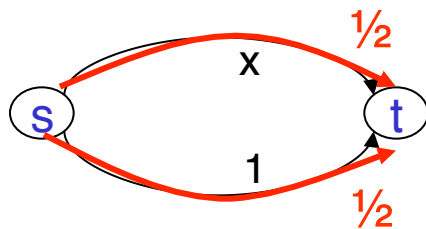


The Price of Anarchy

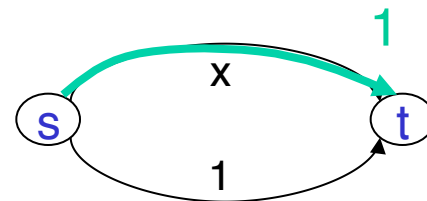
Defn: price of anarchy of a game = $\frac{\text{obj fn value of worst equilibrium}}{\text{optimal obj fn value}}$

– definition from [Koutsoupias/Papadimitriou 99]

Example: POA = 4/3 in Pigou's example



Cost = 3/4

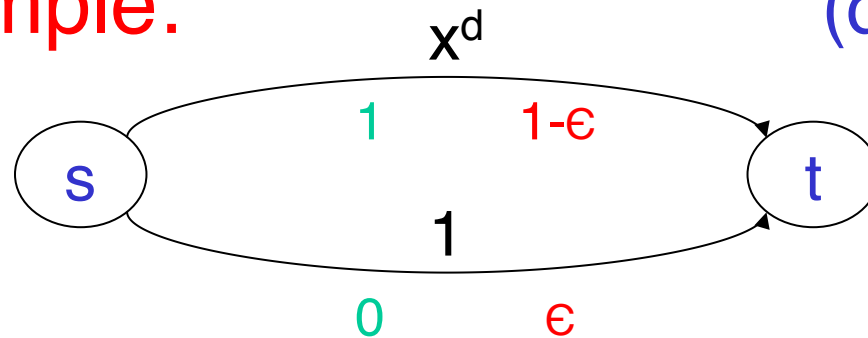


Cost = 1

A Nonlinear Pigou Network

Bad Example:

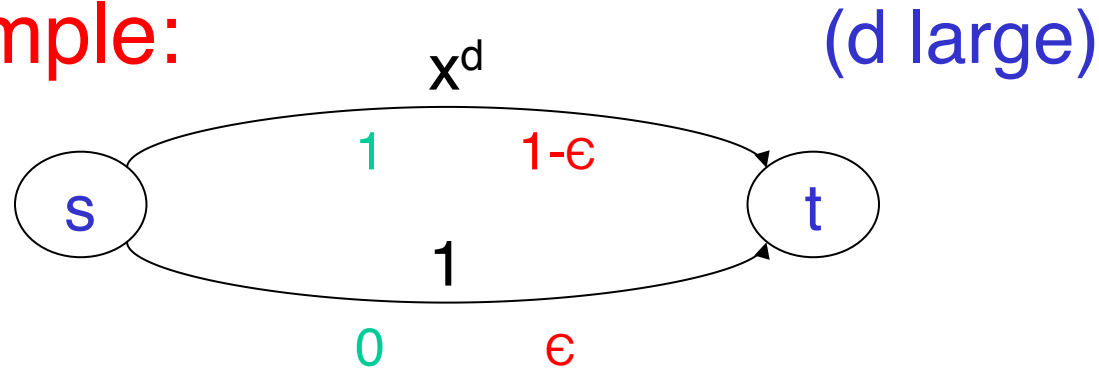
(d large)



equilibrium has cost 1, min cost  0

A Nonlinear Pigou Network

Bad Example:



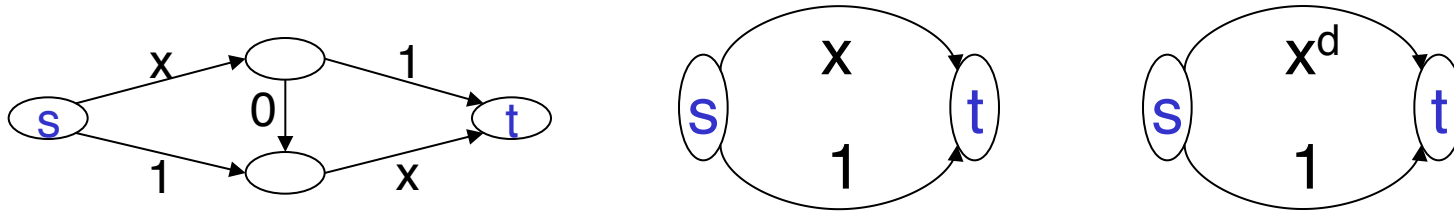
equilibrium has cost 1, min cost  0

 price of anarchy unbounded as $d \rightarrow \infty$

Goal: weakest-possible conditions under which the price of anarchy is small.

When Is the Price of Anarchy Bounded?

Examples so far:



Hope: imposing additional structure on the cost functions helps

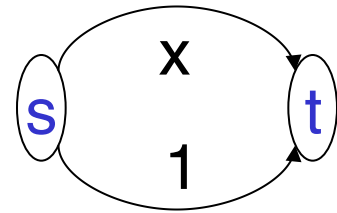
- worry: bad things happen in larger networks

Polynomial Cost Functions

Defn: linear cost function is of form $c_e(x) = a_e x + b_e$

Theorem: [Roughgarden/Tardos 00] for every network with linear cost functions:

$$\text{cost of Nash flow} \leq 4/3 \times \text{cost of opt flow}$$

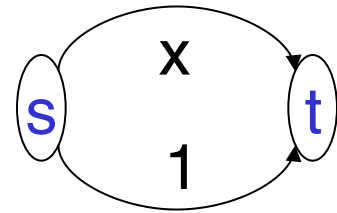


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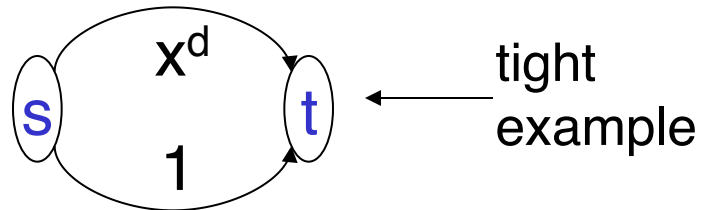
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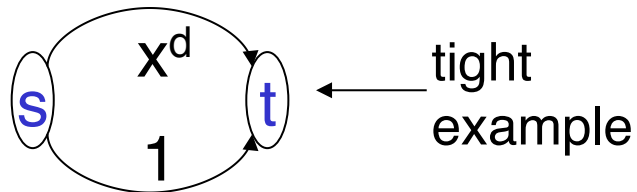


Bounded-degree polynomials: replace $4/3$ by $\approx d/\ln d$



A General Theorem

Thm: [Roughgarden 02], [Correa/Schulz/Stier Moses 03]
fix any set of cost functions. Then, a Pigou-like
example --- 2 nodes, 2 links, 1 link w/a constant
cost function --- achieves the worst P.O.A.



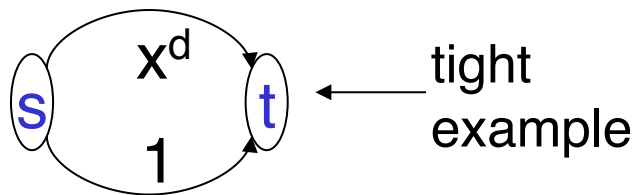
Interpretation

Bad news: inefficiency of selfish routing grows as cost functions become "more nonlinear".

- think of "nonlinear" as "heavily congested"
- recall nonlinear Pigou's example

Good news: inefficiency does not grow with network size or # of source-destination pairs.

- in lightly loaded networks, no matter how large, selfish routing is nearly optimal

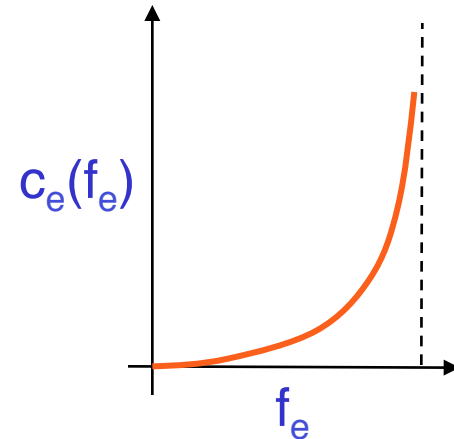


Benefit of Overprovisioning

Suppose: network is overprovisioned by $\beta > 0$ (i.e., β fraction of each edge unused).

Then: Price of anarchy is at most $\frac{1}{2}(1 + 1/\sqrt{\beta})$.

- arbitrary network size/topology, traffic matrix



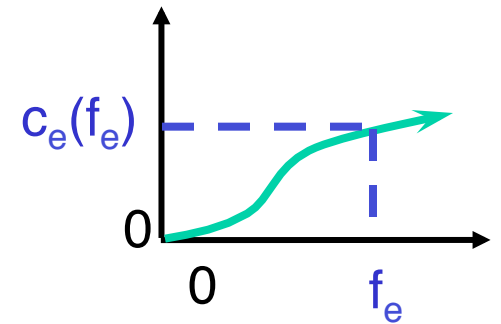
Moral: Even modest (10%) over-provisioning sufficient for near-optimal routing.

Potential Functions

- potential games: equilibria are actually optima of a related optimization problem
 - has immediate consequences for existence, uniqueness, and inefficiency of equilibria
 - see [Beckmann/McGuire/Winsten 56], [Rosenthal 73], [Monderer/Shapley 96], for original references
 - see [Roughgarden ICM 06] for survey

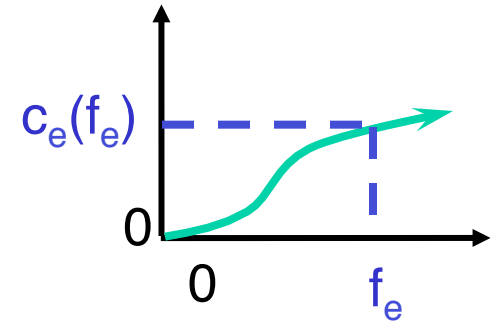
The Potential Function

Key fact: [BMV 56] Nash flows minimize “potential function”
 $\int_{\Theta} \int_e^f c_e(x) dx$ (over all flows).



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Lemma 1: locally optimal solutions are precisely the Nash flows (derivative test).

Lemma 2: all locally optimal solutions are also globally optimal (convexity).

Corollary: Nash flows exist, are unique.

Consequences for the Price of Anarchy

Example: linear cost functions.

Compare cost and potential functions:

$$C(f) = \sum_e f_e \cdot c_e(f_e) = \sum_e [a_e^2 f_e + b_e f_e]$$

$$PF(f) = \sum_e \int_0^{f_e} c_e(x) dx = \sum_e [(a_e^2 f_e)/2 + b_e f_e]$$

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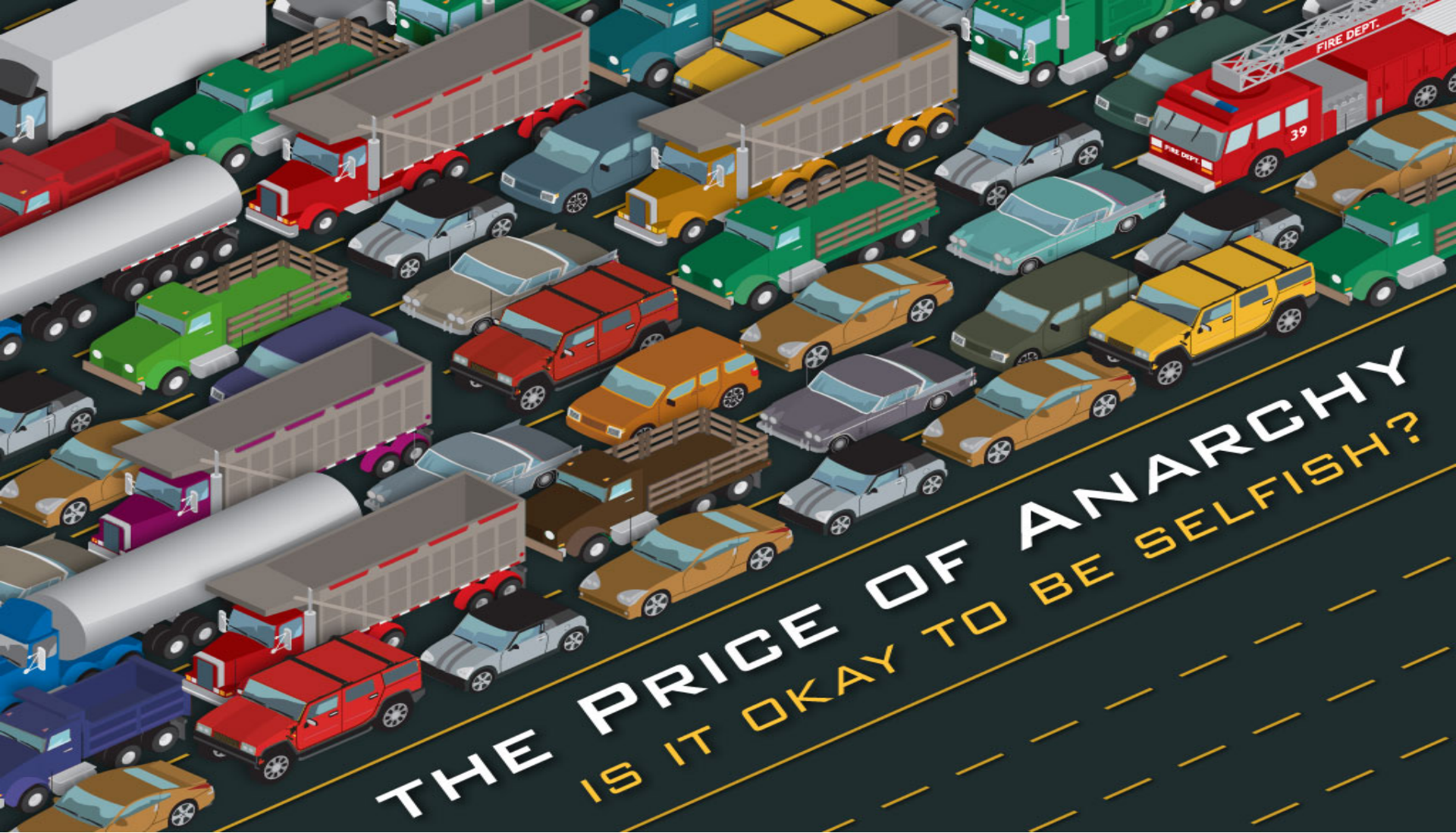
$$PF(f) = \sum_e \int_0^{f_e} c_e(x) dx = \sum_e [(a_e f_e)/2 + b_e f_e]$$

- cost, potential functions differ by factor of ≤ 2
- gives upper bound of 2 on price on anarchy
 - $C(f) \leq 2 \times PF(f) \leq 2 \times PF(f^*) \leq 2 \times C(f^*)$

Better Bounds?

Similarly: proves bound of $d+1$ for degree- d polynomials (w/nonnegative coefficients).

- not tight, but qualitatively accurate
 - e.g., price of anarchy goes to infinity with degree bound, but only linearly
- to get tight bounds, need "variational inequalities"
 - see my ICM survey for details



PRICE OF ANARCHY (2G)

POA Bounds Without Convergence

Meaning of a POA bound: *if* the game is at an equilibrium, *then* outcome is near-optimal.

Problem: what if can't reach an equilibrium?

- non-existence (pure Nash equilibria)
- intractability (mixed Nash equilibria)

[Daskalakis/Goldberg/Papadimitriou 06],
[Chen/Deng/Teng 06]

Worry: are our POA bounds “meaningless”?

POA Bounds Without Convergence

Theorem: [Roughgarden STOC 2009] most known POA bounds hold *even if the game is not at Nash equilibrium!*

- e.g., if game is played repeatedly, no-regret conditions or a few myopic best responses are enough

Concluding Remarks

- lens of approximation gives new insights into fundamental mathematical models
- good bounds for many games of interest, even out-of-Nash-equilibrium
 - refutes non-existence/intractability critiques
- routing games: worst-case price of anarchy depends only on “nonlinearity” of cost functions
 - parameterize POA bounds via worst-case examples
 - equilibria inadvertently optimize a potential function