

Ridesourcing for the first/last mile: How do transfer penalties impact travel time savings?

Daniel J. Reck^a
daniel.reck@ivt.baug.ethz.ch
Corresponding author

Kay W. Axhausen^a
axhausen@ivt.baug.ethz.ch

^a Institute for Transport Planning and Systems (IVT), ETH Zürich, Stefano-Franscini-Platz 5, 8093 Zürich, Switzerland

Keywords: first/last mile, transfer penalty, ridesourcing

Problem statement

The first and last mile of public transportation (PT) trips are a long known problem to planners: low and dispersed spacio-temporal demand is expensive to serve with large-capacity vehicles, yet they deter many potential passengers from using PT. Demand-responsive feeders have been suggested as a remedy (see Chandra and Quadrioglio, 2013, for an overview) in three phases:

In the 20th century ('phase 1'), demand-responsive transportation generally faced technological constraints (manual routing, scheduling and dispatching, corresponding high labor costs, long lead times), resulting in low levels of ridership and/or high expenditures (Mageean and Nelson, 2003; Davison et al., 2014).

The dissemination of GPS-enabled smartphones, advances in routing algorithms and computing power, and regulatory voids have enabled new (cost-)efficiencies in demand-responsive transportation and led to the popularity of ridesourcing companies such as Uber or Lyft ('phase 2'). Their use as first/last mile feeders has often been suggested (e.g., Feigon and Murphy, 2016; Westervelt et al., 2017; Shaheen and Chan, 2018) and many US transit agencies have engaged in partnerships to subsidize first/last mile rides (e.g., Charlotte, Austin, Centennial, Pinellas County) or are planning to do so (e.g., Los Angeles, Chicago). Ridership, however, has so-far been low and operations of ridesourcing companies remain deficient.

Perhaps most importantly, the first and last mile is seen as one area of application where automated taxis could complement PT ('phase 3') (Chong et al., 2011; Liang et al., 2016; Cervero, 2017; Moorthy et al., 2017; Shen et al., 2018). While profitable operations can be expected (Loeb and Kockelman, 2017; Boesch et al., 2018), it is unclear whether ridership on the first/last mile will finally meet expectations or whether a conceptual barrier to demand-responsive feeders for the first/last mile persists.

Literature review

So-far, mostly *operational* explanations for low ridership of first/last mile ridesourcing services have been identified (e.g., sparse marketing, short pilot duration, small pilot area, high costs) (City of Centennial, 2017; PSTA, 2018).

Despite a long history of research into transfers and associated disutilities ('transfer penalty') (Algers et al., 1975; Alter, 1976; Allen and DiCesare, 1976; Newell, 1979; Horowitz, 1981), the additional transfers caused by first/last mile demand-responsive feeders have not been considered as a *conceptual barrier* to their use. Yet, this seems important as

55 passengers prefer to avoid additional transfers due to factors such as anxiety to reach the
56 subsequent connection, security, activity disruption and comfort (Currie, 2005; Iseki and
57 Taylor, 2009; Cheng, 2010).

58 Studies investigating the general size of the transfer penalty exhibit wide value
59 ranges. Currie (2005) provides a review finding an average transfer penalty for bus-bus
60 transfers of 22 min of in-vehicle travel time (ranging between 5 and 50 minutes). Reasons
61 for these wide ranges are context-sensitivity (e.g., climate, security, local amenities, type of
62 vehicle) (Iseki and Taylor, 2010; Guo and Wilson, 2011) and measurement scope (e.g.,
63 waiting time, walking time to the subsequent vehicle, and/or the disutility of the transfer itself)
64 (Garcia-Martinez *et al.*, 2018). In a recent effort to improve comparability, Garcia-Martinez *et al.*
65 (2018) investigate the 'pure transfer penalty' (i.e., without walking or waiting times). Using
66 SP data in Madrid, they find the pure transfer penalty to average 15.2 min.

67 Yan *et al.* (2018) are the first to consider a transfer penalty in their survey-based
68 investigation of traveler responses to a potential first/last mile ridesourcing service on the
69 University of Michigan Ann Arbor campus. Despite finding a transfer penalty of 10.9 min in-
70 vehicle travel time, they conclude: "when used to provide convenient last-mile connections,
71 ridesourcing could provide a significant boost to transit". (p. 1)

72

73 **Research objectives**

74

75 Complementing popular *operational* explanations, we argue that the additional transfer and
76 associated penalty provide a *conceptual* explanation for low ridership of current first/last mile
77 ridesourcing services as well as future first/last mile usage of automated taxis. In this study,
78 we aim to quantify the relative impact of transfer penalties on the total time travel time
79 savings using first/last mile demand-responsive feeders empirically.

80

81 **Methodological approach**

82

83 As a case study, we chose Pinellas County, Florida, which is home to the longest operating
84 first/last mile ridesourcing partnership ('PSTA Direct Connect'). We obtain block-group level
85 origin-destination commuting trip information from the 2015 US Census Origin-Destination
86 Employment Statistics (99 470 observations). For each, we construct PT travel times
87 including access/egress walking times and intermediate wait times using the Google
88 Directions API (Alternative A). We then obtain the coordinates of the first and last PT station
89 used and, using the Google Directions API, construct first/last mile car trips from the origin to
90 the first PT station used, and from the last PT station to the destination (Alternative B). We

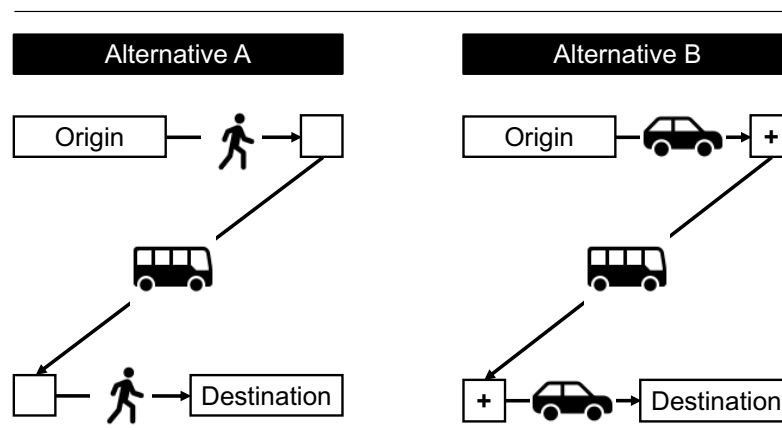


Fig. 1: Alternatives without (A) and with (B) first/last mile DRF, for which travel times are being compared. Transfer penalties are added to Alternative B.

91 then compare weighed travel times for A and B adding a transfer penalty between 5 and 15
92 minutes for the first/last mile transfer (Figure 1).

93

94 Results

95

96 We find that a first/last mile service leads to average travel time savings of 15.7 minutes.
97 However, transfer penalties of 5, 10 and 15 minutes diminish travel time savings by 54%,
98 82% and 95%, respectively (Figure 2). Thus, even at small values the transfer penalty
99 presents an important conceptual barrier to first/last mile demand-responsive feeders.

100

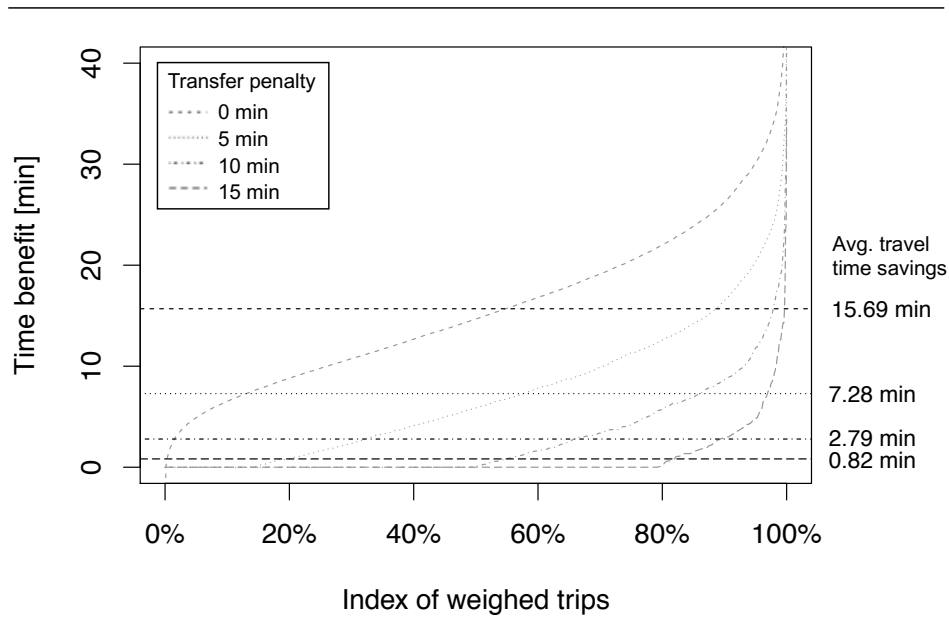


Fig. 2: Travel time savings for first/last mile trips after applying transfer penalties.

101

102 Discussion

103

104 Our results not only help to explain the low ridership of current first/last mile ridesourcing
105 services, they also help to explain why a significant and substantive positive relationship
106 between ridesourcing and public transit ridership for urban areas has not been found yet.
107 Furthermore, they conceptually question the usefulness of demand-responsive feeders on
108 the first/last mile, including automated taxis.

109

110 Future work investigating ridesourcing or automated taxis as potential first/last mile
111 solutions similar to Moorthy *et al.* (2017) and Shen *et al.* (2018) might come to a different
112 conclusion once considering transfer penalties. Taking into account a distribution of transfer
113 penalties, however, might be more accurate to reproduce real-world preferences than our
114 simplistic, yet illustrative approach of considering averages. As values are highly context-
115 dependent, it seems important to study local factors such as the built environment, safety
116 and weather conditions carefully to make meaningful assumptions.

117

118 Our results finally suggest the following policy implication. Vehicle-based first/last
119 mile services in general (including automated taxis) appear to decrease perceived travel
120 times (including the transfer penalty) only in areas with particularly long ingress/egress
121 distances. Even in suburban Pinellas County with an average population density of
122 1368/km² and an average first/last mile of 900m, distances seem too close for a first/last
123 mile demand-responsive feeder to improve perceived travel times substantially. Thus, in
124 contrast to current studies, first/last mile services appear more relevant in less urbanized /
rural areas or for connections to (sub)urban high-speed PT such as rail or BRT.

124

125 **References**

126
127
128
129
130
131
132
133
134
135
136
137
138
139
140
141
142
143
144
145
146
147
148
149
150
151
152
153
154
155
156
157
158
159
160
161
162
163
164
165
166
167
168
169
170
171
172
173
174
175
176
177
178

Algers, S., S. Hansen and G. Tegner (1975) Role of waiting time, comfort, and convenience in modal choice for work trip, *Transportation Research Record*, **534**, 38-51.

Allen, W. G. and F. DiCesare (1976) Transit Service Evaluation: Preliminary Identification of Variables Characterizing Level of Service, *Transportation Research Record*, **606**, 41-47.

Alter, C. H. (1976) Evaluation of Public Transit Services: The Level-of- Service Concept, *Transportation Research Record*, **606**, 37-40.

Boesch, P. M., F. Becker, H. Becker and K. W. Axhausen (2018) Cost-based analysis of autonomous mobility services, *Transport Policy*, **64**, 76-91.

Cervero, R. (2017) Mobility Niches: Jitneys to Robo-Taxis, *Journal of the American Planning Association*, **83** (4) 404-412.

Chandra, S. and L. Quadrioglio (2013) A model for estimating the optimal cycle length of demand responsive feeder transit services, *Transportation Research Part B: Methodological*, **51**, 1-16.

Cheng, Y. H. (2010) Exploring passenger anxiety associated with train travel, *Transportation* **37** (6) 875–896.

Chong, Z. J., B. Qin, T. Bandyopadhyay, T. Wongpiromsarn, E. S. Rankin, M. H. Ang Jr., E. Frazzoli, D. Rus, D. Hsu and K. H. Low (2011) Autonomous personal vehicle for the first- and last-mile transportation services, *Proceedings of the 2011 IEEE 5th International Conference on Cybernetics and Intelligent Systems*, 253-260.

City of Centennial (2017) goCentennial: Final Report, Centennial, CO, 2017, June. Available online: https://www.centennialco.gov/uploads/files/Government/Iteam/Go%20Centennial%20Final%20Report_for%20web.pdf (accessed 25 January 2019).

Currie, G. (2005) The demand performance of bus rapid transit, *Journal of Public Transport*, **8**, 41–55.

Davison, L., M. Enoch, T. Ryley, M. Quddus, and C. Wang (2014) A survey of demand responsive transport in Great Britain, *Transport Policy*, **31**, 47-54.

Feigon, S. and C. Murphy (2016) Shared mobility and the transformation of public transit, *TCRP Research Report*, **188**.

Garcia-Martinez, A., R. Cascajo, S. R. Jara-Diaz, S. Chowdhury and A. Monzon (2018) Transfer penalties in multimodal public transport networks, *Transportation Research Part A: Policy and Practice*, **114**, 52-66.

Guo, Z. and N. H. Wilson (2011) Assessing the cost of transfer inconvenience in public transport systems: A case study of the London Underground, *Transportation Research Part A: Policy and Practice*, **45** (2) 91-104.

Horowitz, A. J. (1981) Subjective value of time in bus transit level, *Transportation*, **10** (2) 149-164.

179 Iseki, H. and B. D. Taylor (2009) Not all transfers are created equal: towards a framework
180 relating transfer connectivity to travel behaviour, *Transport Reviews*, **29** (6) 777–800.
181

182 Liang, X., G. H. de Almeida Correia and B. Van Arem (2016) Optimizing the service area
183 and trip selection of an electric automated taxi system used for the last mile of train trips,
184 *Transportation Research Part E: Logistics and Transportation Review*, **93**, 115-129.
185

186 Loeb, B. and K. M. Kockelman (2017) Fleet Performance & Cost Evaluation of a Shared
187 Autonomous Electric Vehicle (SAEV) Fleet: A Case Study for Austin, Texas, Under review
188 for publication in *Transportation Research Part A – Policy and Practice*. Available online:
189 http://www.caee.utexas.edu/prof/Kockelman/public_html/TRB18SAEVFinancialAnalysis.pdf
190 (accessed 25 January 2019).
191

192 Mageean, J. and J. D. Nelson (2003) The evaluation of demand responsive transport
193 services in Europe, *Journal of Transport Geography*, **11** (4) 255-270.
194

195 Moorthy, A., R. De Kleine, G. Keoleian, J. Good and G. Lewis (2017) Shared Autonomous
196 Vehicles as a Sustainable Solution to the Last Mile Problem: A Case Study of Ann Arbor-
197 Detroit Area, *SAE International Journal of Passenger Cars - Electronic and Electrical*
198 *Systems*, **10**, 328-336.
199

200 Newell, G. F. (1979) Some issues relating to the optimal design of bus routes,
201 *Transportation Science*, **13** (1) 20-35.
202

203 PSTA (2018) Several interviews of the corresponding authors with PSTA staff. Transcripts
204 available upon request.
205

206 Shaheen, S. and N. Chan (2016) Mobility and the sharing economy: Potential to facilitate the
207 first-and last-mile public transit connections, *Built Environment*, **42** (4) 573-588.
208

209 Shen, Y., H. Zhang and J. Zhao (2018) Integrating shared autonomous vehicle in public
210 transportation system: A supply-side simulation of the first-mile service in Singapore,
211 *Transportation Research Part A: Policy and Practice*, **113**, 125-136.
212

213 Westervelt, M., J. Schank and E. Huang (2017) Partnerships with Technology-Enabled
214 Mobility Companies: Lessons Learned, *Transportation Research Record*, **2649**, 106-112.
215

216 Yan, X., J. Levine and X. Zhao (2018) Integrating ridesourcing services with public transit:
217 An evaluation of traveler responses combining revealed and stated preference data,
218 *Transportation Research Part C: Emerging Technologies*, in Press.