



A comparative study of contact frequencies among social network members in five countries

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Transportation Research Record

A Comparative Study of Contact Frequencies Among Social Network Members in Five Countries

--Manuscript Draft--

Full Title:	A Comparative Study of Contact Frequencies Among Social Network Members in Five Countries
Abstract:	<p>As face-to-face and ICT-mediated social interaction patterns are relevant to explain (social) travel behavior, the objective of this paper is to study comparatively the factors that influence social interaction frequency via different communication modes. The analysis is based on seven recent data collections on personal social networks from Canada, Chile, Switzerland, the Netherlands, and Japan. A multilevel-multivariate mixed model that explicitly accounts for the hierarchical nature of the data is used to jointly analyze contact frequency patterns across all samples. We show the existence of very consistent associations across samples between individual and relational characteristics and social interactions such as age, network size, distance and emotional closeness. At the same time, for other characteristics such as gender and relationship type, among others, effect patterns were less clear, differences that might be explained by intrinsic contextual characteristics as well as methodological differences among studies.</p>
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1 **A Comparative Study of Contact Frequencies Among Social Network Members in Five**
2 **Countries**

3

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1 **Abstract**

2 As face-to-face and ICT-mediated social interaction patterns are relevant to explain (social)
3 travel behavior, the objective of this paper is to study comparatively the factors that influence
4 social interaction frequency via different communication modes. The analysis is based on
5 seven recent data collections on personal social networks from Canada, Chile, Switzerland,
6 the Netherlands, and Japan. A multilevel-multivariate mixed model that explicitly accounts
7 for the hierarchical nature of the data is used to jointly analyze contact frequency patterns
8 across all samples. We show the existence of very consistent associations across samples
9 between individual and relational characteristics and social interactions such as age, network
10 size, distance and emotional closeness. At the same time, for other characteristics such as
11 gender and relationship type, among others, effect patterns were less clear, differences that
12 might be explained by intrinsic contextual characteristics as well as methodological
13 differences among studies.

14

15 **Keywords:** Social networks, personal networks, social interaction, communication
16 frequency, social activity travel behavior.

17

18 **1. INTRODUCTION**

19 Social activities are responsible for a substantial share of trips (1). These social trips are
20 influenced by new ways of social interaction made possible by the rise of new information
21 and communication technologies (ICT). From a transportation perspective, the study of both
22 face-to-face and ICT-mediated interactions is of importance since these changing
23 communication patterns might affect demand dynamics on urban environments and
24 transportation systems. To study this, data collection efforts have been made in Toronto,
25 Canada (2), in Zurich (3, 4), and Switzerland nationwide (5), in Eindhoven, the Netherlands
26 (6), in Concepción, Chile (7) and in the Greater Tokyo Area, Japan (8), among others. In all
27 seven studies, the personal networks approach has been used to capture information on
28 individuals (egos) and their social network members (alters). Analyses from each of these
29 studies have indicated that incorporating social network characteristics is crucial in studying
30 social activity-travel behavior. However, there might be substantial differences between
31 countries and cultures in terms of communication patterns, social activity and travel behavior.
32 The aim of this paper, is therefore, to study comparatively the factors that influence social
33 interaction frequency among social network members via different communication modes in
34 five different countries on three continents.

35 The rest of the paper is structured as follows. The next section discusses the
36 theoretical framework. Section 3 summarizes data collection efforts and its respective
37 descriptive statistics. Section 4 describes methods and results of the statistical model used.
38 The final section discusses the key findings of this study and future avenues of research.

39

40 **2. THEORETICAL FRAMEWORK**

41 **Social networks and travel**

42 In general, social networks research in the transportation field can be classified into two
43 streams: (i) the relationship between social influences and activity travel decisions, and (ii)

1 the relationship between social networks and social-activity travel (9). The discussion below
2 will be largely focusing on the second stream of research, as it is the most relevant to the
3 analysis presented in this article.

4 To analyze social networks and social-activity travel, researchers have largely relied
5 on the social networks approach, developed and used for decades in sociology. Particularly
6 relevant to this study is the egocentric approach, in which personal network members (alters)
7 are elicited for the given individuals (egos), through the use of name generators, that is,
8 questions designed to help egos elicit the names of these alters (10). Using this approach,
9 several data collections efforts have been carried out (2–8, 11, 12)

10 To collect more information on each ego-alter relationship, additional questions
11 (name interpreters) are used. These questions may include age and gender of the alter, to
12 assess homophily (the tendency to interact with people similar to you). Other questions
13 include tie strength, duration, and relational roles. Social network data collection in the
14 transportation field also include geographical distance between the homes of egos and alters
15 and contact frequency by different modes (face-to-face, telephone, SMS, email, SNS), as
16 these are important aspects of social activity-travel behavior. An in-depth comparison of the
17 distance patterns between social network members in Toronto, Zurich, Eindhoven,
18 Switzerland, and Concepción is reported in Kowald et al. (13). In addition to adding two new
19 datasets, this article will complement the study by Kowald et al. (13) by analyzing social
20 interaction frequency in different socio-cultural contexts.

21 **Factors influencing communication frequency**

22 To maintain their social networks, people need to communicate and meet each other. This
23 contact can include face-to-face meetings as well as contacts mediated by different ICT tools.
24 Previous studies have shown that for leisure or social activities, the effect of ICT is generally
25 complementary, in that ICT use might generate new travel by stimulating the demand for
26 new activities (14–19).

27 Communication modes and frequencies, however, vary among individuals and
28 relationships. Past research has indicated that contact frequencies are influenced by
29 individual and household characteristics, time constraints, and availability (and costs) of
30 mobility and communication tools.

31 Being young, highly-educated and male has been associated with higher contact
32 frequencies with ICT-mediated modes such as e-mail and SMS (15, 17). Household income
33 has been associated with higher face-to-face interactions (6, 20, 21), a rather intuitive
34 findings given larger budgets to spend on interactions. On the other hand, the presence of
35 children in the household has been associated with shorter out-of-home, face-to-face social
36 activities (20), and face-to-face interaction frequencies (6, 22), likely as a result of time-
37 budget constraints associated with childrearing.

38 With respect to access to mobility tools Banister and Bowling (23) found that elderly
39 people with access to a vehicle (and people with access to adequate local transport) were
40 likely to undertake more (face-to-face) social activities. On the other hand, Farber and Páez
41 (24) found that people who are more automobile-reliant tend to participate in fewer social
42 activities.
43

1 Social network characteristics have also been associated with communication
2 frequencies. People with larger networks have been found to have more social interactions
3 (20, 25). Other studies, however, have suggested that communication frequencies might
4 actually be reduced in order to maintain larger social networks (26).

5 Regarding ego-alter relationships, emotional closeness has been positively associated
6 with interaction frequencies irrespective of modes (17, 27), while in terms of relational roles,
7 lower face-to-face and telephone contact frequencies between work mates, and higher
8 telephone contact frequencies between relatives have been reported (15). Van den Berg et al.
9 (17) also found higher telephone contact frequencies between relatives, and lower face-to-
10 face and email frequencies compared to friends.

11 In terms of homophily, analyzing contact frequency by different modes, Carrasco
12 (18) only found a significant effect for gender homophily on face-to-face interaction
13 frequency. Sharmeen et al. (22), analyzing the frequency of face-to-face social interaction,
14 found a positive effect for gender and education level homophily. They found however a
15 negative effect for age homophily. Parady et al. (19) found negative effects for both age and
16 gender homophily on face to face and ICT contact propensity.

17 Another important variable is the geographical distance between egos and alters. In
18 spite of the development of new ICT tools and decreasing prices for telecommunication,
19 geographical distance is still an impediment for interaction in the post-Internet era (28).
20 Perhaps the most consistent finding in the literature is the negative effect of distance on face-
21 to-face interaction frequency (3, 11, 17, 19, 29, 30). These studies also found a negative effect
22 of distance on telephone contact frequencies, except for Boase et al. (30) who found no
23 relationship. On the other hand, email contact frequencies have been found to increase with
24 geographical distance by all studies except for Frei and Axhausen (15) who found no
25 significant relationship.

26 The literature shows that the study of face-to-face and ICT mediated communication
27 and the link with social activity-travel has received some interest in recent years. Although
28 these studies have indicated a number of factors that influence social interaction behavior,
29 the review has also shown that communication patterns differ among countries and cultures.
30 Therefore, an in-depth comparison of the factors that influence the frequency of interaction
31 through various communication tools in different socio-cultural contexts will increase our
32 understanding of the maintenance of social networks and related social travel demand. As
33 such, the analysis presented in this article is guided by the following research questions:

- 34 1. What are the differences in social interaction patterns between societies and
35 nation-states?
- 36 2. What role does distance play in these patterns?
- 37 3. How are personal and network characteristics associated with social interaction
38 patterns?
- 39 4. To what extent are these associations contextual- or culture-specific, or consistent
40 across societal and cultural backgrounds?

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1 **3. DATA COLLECTION**

2 **Toronto, Canada - TOR**

3 As part of the “Connected Lives Study”, data were collected between May 2004 and
4 April 2005 in the East York area of Toronto, Canada. The study consisted of two stages: a
5 survey of a random sample of 350 people from East York, and interviews of a sub-sample of
6 87 respondents (86 valid responses). The interviews were used to elicit respondents’ social
7 networks. Respondents were asked to name the people who live outside their household, with
8 whom they felt very close and somewhat close. “Very close” alters consist of those persons
9 with whom the respondent discusses important matters or regularly keeps in touch with or
10 are there for them if they need help. “Somewhat close” alters consisted of “more than just
11 casual acquaintances, but not very close”. Data on alters’ age, gender, type of relationship,
12 and home location was collected. Contact frequency by mode was measured on an ordinal
13 scale. See Hogan et al. (2) and Carrasco et al. (31) for more information.
14

15 **Zurich, Switzerland (1) – ZUR (1)**

16 In Zurich, Switzerland, social networks data were collected between December 2005
17 and December 2006. A random sample of respondents was recruited by telephone. The data
18 collection itself consisted of a written questionnaire containing socio-demographic and travel
19 related questions and a face-to-face interview to collect information on social networks. To
20 elicit very close social network ties, a similar name generator was used as in the Connected
21 Lives Study in Toronto. A second name generator asked for persons with whom the
22 respondents plan to and actually spend leisure time. Data on 307 respondents was collected.
23 Data on alters’ relationship type and duration, and home location was collected. Contact
24 frequency by mode was measured on a metric scale. For more information, see Axhausen
25 and Frei (3)
26

27 **Eindhoven, The Netherlands - EIN**

28 Between January and June 2008, social network data were collected in Eindhoven
29 and a number of surrounding villages. The data were collected as part of a larger study that
30 consisted of a two-day social interaction diary and a paper-based questionnaire to capture the
31 respondents’ social networks. The social interaction study involved 747 respondents. The
32 social network questionnaire was completed by a subsample of 116 respondents. The name
33 generating questions used in this study are similar to those in Toronto. Respondents could
34 record up to 25 very close and 40 somewhat close social network members. Data on alters’
35 age, gender, type of relationship, and distance between home locations of ego and alter was
36 collected. Contact frequency by mode was measured on an ordinal scale. For more
37 information, see van den Berg et al. (6).
38

39 **Concepción, Chile - CCP**

40 In Concepción, Chile, characteristics of social activity travel and personal networks
41 were collected between August 2008 and April 2009. Data about personal networks were
42 collected through semi-guided interviews with 241 respondents in four distinctive
43 neighborhoods. The same name generators were used as in Toronto. Data on alters’ gender,

1 age, type of relationship and home location was collected. Contact frequency by mode was
2 measured on an ordinal scale. For more information, see Carrasco and Cid-Aguayo (7).

3 4 **Switzerland - SWI**

5 In Switzerland, social networks data was collected between January 2009 and March
6 2011. In this study, snowball sampling was used. A stratified random sample of the Zurich
7 population was used to recruit 40 initial respondents. In a paper-based questionnaire,
8 respondents were asked to record the alters with whom they make plans to spend free time.
9 In addition, a second name generator was used, asking about other people with whom
10 important problems are discussed. The elicited social network members were then asked to
11 fill out the social network questionnaire as well. After five iterations a total of 743 responses
12 were collected. The attributes collected for each alter included age, gender, type and duration
13 of relationship and home location. Contact frequency by mode was collected on a metric
14 scale. For more information, see Kowald and Axhausen (5).

15 16 **Zurich, Switzerland (2) – ZUR (2)**

17 A second social network survey in Zurich, Switzerland was conducted as part of a
18 larger mobility survey in 2017. The survey consisted of two parts: a first part focusing on
19 mobility behavior, mobility tool ownership, etc. The second part focused on social networks.
20 For the second part of the survey, data on 1,536 respondents was collected. The attributes
21 collected for each alter include age, gender, type and duration of relationship and home
22 location. Contact frequency by mode was collected on a metric scale. For more information,
23 see Guidon et al. (4).

24 25 **Greater Tokyo, Japan - TYO**

26 In the Greater Tokyo Area (the Tokyo metropolis and the prefectures of Saitama,
27 Chiba and Kanagawa) a personal networks survey was conducted in early 2019 on 1000
28 randomly sampled residents of six municipalities in the Greater Tokyo. The survey mediums
29 were web and paper. The name generators used to elicit network members were similar to
30 the ones used in the Switzerland data. Data on 217 respondents was collected. The attributes
31 collected for each alter included age, gender, employment and marital status, type and
32 duration of relationship, tie strength, and home location. Contact frequency by mode was
33 collected on an ordinal scale. For more information, see Parady et al. (8).

34 35 **4. DESCRIPTIVE STATISTICS**

36 General statistics of each survey location are presented in Table 1. Statistics from
37 Canada, The Netherlands, Switzerland and Japan show relatively small differences in wage
38 level and transportation costs, although the city of Zurich exhibits much higher wage levels
39 and transportation costs. In the case of Concepción, Chile, wage levels are considerably lower.

40 Table 2 summarizes ego characteristics for each dataset. Although the table is self-
41 explanatory, some differences among samples are worth highlighting. In particular,
42 differences in internet and mobile phone access are likely related with the year the surveys
43 were conducted, which explains the low market penetrations rates in the Toronto sample, and

1 Zurich (1). Although internet and mobile access was not measured in the Zurich (2) and
 2 Greater Tokyo samples, given that these surveys are very recent, it can be expected that
 3 access is very high.

4
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Table 1: Country/city specific statistics

Variable	TOR ^a	ZUR ^b	EIN ^a	CCP ^a	SWI ^a	TYO ^c
Population (1,000s)	5,500	394	213	292	7,900	36,131
Population density	3,972	4287	2,407	1,318	188	2,664
Wage level Gross (Base: New York=100)	74.2	131.1	77	21.2	115.1	79.2
Wage level Net (Base: New York=100)	80.4	141.8	72.7	24.3	124.2	85.3
Bus/Tram/Metro (US\$, 10 km trip ticket)	2.4	3.8	2.6	0.7	2.7	2.3
Taxi (US\$ per 5km)	8.2	27.6	17.2	7	21.2	17
Train (US\$, 200 km single ticket)	45.4	73.4	31.2	11.7	44.8	32.9
Average cost of fuel per liter (US \$)	0.89	2.0	1.72	1.06	1.22	1.6
Average Mid-Price Car (US \$)	19,933	45,200	21,140	11,416	22,240	26,300
Tax on Car (US \$/Year)	64	426	289	210	255	495

Source: UBS, Wealth Management Research (2006, 2012, 2015, 2017)

^a Wage level and transport costs compiled from 2006 data at country level

^b Wage level and transport costs compiled 2012, 2015 data for Zurich City;
 population data from 2012

^c Wage level and transport costs compiled from 2012, 2017 data for Tokyo Metropolis;
 population data from 2015

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 7

Table 2: Survey information and socio-demographic characteristics of the respondents

	TOR	ZUR (1)	ZUR (2)	EIN	CCP	SWI	TYO
Year of the survey	2004-5	2005-6	2017	2008	2008-9	2009-11	2019
Survey medium	Paper/ Interview	Paper/ Interview	Web/ paper	Paper	Paper/ Interview	Paper	Web/ paper
Number of respondents	84	265	1,536	106	241	426	217
Variable	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)
Male	39.8	42.3	50.4	31.1	39.8	41.1	54.4
Young (<30)	9.8	19.6	12.6	7.5	24.1	5.9	12.4
Middle (30-60)	69.5	44.2	56.9	59.4	58.5	73.5	52.5
Old (>60)	20.7	36.2	30.6	33.0	17.4	20.6	35.0
Living with partner	61.9	48.3	85.3	72.6	58.1	80.3	55.8
Child(ren) under 18	46.4	-	-	34.0	61.8	36.6	20.7
Secondary education or lower	18.3	7.9	16.8	17.0	44.8	1.6	30.9
Technical education	28.0	70.2	42.7	34.0	24.5	48.6	23.0

	TOR	ZUR (1)	ZUR (2)	EIN	CCP	SWI	TYO
University education	53.7	21.9	40.5	49.0	30.7	49.8	46.1
Low HH-income	29.4	24.5	36.8	36.8	43.8	10.1	22.4
Medium HH-income	41.2	47.9	37.5	25.5	25.1	39.2	57.9
High HH-income	29.4	27.5	25.7	37.7	31.1	50.7	19.7
1 or more cars	-	63.4	84.7	83.0	56.4	89.0	57.4
Season ticket	-	38.9	64.9	42.5	-	82.9	-
Internet access	79.8	67.9	-	90.6	63.9	97.9	-
Mobile phone access	42.3	65.7	-	94.3	86.3	95.8	-
	mean	mean	mean	mean	mean	mean	mean
Age	50.3	50.7	50.5	51.6	42.8	50.1	50.1
Work hours	22.5	-	40.0	14.6	21.8	-	-
Years in current location	12.8	-	20.7	13.3	16.4	25.4	-

1
2 Table 3 summarizes the social network characteristics in each dataset and contact
3 frequencies by mode. In terms of social network size, networks in the Toronto, Eindhoven,
4 Concepción and Switzerland samples are quite similar, while networks in the Zurich (1),
5 Zurich (2) and the Greater Tokyo samples are considerably smaller. Although socio-cultural
6 factors might partly explain these differences, there are several methodological factors that
7 might also affect the number of elicited alters, such as the name generators, survey mediums
8 (i.e. in-person interview, self-response paper, or self-response web) and sampling methods
9 used.

10 In terms of alter characteristics, the distribution of alters across age groups shows
11 similar patterns to the distribution of the respondents' age, except for the Concepción study
12 where the share of alters in the youngest age class is higher than the share of egos in that
13 class. The Zurich (1) study did not include questions on alters' sex and age. Regarding
14 relationship type, average shares of immediate family in the network vary between 13.4%
15 and 32.5%, being the lowest in the Switzerland sample. The average shares of extended
16 family range between 11% and 25 %, with higher shares in Eindhoven (25.1%) and
17 Concepción (23.6%). In terms of relationship duration, the Concepción sample has a
18 relatively large proportion of 'new' ego-alter relationships (known each other less than one
19 year). This may be related to the larger shares of young egos and alters in the sample. Finally,
20 regarding the distance between home locations of ego and alter the results show different
21 patterns for each dataset as well. The proportion of local contacts (<2 km) range between
22 17.9% and 30.1%. Toronto shows a relatively large share of long-distance social ties (>100
23 km), whereas this share is lowest in case of Switzerland (6.8%), followed by the Greater
24 Tokyo (8.7%).

25 Interaction frequency is summarized in terms of three communication modes: face-
26 to-face, by telephone, and via Internet. Internet includes communication means such as e-
27 mail, short-messaging services (SMS), and for the two most recent datasets (Zurich (2) in
28 2017 and Greater Tokyo in 2019), also social networking services (SNS), and video chats.
29 Regarding face-to-face contact, in the Concepción sample more than half (55.1%) of the ego-
30 alter pairs meet each other weekly, whereas this share is roughly one third for Toronto, Zurich
31 (2), Eindhoven and the Greater Tokyo (24.3% to 32.4%), 19% for Zurich (1) and only 8.2%

1 for Switzerland. In the Swiss dataset the largest share of ties meets less than once a month.
 2 There is also great variation in the share of ego-alter pairs that meets less than once a year,
 3 with lowest values in the Eindhoven sample (1.3%) and Zurich (2) showing the highest
 4 percentage at 35.2%.

5

6 **Table 3: Characteristics of personal networks**

	TOR	ZUR (1)	ZUR (2)	EIN	CCP	SWI	TYO
Number of alters	1,019	3,156	19,231	2,452	5,038	6,982	1,448
Mean network size ^a	23.8	11.9	15.1	23.9	20.9	21.6	9.9
Personal network variables	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)
Very close	54.0	52.4	88.7	43.0	51.3	28.9	38.0
Somewhat close	46.0	47.6	11.3	57.0	48.7	71.1	62.0
Male	42.1	-	45.7	41.4	45.9	43.1	40.9
Young (<30)	11.8	-	14.5	13.0	32.2	9.2	17.3
Middle (30-60)	65.3	-	54.0	53.7	52.9	67.2	45.3
Old (>60)	22.9	-	31.5	33.3	14.9	23.6	32.7
Immediate family	25.4	18.3	32.5	18.8	20.2	13.4	21.8
Extended family	11.0	12.7	6.8	25.1	23.6	11.1	11.3
Friend or other	63.6	69.0	60.8	56.1	56.2	75.5	66.9
Known <1 year	-	0.6	0.4	1.4	14.6	2.0	3.9
1-10 years	-	36.7	25.7	37.4	25.8	28.5	32.5
>10 years	-	62.7	73.9	61.2	59.6	69.5	63.6
Distance < 2km	19.3	26.0	30.1	17.9	36.1	24.6	29.7
Distance 2 – 30km	42.9	43.2	44.7	47.6	41.3	54.1	45.4
Distance 30 – 100km	9.5	12.4	13.9	17.3	2.4	14.5	16.2
Distance > 100km	28.3	18.4	11.3	17.2	20.2	6.8	8.7
Contact frequencies by mode	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)	Sample (%)
Face-to-face							
Weekly	32.4	19.0	24.3	28.8	55.1	8.2	27.9
Monthly	25.1	25.5	21.1	31.0	24.9	27.6	25.5
Less than monthly	31.5	44.5	19.2	38.9	14.4	57.8	44.1
Less than yearly	11.0	11.0	35.3	1.3	5.7	6.4	2.6
Telephone							
Weekly	35.9	13.0	14.4	16.1	34.5	6.5	10.7
Monthly	32.2	29.9	16.5	30.2	20.0	20.6	16.2
Less than monthly	18.1	34.8	9.5	43.1	15.4	52.1	31.6
Less than yearly	13.7	22.2	59.6	10.6	30.1	20.8	41.5
Internet							
Weekly	16.4	4.3	25.3	4.8	12.5	3.5	27.3
Monthly	17.4	10.0	16.4	19.3	5.3	12.1	23.6
Less than monthly	8.5	18.1	14.1	26.7	5.1	35.7	26.5
Less than yearly	57.7	67.5	51.0	49.2	77.2	48.8	22.6

7

^a Values exclude isolates (egos who did not report any alters). Small differences between networks sizes reported in this table and values reported in individual studies are due to the difference in the number of observations used.

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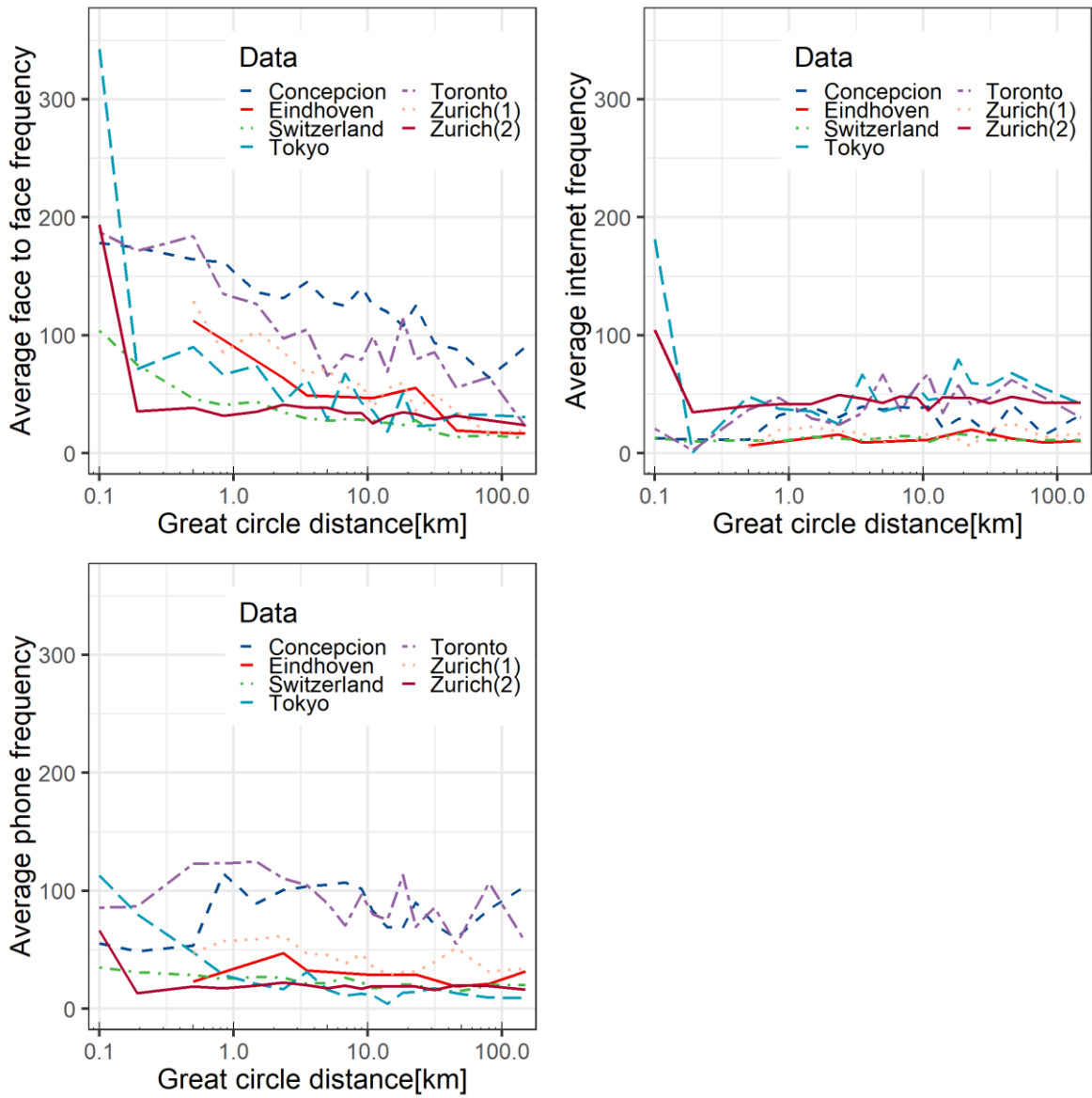
10

1 With respect to telephone and internet contact frequencies, patterns not only differ by
2 country but also by survey year. For the two most recent datasets, a considerable increase in
3 weekly internet contact frequency can be observed (>25%). Similarly, these samples exhibit
4 the largest shares of no contact by phone less than once per year (> 41%). These differences
5 highlight the increasing market penetration of internet-based communication in recent years,
6 and the move away from phone-based contact.

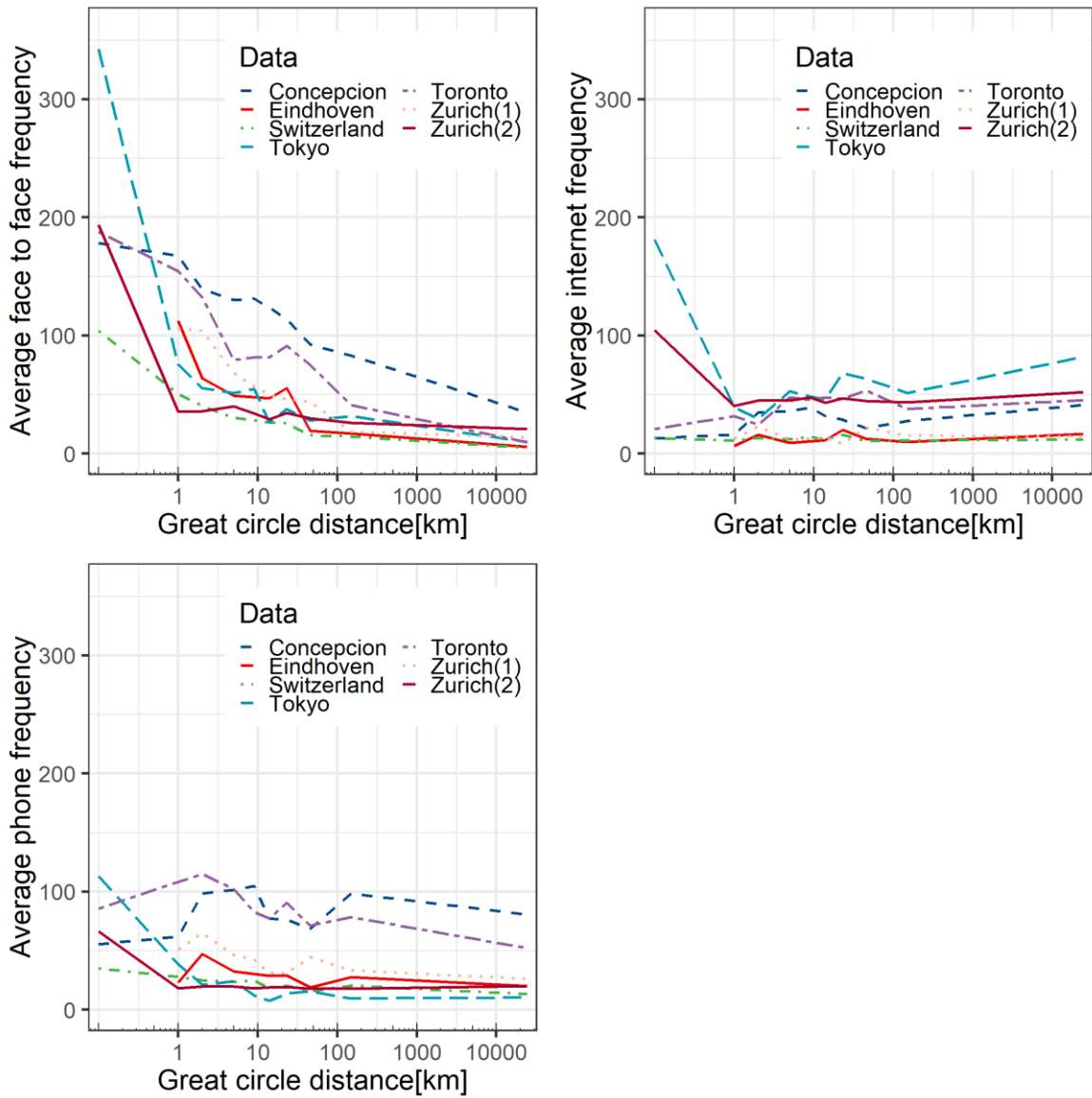
7 Figure 1 shows the mean face-to-face, telephone and Internet interaction frequencies
8 for egos and alters with up to 100km distance between them for each of the seven datasets.
9 This is of importance for the analysis, as the dominant part of all ego-alter distances in each
10 dataset is shorter than that. The results show that for all datasets the face-to-face interaction
11 frequencies decrease with increasing geographical distance between homes of ego and alter.
12 Regarding telephone and internet interaction frequencies the results in Figure 1 show no clear
13 relation between frequency of interaction and geographical distance in the 100km range,
14 which is expected given the low marginal costs of these modes given distance. Similarly,
15 Figure 2 shows mode-specific communication frequencies for distances of up to 10,000 km.
16 The plot for face-to-face meetings clearly shows a decreasing trend for all datasets with
17 increasing distance. On the other hand, telephone and internet-based contact frequencies are
18 largely unaffected by distance.

19 Figure 3 shows the mode specific market shares by distances of up to 10,000 km. All
20 three communication modes are clearly affected by increases in distance, and patterns are
21 very similar among all datasets, with some clear temporal differences. The shares of face-to-
22 face meetings are dominant for distances of up to 10 km and stay very important even at long
23 distances of 1,000 km and above. However, the share of face-to-face interactions clearly
24 decreases with increasing distance in all data sets. For the older datasets (all datasets except
25 Zurich (2) and Greater Tokyo), at distances of around 10 km, similar telephone contact shares
26 can be observed, increasing up to the 100 km range, where they flatten. For the most recent
27 datasets, a considerable drop in phone market share is observed with both curves almost flat,
28 and below a 30% share irrespective of distance. On the other hand, market shares of internet-
29 based communication modes increase consistently with increasing distances, with
30 considerable differences in share for the most recent datasets. In case of the Greater Tokyo
31 sample, at 10 km the market share is almost 50%.

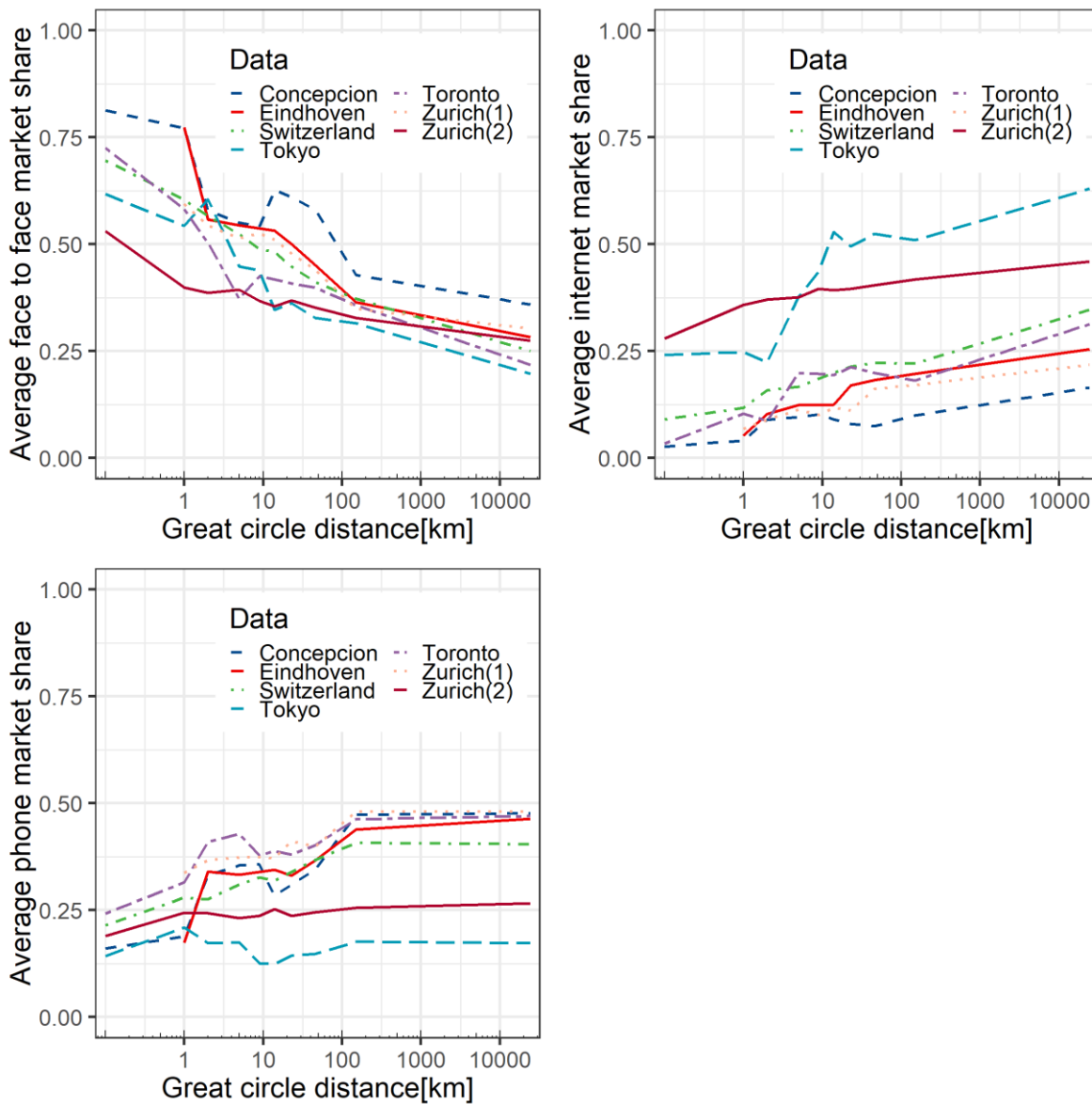
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33



1
 2 **Figure 1** Yearly average interaction frequency by geographical distance of up to 100
 3 **km**
 4



1
 2 **Figure 2** Yearly average interaction frequency by geographical distance of up to 10,000
 3 **km**
 4



1

2 **Figure 3: Mode specific market shares by geographical distance**

3 **5. MODEL BASED COMPARISONS AND RESULTS**

4 To further evaluate the effects on communication frequencies and social travel, a multilevel
 5 regression model of yearly contact frequencies by mode is estimated. Multilevel linear
 6 regression modelling is used given the hierarchically clustered nature of the data. Note that
 7 these regression techniques require models to have complete case observations without
 8 missing values, as such, only observations with complete information are used for estimation.
 9 This also restricts the potential set of explanatory variables to variables that can be found in
 10 all seven datasets.

1 The datasets studied in this paper can be structured in three levels. Level 1 includes
2 information on the characteristics of each alter and individual ego, level 2 includes the ego's
3 socio-demographic as well as aggregated personal network characteristics, level 3 includes
4 information about the study area where each data collection took place. In this analysis we
5 use random intercepts and slopes to account for intrapersonal and interpersonal variations as
6 well as correlation among contact modes. A specific characteristic of our study is the right
7 skewed distribution of the dependent variables, because mode-specific communication
8 frequencies between egos and alters include predominately low and only few (very) high
9 values. The residual maximum likelihood estimator (REML) was used for model estimation.

10 A logarithmic transformation was employed to address the skewed distance and
11 communication-frequency distribution in each dataset. Distances with zero km were recorded
12 due to rounding and aggregation, considering that not all data bases are geocoded at the
13 household level, but to higher level units such as blocks and street corners. In order to avoid
14 infinite values and the loss of observations, an empirically calculated constant was added to
15 the smallest 2% of all distances in each dataset. Table 4 presents a comparative analysis of
16 effects on log yearly contact frequencies by mode. For legibility reasons only coefficients
17 significant at least at the 0.1 level are reported.

18 Regarding ego-level variables, a higher education degree is associated with higher
19 internet contact frequencies, although the coefficients are not significant for the Switzerland,
20 Zurich (1) and Greater Tokyo samples. This positive association might be a result of more
21 geographically dispersed networks. Kowald et al. (13) reported positive associations between
22 higher education and greater distances between contacts, arguing that higher education both
23 enables and demands more travelling in professional life and might result in a more diverse
24 geographical biography. As such, these ties are likely maintained through low-cost internet-
25 based communication.

26 The effect directions among significant age coefficients is consistent across all
27 samples given contact modes. On average, people in age cohorts of 30 and over tend to have
28 lower face-to-face contact frequencies, than their "under 30 years old" counterparts. This
29 effect can be explained by the fact that younger cohorts tend to have larger discretionary time
30 budgets, due to factors such as not having children, etc. No significant effects, however, are
31 observed in the Concepción and Greater Tokyo samples. The clearest age pattern, however,
32 is that of internet contact frequency, where the older the cohort, the larger the magnitude of
33 the effect. This indicates a decreasing willingness to adopt new technologies.

34 Household income was positively associated with higher face-to-face frequencies in
35 the Toronto, Eindhoven and Zurich samples, consistent with findings from the literature that
36 suggests higher monetary budgets might facilitate physical interaction. Larger networks are
37 negatively associated with contact frequencies irrespective of mode, although in some cases
38 the coefficients are not significant. Although effect magnitudes are not very large, this
39 supports the hypotheses that to some extent, larger networks impose heavier burdens on
40 individuals to maintain ties, which results in lower contact frequencies (26).

41 In terms of ego-alter relational characteristics, mixed associations are observed
42 regarding relationship type. These effects, however, need to be interpreted with care because
43 of the interactions with distance and emotional closeness. More specifically, these

1 coefficients capture the effects of “somewhat close” relations at a 1km distance, compared
2 against “somewhat close” non-relatives. While these effect directions differ among datasets
3 and contact modes, the significant effects of emotional closeness are very consistent. With
4 the exception of the Zurich (1) sample, emotional closeness is positively associated with
5 contact frequencies irrespective of mode, with differences in magnitude given relationship
6 type. Furthermore, when looking at effect magnitudes, emotional closeness has relatively
7 larger effects on phone contact frequencies, suggesting that people are willing to incur in
8 higher costs for communication with emotionally closer members of their networks. In the
9 particular case of the Greater Tokyo, emotional closeness is only significant for internet
10 contact with immediate family.

11 Distance is negatively associated with face-to-face contact, with variations in
12 magnitude given relationship type. Phone contact frequencies also tend to decrease with
13 increasing distances. Contact with non-relatives in Concepción is the exception, where the
14 phone might serve as a substitute for face-to-face contact, given increasing distances. A
15 clearer substitution pattern, however, is observed for internet contact. With the exception of
16 Zurich (2), increasing distances are associated with higher non-relative contact frequencies.
17 This was also the case for contact with relatives in Concepción, Switzerland, Toronto and
18 Zurich (2) samples. On the other hand, in the Greater Tokyo area, people tend to have lower
19 internet contact with relatives given increasing distances, that is, no substitution effect is
20 observed. Given that blood ties are more likely to be maintained regardless of distance (13),
21 it is plausible that in the Greater Tokyo case, egos only use internet communication to
22 maintain non-relative ties, which might be more sensitive to distance.

23 Although, the findings above suggest the existence of some consistent associations
24 between individual and relational attributes with contact frequencies irrespective of socio-
25 cultural contexts, there were several variables for which no clear patterns were observed. At
26 the ego level, the associations between education with face-to-face and phone contact, gender
27 with face-to-face and internet contact, and household income with ICT contact in general,
28 were not clear. At the ego-alter level, the differences observed among relationship types, as
29 well as effect magnitude variations given relationship type also exhibited diverse patterns.

30 Some results might be explained by context-specific socio-economic factors. For
31 example, economic affluence might be more crucial a factor in facilitating ICT
32 communication in the Concepción case than other contexts, which might help explain the
33 association between higher incomes and higher phone and internet contact frequencies. This
34 would, however, not explain the negative coefficient observed for phone contact in Zurich
35 (1), and for internet in Switzerland, or the positive coefficients observed for internet contact
36 in the Zurich (2) sample.

37 Methodological differences in terms of sampling, survey medium, question
38 formulation, etc., as well as cultural factors might also account for observed differences. The
39 precise source of these variations remains, however, an open question, since it is very difficult
40 to precisely discriminate between these effects from these data alone.

41

Table 4: Multilevel-multivariate model of contact frequencies by mode - standardized coefficients

Fixed effects	Dataset							Contact modes																		
	CCP	EIN	SWI	TOR	ZUR1	ZUR2	TYO	Phone	Internet																	
Intercepts	1.98	1.96	1.95	2.09	1.69	1.24	1.47	-1.31	-1.80																	
	Face to face							Phone							Internet											
	CCP	EIN	SWI	TOR	ZUR1	ZUR2	TYO	CCP	EIN	SWI	TOR	ZUR1	ZUR2	TYO	CCP	EIN	SWI	TOR	ZUR1	ZUR2	TYO					
Dataset x contact mode	-	-	-	-	-	-	-	-	0.97	0.83								-	0.38	1.63	1.65					
Ego characteristics																										
Education level																										
Secondary or lower (r)																										
Technical degree	0.11							0.27							0.32					0.33						
University degree or higher															-0.07							-0.26	0.47	0.38	0.41	0.09
Gender																										
Female (r)																										
Male			0.07	-0.16			0.05									0.18	-0.14									
Age																										
Under 30 (r)																										
30 to 60			-0.20	-0.29	-0.39	-0.17	-0.16	0.14				0.28			-0.19	-0.43	-0.28	-1.08	-0.53	-0.26	-0.20					
Over 60			-0.28	-0.31			-0.16	-0.21				0.26	0.42	-0.43	-0.43	-0.28	-1.08	-0.53	-0.69	-0.81						
Household income																										
Low income (r)																										
Medium income																										
High income			0.12			0.20	0.13	0.05	0.55				-0.28			0.31	-0.11			0.16	0.18					
Network size	-0.01	0.00*	0.00*			-0.01	-0.01	-0.01			-0.01			-0.01			-0.01	-0.04			-0.01					
Ego-alter relational characteristics																										
Relationship type																										
Non-relative (r)																										
Extended family			-0.26	-0.40			-0.27	0.48				0.25	-0.61			0.23			0.50	-0.22	0.11	-0.43	-0.17			
Immediate family	-0.17			-0.29			0.45	0.27	0.63	-0.15	0.46	-0.41			0.60	0.40	0.86	-0.18	0.21			-0.31	0.20	0.38		
Emotionally close																										
Non-relative	0.08	0.07	0.26	0.17	-0.05			0.31	0.39	0.33	0.37	0.73	-0.15			0.46	0.16	0.27	0.28	0.12	-0.07			0.55		
Extended family	0.16	0.33	0.24	0.20	-0.20			0.28	0.41	0.29	0.52	0.62			0.51	0.21	0.19	0.56	-0.18			0.43				
Immediate family	0.31	0.39	0.24	0.23			0.58	0.80	0.57	0.49	0.74	-0.20			0.71	0.21	0.12	0.40			0.68	0.31				
Log of distance																										
Non-relative	-0.27	-0.34	-0.44	-0.36	-0.31	-0.10	-0.28	0.04			-0.15	-0.12			-0.04	0.10	0.16	0.05	0.18	0.12			0.09			
Extended family	-0.31	-0.30	-0.28	-0.27	-0.32	-0.12	-0.62	-0.05	-0.11	-0.10	-0.11	-0.11	-0.11	-0.21	0.04			0.05			-0.16					
Immediate family	-0.28	-0.42	-0.33	-0.30	-0.43	-0.35	-0.54	-0.05	-0.19	-0.15	-0.14	-0.05	-0.17	0.08			0.07			0.04	-0.21					

Random Effects							Goodness of fit statistics		
Groups	Var.	S.D.	Correlation	Test	AIC	Likelihood ratio test (p-value)			
Ego (n=2636)				None	198,440	reference	Number observations 94485		
Face-to-face	0.05	0.21		Ego	202,558	9990.3 (<0.01, df=5)	REML criterion 197710.30		
Internet	0.26	0.51	0.12	Ego-alter	208,421	4119.9 (<0.01, df=1)	Marginal R-square 0.37		
Phone	0.19	0.43	0.23	Intra-class correlation :			Conditional R-square 0.56		
Ego-alter (n=31495)				Adjusted	0.31				
Intercept	0.10	0.32		Conditional	0.19				
Residual	0.34	0.58							

* Coefficient is smaller than -0.01. (r) indicates the reference category

1 In terms of random effects, their significance was tested via a single-term-deletion
2 likelihood ratio test. That is, for each test, one random effect is eliminated from the model and the
3 likelihood ratio test calculated. All random effects were statistically significant. Furthermore, the
4 positive correlations between modes suggests, as reported in the literature, that face to face and
5 ICT modes are complementary rather than substitutes. However, when accounting for the effect of
6 distance discussed above, this complementarity tends towards substitution for internet-mediated
7 contact given increases in distance.

8 Intraclass correlation levels also support the multilevel structure used in this analysis.

9 Finally, regarding the limitations of this study, it is important to note that since the model
10 required complete case observations without missing values, only variables available to all datasets
11 could be incorporated. As such, homophily effects, which have been well documented in the
12 literature could not be evaluated.

13 14 **6. CONCLUSIONS**

15 In recent years, there has been an increasing awareness of the importance of the relation between
16 social networks and travel. Face-to-face and ICT mediated social interaction patterns are of interest
17 to transportation researchers, as changes in these interaction patterns might impose new demands
18 on urban environments and transportation services. This subject, however, has until recently
19 received little attention. This article compared social interaction frequencies by mode among seven
20 datasets from five different countries. It complements the study by Kowald et al. (13) on the relation
21 between social network distance patterns with personal network and relational characteristics and
22 shows that these characteristics are also strongly associated with social interaction patterns.

23 Analysis of social interaction against distance showed very similar patterns across samples,
24 with face-to-face frequencies showing high sensitivity to distance. In terms of modal share, a clear
25 substitution effect was observed between face-to-face contact and ICT across all samples given
26 increasing distances. For the older datasets (before 2012), the telephone dominated modal shares
27 at distances over 100 km. On the other hand, for the more recent surveys (after 2017), a clear
28 transition in ICT preference from phone to internet-based modes was observed.

29 A Multilevel-multivariate mixed regression analysis results showed the existence of very
30 consistent associations between individual and relational characteristics and social interactions
31 irrespective of socio-cultural context, such as age, network size, distance and emotional closeness.

32 Face-to-face and ICT contact were found to be complementary, a relation that tends
33 towards substitution for internet-mediated contact given increases in distance.

34 While consistent associations across samples were indeed found, for ego-level
35 characteristics such as gender and ego-alter level characteristics such relationship type, among
36 others, effect patterns were less clear. These differences might be explained not only by socio-
37 economic and cultural factors, but also by methodological differences in terms of sampling, survey
38 medium, question formulation. While discriminating between these effects is difficult, it is clear
39 that intrinsic contextual characteristics play an important role and need further understanding.

40 Regarding future research, similar to the study by Calastri et al. (12), new data collection
41 efforts should explicitly measure travel behavior and companionship in a disaggregate manner, in
42 addition to the social network and interaction characteristics measured in this study. This would
43 allow researchers to move beyond descriptive models of social interaction patterns towards the
44 explicit incorporation of social network characteristics into different dimensions of travel behavior
45 such as joint activity generation, companion choice and/or joint activity destination choice.

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9 AUTHOR CONTRIBUTIONS

- 10 ■ GP, KT and NH were involved in the collection of, and provided the Greater
11 Tokyo Area data
- 12 ■ AF and KWA were involved in the collection of, and provided the Zurich (1) data
- 13 ■ SG, MW and KWA were involved in the collection of, and provided the Zurich (2) data
- 14 ■ MK, and KWA were involved in the collection of, and provided the Switzerland data
- 15 ■ JCA and BW were involved in the collection of, and provided the Toronto data
- 16 ■ PVDB, TA and HT were involved in the collection of, and provided the Eindhoven data
- 17 ■ JCA were involved in the collection of, and provided the Concepción data
- 18 ■ GP, AF, and MK conducted the statistical analyses
- 19 ■ GP, AF, MK and PVDB wrote the manuscript
- 20 ■ All authors provided comments and revisions to the analysis results, drafts and approved the
21 final version

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