

Chennai, India



In-vehicle congestion

Sao Paulo Metro Station



Platform congestion

London Tube



Bus stop congestion

Problem

$$\text{Given } \left\{ \begin{array}{ll} \text{transit network} & (V, A) \\ \text{travel demands} & g_i^d \geq 0 \\ \text{arc travel times} & t_a = s_a(v_a) \end{array} \right.$$

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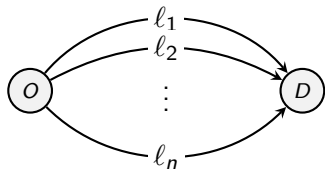
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Determine:

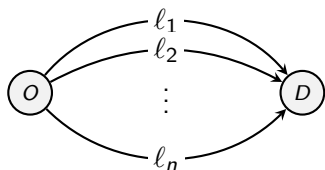
- passenger flows on line segments
- travel times for each OD pair
- waiting times at bus stops

Common lines – uncongested (Chriqui-Robillard'75)



Travel times $t_1 \leq \dots \leq t_n$
Frequencies μ_1, \dots, μ_n
Strategies $s \subseteq L$

Common lines – uncongested (Chriqui-Robillard'75)



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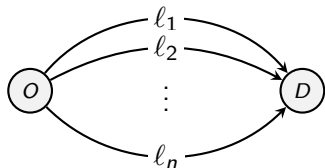
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Strategies $s \subseteq L$

$$T_s = W_s + \sum_{a \in s} t_a \pi_s^a = \frac{1 + \sum_{a \in s} t_a \mu_a}{\sum_{a \in s} \mu_a}$$

$$\tau = \min_{s \subseteq L} T_s$$

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Theorem (Chriqui-Robillard'75)

Optimal strategy: $s^ = \{1, 2, \dots, k^*\}$ take the k^* fastest lines.*

\Rightarrow linear time algorithm: $s^* = \{a : t_a \leq T_{s^*}\}$

Extension to networks

Spiess'1984, Gendreau'1984, Nguyen-Pallottino'1988, Spiess-Florian'1989,
Fernández-De Cea'1989, Wu-Florian-Marcotte'1994,
Bouzaiene-Gendreau-Nguyen'1995

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These models overlook capacity effects.

- ✓ On-board: increased discomfort
- ✓ Bus stops: increased waiting times
- ✓ Boardings & alightings: increased travel times
- ✗ Boarding probabilities: changes in flow distribution

Key point: estimate correctly the flows !



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AANB07
www.alamy.com

Denied boardings deferred to other services

Sydney

“Sydney bus and train commuters say overcrowding is still their main public transport concern.”

<http://www.abc.net.au/news/stories/2010/12/31/3104302.htm?site=sydney>

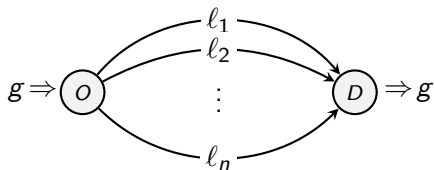
“Buses in Sydney on the busiest routes are often overcrowded and do not stop for passengers, with an extraordinary 22% of people missing their service.”

<http://www.2ue.com.au/blogs/2ue-blog/crowded-buses-just-not-stopping/20111130-1o6f1.html>



Emma Freijinger, November 2013

Common lines with congestion



Split the demand $g = \sum_{s \in L} x_s$ so that only optimal strategies are used

$$x_s > 0 \Rightarrow T_s \triangleq W_s + \sum_{a \in S} t_a \pi_s^a \text{ is minimal}$$

$$\begin{cases} W_s = W_s(x) & ? \\ \pi_s^a = \pi_s^a(x) & ? \end{cases} \quad \dots \text{ queueing theory}$$

Bulk queues $M|M|1$

$$v \longrightarrow \bigcirc \longrightarrow \mu, c \quad (\lambda < \mu C)$$

Little's formula yields the waiting time

$$W(v) = \frac{1}{v} \mathbb{E}(L) = \frac{1}{v} \frac{\rho}{1 - \rho}$$

where $\rho = \rho(v) \in [0, 1]$ is the solution of

$$\mu(\rho + \rho^2 + \dots + \rho^c) = v.$$

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Definition

Effective frequency is defined as $f(v) \triangleq \frac{1}{W(v)}$

REMARK: for light demand $v \sim 0$ we have $f(v) \sim \mu$.

Strategies with congestion

Let $s \subseteq L$ and denote $f_a(v_a)$ the effective frequency of each line $a \in L$.

Lemma (C-Correa'2001)

Assume Poisson arrivals with rate x_s . Then the expected flows v_a on the lines $a \in s$ are the unique solution of the system

$$(E) \quad v_a = x_s \frac{f_a(v_a)}{\sum_{b \in s} f_b(v_b)}$$

Moreover, setting $f_a = f_a(v_a)$ we have

$$\begin{aligned} W_s &= 1 / \sum_{b \in s} f_b \\ \pi_s^a &= f_a / \sum_{b \in s} f_b \\ T_s &= \frac{1 + \sum_{a \in s} t_a f_a}{\sum_{a \in s} f_a} \end{aligned}$$

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When there are several x_s 's we have $v_a = v_a(x)$, $f_a = f_a(x), \dots$

No analytic expressions available... yet !

Common lines – Equilibrium model

We postulate $f_a = f_a(v_a)$ as a decreasing function of the line's load, and we denote $v = v(x)$ the unique solution of the system

$$(E) \quad v_a = \sum_s x_s \pi_s^a = \sum_{s \ni a} x_s \frac{f_a(v_a)}{\sum_{b \in s} f_b(v_b)}$$

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Equilibrium: Split the demand $g = \sum_s x_s$ with $x_s \geq 0$ so that

$$(W) \quad x_s > 0 \Rightarrow T_s(v) = \tau(v)$$

where

$$T_s(v) = \frac{1 + \sum_{a \in s} t_a f_a(v_a)}{\sum_{a \in s} f_a(v_a)}$$

$$\tau(v) = \min_{s \subseteq L} T_s(v)$$

Characterization & Existence

Theorem (C-Correa'2001)

Let $\bar{v}_a(\alpha)$ be the inverse function of $v_a \mapsto \frac{v_a}{f_a(v_a)}$ and denote $\tau_\alpha \triangleq \tau(\bar{v}(\alpha))$. Then the equilibrium flows v_a are the optimal solutions of

$$\begin{aligned} \min_{(\alpha, v)} \quad & \sum_{a \in L} \left[t_a v_a + \int_0^\alpha [\tau_\xi - t_a]_+ \bar{v}'_a(\xi) d\xi \right] \\ & \sum_{a \in L} v_a = g \\ & 0 \leq v_a \leq \bar{v}_a(\alpha). \end{aligned}$$

In particular, there exists an equilibrium.

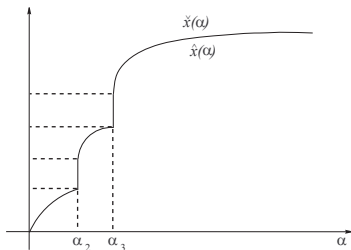
REMARK: Once α is known, the flows v are easily obtained.

Characterization & Existence

Theorem (C-Correa'2001)

Let α be the unique solution of $\sum_{t_a < \tau_\alpha} \bar{v}_a(\alpha) \leq g \leq \sum_{t_a \leq \tau_\alpha} \bar{v}_a(\alpha)$.
Then v is an equilibrium iff $\sum_{a \in L} v_a = g$ with

$$\frac{v_a}{f_a(v_a)} \begin{cases} = \alpha & \text{if } t_a < \tau_\alpha \\ \leq \alpha & \text{if } t_a = \tau_\alpha \\ = 0 & \text{if } t_a > \tau_\alpha \end{cases}$$

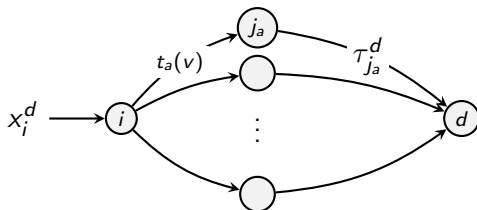


Comments

- Existence + characterization + conditions for uniqueness
- Constant time for some ranges of demand
- Co-existence of multiple equilibrium strategies
- Inefficient equilibria... Braess-type paradox
- Model consistent with simulations

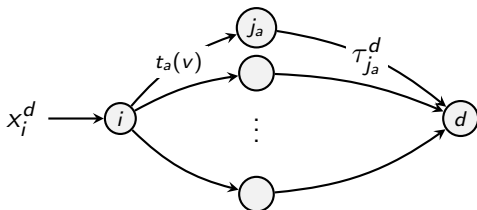
Network Transit Equilibrium

Family of common line problems coupled by flow conservation.



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Family of common line problems coupled by flow conservation.



Generalized Bellman Equations:

$$\tau_d^d = 0 ; \tau_i^d = \min_{s \subseteq A_i^+} \frac{1 + \sum_{a \in s} [t_a(v) + \tau_{j(a)}^d] f_a(v)}{\sum_{a \in s} f_a(v)}$$

Flow Conservation:

$$x_i^d \triangleq g_i^d + \sum_{A_i^-} v_a^d = \sum_{A_i^+} v_a^d$$

Existence of equilibria: Kakutani's Fixed Point Theorem

Characterization

Theorem (Cepeda-C-Florian'2006)

(v_a^d) are equilibrium flows iff there exist (α_i^d) such that

$$\frac{v_a^d}{f_a(v)} \begin{cases} = \alpha_i^d & \text{if } t_a(v) + \tau_{j(a)}^d(v) < \tau_i^d(v) \\ \leq \alpha_i^d & \text{if } t_a(v) + \tau_{j(a)}^d(v) = \tau_i^d(v) \\ = 0 & \text{if } t_a(v) + \tau_{j(a)}^d(v) > \tau_i^d(v) \end{cases}$$

These are precisely the optimal solutions of

$$\begin{aligned} \min_{(v_a^d)} \sum_d & \left[\sum_{a \in A} t_a(v) v_a^d + \sum_{i \in V} \max_{a \in A_i^+} \frac{v_a^d}{f_a(v)} - \sum_{i \in V} g_i^d \tau_i^d(v) \right] \\ \text{s.t. } & g_i^d + \sum_{A_i^-} v_a^d = \sum_{A_i^+} v_a^d \end{aligned}$$

Applications

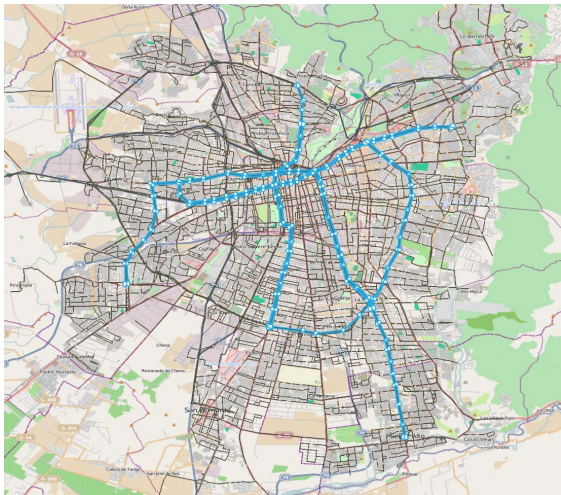
Model implemented as a macro within EMME software (INRO).

Has been used in

- 2005: Sao Paulo
- 2010: San Francisco, Bangkok
- 2011: Sidney, Mexico City, Los Angeles,
- 2012: Santiago
- 2013: Brisbane, Rio de Janeiro
- 2014: Stockholm
- 2016: Toronto

Source: Michael Florian (INRO)

Application: Santiago (developed by SECTRA, 2012)



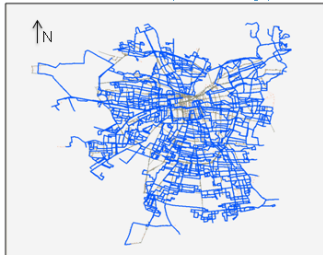


Emme Model – Model Construction

Trunk Routes (2011 coverage)



Feeder Routes (2011 coverage)



Regular nodes: 20,093
 Stops and stations: 12,500
 Centroids: 2,916
 Regular links: 60,528
 Connectors: 24,714

(software size and license)

Metro Lines (2011 coverage)



Trunk routes: 390 service-direction / 3,700 vehicles / 7,800 km

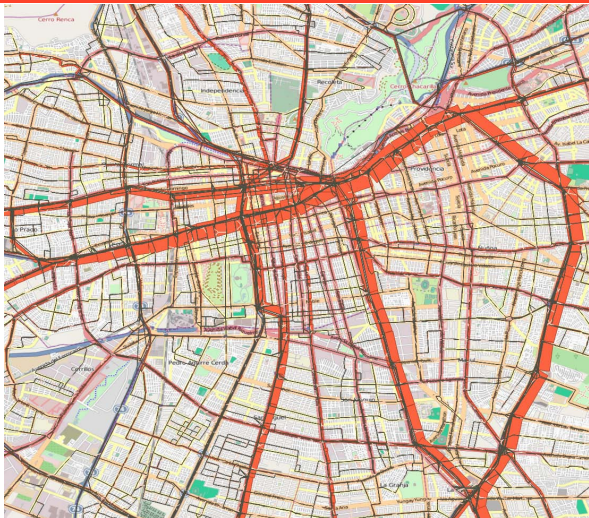
Feeder routes: 430 service-direction / 2,200 vehicles / 5,100 km

Source: SECTRA

Metro: 150 vehicles (trains) / 5 lines / 200 km

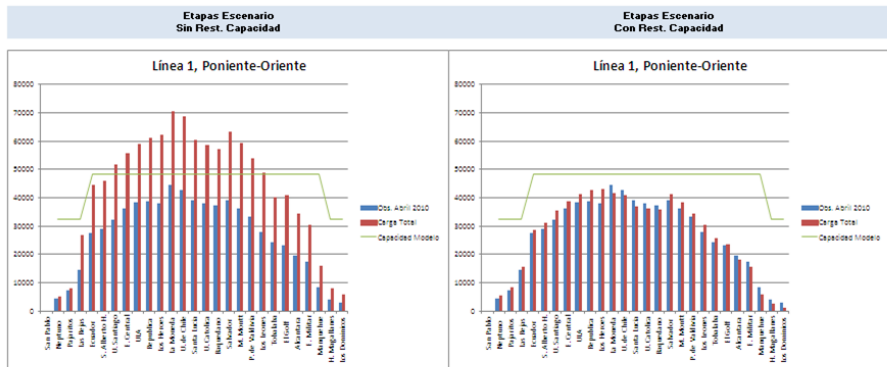


Flows on all transit modes





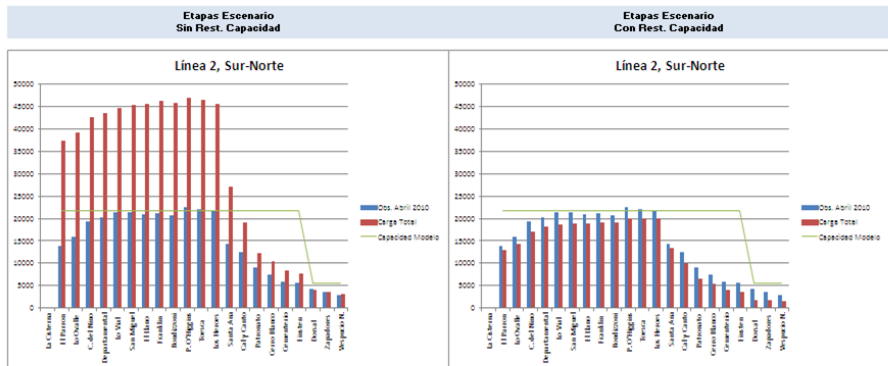
Emme Model – Results with Capacity Considerations



Source: SECTRA



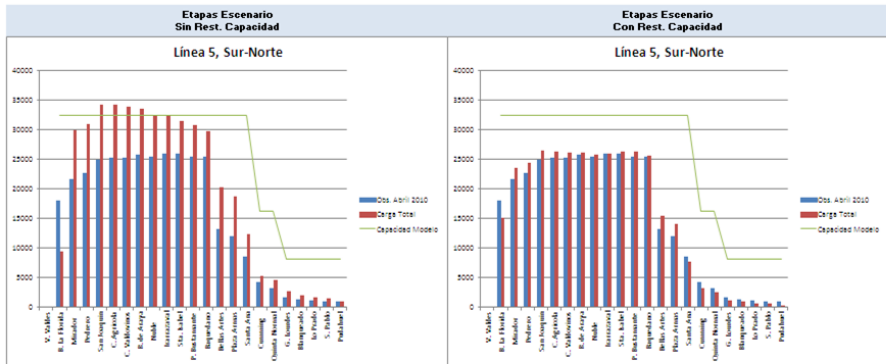
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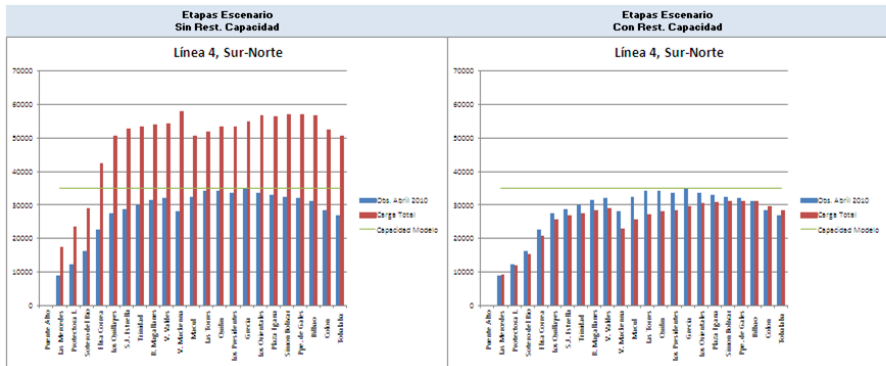
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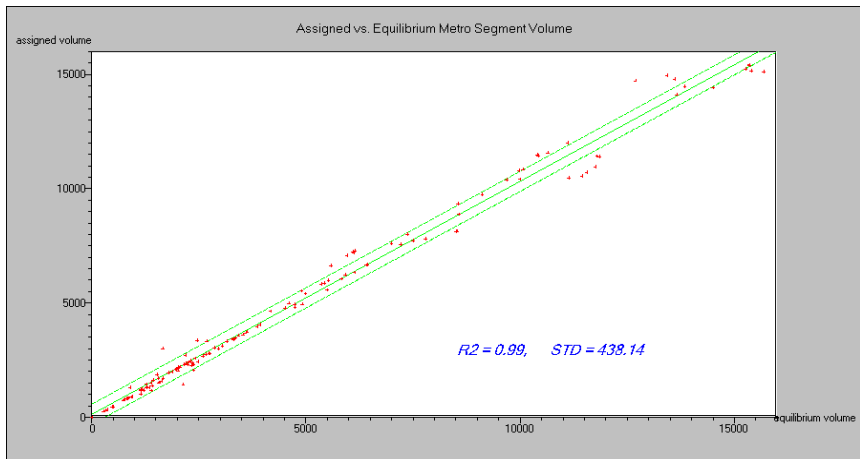


Emme Model – Results with Capacity Considerations



Source: SECTRA

Assigned vs. metro counts segment volume



Emma Freijinger, November 2013

③ Research opportunities

- Adaptive dynamics
- Dynamic equilibrium
- TCP/IP multipath routing

Adaptive dynamics

- Are drivers fully rational? Do they have full information?
- Do myopic adaptive dynamics support equilibrium?
- Recent results by C-Melo-Sorin'2010 and Bravo'2011 provide partial answers and lead to other notions of equilibria

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- Do myopic adaptive dynamics support equilibrium?
- Recent results by C-Melo-Sorin'2010 and Bravo'2011 provide partial answers and lead to other notions of equilibria
- Many open questions remain !
 - almost sure convergence under small noise
 - speeds of convergence of stochastic adaptive dynamics
 - multiplicity and bifurcations of equilibria
 - large population asymptotics
 - more realistic adaptive dynamics
 - robustness under model specification

Dynamic equilibrium

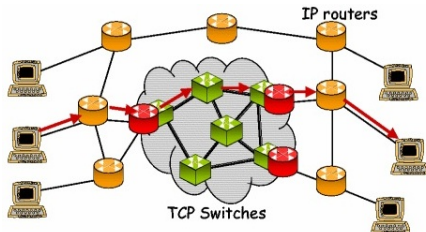
- Traffic equilibrium under automated guidance software tools
- Dynamic equilibrium more appropriate than static equilibrium
- Recent progress for fluid queuing networks:
 - *“derivatives of equilibria \equiv normalized thin flows with resetting”*
 - characterization & computation of equilibria: Koch-Skutella'2010
 - characterization & existence of equilibria: C-Correa-Larré'2015
 - long-term behavior of equilibria: C-Correa-Olver'2017

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- Many open questions remain !
 - finite convergence of reconstruction algorithm
 - efficient computation of NTFR's and equilibria
 - multiple origin-destination networks
 - other link dynamics: LWR, spillbacks

TCP/IP multipath routing

- Each source $s \in S$ transmits packets from o_s to d_s
- TCP = At which rate? / IP = Along which route?
- Random delays $\tilde{t}_a = \text{queuing} + \text{transmission} + \text{propagation}$
- Finite queuing buffers \Rightarrow packet loss probabilities p_a



- Congestion control – TCP Reno/Tahoe/Vegas
Sources adjust transmission rates in response to congestion.
Feedback mechanism: *higher congestion* \Leftrightarrow *smaller rates*.

Congestion control and multipath routing

- Kelly *et al.* 1998: Steady states of TCP protocols are
 - Equilibria for an associated potential game
 - TCP reverse-engineered as a decentralized asynchronous algorithm that solves a network optimization problem

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- Many open questions remain !
 - Convergence analysis accounting for stochastics & delays
 - Increase transmission rates: multi-path routing
 - Combine NUM with MTE (C-Guzmán'2014)
 - Design stable packet-level protocols

Thanks!

Reprints available at

<https://sites.google.com/site/cominettiroberto/>