

1 **Associations between individual characteristics, availability of bicycle**
2 **infrastructure, and city-wide safety perceptions of bicycling: a cross-sectional**
3 **survey of bicyclists in 6 Canadian and U.S. cities**

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19 **Abstract**

20 Safety concerns are a primary deterrent to bicycling. Bicycle infrastructure is both preferred and
21 safer for bicycling. In this paper, we examine the association between availability of bicycle
22 infrastructure and perceptions of bicycling safety amongst over 3,000 bicyclists living in six
23 large Canadian and US cities. In three repeat cross-sectional surveys (2012, 2013 and 2014),
24 adults living in Boston, Chicago, New York, Montreal, Toronto, and Vancouver were surveyed
25 about their bicycling habits, safety perceptions, and demographic characteristics as part of the
26 International Bikeshare Impacts on Cycling and Collisions Study (n = 16,864). Participants were
27 assigned a measure for the availability of bicycle infrastructure (a component of Bike Score[®]
28 called Bike Lane Score, range 0-100) based on their residential postal code. We used weighted
29 multinomial regression models to examine associations between perceived bicycling safety and
30 spatial access to bicycle infrastructure accounting for sociodemographic characteristics amongst
31 those who report bicycling in the past month (n = 3,446; weighted n = 3,493). Overall, 57.9%
32 perceived bicycling in their city as safe, 15.1% as neutral, and 27.0% as dangerous. Our model
33 indicates that, within cities, bicyclists with greater bicycle infrastructure availability had
34 improved odds of perceiving bicycling as safe. Specifically, a 10-unit increase in Bike Lane
35 Score was associated with 6% higher odds of a bicyclist perceiving the safety of bicycling as safe
36 compared to neutral. Bicyclists who are male, younger, lower income, have young children, have
37 a high-school education, and bicycle more frequently are predicted to be more likely to perceive
38 bicycling in their city to be safe. These results suggest that increasing the availability of bicycle
39 facilities by expanding bicycling networks may result in increases in perceptions of bicycling
40 safety for existing bicyclists, but also that individual characteristics play a substantial role in
41 bicycling safety perceptions.

42 **Keywords**

43 Bicycling; Bicycle infrastructure; Perceived safety; Bike Score; Built environment

44 **Highlights**

- 45 • Modelled bicyclists' safety perceptions using large population survey in Canada and US
- 46 • Access to bicycle infrastructure is associated with higher perceived bicycling safety
- 47 • Male and/or younger bicyclists are associated with higher perceived bicycling safety
- 48 • Expanding bicycling infrastructure may result in increased bicyclists' perceptions of
49 safety

50

51 **1.0 Introduction**

52 Research indicates that there are significant societal benefits for bicycling, primarily due to
53 health benefits of increased physical activity (de Hartog et al., 2010; Götschi et al., 2016;
54 Mueller et al., 2015). Within cities in Canada and the United States (US), bicycling uptake is
55 generally low, with only 1.3% and 0.6% of workers reporting that they commute by bicycle to
56 work, in Canada and the US respectively (McKenzie, 2014; Statistics Canada, 2013a). Ridership
57 levels in European cities are much higher than in Canada and the US (Bassett et al., 2008; Pucher

58 and Buehler, 2008; Pucher and Dijkstra, 2003), suggesting there is substantial potential for
59 increased bicycling.

60 Safety concerns are a primary deterrent to bicycling (Heinen et al., 2010; Willis et al., 2014).
61 Previous research has shown that the perceived safety of bicycling varies by age, gender and
62 bicycling experience, across a range of different bicycling environments (Bill et al., 2015;
63 Chataway et al., 2014; Hels and Orozova-Bekkevold, 2007; Lawson et al., 2013; Manton et al.,
64 2016; Møller and Hels, 2008; Parkin et al., 2007). Bicycling environments that provide bicycle
65 infrastructure tend to be perceived as safer than those that require bicycling in mixed traffic
66 (Chataway et al., 2014; Manton et al., 2016; Parkin et al., 2007; Winters et al., 2011). Increasing
67 access to bicycle infrastructure has been promoted as a potentially effective means of increasing
68 bicycling mode share in cities with low bicycling uptake (Buehler and Pucher, 2012; Dill and
69 Carr, 2003; Pucher and Buehler, 2008, 2006). Implementing bicycle infrastructure has the
70 potential to increase trips from new bicyclists, as well as existing (National Institute for
71 Transportation and Communities, 2014; Noland, 1995).

72 Studies of the association between bicycling environments and perceived safety tend to focus on
73 comparing different infrastructure or routes within the road network. These studies are important
74 for understanding how specific infrastructure designs may improve perceived personal safety at a
75 specific time and place, but do not provide insight into their associations with general
76 perceptions of bicycling safety. More generalized perceptions of bicycling safety across a larger
77 geographic context (e.g., across a neighbourhood or city) may be associated with individuals
78 bicycling, or how often they bicycle (Lawson et al., 2013). In this study we aim to examine the
79 associations between individual characteristics, bicycling infrastructure availability, and city-
80 wide perceptions of bicycling safety across 6 major cities in the US and Canada. We use survey
81 data from the International Bikeshare Impacts on Cycling and Collisions Study (IBICCS) (Fuller
82 et al., 2014) to measure perceived safety and individual characteristics, and Bike Lane Score, a
83 component of the Bike Score Index (Winters et al., 2016), to measure bicycle infrastructure
84 availability.

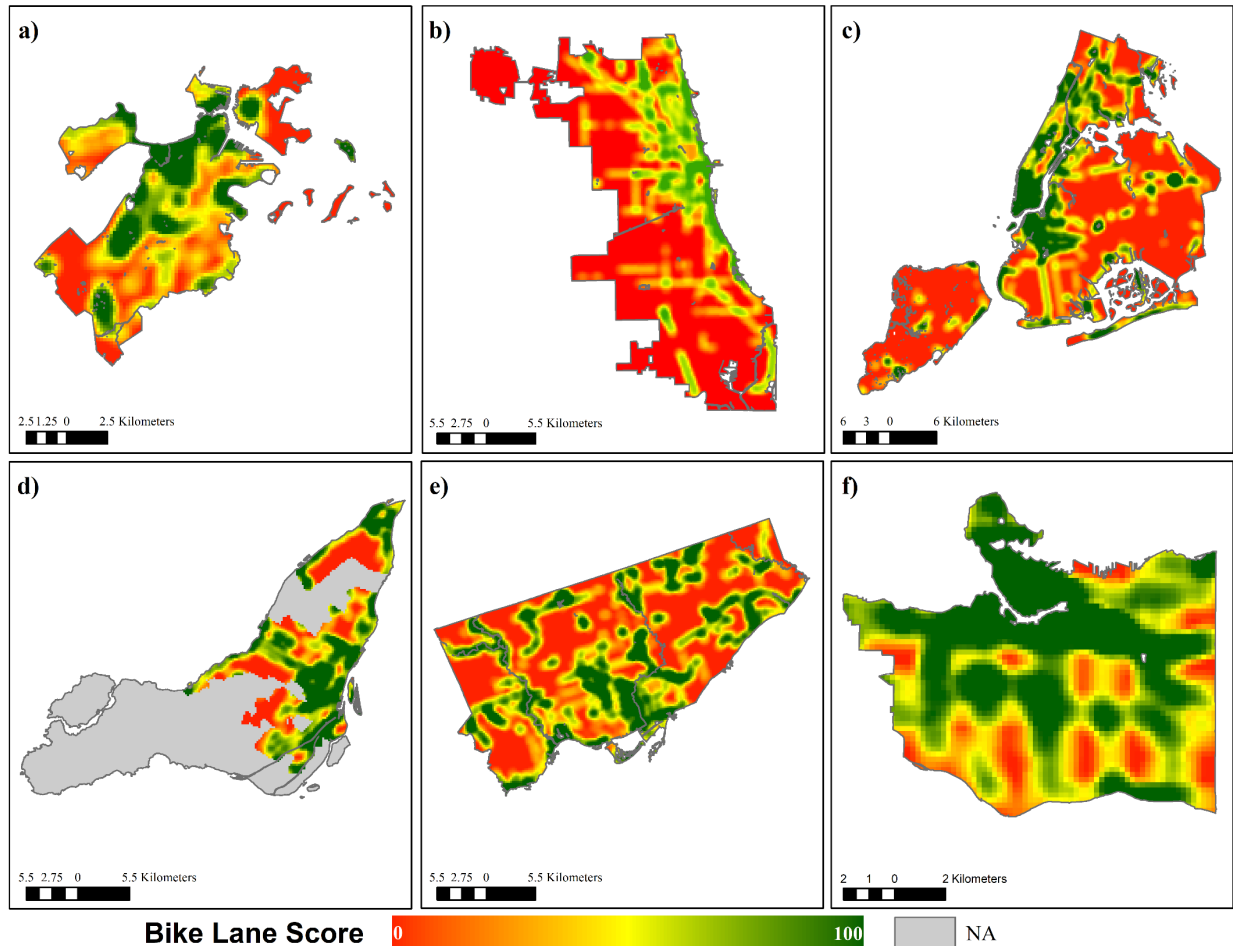
85 2.0 Materials and Methods

86 2.1 IBICCS

87 IBICCS was designed to evaluate the health impacts of public bicycle share programs. The study
88 protocol is published elsewhere (Fuller et al., 2014). Briefly, it included three repeat, cross-
89 sectional online panel surveys (in Fall of 2012, 2013, and 2014) of residents across eight cities:
90 Montreal, Toronto, Boston, New York, Vancouver, Chicago, Detroit and Philadelphia
91 (n=23,901). Each survey collected information on sociodemographic characteristics, health and
92 travel behaviour, safety perceptions, and residential and work locations (post codes). The study
93 aimed to survey participants within a defined sampling area, which were generally smaller
94 subsets of the city proper based on where public bike share stations were (or potentially would
95 be) implemented (see Fuller et al., 2014 for details). In this analysis we include all participants
96 living within the city boundaries. The IBICCS data collection was approved by the Research
97 Ethics Committee of the Centre Hospitalier de l'Université de Montréal.

98 2.2 Study Area

99 Our study area includes the six IBICCS cities for which we have comparable data on bicycle
 100 infrastructure: Boston, Chicago, New York, Montreal, Toronto and Vancouver (Figure 1). In
 101 2012 each city had a low bicycling rate, with between 0.8 to 4.4% of workers reporting that they
 102 commute by bicycle (Table 1).



104 *Figure 1: IBICCS study area boundaries and the spatial distribution of bicycle infrastructure,*
 105 *using Bike Lane Score as a proxy. The study areas include the cities of: a) Boston, b) Chicago, c)*
 106 *New York, d) Montreal, e) Toronto, f) Vancouver. Bike Lane Score can range from 0-100, where*
 107 *100 indicates close proximity, or greater availability of bicycle infrastructure and 0 indicates no*
 108 *bicycle infrastructure within 1-km. NA indicates an area where Bike Lane Score data were not*
 109 *available. Note that cities are of different geographic extents as indicated by scale bars.*

City	Boston	Chicago	New York	Montreal	Toronto	Vancouver
Area (km ²)	125.1	589.3	783.5	366.4	630.6	114.9
Total population ^a	619,662	2,702,471	8,199,221	1,649,519	2,615,060	603,502
Working population ^b	317,930	1,213,901	3,685,786	727,455	1,174,610	294,790
Proportion who commute by bicycle ^b	1.7%	1.3%	0.8%	3.2%	2.2%	4.4%

Average fatalities per million bicycle to work trips (2007-2012) ^c	1.16	1.38	2.94	0.60	NA	1.10
City-wide Bike Lane Score (2012)						
Mean	46	20	29	56	39	68
Median	40	6	11	61	30	74
Q1	14	0	0	19	1	44
Q3	78	33	54	100	72	100
IQR	64	33	54	81	71	56
Range	0-100	0-100	0-100	0-100	0-100	0-100

^a Boston, Chicago and New York, based on 2008-2012 American Community Survey 5-year estimates (US Census Bureau, 2017a), Montreal, Toronto, Vancouver, based on Census 2011 (Statistics Canada, 2012a, 2012b, 2012c);^b Boston, Chicago and New York, based on 2008-2012 American Community Survey 5-year estimates (US Census Bureau, 2017b), Montreal, Toronto, Vancouver statistics are based on 2011 National Household Survey (Statistics Canada, 2013b, 2013c, 2013d);^c Based on previous study which, for Canadian cities compiled fatality and exposure data directly from the cities, and for American cities obtained data from National Highway Traffic Safety Administration Fatality Data (Urban Systems, 2015);NA: Not Available

110 *Table 1: Descriptive information regarding city populations, bicycling mode shares, and*
111 *distributions of Bike Lane Score for 6 cities IBICCS participants were recruited.*

112 2.3 Measures

113 *Safety Perceptions*

114 Only those participants who reported bicycling in the previous month were asked their
115 perceptions of bicycling safety. Bicycling safety perceptions were measured based on a question
116 which asked bicyclists “Overall, how safe do you think bicycling is in your city?” Respondents
117 could answer based on a 5-point scale: 1-“Very Safe”, 2- “Somewhat safe”, 3-“Neutral”, 4-
118 “Somewhat dangerous” and 5-“Very Dangerous.” We collapsed these responses into a 3-point
119 scale consisting of “Safe” (1+2), “Neutral” (3) and “Dangerous” (4+5).

120 *Bicycling Infrastructure*

121 We use a component of the Bike Score index, called Bike Lane Score, to measure the availability
122 of bicycling infrastructure at a given location within the study area (Figure 1). Bike Lane Score
123 (2012, 100 m grids) was provided by RedFin Real Estate and represents the only internationally
124 available standardized infrastructure dataset at the time the IBICCS survey data were collected.
125 Bike Lane Score is a normalized index of a location’s proximity to bicycle infrastructure.
126 Bicycle infrastructure data used in this index were provided by municipal governments, and
127 includes the following: on-street painted bicycle lanes, off-street trails, separated bicycle paths,
128 and neighbourhood bikeways (Winters et al., 2016). To compute Bike Lane Score for a given
129 location, the length of all infrastructure nearby were summed and weighted based on a distance
130 decay function where infrastructure outside 1 kilometre were weighted 0. Infrastructure types
131 that are separated from traffic are weighted double compared to those that are not. The raw
132 weighted lengths were then normalized to a score between 0-100, with higher Bike Lane Scores
133 indicating greater availability of bicycle infrastructure and a Bike Lane Score of 0 indicating no
134 infrastructure within 1 kilometre. Past work shows that Bike Lane Score is positively correlated
135 to journey-to-work bicycle mode share in 24 Canadian and US cities (Winters et al., 2016).

136 The distribution of Bike Lane Scores within each city indicated that many areas still lacked
137 bicycle infrastructure. The median Bike Lane Score within each city ranged from a low of 6 to a
138 high of 74 (Table 1). Chicago, New York and Toronto had lower median Bike Lane Scores as
139 compared to Boston, Montreal or Vancouver, but the spatial extent of Bike Lane Score in these
140 cities was much greater (Figure 1). Each city tended to have higher Bike Lane Scores in the core
141 area as compared to the surrounding area. Survey participants were assigned a Bike Lane Score
142 based on the location of their residence as derived from postal codes.

143 *Covariates*

144 Potential confounders were identified a priori based on sociodemographic characteristics that
145 could influence both choosing to live in bicycling supportive areas, as well as to differences in
146 the perceived safety of bicycling. Selected variables were based on a consideration of individual
147 level confounding variables previously used in research on the effect of proximity to bicycle
148 infrastructure and bicycle use (Krizek and Johnson, 2006) and include age (continuous), gender
149 (male or female), having young children (yes or no), income (under \$50 K, \$50-100 K, over
150 \$100 K, refuse), education (high school or less, or any college/university), and ethnicity
151 (White/Caucasian or Asian / Insular of the Pacific or Black/African/African-American or
152 Hispanic/Latino/Spanish or Other).

153 We calculated average daily bicycling frequency as a measure of individual bicycling exposure.
154 Specifically, participants were asked to recall the average number of days per week in the last
155 month they used a given mode for at least 10 minutes at time to go from place to place, and then
156 how much time they usually spent on one of those days using that mode. An average daily
157 bicycling frequency measure was calculated for each participant, and outliers (n = 56) were
158 truncated to a maximum of 3 hours per day.

159 2.4 Sample

160 Overall 16,864 of 23,901 IBICCS survey participants lived within the boundaries of the six
161 cities, and these participants were assigned post-stratification weights based on comparing the
162 age and sex distribution of each city as defined by census data, to the age and sex distribution of
163 our sample within each city. We removed participants that did not have a valid Bike Lane Score
164 (n=374; weighted n = 356) and those that reported spending a combined total of greater than 16
165 hours per day using any form of transportation (n=53; weighted n=55). Of the remaining
166 participants, those who report bicycling in the past month had responses for bicycling safety
167 perceptions and were eligible to be studied (n=3,561; weighted n = 3,608). Finally, participants
168 that had missing data on sociodemographic characteristics apart from income were removed
169 (n=113; weighted n = 111). We maintained those who refused to provide income (n=243;
170 weighted n = 238). Our final sample of bicyclists included 3,446 participants, representing a
171 weighted population of 3,493.

172 2.5 Statistical Analysis

173 Weighted descriptive statistics were generated for all variables, overall and stratified by city.
174 Weighted multinomial logistic regression was used to analyze the association between bicyclists'
175 spatial access to bicycling infrastructure (as represented by Bike Lane Score) and perceptions of
176 bicycling safety (ie., safe, neutral, dangerous), controlling for sociodemographic characteristics.

177 We included a fixed effect for city which is appropriate when adjusting for clustering with small
178 numbers, allowing us to pool our sample rather than run city-specific models (Cerin, 2011).
179 Modeling was done using Proc SURVEYLOGISTIC (SAS 9.4) with link=GLOGIT to specify a
180 multinomial outcome. We applied post-stratification weights. We fit individual models for each
181 covariate, followed by an adjusted model containing all variables, with ‘Neutral’ set as the
182 reference category. Each exponentiated coefficient represents the within-city effect of a variable
183 on the odds of rating bicycling as neutral compared to safe, or dangerous compared to safe. We
184 tested all two-way interaction effects between Bike Lane Score and sociodemographic variables
185 and found no significant interactions.

186 We plotted the marginal effects of each covariate to visualize the predicted relationship between
187 each covariate and perceived bicycling safety. Here we define marginal effects as the adjusted
188 model’s prediction of perceived safety over the range of values for a specific covariate, when
189 other covariates are held to a representative value (mean or mode). These plots can be
190 conceptualized as the predicted effect of a given variable on the perceived safety of bicycling for
191 an *average* bicyclist.

192 Since infrastructure availability is a readily modifiable factor, we create a scenario to quantify
193 the predicted effect of increasing the bicycling infrastructure availability on perceptions of
194 bicycling safety. In this scenario we first defined a hypothetical sample of bicyclists where each
195 individual was characterized by a unique combination of the levels of the independent variables
196 in the adjusted model (including gender, age, bicycling frequency, income, education, having
197 children, ethnicity, and city of residence). We then used the adjusted model to predict the
198 probability that each bicyclist in this sample perceived bicycling in their city to be safe, neutral
199 and dangerous if they resided in an area with no bicycling infrastructure available within 1
200 kilometre (Bike Lane Score of 0) and also if they resided in an area with high quality bicycling
201 infrastructure available nearby (Bike Lane Score of 100). We then subtract the predicted
202 probabilities of safe, neutral and dangerous when Bike Lane Score is 0, from when Bike Lane
203 Score is 100, for each individual. This difference represents the predicted effect that increasing
204 Bike Lane Score from 0 to 100 would have on perceived bicycling safety, where a positive value
205 represents an increase in probability of a given perceived safety rating, and a negative value a
206 decrease. The absolute difference in probabilities when Bike Lane Score is 100 versus 0, is
207 repeated across all individuals in our hypothetical sample and plotted in a boxplot to visualize
208 the predicted effect of increasing the availability of bicycling infrastructure of an area. All plots
209 were created in R 3.4.1 (R Core Team, 2017).

210 3.0 Results

211 3.1 Sample Characteristics

212 Across the six cities, most bicyclists were male (61.1%), employed full-time (57.7%), had at
213 least some post-secondary education (90.9%), and did not have any children aged 17 years or
214 under (74.4%) (Table 2). A majority of bicyclists had a driver’s license (87.8%), as well as had
215 access to a motor vehicle (69.4%). Most had cycled 1-2 days per week in the previous month
216 (56.6%) (Table 2). Compared to the full IBICCS sample (bicyclists and non-bicyclists) bicyclists
217 were disproportionately male (61.1% compared to 47.7% of full sample) and under the age of 55

218 (81.0% compared to 69.6% of full sample) but otherwise were similar in sociodemographic
 219 characteristics. Overall, 22.1% report bicycling in the past month; 20.8% in Boston, 22.9% in
 220 Chicago, 23.4% in Montreal, 20.3% in New York, 22.6% in Toronto, and 22.2% in Vancouver.

221

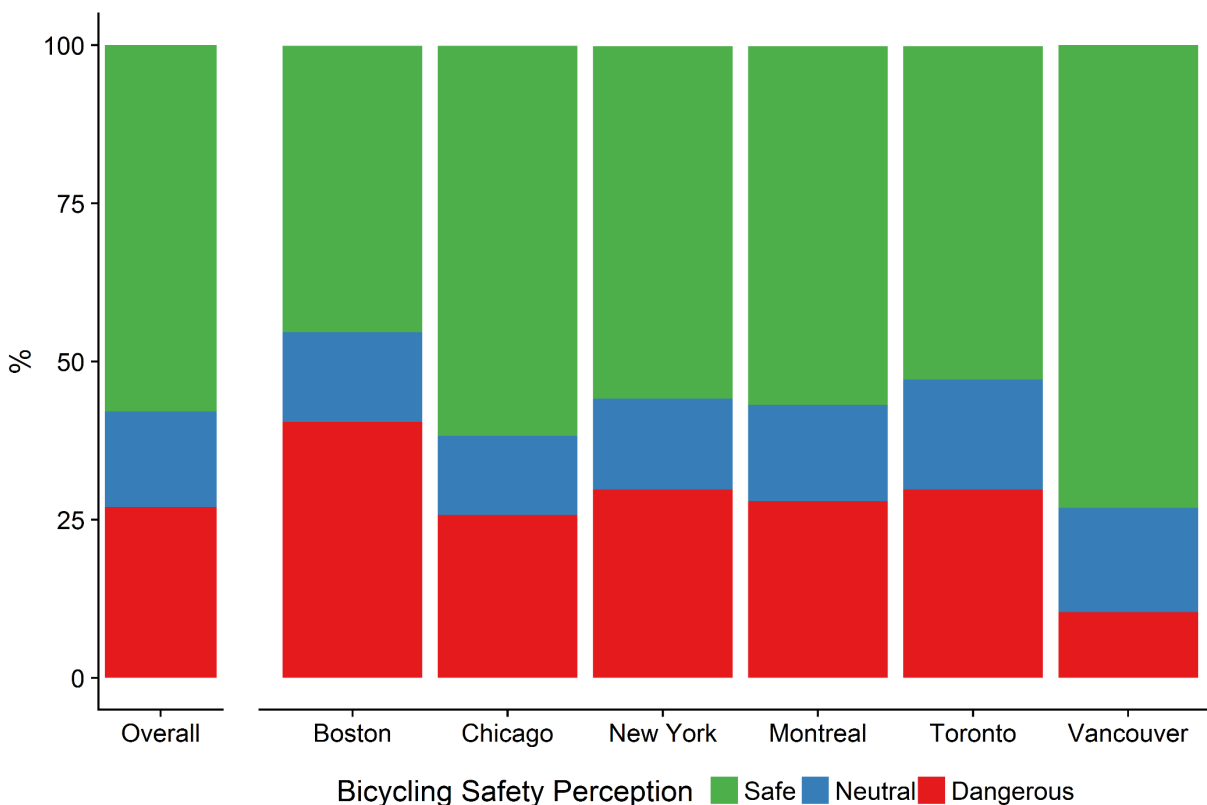
		Bost on	Chica go	New York	Montr eal	Toro nto	Vancou ver	Over all
Weighted Total (n)		222	785	715	467	887	417	3493
		% of n						
Safety Perception	Safe	45.3	61.7	55.7	56.7	52.7	73.2	57.9
	Neutral	14.2	12.5	14.4	15.2	17.4	16.5	15.1
	Dangerous	40.5	25.8	29.8	28.0	29.8	10.4	27.0
Residential Bike Lane Score	25 or under	4.4	33.2	12.4	16.1	30.6	11.1	21.5
	26 - 50	20.4	28.5	16.8	12.2	22.0	11.0	19.7
	51 - 75	11.0	22.5	20.7	18.1	16.8	19.0	19.0
	75-100	64.3	15.8	50.0	53.6	30.6	58.8	39.8
Gender	Female	45.2	37.1	40.5	39.8	38.2	36.4	38.9
	Male	54.8	62.9	59.5	60.2	61.8	63.6	61.1
Age	18-24	13.1	25.0	23.3	14.8	22.8	24.4	21.9
	25-34	14.4	19.2	24.3	30.3	22.3	23.9	22.8
	35-44	15.8	17.9	20.0	16.9	19.3	16.9	18.3
	45-54	18.1	17.2	17.3	21.9	17.6	17.1	18.0
	55 or older	38.5	20.7	15.0	16.2	18.0	17.8	19.0
Income	Under \$50 K	21.4	33.8	29.7	40.4	30.9	31.1	32.0
	\$50-100 K	33.5	35.8	34.0	37.5	33.3	37.2	35.0
	Over \$100 K	38.0	25.6	30.4	14.6	27.4	24.1	26.2
	Refuse	7.1	4.8	5.9	7.5	8.4	7.6	6.8
Ethnicity	White/Caucasian	78.3	68.5	64.6	83.5	71.2	58.0	69.8
	Black/African/African-American	7.3	12.2	9.9	1.9	2.6	0.8	6.3
	Hispanic/Latino/Spanish	4.3	9.4	11.3	2.1	2.0	5.1	6.1
	Asian / insular of the Pacific	9.3	8.8	12.1	7.4	17.6	30.9	14.2
	Other	0.9	1.1	2.1	5.0	6.7	5.1	3.7
Children Under 18	Children	17.0	23.7	29.6	23.1	26.4	28.2	25.6
	No Children	83.0	76.3	70.4	76.9	73.6	71.8	74.4
Education	Any College or University	97.2	92.0	92.2	88.9	91.2	85.2	90.9
	High School or Less	2.8	8.0	7.8	11.1	8.8	14.8	9.1
Employment	Full-Time	65.9	59.0	57.5	51.0	59.2	56.1	57.7
	Part-Time	7.6	9.5	9.4	7.1	9.9	11.8	9.4
	Self-employed	5.4	6.9	12.3	11.5	9.9	7.1	9.3
	Student	8.6	10.3	11.9	13.1	8.9	10.2	10.5
	Unemployed/Retired/Other	12.6	14.1	8.6	17.3	11.8	14.6	12.8
	Refuse	0.0	0.2	0.2	0.0	0.4	0.2	0.2
Driver's License	Yes	96.6	89.7	84.6	83.3	87.8	90.0	87.8
	No	3.4	9.8	15.2	16.6	11.9	9.6	11.9
	Refuse	0.0	0.5	0.3	0.1	0.3	0.3	0.3

Access to Motor Vehicle	Yes	77.1	76.7	54.4	63.3	70.1	82.9	69.4
	No	22.8	22.3	45.2	36.3	29.0	16.7	29.9
	Refuse	0.1	1.0	0.4	0.4	0.8	0.4	0.6
Average Days per Week Cycled in Previous Month	1 to 2	59.2	60.8	57.0	48.0	55.7	58.2	56.6
	3 to 5	30.0	31.5	36.3	44.4	33.6	33.7	34.9
	6 to 7	10.8	7.7	6.7	7.7	10.7	8.1	8.5
Average Daily Bicycling Minutes in Previous Month	15 minutes or less	49.7	48.4	43.4	48.0	47.8	45.5	46.9
	15-30 minutes	31.2	23.4	24.8	26.5	21.3	25.1	24.3
	30-60 minutes	14.4	17.0	18.1	16.6	21.4	19.4	18.4
	>60 minutes	4.8	11.3	13.7	8.9	9.4	10.0	10.4
	Mean (minutes)	23.5	28.2	31.7	27.3	29.0	28.0	28.7
	Minimum (minutes)	0.1	0.1	0.1	0.3	0.3	0.1	0.1
	Maximum (minutes)	180.	180.	180.0	180.0	180.	180.0	0.0

222 *Table 2: Weighted values for sociodemographic characteristics of the subsample of respondents*
 223 *from the IBICCS sample who reported bicycling in the previous month.*

224 A Bike Lane Score lower than 50 indicates minimal bike infrastructure was available near a
 225 bicyclists' residence (Walk Score, 2018). A considerable proportion of bicyclists (41.2%) had a
 226 Bike Lane Score lower than 50, with variability by city. Toronto and Chicago had the highest
 227 proportion of bicyclists with limited access (52.6 % and 61.7% with a Bike Lane Score lower
 228 than 50, respectively), while the remaining cities had similar proportions ranging between 22.1%
 229 and 29.2%.

230 Overall, most bicyclists perceived bicycling in their city to be safe. Specifically, 57.9% of
 231 bicyclists reported bicycling in their city as safe, 15.1% as neutral, and 27.0% as dangerous
 232 (Figure 2). A larger proportion of bicyclists in Boston perceived bicycling to be dangerous
 233 (40.5%) and Vancouver was perceived as safer (73.2%) (Figure 2).



234

235 *Figure 2: Weighted distribution of perceived bicycling safety, overall and by city.*

236 3.2 Bicycling infrastructure and perceived bicycling safety

237 Within the six cities, bicycling infrastructure availability was significantly associated with the
 238 perceived safety of bicycling in that city (Table 3). For every 10-unit increase in Bike Lane
 239 Score the odds of rating bicycling as safe compared to neutral increased by 6% (OR = 1.06, 95%
 240 confidence interval (CI): 1.02 to 1.10), and the odds of rating bicycling as dangerous compared
 241 to neutral were estimated to increase by 4% (OR = 1.04, 95% CI: 1.00 to 1.08).

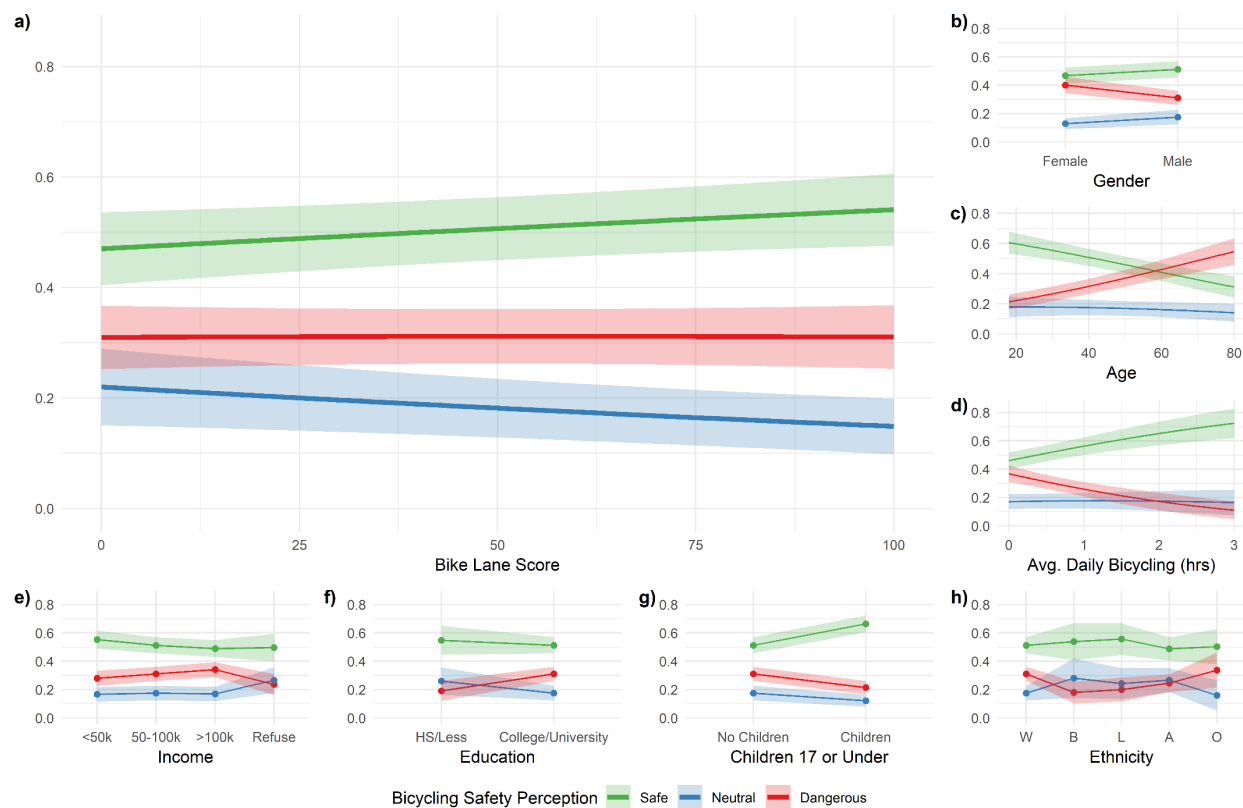
Variable	Outcome	Unadjusted OR ^a (95% CI)	Adjusted OR ^b (95% CI)
Residential Bike Lane Score			
10 unit change	Safe	1.05 (1.01, 1.09)	1.06 (1.02, 1.10)
	Neutral	Reference	
	Dangerous	1.06 (1.02, 1.11)	1.04 (1.00, 1.08)
Gender [ref=Female]			
Male	Safe	0.81 (0.63, 1.03)	0.82 (0.64, 1.04)
	Neutral	Reference	
	Dangerous	0.57 (0.43, 0.74)	0.57 (0.44, 0.74)
Age			
10 year change	Safe	0.95 (0.85, 1.05)	0.94 (0.84, 1.04)
	Neutral	Reference	
	Dangerous	1.29 (1.15, 1.44)	1.21 (1.08, 1.35)
Income [ref=<50 K]			
50k – 100 K	Safe	0.94 (0.69, 1.28)	0.88 (0.64, 1.22)
	Neutral	Reference	

	Dangerous	1.42 (1.01, 1.99)	1.07 (0.75, 1.52)
	Safe	1.01 (0.73, 1.42)	0.88 (0.62, 1.26)
>100 K	Neutral	Reference	
	Dangerous	1.84 (1.29, 2.62)	1.21 (0.82, 1.78)
	Safe	0.60 (0.38, 0.96)	0.57 (0.35, 0.93)
Refuse	Neutral	Reference	
	Dangerous	0.82 (0.50, 1.36)	0.54 (0.31, 0.92)
Ethnicity [ref = White/Caucasian]			
	Safe	0.68 (0.48, 0.97)	0.62 (0.44, 0.88)
Asian / insular of the Pacific	Neutral	Reference	
	Dangerous	0.42 (0.29, 0.62)	0.52 (0.35, 0.77)
	Safe	0.74 (0.44, 1.24)	0.66 (0.38, 1.14)
Black/African/African-American	Neutral	Reference	
	Dangerous	0.30 (0.16, 0.55)	0.36 (0.19, 0.68)
	Safe	0.92 (0.54, 1.57)	0.78 (0.47, 1.31)
Hispanic/Latino/Spanish	Neutral	Reference	
	Dangerous	0.33 (0.17, 0.63)	0.47 (0.24, 0.89)
	Safe	1.24 (0.61, 2.54)	1.08 (0.51, 2.27)
Other	Neutral	Reference	
	Dangerous	1.02 (0.46, 2.24)	1.19 (0.52, 2.72)
Have Children Under 18 [ref = None]			
	Safe	1.77 (1.34, 2.35)	1.86 (1.40, 2.48)
At least 1	Neutral	Reference	
	Dangerous	0.85 (0.63, 1.16)	0.99 (0.72, 1.36)
Education [ref = Highschool or Less]			
	Safe	1.34 (0.87, 2.04)	1.38 (0.90, 2.12)
Post-Secondary	Neutral	Reference	
	Dangerous	3.07 (1.85, 5.08)	2.43 (1.43, 4.10)
Average Daily Bicycling in Previous Month			
	Safe	1.17 (0.94, 1.47)	1.18 (0.94, 1.47)
1 hour change	Neutral	Reference	
	Dangerous	0.61 (0.46, 0.80)	0.67 (0.50, 0.90)

^a Unadjusted refers to a model consisting of the variables + City Term, ^b Adjusted refers to a model consisting of all variables + City Term; Bold: p<0.05

242 *Table 3: Results of multinomial logistic regression models estimating associations between Bike*
 243 *Lane Score (proxy for spatial access to bicycle infrastructure around one's residence),*
 244 *sociodemographic characteristics, and perceived safety of bicycling among 3493 IBICCS*
 245 *respondents reporting having cycled in the previous month.*

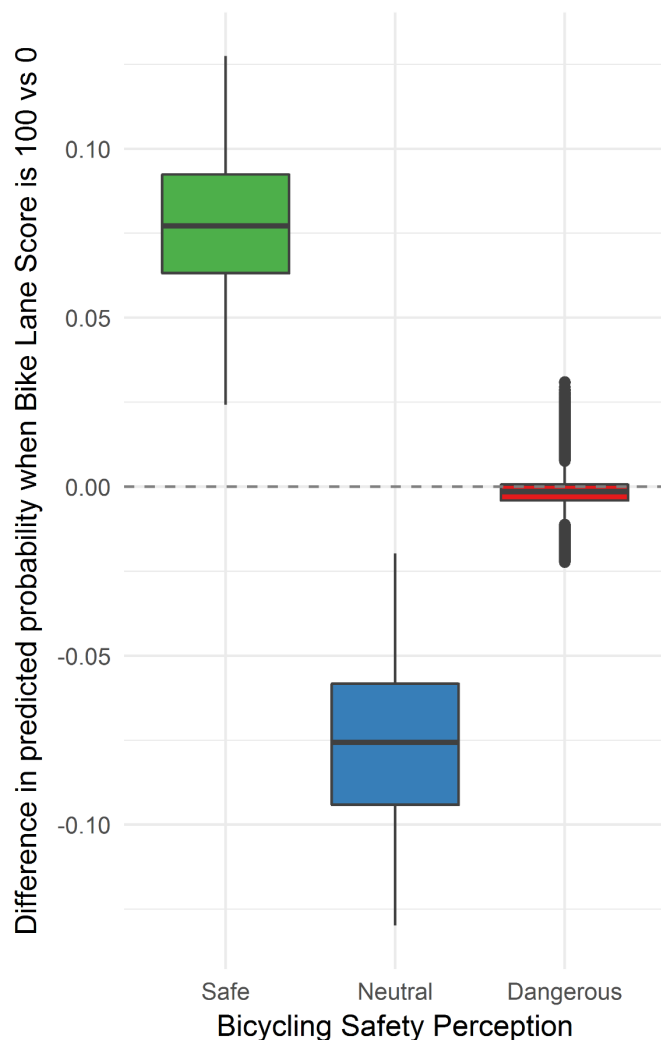
246 The marginal effects plot visualises the estimated relationship between bicycling infrastructure
 247 availability and perceived bicycling safety (Figure 3a). Model predictions indicate that with
 248 larger Bike Lane Score values, there is an increase in the predicted probability that the average
 249 bicyclist perceives bicycling as safe, a decrease in the probability they perceive bicycling to be
 250 neutral, and virtually no effect on the probability they perceive bicycling to be dangerous (Figure
 251 3a).



252

253 *Figure 3: Marginal effects plots visualize the predicted probability that a bicyclist perceives*
 254 *bicycling in their city as safe, neutral, or dangerous across different values for each variable,*
 255 *when all other independent variables are held to their mean or mode. Each plot describes how*
 256 *the perception of bicycling safety would be expected to change given a particular variable's*
 257 *value was changed for bicyclists who, in all other aspects, are average with respect to*
 258 *sociodemographic characteristics within the sample. For example, in 3b) the average female*
 259 *bicyclist is less likely to rate bicycling as safe or neutral and more likely to rate bicycling as*
 260 *dangerous, when compared to the average male bicyclist. HS: High School education; W:*
 261 *White/Caucasian; B: Black/African/African-American; H: Hispanic/Latino/Spanish; A:*
 262 *Asian/insular of the Pacific; O: Other.*

263 The predicted effect of increasing bicycling infrastructure availability for a hypothetical sample
 264 of bicyclists, such that the Bike Lane Score increases from 0 to 100, suggests that perceptions
 265 shifted from neutral to safe, but the perception of bicycling as dangerous was virtually
 266 unchanged (Figure 4). The hypothetical sample were, on average, 7.8% more likely to perceive
 267 bicycling as safe (predictions ranged between a 2.4% and 12.7% increase), 7.7% less likely to
 268 perceive bicycling as neutral (predictions ranged between a 2.0% and 13.0% decrease) and 0.1%
 269 less likely to perceive bicycling as dangerous (predictions ranged between 2.2% decrease and
 270 3.1% increase).



271
 272 *Figure 4: The distribution of the predicted change in probability of perceiving bicycling as safe,*
 273 *neutral and dangerous for a hypothetical sample of bicyclists when the bicycle infrastructure*
 274 *availability was increased such that the Bike Lane Score went from 0 to 100.*

275 3.3 Sociodemographic characteristics and perceived bicycling safety

276 Adjusted models indicate that sociodemographic characteristics are statistically associated with
 277 bicycling safety perceptions (Table 3). Bicyclists who are male, younger, lower income, have
 278 young children, have a high-school education, and bicycle more frequently are predicted to be
 279 more likely to perceive bicycling in their city to be safe (Figure 3b-d, 3f-h). Ethnic background
 280 does not appear to be associated the likelihood with perceive bicycling as safe, but rather the
 281 likelihood that bicycling is rated neutral or dangerous, with non-Caucasian bicyclists being more
 282 likely to rate bicycling as neutral compared to dangerous (Figure 3e).

283 4.0 Discussion

284 In this study, we examined the association between perceived bicycling safety and spatial access
 285 to bicycle infrastructure accounting for sociodemographic characteristics, amongst over 3,000

286 bicyclists across 6 geographically diverse major Canadian and US cities, including Boston,
287 Chicago, New York, Toronto, Montreal and Vancouver. These cities covered a wide range of
288 bicycling conditions and safety contexts with mode share ranging from 0.8% to 4.4% of
289 commuters who travel by bicycle (Statistics Canada, 2013c, 2013d, 2013b; US Census Bureau,
290 2017b), bicyclist fatality rates varying between 0.60 to 2.94 deaths per million bicycle trips
291 (Urban Systems, 2015) and the proportion of bicyclists who perceive cycling to be safe between
292 45.3% to 73.2%. In multinomial regression analyses, bicycle infrastructure was positively
293 associated with the likelihood of perceiving bicycling as safe compared to neutral. Bicyclists
294 who are male, younger, lower income, have young children, have a high-school education, and
295 bicycle are predicted to be more likely to perceive bicycling in their city to be safe. Our adjusted
296 model indicated that increasing the availability of bicycling infrastructure such that the Bike
297 Lane Score increased from 0 to 100, would result in a 7.8% average increase in the probability a
298 bicyclist perceives bicycling to be safe. These results highlight that amongst a population that
299 report bicycling, the availability of nearby bicycle infrastructure, in combination with individual
300 sociodemographic characteristics, plays a role in shaping overall perceptions of bicycling safety.

301 Our findings support the hypothesis that bicycling infrastructure around one's residence is
302 associated with more favourable perceptions of bicycling safety. The finding that greater
303 availability is associated with a greater likelihood of rating bicycling as safe is consistent with
304 previous research which suggests that at the network level, bicyclists perceive bicycle facilities
305 to be safer than bicycling in mixed traffic (Manton et al., 2016; Møller and Hels, 2008; Parkin et
306 al., 2007; Winters et al., 2012b). Bike Lane Score is a generalized metric of bicycling
307 infrastructure availability, and thus our work complements this past work investigating how
308 perceived safety varies for specific route characteristics (Manton et al., 2016; Møller and Hels,
309 2008; Parkin et al., 2007; Winters et al., 2012b). Our results show a positive relationship between
310 availability of bicycle infrastructure around one's residence and perceiving bicycling as safe in
311 one's city, but no difference in perceiving bicycling as dangerous. This result suggests that
312 increasing the availability of bicycling facilities may lead to improved safety perceptions
313 amongst those who feel neutral, but may not be effective way to improve perceptions of
314 bicycling safety for those who already consider bicycling to be dangerous.

315 Implementing bicycle infrastructure may also be a means of promoting gender equity in a
316 traditionally male dominated activity, at least in cities with low ridership levels (Aldred et al.,
317 2016a; Garrard et al., 2008). In our sample, females are underrepresented as only 38.9% of
318 bicyclists are female compared to 52.3% of participants in our full sample. The sex imbalances
319 often found in bicycling populations within low bicycling contexts are often attributed to
320 differences in risk aversion where women are more likely to perceive bicycling to be a risky
321 mode of a transport and choose not to bicycle (Garrard et al., 2008; Heinen et al., 2010). Our
322 research indicates that after controlling for confounders such as bicycling frequency, that there
323 are still significant differences in perceptions of bicycling safety between males and females who
324 already bicycle (at least once a week in the previous 30 days). Previous research has shown that
325 female bicyclists, when compared to males, are more likely to perceive bicycling on specific
326 routes to be too unsafe to bicycle on (Parkin et al., 2007), consider various situations while

327 bicycling in traffic to be more risky (Bill et al., 2015; Chataway et al., 2014; Møller and Hels,
328 2008), and report that road segments along their commute are dangerous (Manton et al., 2016).

329 In cities where bicycling prevalence is low, not only do bicyclists tend to be disproportionately
330 male, but they are also generally much younger (Aldred et al., 2016b). We find a similar pattern
331 in our sample with participants over the age of 55 years make up only 19.0% of bicyclists but
332 30.4% of our full sample. Similar to that of the gendered trends, the literature suggests that older
333 adults are underrepresented in low bicycling environments due to a higher aversion to bicycling
334 in mixed traffic, combined with more limited physical abilities (Aldred et al., 2016b, 2016a). Our
335 sample is limited only to those persons that already report bicycling, but still shows that younger
336 bicyclists are still more likely to perceive bicycling as safe, compared to older bicyclists, even
337 after controlling for bicycling regularity and the bicycling environment. Previous research has
338 found mixed results on the associations between age and perceptions of bicycling safety within
339 bicycling populations in Western Europe. Some research has indicated that older bicyclists are
340 less likely to perceive bicycling as safe compared to their younger counterparts in Ireland and
341 Denmark (Lawson et al., 2013; Møller and Hels, 2008), while others have found that older age
342 was associated with an increased likelihood of perceiving bicycling as safe in Ireland and the UK
343 (Manton et al., 2016; Parkin et al., 2007). We contribute to the literature in that we provide
344 evidence that in a Canadian and US context, older bicyclists tend to perceive bicycling to be less
345 safe than younger bicyclists.

346 Perceptions of safety may inform transportation choices (Aldred, 2016; Heinen et al., 2010), but
347 safety perceptions don't necessarily align with observed safety (Winters et al., 2012a). We were
348 not able to assess observed safety (e.g. a measure of injuries or crashes per unit of exposure), as
349 spatially and temporally resolved crash data are not readily available across municipalities. At
350 the city level, there is no correlation between the proportion of bicyclists who perceive bicycling
351 to be dangerous in our sample and city-wide bicycling fatality rates. For example, Boston and
352 Vancouver have similar bicyclist fatality rates, but Boston had the highest proportion reporting
353 bicycling to be dangerous in our study, and Vancouver the lowest. Overall perceptions of
354 bicycling safety are likely influenced not only by fatal and serious injuries that occur in a city,
355 but also by minor injuries and near miss events which occur much more frequently and can be
356 formative negative experiences (Aldred, 2016; Branion-Calles et al., 2017; Sanders, 2015).

357 Our study has various strengths and limitations. The IBICCS study provided a large sample of
358 spatially located survey data over diverse contexts in six large Canadian and US cities, and used
359 a standardized a measure of the bicycle environment, shown to be associated with bicycling
360 (Winters et al., 2016). We also use an outcome with three levels to measure perceived safety
361 (safe vs neutral vs dangerous), rather than a binary category (safe vs dangerous), which enables a
362 more nuanced understanding of the association between access to bicycle infrastructure and
363 perceived safety. Bike Lane Score is a standardized measure that captures the availability of
364 bicycle infrastructure, and was the best available proxy for bicycle infrastructure consistent
365 across cities at the time of the study. The development of Bike Score, including the weighting
366 and decay functions, were informed by research but led by a private company. Bike Score uses a
367 proprietary distance decay function to weight nearby infrastructure higher than distant

368 infrastructure, but the score does not provide the distance to nearest infrastructure, a metric is
 369 often used in research studies (Panter et al., 2016). To note, the methodology of Bike Score has
 370 recently changed, and the description in this paper differs from the Redfin website (Walk Score,
 371 2018). Motor vehicle traffic volumes may also play a role in safety perceptions (Parkin et al.,
 372 2007; Winters et al., 2012b), but standardized, spatially resolved data on traffic volumes was not
 373 available. Of note, the administrative city boundaries vary in terms of the geographic extent with
 374 some cities covering a much larger area (e.g., New York, Chicago). The IBICCS sample was an
 375 online panel survey, and both a large sample size and use of post-stratification weights on age
 376 and sex improve the generalizability of our results.

377 5.0 Conclusions

378 Our results show that within six major Canadian and US cities, greater availability of bicycle
 379 infrastructure as represented by the Bike Lane Score of a bicyclist's residence, is associated with
 380 greater odds of perceiving the broader city-wide bicycling environment as safe. We suggest that
 381 municipalities who wish to expand their bicycling network, thereby increasing spatial access to
 382 bicycle infrastructure, could see the perceived safety of bicycling in their city increase amongst
 383 bicyclists. These findings can be complemented by natural experiment studies which track
 384 changes in perceived safety as bicycle networks are expanded.

385 **Competing interests**

386 The authors have no financial or non-financial competing interests to declare.

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