



Revenue Protection – National and International Research Perspectives

TfL Meeting 19 January 2065

Professor Graham Currie Director

Public Transport Research Group (PTRG) Institute of Transport Studies Monash University, Australia



Outline

- 1 Introduction
- 2 Fare Evasion Psychology Melbourne
- 3 Fare Evasion Psychology International
- 4 Research Developments





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This paper reviews recent research on revenue protection & fare evasion psychology



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Psychology of Fare Evasion (Melbourne) - AIMS

- Overall project objective:
 - to understand the psychology behind fare evasion and provide actionable recommendations for use in improving compliance.
- Aims
 - 1.To understand what motivates people to fare evade
 - What is the prevalence and distribution of unintentional, opportunistic and purposeful fare evasion?
 - 2. To develop an empirical model that will suggest strategies to reduce fare evasion



Four 'rationales' for Fare Evasion were found...

			Fare Evasion	Rationales	
	Perspective	1. Its wrong – the accidental evader	2. The 'it's not my fault' evader	3. The calculated risk-taker evader	4. Career evaders
Strong view that Fare Evasion Is	Occurrence	Rare	Occasional	Fairly Often	Always
about INTENT. Feeling of INJUSTICE about being caught if you intended to buy a ticket – feel "THE SYSTEM IS WRONG" if this happens	Intentions	No Intention – Evasion by Accident	No Intention – Evasion due to payment barriers	Intention – Evasion due to low risk	Entirely Intentional
	Feelings	Guilt/ Embarrassm ent	Nervous, worried but no guilt	Dispassionate, vigilant, no guilt	Pride
	View of Fare Evaders	Condemnati on	Empathy - sense of injustice to condemnati on	Understanding to condemnation	Empathy

Source: Monash User Focus Groups and Discussion Groups





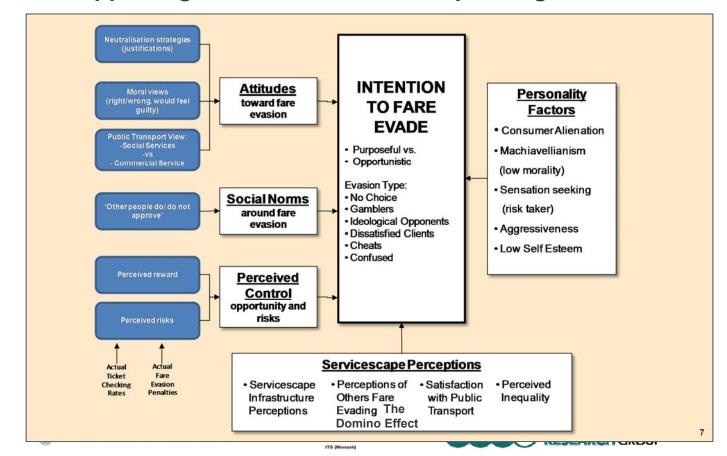


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... supporting a theoretical model explaining FE choice



Key Finding: most fare loss is a few frequent users..

Table 5.3: Estimated Volume of Trips Made by Fare Evasion Frequency and Public Transport Trip Frequency Groups

		Estimated Fare Evasion Trips Made by People in Each Evasion Frequency Group (M p.a.)				n Each		-		
Estimated Sha	are of Trips	6-7 days a week	5 days a week	3-4 days a week	1-2 days a week	> monthly	Less often	Total Trips (M)	Share of Total Travel	Share of Evasion Trips
Always	100.0%	1.2	2.9	-	-	-	0.0	4.1	0.8%	16%
Almost Always Mostly	95.0% 75.0%	1.1 0.9	4.6 3.7	- 2.7	- 0.6	0.0 0.1	0.0 0.0	5.8 7.9	1.1% 1.5%	22% 30%
Regularly	37.5%	0.4	0.7	0.8	0.3	0.1	0.0	2.3	0.4%	9%
Occasionally	12.5%	0.1	2.8	1.3	0.4	0.1	0.0	4.8	0.9%	18%
Rarely	1.0%	0.0	0.6	0.4	0.2	0.0	0.0	1.2	0.2%	5%
Never	0.0%	-	-	-	-	-	-	0	0.0%	
Sub-Total: Fa	are Evasion ips (M p.a.)	3.8	15.4	5.2	1.4	0.4	0.1	26.2	5.1%	100%
Share of To	tal Evasion	14.3%	58.7%	19.9%	5.4%	1.4%	0.3%			
Sub-Total: Fa	are Evasion ips (M p.a.)		15.4			0.4				

<u>Recidivists</u>

- 68% of all FE trips
- 65,400 people
- 81% high frequency PT
 users

High Frequency Users who Fare Evade

- 73% of all FE trips
- 285,900 people
- 75% Recidivists
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All Fare Evaders

- 822,200 people (20.6% of Melbourne population)
- 71% (580,000 people) a one off occurrence never
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..."recidivists" contrast with accidental evaders

Measure	Fare Evader Type				
	Recidivists	Meant to pay,	Deliberate	Unintentional	
		accident, one off			
Share of people fare	8%	70%	41.0%	44.0%	
evading at least					
once p.a.					
Share of revenue	68%	5%	77.4%	15.5%	
lost/fare evasion					
trips					
Estimated Value of	\$54M	\$4M	\$47.8M	\$9.6M	
Revenue Lost p.a.					
Number of People	65,400	580,000	702,240	1,388,520	
Share of Melbourne	1.6%	14.5%	17.6%	34.8%	
population					
Lost Revenue per	\$826	\$6.90	\$68.00	\$6.90	
person p.a.					

Contrasting Fare Evader Metrics

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3 valid FE clusters were identified

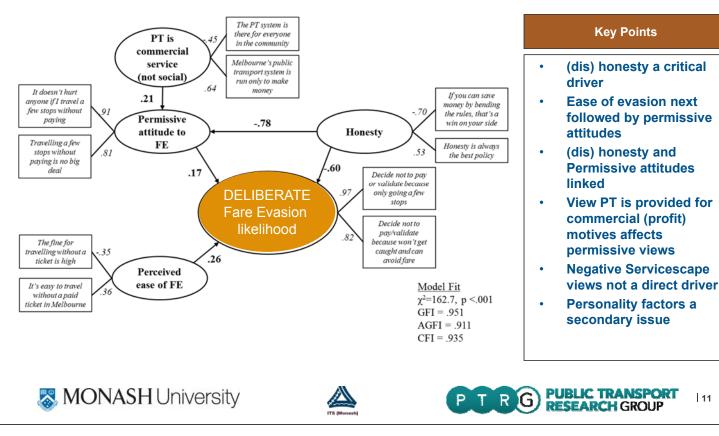
Deliberate Evaders	Unintentional Evaders	Never Evaders
17.6% of market	34.8% of market	47.6% of market
 Most likely to repeat FE and intend to FE in future High frequency PT user, full-time worker or student, age 17-34 Lower self esteem, higher sensation seeking, less honest More influenced by the 'domino effect' Most likely to have been caught for FE (8%p.a.) Have a poorer opinion of PT Think PT is run for commercial profit 	 One-off FE and low future intent Range of PT use (frequent to infrequent) Range of demographics (no standout features) Higher self esteem, lower sensation seeking, more honest Strongest worry about being caught (5% caught in last year) Stronger view that PT is for social benefit not commercial 	 Almost no FE and very low future intent Lower frequency PT users Range of demographics but higher older and retired Highest self esteem, lowest sensation seeking, highest honesty rating, stronger social beliefs Stronger view that PT is for social benefit not commercial
Biggest revenue loss	Very little revenue loss	Almost no revenue loss



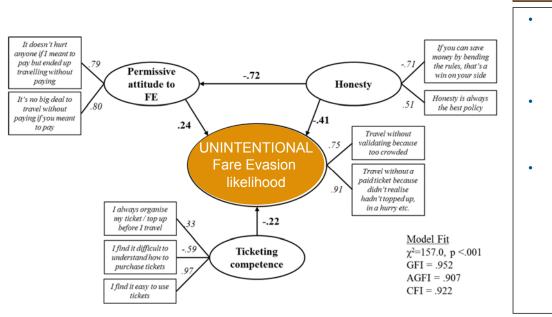




Deliberate FE is driven by (dis)honesty, (weak) perceived control and permissive views



Accidental FE is driven by (dis)honesty permissive views and (poor) ticketing competence



Key Points

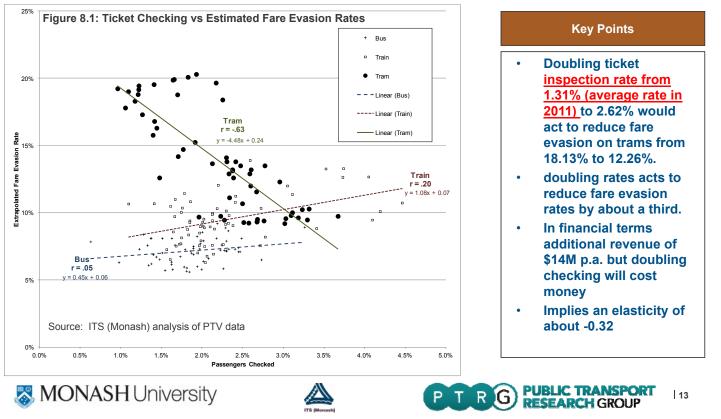
- (dis) honesty a main driver followed by permissive attitudes then ticketing competence
- Ease of evasion is not an issue since evasion is accidental/ unintended
- Ticketing competence a valuable concept in understanding accidental fare evasion

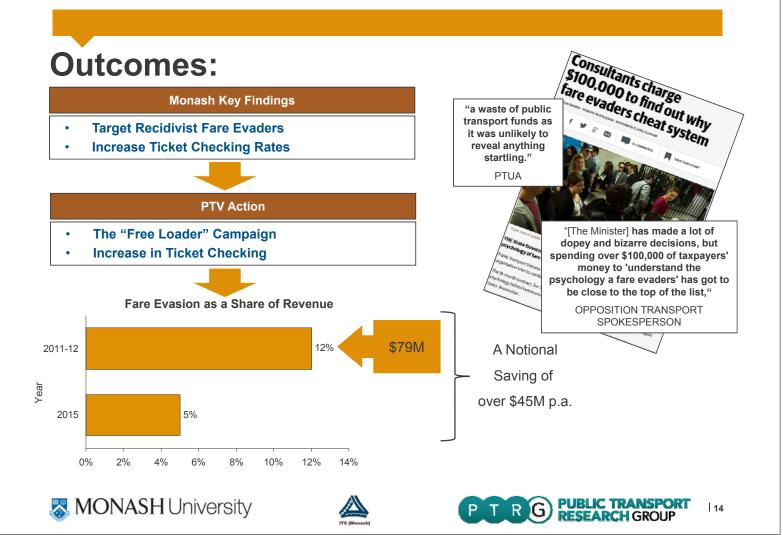






Key Finding: FE Sensitivity Analysis suggests ticket check rates can reduce tram FE....





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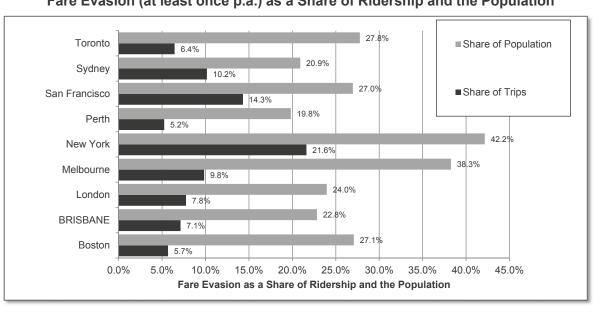
Cross National follow-on study - AIMS

- Overall project objective:
 - Cross national study of 9 international cities including Melbourne, London, Sydney and Perth
- Aims
 - Implement web survey method for fare evasion metrics on a sample on international cities (including London) to estimate broad levels of:
 - Fare evasion (trip share, population share)
 - Recidivism rates
- Approach
 - 200 randomised PT users living in target cities





OZ Cities; FE rate of PT trips 5-10%; share of residents have FE'd in last year 20-38%



Fare Evasion (at least once p.a.) as a Share of Ridership and the Population

Source: Monash Cross National Study

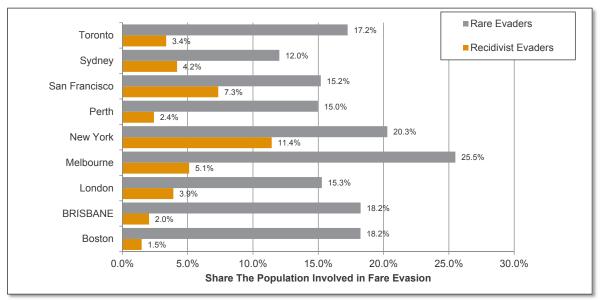
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OZ Cities; share of pop who are recidivists; 2-5% - rare **FEdrs 12-26%**

Share of the Population Engaged in Fare Evasion (at least once p.a.)

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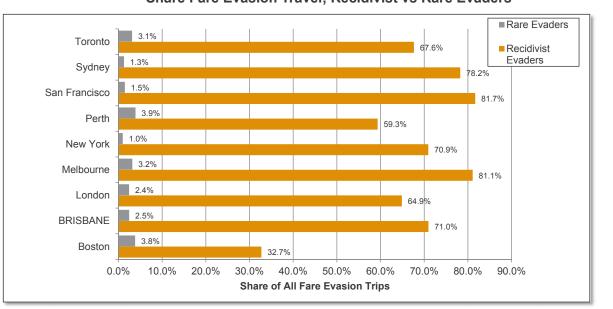
Source: Monash Cross National Study





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OZ Cities – Share FE trips/ revenue lost due to recidivists 59-81% - <u>RECIDIVISM IS A GLOBAL PROBLEM</u>



Share Fare Evasion Travel; Recidivist vs Rare Evaders

Source: Monash Cross National Study

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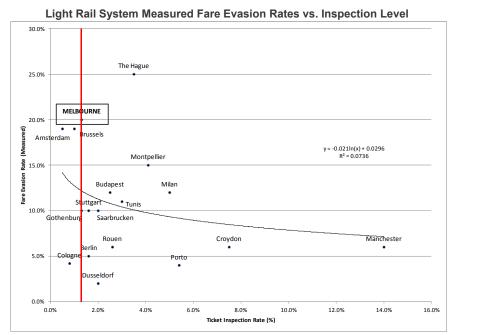
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Melbourne's tram proof of payment ticket inspection rate (1.3%) was low compared to other cities



Source: <u>ITS (Monash) analysis of Dauby and Kovacs 2006 data and Melbourne data from Tables 3.1 and 3.2</u> Note: Mid range of data points used where a range is shown

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Barabino et al (2013)

Context & Data

- Fare evasion on buses in Sardinia, Italy
- 98 days of ticket checks
- 3,659 on-board interviews

Barabino et al (2014)

Context & Data

- Fare evasion on buses in Sardinia, Italy
- 3 years of ticket checks (total of 27,514 checks)
- 10,586 on-board
- interviews

Approach

- Economic model (focus on profit maximisation)
- Costs of fare evasion control (inspectors, administration)
- Increase revenue yield from lower fare evasion

Optimal Inspection Rate

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Approach

- Profit maximisation model
- Costs of fare evasion control
- (inspectors, administration) Increase revenue yield from
- lower fare evasion

Optimal	Inspection	Rate

4.5%







Fare evader profiles, again from Italy, profile young, unemployed males, and those taking short trips

	Key determinants of fare evaders in Italy
•	Male
•	Less than 26 years old
•	Low education level
•	Unemployed and/or students without other means of transport
•	Those undertaking trips less than 15 mins
•	Systematic users not satisfied with the service
•	Passengers on routes with low inspection rates
•	Passengers with fines and previous ticket violations

Source: Barabino, B., Salis, S. & Useli, B. (2015) 'What are the determinants in making people free riders in proof-of-payment transit systems? Evidence from Italy'. *Transportation Research Part A*, Vol. 80, pp. 184-196







Santiago, Chile model FE influences; key are proximity to intermodal stations, ticket inspections & time of day

Modelling of Factors Linked to Higher Fare Evasion Rates

Variables affecting fare evasion	% change in fare evasion rate
Proximity to intermodal station	-89.8%
Ticket inspections	-45.8%
Morning weekday	-29.6% DECREAS
Area with high income level (>US\$1,674)	-28.9% fare evasi
Proximity to metro station	-16.4%
Area with moderate income level (US\$1,065-1,674)	-14.2%
Bus occupancy	+0.8%
Number of passengers alighting	+1.8%
Number of bus doors	+5.9% fare evasi
Afternoon weekday	+19.6%

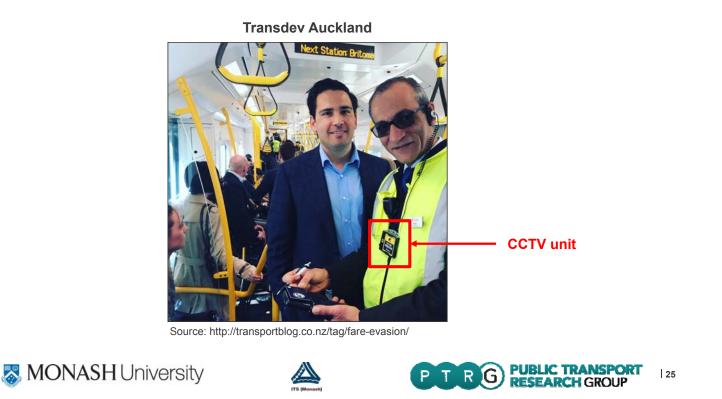
Source: Guarda, P., Ortuzar, J., Galilea, P., Handy, S. & Munoz, J. (2015) 'Decreasing fare evasion without fines? A microeconomic analysis'. Presented at Thredbo 14 Conference, Santiago, Chile.



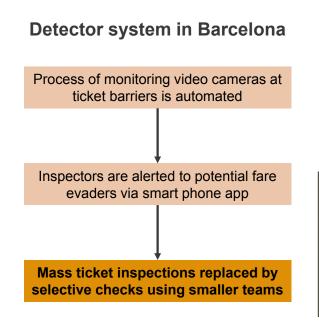




Emerging technologies: range from ticket inspectors fitted with CCTV on their jackets...



...to sophisticated camera technology at ticket barriers...







Source: http://www.railway-technology.com/





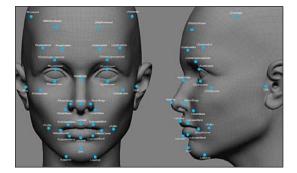


...and even facial recognition (biometric technology), although applications are yet to be seen in this area

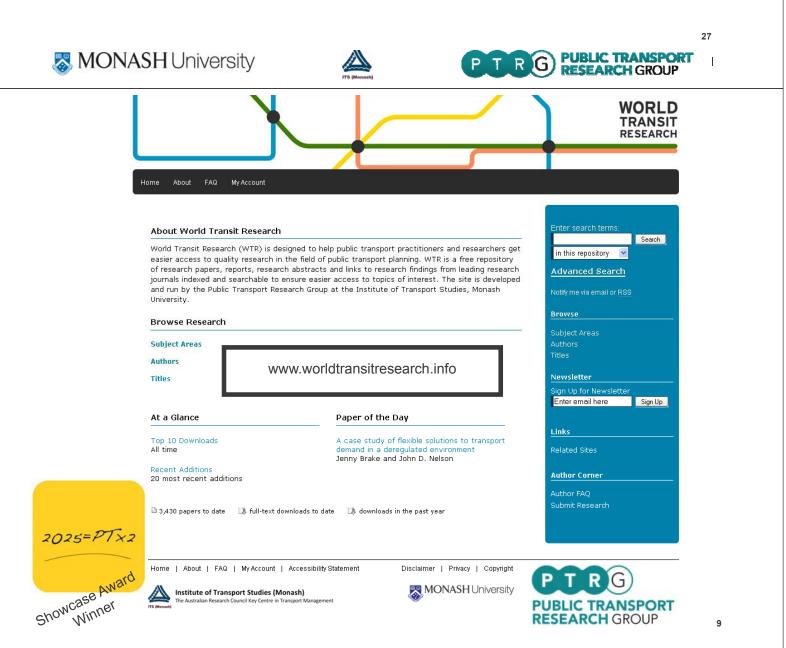
UITP Survey Results (2015)

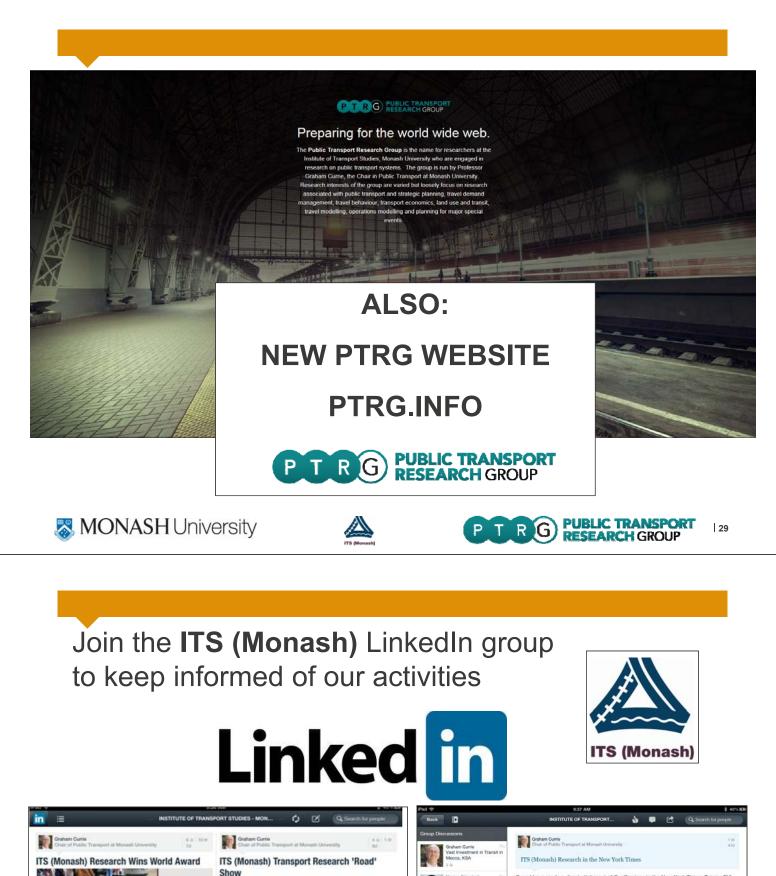
- 74 public transport organisations in 30 countries
- <u>None</u> have used facial recognition technology yet
- Half (50%) are interested in using facial recognition technology in the future

Source: UITP (2015) Video Surveillance in Public Transport: International Trends 2015-16, Full Report









Guys I have made a direct attchment of Geoffs piece in the New York Times; Team - FYI linking a newspaper articule directly improves its look on LinkedIn but can only be done on first post of the discussion, not commenting on an existing one.

ITS (Monash) Research in the New York Times

m · AUSTIN, TEXAS - Kelly Hunka tried out an electric bicycle for the first time last week Hours later, she was back for another ride. "These are so cool," Ms. Hunka, who works for an Austin technology company, said as she paced.

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nt in Transit in Meyea, KSA

Saudi Arabia approves \$16.5

bin Mecca transport revamp

O mutara - JEDDAH, Saudi Arabia | Tue Aug 14, 2012 6:17am EDT JEDDAH, Saudi Arabia

(Reuters) - Saudi Arabia has approved a 62 billion rival plan to modernise the transport system in its holy city of Mecca, including

building a bus network and metro system

A Londel Schwitt PhD Candidate at Ma

110A) (map available at

http://www.monash.edu.au/pubs

TRW Session with Christain

Monash University's Clayton campus on

Bode - Case Studies on Bus...

We will be hosting our next TRW Session at

Monday 27 August at 11:00em. Details below Details: Where: MU Clayton Campus - Civil

Engineering Seminar Room (Building 60, Room

Claytoncolour.pdf) When: Monday 27 August

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Mirajan Shiwakoti Lacharer pr Manael

In a forth

Interested in learning how

animal models can be used to study pedestrian crowd panic?

published in Transport Geography Journal entitled "Understanding pedestrian crowd

panic: a review on model organisms

approach", we will be reviewing our eriences of using animal mo

ming paper from ITS (to be

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