

# Casino proximity, visit frequency, and gambling problems

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## 17 Abstract

18 The geospatial impact of casinos on gambling problems is poorly understood, despite its  
19 importance to policy decisions. In this study, we propose a conceptual model to describe how  
20 access relates to gambling problems and we test whether access convenience increases risk. We  
21 collect a large sample of Canadian gamblers ( $n=6,234$ ) and geolocate each individual relative to  
22 domestic casino locations ( $N=110$ ), using their home addresses. Our analysis suggests that  
23 nearby casinos increase risk for residents. We further find that frequency of play mediates the  
24 relationship, implying an indirect link between access convenience and gambling problems. The  
25 results are robust to several estimation strategies that address endogeneity issues found in the  
26 empirical literature.

27 *Keywords:* casino; gambling exposure; access convenience; travel time; adaptation;  
28 gambling disorder

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## 30 INTRODUCTION

31 The prospect of a new casino in a community is often a source of controversy for nearby  
32 residents. Economic benefits of casino expansion, including tax revenue, local employment, and  
33 tourism activity are typically welcomed by an array of stakeholders (Eadington, 1998, 1999;  
34 Henderson, 2006; Ishihara, 2017; Philander et al., 2015). Conversely, social costs are often the  
35 focus of parties opposed to gambling. Categories of perceived costs are numerous, including a  
36 loss of community values, increased risk of crime, and increased traffic congestion, but the most  
37 cited concern is typically the harms from gambling-related addiction (Ishizaka et al., 2013; Korn  
38 et al., 2003; C. K. Lee et al., 2010; C. K. Lee & Back, 2003, 2006; Philander et al., 2017; Wan,  
39 2012).

40 Some of the smallest and largest casino projects involve debates about gambling-related  
41 addiction in the local population. For example, Japan's casino resorts, which are expected to  
42 involve capital investments in excess of \$10 billion per property, have generated considerable  
43 debate about future gambling problems (Murase, 2018; Philander et al., 2017; Roberts &  
44 Johnson, 2016). Resident attitudes about casinos are closely shaped by the direct and indirect  
45 impacts of gambling addiction on the community (C. K. Lee et al., 2010; Vong, 2009; Wu &  
46 Chen, 2015) and policy makers therefore attempt to minimize the former and maximize the latter  
47 by geographically restricting property locations to island or border communities, in an attempt to  
48 concentrate consumption among out-of-jurisdiction visitors (Eadington, 1999).

49 Despite the importance of understanding casino impacts on nearby residents, the  
50 influence of access convenience on gambling problems remains an open question. There was a  
51 significant expansion of legal gambling in the late-20<sup>th</sup> and early-21<sup>st</sup> centuries, but the

52 prevalence of problem gambling in related populations has tended to remain unchanged (M. W.  
53 Abbott et al., 2014; Welte et al., 2015). This seemingly paradoxical phenomenon has led to  
54 conflicting narratives about the influence of access by residents. Some research posited that  
55 expansion of gambling increases risk for local populations (Korn, 2000b), while other research  
56 proposes that individuals will adapt to the risks after an adjustment period (LaPlante & Shaffer,  
57 2007; Prentice & Zeng, 2018). To thoughtfully develop host community plans for casinos, a  
58 more rigorous understanding of how casino access will contribute to the local populations'  
59 addiction risk is needed.

60 In the current study, we develop a conceptual model of casino patronage that helps  
61 explain the conflicting narratives about the casino access and local gambling problems. Using a  
62 large sample (n=6,234) of active gamblers from Canada, we then test our hypothesized pathway  
63 using geolocation coordinates for each of the survey respondents' homes, which is merged into a  
64 database containing the location and age of every casino in Canada. We address potential biases  
65 in our identification strategy by also providing robustness tests in the form of an instrumental  
66 variable model, which we used to test assumptions important to the underlying validity of our  
67 approach.

## 68 LITERATURE

69 There is a well-developed literature on the mechanisms by which casinos can contribute  
70 to local economies (Eadington, 1998). Empirical evidence suggests that casinos positively  
71 contributed to economic output in Macau (Zheng & Hung, 2012), Korea (C. K. Lee & Kwon,  
72 1997), Australia (T. J. Lee, 2011), and a diverse range of U.S. jurisdictions (Evans & Topoleski,  
73 2002; Walker & Jackson, 2007). Studies of resident perceptions find that locals recognize  
74 positive tourism-based impacts of casinos, but have mixed attitudes to casinos overall due to the

75 perceived social harms (Giacopassi et al., 1999; C. K. Lee et al., 2010; Nichols et al., 2002;  
76 Vong, 2009; Wu & Chen, 2015).

77         The most salient casino-related cost is problem gambling (e.g. Tan et al., 2017). Problem  
78 gambling is a behavioral addiction, characterized by continued gambling despite negative  
79 personal, social, and/or financial outcomes (Hodgins et al., 2011). Among other criteria,  
80 impacted individuals are preoccupied with gambling, have unsuccessful efforts to control or stop  
81 gambling, and return to gambling another day to recoup losses (Potenza, 2014). Individuals tend  
82 to have excessive involvement in gambling, and this may manifest as a significant amount of  
83 time and/or money spent at gambling outlets, and/or participation in many different forms of  
84 gambling (LaPlante et al., 2014).

85         According to the Pathways Model, a prominent theory about the etiology of gambling  
86 problems (Blaszczynski & Nower, 2002; Milosevic & Ledgerwood, 2010; Moon et al., 2017;  
87 Nower & Blaszczynski, 2017), increased availability and accessibility of gambling is a factor in  
88 the development of gambling related problems. It is proposed in the model that access for  
89 individuals will enable increased frequency of play, which subsequently can lead to behavioral  
90 conditioning and development of problems (Blaszczynski & Nower, 2002).

91         Empirical studies of population-wide prevalence rates fail to align with the intuition  
92 provided by the Pathways Model. In general, prevalence studies find that rates of gambling  
93 disorders tend stay level or fall after a casino is introduced in a community, despite the implied  
94 increase in availability (M. W. Abbott et al., 2014; Welte et al., 2015). Many scholars propose  
95 that this phenomenon occurs because the influence of exposure on gambling disorders can be  
96 reduced through adaptive processes, whereby individuals change their behavior to mitigate risk  
97 (M. Abbott, 2006; LaPlante & Shaffer, 2007; Prentice & Zeng, 2018). However, these proposed

98 mechanisms are propositional or explanatory. It is unclear if there is an “exposure effect” which  
99 can then be followed by individual “adaptation”. Resolving the nature of this relationship is  
100 important, as it will provide clarity as to the appropriate policy decisions that can be made about  
101 casino availability, expansion, and harm mitigation.

102         Some empirical work has attempted to provide clarity, but these studies have been  
103 affected by different endogeneity issues. In a meta-analysis of problem gambling prevalence  
104 surveys from Australia and New Zealand, Storer et al. (2009) found that higher density of  
105 electronic gambling machines is correlated with higher rates of gambling problems. Philander  
106 (2019) found similar effects in a large representative sample of Canadians, but both studies were  
107 based on cross-sectional data sets. Two noteworthy papers by Jacques and colleagues (Jacques et  
108 al., 2000; Jacques & Ladouceur, 2006) compared rates of disordered gambling from a survey  
109 conducted before and after a casino opening with rates in a similar control municipality without  
110 expansion. Both municipalities had a decrease in prevalence rates, but the control municipality  
111 had a greater drop. Although this appeared to be a useful natural experiment to inform policy,  
112 because of the length of time required for effects to emerge, the difference-in-difference  
113 approach may have been biased by the opening of a new casino less than 20 miles from the  
114 original casino, and a U.S. tribal resort-casino opened less than 75 miles away (Morrison, 2019).  
115 A bias in sample attrition may also explain the results, as gamblers with problems are more likely  
116 to leave longitudinal research studies (Wohl & Sztainert, 2011).

117         In addition to methodological challenges observed in the empirical literature, a gap in  
118 understanding relates to an absence of generalizable mechanisms that would explain the  
119 connection between the nearby presence of a casino and development of gambling problems. It

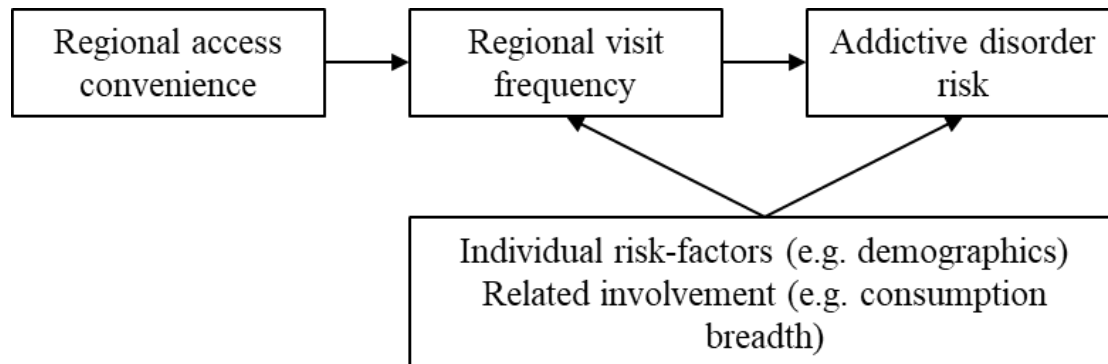
120 seems likely that a behavioral response must occur. In the Pathways Model, consumption  
121 changes are proposed to result from access changes (Nower & Blaszczynski, 2004).

122 Intuitively, frequency of visit appears to be the most likely mechanism. Increased  
123 frequency of consumption due to closer access aligns with retail management theory. That  
124 literature suggests convenience of access is a determining factor in purchase behavior  
125 (Beauchamp & Ponder, 2010; Jiang et al., 2013; Seiders et al., 2007). Berry et al., (2002)  
126 describes access convenience as consumers' perceived expenditures of time and effort to initiate  
127 consumption behavior, and this seems particularly important for inseparable services like casino  
128 gambling, which must be consumed while they are produced.

129 A growing number of observational studies suggest that gambling frequency is positively  
130 correlated with risk of developing a gambling problem (Binde et al., 2017; Gainsbury, 2015;  
131 LaPlante et al., 2011, 2014). Most applicable to our study, an examination resort casino  
132 customers found that frequency of play, rather than any game or set of games played during their  
133 casino visits, predicted a history of gambling-related problems (LaPlante et al., 2013). However,  
134 the authors note that their modeling approach is non-causal. Although the Pathways Model  
135 proposes a causal role for behaviorally conditioned gamblers who develop habits and subsequent  
136 gambling problems (Blaszczynski & Nower, 2002; Nower & Blaszczynski, 2017), reverse  
137 causality also seems plausible, considering one of the defining characteristics of gambling  
138 problems is returning to 'chase losses' on a subsequent day (Potenza, 2014).

139 Overall, we reconcile casino access literature with the retail access literature, in a  
140 conceptual model that explain observed and hypothesized phenomena (see Figure 1). We propose  
141 that increased access convenience to casino locations is associated with an increase in the  
142 prevalence of gambling disorder, and that this increase is indirect. Specifically, we hypothesize

143 that increased proximity of casino outlets is positively associated with frequency of play, which  
 144 in turn, is positively associated with gambling disorder. In addition, individual risk-factors such  
 145 as age, gender, and income levels (Johansson et al., 2009), or consumption of other forms of  
 146 gambling may influence both visit frequency and disorder risk.



147

148 *Figure 1 – Proposed model of the indirect role of access convenience in impacting addictive disorder*  
 149 *risk. The hypothesized relationship between casino access and gambling disorder risk is mediated by*  
 150 *frequency of visits to casinos. Individual factors contribute to both visit frequency and personal*  
 151 *vulnerabilities to developing a gambling disorder. Other risk-factors or measures of involvement in*  
 152 *gambling contribute visit frequency and risk of developing a gambling disorder.*

153 Although this is a straightforward linking of ideas from consumer behavior and problem  
 154 gambling etiology, a similar conceptual model has not been articulated (or tested) in the tourism  
 155 or broader gambling literatures. Validating this conceptual framework provides substantial clarity  
 156 to policy discussions. In our first hypothesis, we contend that there is behavioral response,  
 157 whereby closer casino access leads to increased frequency of consumption.

158 *H1: Frequency of casino play is positively related to increased casino access convenience*

159 We next operationalize potential for behavioral conditioning described by the Pathways

160 Model, and test for the role of frequency of play in contributing to gambling-related problems.

161 *H2: Gambling disorders are positively related to frequency of casino play*



162 We anticipate that findings provide evidence of the pathway from casino access  
163 convenience to gambling problems via frequency of play, as well as related effect sizes, which  
164 leads to the final hypothesis:

165 *H<sub>3</sub>*: Frequency of casino play mediates an indirect relationship from casino access  
166 convenience to gambling disorders

167 We hypothesized that in addition to the direct relation described in *H<sub>1</sub>* and *H<sub>2</sub>*, we would  
168 observe an indirect relation that connects casino access convenience to risk of developing a  
169 gambling disorder. We further test whether there is a direct effect of access on risk.

## 170 **EMPIRICAL APPROACH**

### 171 *Mediation model*

172 We estimated a mediation model like that shown in Figure 1, to test whether greater  
173 casino proximity predicts greater gambling disorder risk indirectly via greater visit frequency.  
174 We followed the procedure outlined by Zhao et al. (2010), and tested for the presence of an  
175 indirect-only mediation effect using a bootstrap test (Preacher & Hayes, 2008). We used average  
176 travel time from respondents' home to their nearest casino as our proxy of access convenience. In  
177 a meta-study of retail patronage, Pan & Zinkhan (2006) concluded that location convenience is  
178 an important determinant for consumers, noting that distance and drive time meaningfully impact  
179 shopping decisions. All models were estimated using the generalized structural equation model  
180 (gsem) in StataMP 15 (StataCorp LLC, 2017).

### 181 *Instrumental variable model*

182 As mentioned previously, there are potential endogeneity issues in the mediation model,  
183 as individuals at higher risk of gambling problems may choose to locate nearer casinos, leading  
184 to potential reverse causality. To further support the validity of our findings, we estimated a set

185 of instrumental variable models that provided a stronger framework for inferring causality  
186 between visit frequency and gambling disorder risk (Angrist et al., 1996; Angrist & Imbens,  
187 1995; Baiocchi et al., 2014; Rassen et al., 2009). We also used this approach to perform a  
188 Hausman test (Hausman, 2006) on our travel time variables, to support our claim that they are  
189 not directly related to gambling disorder risk and are indirectly related through visit frequency.  
190 We used a two-stage least squares (2SLS) approach that estimated the reduced form equations:

$$191 \quad (1) \quad DG = f(CF, OG, DE, \varepsilon)$$

$$192 \quad (2) \quad CF = f(T, OG, DE, \mu)$$

193 Where  $DG$  is disordered gambling risk;  $CF$  is casino gambling frequency;  $T$  is travel time  
194 to the nearest casino;  $OG$  is other gambling variables including duration of casino presence  
195 (Shaffer et al., 2004), number of other gambling formats played (LaPlante et al., 2014), highest  
196 frequency of non-casino play (LaPlante et al., 2014);  $DE$  are demographic controls, and  $\varepsilon$  and  $\mu$   
197 are respective error terms. We estimated linear and ordinal versions of equation (2). We report  
198 standard errors that account for clustering of observations by forward sortation area (discussed in  
199 the next section) because that location is implicated in the “assignment” of estimated travel time  
200 (Abadie et al., 2017). Individuals living in the same forward sortation area have the same  
201 estimated travel distance to their nearest casino.

## 202 ***Data***

### 203 *Respondents*

204 This study was determined to be exempt from review by the first author’s Institutional  
205 Review Board. We conducted a secondary analysis of a sample of active gamblers from Canada  
206 (Tabri et al., 2020). Casinos were not legal in Canada until 1985 (Korn, 2000a), but expanded to  
207 110 by the time of data collection. We exploited the variation in casino locations by geocoding

208 each respondent and each casino through postal code information, and computing travel times to  
209 each property for each respondent.

210         Since gambling disorders often afflict no more than 1-2% of the adult population at a  
211 given time, a large sample is needed (Welte et al., 2015). The data set includes 7,980 Canadians  
212 recruited by a third-party survey company in 2018 from their established online panel of 500,000  
213 people. Participants were compensated with points that were redeemable for gift cards. Roughly  
214 1,000 participants were recruited from each of the managed casino regions in Canada: British  
215 Columbia, Alberta, Saskatchewan, Manitoba, Ontario, Quebec, Nova Scotia, and a set of  
216 provinces that jointly manage some gambling-related operations (New Brunswick, Prince  
217 Edwards Island, and Newfoundland and Labrador). The firm contacted potential participants and  
218 ended recruitment once the requisite number of individuals were recruited.

219         Within each gambling jurisdiction, quota sampling was used to recruit an equal number  
220 of men and women. Age groups were sampled to be representative of each province.  
221 Participants were adults aged 18 years or older who gambled in the last year, and we subjectively  
222 determined that 80% must have gambled in the last month. Surveying only gamblers increases  
223 the power of the study, as our estimated effects are not impacted by attenuation bias from non-  
224 gamblers. In a representative survey of four Canadian provinces, 49% of individuals aged 18 or  
225 older reported gambling during the past year (*Canadian Community Health Survey - Annual*  
226 *Component Study Documentation*, 2017). Although the non-representative sample may create  
227 concerns that we are omitting part of the demand curve, the sample included non-casino  
228 gamblers. Such participants are likely to be the marginal consumers who may participate in  
229 casino gambling if a property is sufficiently close. That is, it is unlikely that a potential casino  
230 gambler is a non-gambler altogether, and as our sampling strategy only excludes non-gamblers, it

231 seems likely that our approach captures consumers on the margin. Consistent with best practices,  
232 participants who failed one or more attention check questions ( $n = 1,317$ ) were removed (Abbey  
233 & Meloy, 2017), as were those who had relevant data missing ( $n = 360$ ).

234         Of the remaining respondents, 48.4% identified as male ( $n = 3,049$ ), 51.5% identified as  
235 female ( $n = 3,248$ ), and 0.1% provided an alternative gender response ( $n = 6$ ). Participants  
236 ranged in reported age from 18 to 99 years ( $M = 47.80$ ,  $SD = 14.94$ ). Reported household income  
237 ranges were (in Canadian dollars) ‘under \$25,000’ (6.87%), ‘\$25,000 to under \$40,000’  
238 (12.47%), ‘\$40,000 to under \$60,000’ (16.02%), ‘\$60,000 to under \$80,000’ (16.48%), ‘\$80,000  
239 to under \$100,000’ (14.25%), ‘\$100,000 to under \$150,000’ (17.02%), ‘\$150,000 or more’  
240 (7.89%), and 9.00% stated they were not comfortable providing income information. In terms of  
241 highest level of completed education, 25.56% completed high school, 36.77% received some  
242 post-secondary, 14.96% completed an undergraduate degree, 21.21% completed graduate-level  
243 education. Reported marital status were ‘married’ (48.26%), ‘never married’ (21.56%), ‘married  
244 but separated’ (2.81%), ‘divorced’ (8.01%), ‘widowed’ (3.06%), ‘common-law’ (15.48%), and  
245 0.81% declined to provide a response.

246         Problem gambling severity index scores are used to measure gambling disorders (Ferris  
247 & Wynne, 2001). The problem gambling severity index is a nine-question scale used for  
248 measuring the severity of gambling problems in the general population. Scores vary from 0 to  
249 27, and there are four classification categories based on that scale (Currie et al., 2013): non-  
250 gambler/non-problem gambler (0); low-risk gambler (1-4); moderate-risk gambler (5-7); and  
251 problem gambler (8-27).

252         Respondents were asked about their participation in casino and other forms of offline and  
253 online gambling. Gambling frequency variables were measured on a seven-point ordinal scale

254 including: ‘never’ (never), ‘once a year’ (once yearly), ‘more than once a year but less than once  
 255 a month’ (less than monthly), ‘more than once a month but less than once a week’ (monthly),  
 256 ‘once a week’ (weekly), ‘most days’ (most days), and ‘everyday’ (everyday). Two variables for  
 257 gambling involvement were computed, because these behaviors are believed to be implicated in  
 258 gambling risk (LaPlante et al., 2014). The first was the highest frequency of non-casino gambling  
 259 (depth), which ranged from 1 to 7 ( $M = 3.07$ ,  $SD = 1.57$ ). The second was the count of the  
 260 number of gambling formats in which the individual participated (breadth), which ranged from 0  
 261 to 6 ( $M = 1.84$ ,  $SD = 1.57$ ). Other demographic variables were measured, including age, gender  
 262 identity, income level, educational level, and marriage status, as these have been identified as  
 263 risk factors among Canadian gamblers (Afifi et al., 2010). Summary statistics for variables used  
 264 in the analysis are provided in Table 1.

265 *Table 1: Summary Statistics*

	Count	Mean	Std. dev.	Min.	Max.
Problem gambling severity index	6,277	1.81	3.94	0	27
Casino visit frequency	6,299	2.15	1.14	1	7
Travel time (hours)	6,262	1.24	2.90	.02	15.27
Travel time <sup>2</sup>	6,262	9.93	42.16	.0005	233.04
Highest freq. non-casino (depth)	6,303	3.07	1.57	1	7
Count of gambling formats (breadth)	6,303	1.84	1.57	0	6
Log of years with casino	6,303	2.83	0.44	1.10	3.47
Age cat.	6,303	3.79	1.47	1	6
Gender cat.	6,303	1.52	0.50	1	3
Income cat.	6,303	4.46	2.01	1	8
Education cat.	6,303	3.89	1.66	1	8

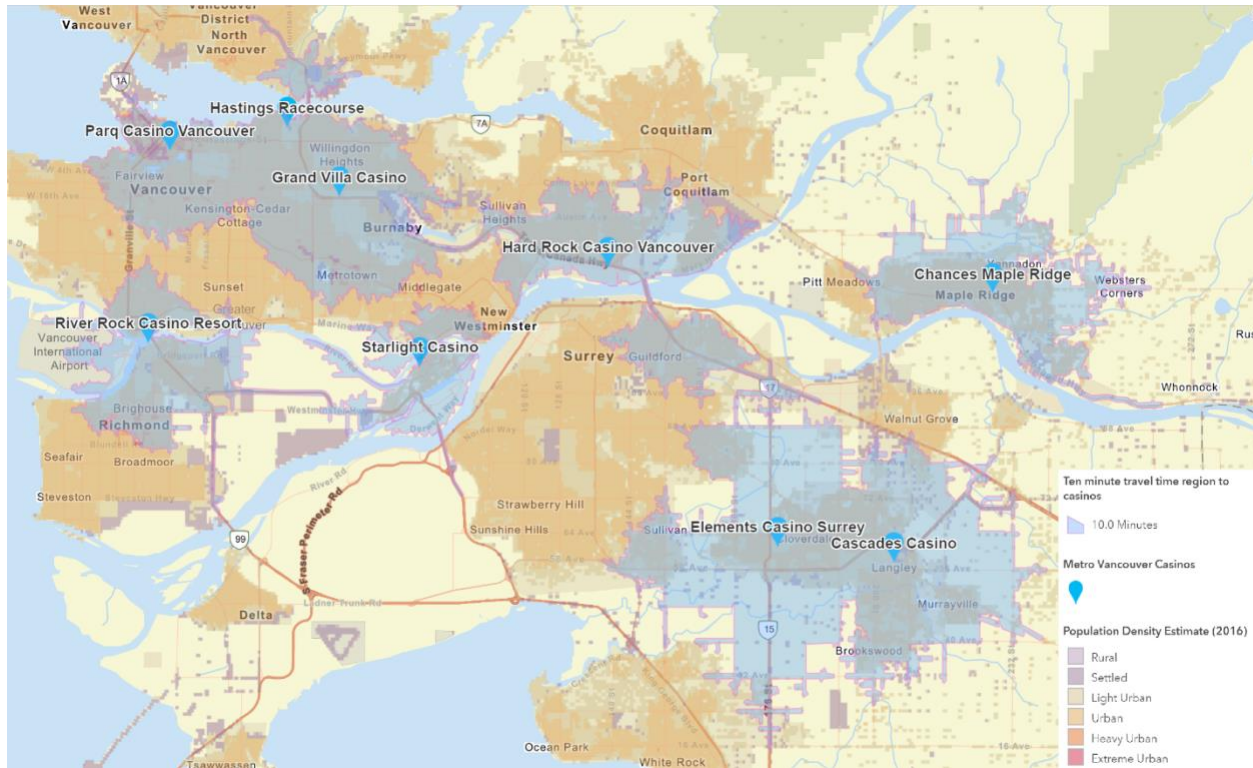
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267 *Access convenience*

268 Access convenience was estimated as the travel time between respondents’ homes and  
 269 their nearest casino. Casino location and opening date information was obtained using data from  
 270 the Alberta Gambling Research Institute (2018), provincial records (*Local Government Share of*  
 271 *Provincial Casino and Community Gaming Centre Revenue*, 2018), and individual property

272 searches. All locations were cross-validated in Google Earth (Gorelick et al., 2017). In total,  
273 there were 110 properties across Canada. Respondents were asked for the first three digits of  
274 their postal code, which is also known as the forward sortation area. Forward sortation areas are  
275 geographically contiguous areas containing an average of 8,000 households, but can include over  
276 60,000 households depending on density (Graebner, 2018). Forward sortation areas were turned  
277 into geolocations by averaging the latitudes and longitudes of all six-digit postal codes in the  
278 area.

279         Euclidian distances were computed for every casino and forward sortation area, and  
280 travel times in hours by car to the nearest casino were then computed for each respondent that  
281 provided a forward sortation area, using the *georoute* module in Stata (Weber & Péclat, 2019). In  
282 Figure 2, we illustrate an example of geographic relationships appearing in this study, using a  
283 market with multiple casinos, the Vancouver metropolitan area (*ArcGIS Online*, 2020). In that  
284 market, casinos appear in many of the high-density population areas, and we illustrate that a  
285 large proportion of the population is within a ten minute drive of one or more casinos.



286

287 *Figure 2 – Sample illustration of geographic relationships appearing in this study: Metro Vancouver*  
 288 *area casino locations, overlaid on population density plots and sub-ten minute travel time*  
 289 *regions by car (ArcGIS Online, 2020).*

290 The age of the nearest casino has also been proposed as a potentially important exposure  
 291 variable (Shaffer et al., 2004), and therefore that value was computed and converted using the  
 292 natural logarithm as a measure of the duration of exposure (log of years with casino). We model  
 293 travel time using a quadratic functional form, which is consistent with prior models of retail  
 294 demand conveying a negative first-order and positive second-order effect of distance on  
 295 patronage (Barrow & Borges, 2014; Converse, 1950; Reilly, 1932).

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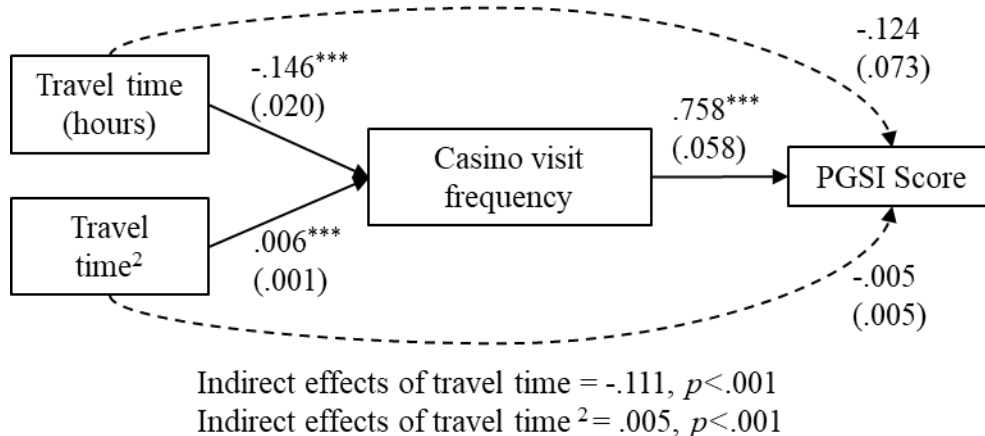
## RESULTS

### 297 *Mediation model*

298 We estimate standard errors using a bootstrap approach, but note that the use of bias-  
 299 corrected confidence intervals do not change our findings (Hayes & Scharkow, 2013). We

300 attempted 5,000 replications and all but nine converged. As shown in Figure 2, travel time has a  
 301 statistically significant effect on casino visit frequency, with a negative first-order effect,  $\beta = -$   
 302 0.146,  $z = -7.18$ ,  $p < .001$ , and a positive second-order effect,  $\beta = 0.006$ ,  $z = 5.07$ ,  $p < .001$ ,  
 303 supporting  $H_1$ . Casino visit frequency has a statistically significant effect on problem gambling  
 304 severity index scores,  $\beta = 0.758$ ,  $z = 13.20$ ,  $p < .001$ , supporting  $H_2$ . The indirect effects of the  
 305 linear,  $\beta = -0.111$ ,  $z = -6.19$ ,  $p < .001$ , and squared,  $\beta = 0.005$ ,  $z = 4.62$ ,  $p < .001$ , travel time terms  
 306 are statistically significant, supporting  $H_3$ .

307 As a robustness check, we tested whether the non-linear travel time terms varied across a  
 308 meaningful range. To do so, we calculated the marginal indirect effects of travel time, combining  
 309 both linear and squared effects of travel time at the 25<sup>th</sup> percentile = 0.202 hours, and 75<sup>th</sup>  
 310 percentile = 0.817 hours. At the 25<sup>th</sup> percentile, the marginal indirect effect of travel time was -  
 311 0.109,  $p < 0.001$ , and at the 75<sup>th</sup> percentile, the marginal indirect effect of travel time was -0.103,  
 312  $p < 0.001$ , two similar results. These estimates suggest that within common travel times, the linear  
 313 term tends to dominate effect sizes.



314

315 *Figure 3 – Mediation model results. Bootstrap standard errors from 5,000 replication attempts*  
 316 *and 4,991 replications shown in brackets; dotted lines indicate non-significant effects; PGSI =*  
 317 *problem gambling severity index.*

318 \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

319



320           The significant indirect effect estimate in our model implies that frequency mediates the  
321 impact of access on problem gambling. Some authors would suggest that the absence of a direct  
322 effect is consistent with a fully mediated effect (Zhao et al., 2010), although others suggest that  
323 focus should remain on magnitude and significance of indirect effects (Rucker et al., 2011). In  
324 either case, our findings are consistent with our proposed mediation model. The related Hausman  
325 test following the process outline by Antonakis et al. (2010) required the exclusion of the  
326 demographic variables in order to enable the model estimator to converge, but despite that  
327 restriction, still supports the mediator as being exogenous:  $\text{cov}(u_{\text{frequency}}, e_{\text{pgsi}}) = -.317, p=0.073$ .  
328 A Hausman test using a linear regression model and using all variables also supports exogeneity:  
329  $\chi^2(40) = 5.75, p=1.000$ ). Despite this empirical evidence, we relaxed the assumption of  
330 exogeneity in the next section as a robustness test.

### 331 *Instrumental variable model*

332           To provide a baseline understanding of effects, Table 2 first presents a series of ordinary  
333 least squares (OLS) models estimating the impacts of casino visit frequency on problem  
334 gambling severity index scores. Models 1-6 all show relatively stable effect sizes, estimating that  
335 a one unit increase in visit frequency along our 7-point scale is related to an average increase of  
336 0.7 to 1.1 in problem gambling severity index score. As we noted, it is possible that these OLS  
337 effect sizes are endogenous, due to reverse causality or omitted variable bias. The estimated  
338 effects of other variables are as expected. Respondents' depth and breadth of gambling are both  
339 positively related to problem gambling severity index scores, with large effect sizes. For  
340 example, according to our estimate in Model 3, a person that participates in all six forms of  
341 gambling that compose our breadth variable would be projected to have a 3-point higher problem  
342 gambling severity index score on average than a person with no participation. The inclusion of

343 demographic and province of residence control variables is indicated. Their addition causes our  
 344 casino age variable to cease being statistically significant. As expected, our travel time variables  
 345 (Models 4-6) were not statistically significant predictors of problem gambling severity index  
 346 score.

347 *Table 2: OLS models (Dependent variable: Problem gambling severity index score)*

	(1)	(2)	(3)	(4)	(5)	(6)
Casino visit freq.	1.082*** (0.059)	0.714*** (0.058)	0.757*** (0.061)	1.107*** (0.060)	0.694*** (0.061)	0.758*** (0.061)
Depth		0.330*** (0.043)	0.310*** (0.043)		0.331*** (0.043)	0.316*** (0.043)
Breadth		0.497*** (0.050)	0.456*** (0.052)		0.516*** (0.052)	0.455*** (0.053)
Log casino age		0.344*** (0.096)	0.193 (0.099)		0.385*** (0.098)	0.160 (0.100)
Travel time (hours)				0.068 (0.078)	-0.013 (0.069)	-0.124 (0.068)
Travel time <sup>2</sup>				-0.002 (0.005)	-0.001 (0.005)	0.005 (0.006)
Province	No	No	Yes	No	No	Yes
Sex	No	No	Yes	No	No	Yes
Age	No	No	Yes	No	No	Yes
Income	No	No	Yes	No	No	Yes
Education	No	No	Yes	No	No	Yes
Marital status	No	No	Yes	No	No	Yes
Observations	6,274	6,274	6,274	6,234	6,234	6,234
Adjusted R <sup>2</sup>	0.098	0.176	0.207	0.099	0.177	0.207

348 Robust standard errors in parentheses

349 \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

350 In Table 3, we provide reduced-form models using a 2SLS approach. Our instruments  
 351 appear to be valid. Both travel time variables were significant in all three models, showing first  
 352 and second order effects. The joint weak identification  $F$ -statistics are all above recommended  
 353 values of 10 (Baum et al., 2018; Staiger & Stock, 1997) and also exceed the computed 10%  
 354 maximal IV size (19.93) of weak instruments (Stock & Yogo, 2005). These results support  $H_1$ ,  
 355 suggesting frequency of casino play is positively related to convenience. Also, the Hansen-J tests  
 356 support the use of our instruments, failing to reject their exclusion from the second-stage

357 equation and further supporting the results of our Hausman tests (Baum et al., 2018; Hansen,  
358 1982).

359 Importantly, the Hansen-J tests imply that casino proximity did not exert a direct effect  
360 on problem gambling severity index scores in our sample, but more likely influenced gambling  
361 risk indirectly, through frequency of visit. The estimate of the effect of casino visit frequency in  
362 our full model (3) was larger than the equivalent OLS model, but both support *H2*. The fully  
363 specified 2SLS model suggests a one unit increase in reported casino frequency was associated  
364 with a 1.6-point increase in problem gambling severity index score.

365 *Table 3: Two-stage least squares models (First stage dependent variable: Casino visit frequency; second*  
366 *stage dependent variable: Problem gambling severity index score)*

	Model 1		Model 2		Model 3	
	Casino Freq.	PGSI Score	Casino Freq.	PGSI Score	Casino Freq.	PGSI Score
Travel time (hours)	-0.238*** (0.028)		-0.241*** (0.022)		-0.146*** (0.022)	
Travel time <sup>2</sup>	0.012*** (0.002)		0.011*** (0.002)		0.006*** (0.001)	
Casino visit freq.		0.598** (0.230)		1.003*** (0.169)		1.620*** (0.472)
Depth			-0.015 (0.011)	0.337*** (0.044)	-0.015 (0.011)	0.329*** (0.045)
Breadth			0.308*** (0.011)	0.418*** (0.066)	0.312*** (0.011)	0.186 (0.159)
Log casino age			0.037 (0.034)	0.350*** (0.097)	0.064* (0.031)	0.105 (0.109)
Province	No	No	No	No	Yes	Yes
Sex	No	No	No	No	Yes	Yes
Age	No	No	No	No	Yes	Yes
Income	No	No	No	No	Yes	Yes
Education	No	No	No	No	Yes	Yes
Marital status	No	No	No	No	Yes	Yes
Weak ID. (F-stat)		126.89		241.68		22.74
Hansen J-stat		0.81		1.10		0.00

Hansen J p-value		0.37		0.29		0.94
Adj. R-Squared		0.08		0.17		0.16
Observations	6,234	6,234	6,234	6,234	6,234	6,234

367 Robust standard errors in parentheses; PGSI = problem gambling severity index  
 368 \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$   
 369

370 Although the models in Table 3 provide several diagnostic tests about the suitability of  
 371 our instruments, they assume that the casino visit frequency variable is linear. In Table 4, we  
 372 relaxed that assumption, and estimated an ordered probit model in the first stage. Ordered probit  
 373 models fit onto regression models with ordinal dependent variables. Larger fitted values  
 374 correspond to higher outcomes. The results were consistent with our other findings. Travel time  
 375 decreased frequency of visits at a decreasing rate, and higher visit frequencies were associated  
 376 with higher risk of a gambling disorder.<sup>1</sup>

377 *Table 4: Linear regression model with ordered endogenous variable (First stage dependent variable:*  
 378 *Casino visit frequency; second stage dependent variable: Problem gambling severity index score)*

	Model 1		Model 2		Model 3	
	Casino Freq.	PGSI Score	Casino Freq.	PGSI Score	Casino Freq.	PGSI Score
Travel time (hours)	-0.253*** (0.033)		-0.285*** (0.031)		-0.186*** (0.028)	
Travel time <sup>2</sup>	0.012*** (0.002)		0.013*** (0.002)		0.008*** (0.002)	
Casino visit freq. (Base: Never)						
Once yearly		0.479** (0.164)		0.531*** (0.152)		0.347** (0.133)
Less than monthly		0.842*** (0.228)		1.114*** (0.220)		0.876*** (0.177)
Monthly		3.418*** (0.394)		3.398*** (0.373)		2.985*** (0.331)
Weekly		4.967*** (0.620)		5.030*** (0.599)		4.400*** (0.565)

<sup>1</sup> We thank a reviewer for identifying that one of the PGSI questions is related to increased visitation and may produce some recursive results, “did you go back another day to try to in back the money you lost?” We re-estimated Table 4 without that question contributing to the PGSI score and produced similar results (not shown but available upon request).

Most days/Everyday			9.309*** (1.419)	9.291*** (1.362)	8.262*** (1.320)	
Depth			-0.014 (0.012)	0.282*** (0.043)	-0.012 (0.013)	0.265*** (0.042)
Breadth			0.314*** (0.011)	0.433*** (0.054)	0.323*** (0.012)	0.435*** (0.054)
Log casino age			0.023 (0.037)	0.333*** (0.096)	0.061 (0.035)	0.173 (0.099)
Province	No	No	No	No	Yes	Yes
Sex	No	No	No	No	Yes	Yes
Age	No	No	No	No	Yes	Yes
Income	No	No	No	No	Yes	Yes
Education	No	No	No	No	Yes	Yes
Marital status	No	No	No	No	Yes	Yes
Observations	6,234	6,234	6,234	6,234	6,234	6,234

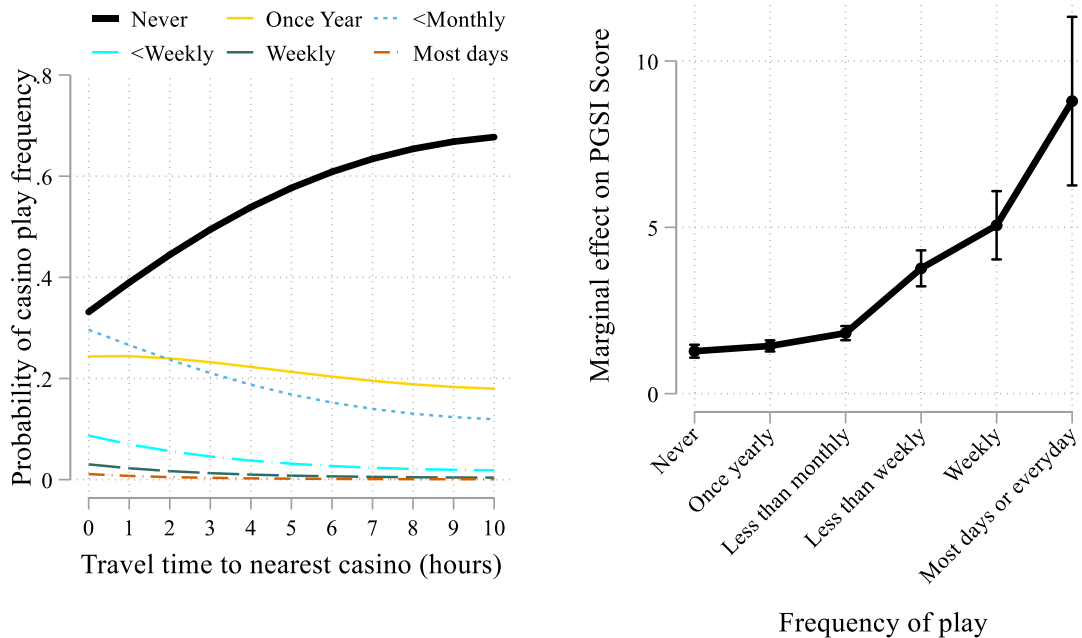
Robust standard errors in parentheses; PGSI = problem gambling severity index

\*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

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To illustrate the risk from higher visitation, we plotted the marginal effects of travel time on visit frequency, and the influence of visit frequency on problem gambling severity index scores (see Figure 4). As predicted, increased travel time was related to reductions in frequency of play, while gambling disorder risk and standard errors increased with higher visitation. Risk increased more quickly above a monthly frequency. At a frequency of ‘most days/everyday’, estimated problem gambling severity index scores were roughly eight points higher than with no casino gambling. An eight point score is a meaningful difference, as it is sufficient to be classified as a ‘problem gambler’ on the index (Ferris & Wynne, 2001).

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 391 *Figure 4 – Estimated impact of travel time on reported casino visit frequency category (left –*  
 392 *“Probability of casino play frequency”) and estimated impact of casino visit frequency on problem*  
 393 *gambling severity index scores, with vertical bars indicating standard errors (right – “Marginal effect on*  
 394 *PGSI Score”). Increases in travel time lead to a higher probability of reporting ‘never’ as a casino play*  
 395 *frequency (left). Higher frequency of play leads to higher PGSI score estimates, with notable rises from*  
 396 *‘less than monthly’ to ‘less than weekly’ and ‘weekly’ to ‘Most days or everyday’.*

## 397 DISCUSSION

398 Casino gambling can make substantial contributions to regional economies by drawing in  
 399 visitors, but negative perceptions of casinos are shaped by potential increases in local gambling  
 400 problems (Eadington, 1998, 1999). Confounding these worthwhile debates is a set of empirical  
 401 findings on the impact of increased access convenience on problem gambling, whose mixed  
 402 results are generally unreliable due to methodological limitations. This study used a novel  
 403 empirical approach to demonstrate reliable relations among casino proximity, casino gambling,  
 404 and problem gambling risk levels, in a large data set from multiple Canadian jurisdictions. We  
 405 observed that after controlling for other gambling behavior, demographics, and selection effects,  
 406 risk of gambling problems increases for individuals who live closer to casinos. This increase in

407 risk was mediated by frequency of visit, and the results further suggest that access convenience  
408 does not exert a direct effect on risk.

409         These findings provide substantially improved clarity to a policy area with frequent  
410 disagreement. Our identification strategy enabled us to provide a stronger argument of a link  
411 between high frequency casino gambling and risk of developing a gambling disorder. Because of  
412 the challenges posed by uncontrollable covariates, past efforts to establish such relations have  
413 been susceptible to statistical bias, and even natural experiments are vulnerable to bias due to the  
414 complexity of contributing environmental factors to gambling disorders. By examining responses  
415 from gamblers using an instrumental variable approach, we were able to estimate effects with  
416 less of a concern about confounding variables. Although we use a cross-sectional survey to  
417 collect respondent data, our variable of interest (travel time) is not identified through subjective  
418 survey responses, but rather location data that has had an impact that precedes our collected  
419 responses.

420         Noteworthy in interpreting these findings is that gambling disorder prevalence rates in  
421 Canada fell between 2002 and 2014, despite the expansion of casino gambling locations  
422 (Philander, 2019). Accordingly, there is reason to believe that prevention and adaptation  
423 strategies by individuals and organizations, were effective at reducing population-level  
424 prevalence of gambling addiction. Casinos may increase risk to nearby residents, but individual,  
425 public, or industrial mitigation strategies to reduce disorder prevalence may have had a greater  
426 aggregate effect. Our observation that risk rises more sharply above monthly frequencies of play  
427 may assist managers and public health advocates in targeting interventions that can reduce the  
428 probability of developing a gambling disorder by focusing interventions on higher frequency  
429 players. For example, many casinos have introduced entry fees for local residents (Philander,

430 2017), and the schedule of payments for entrance can be tiered around frequency levels noted to  
431 be related to higher risks.

432 More broadly, our findings provide an empirical basis for considering alternative location  
433 policies in the placement of casinos. Placing casinos in “tourism zones” that have relatively few  
434 nearby residents may be optimal, depending on policymaker goals. Where residents are  
435 impacted, regional exposure effects should be weighed against positive social outcomes (e.g.  
436 jobs and development) and policymakers should consider interventions that can mitigate harms  
437 from excessive consumption. For instance, strategies such as in-venue responsible gambling  
438 messaging, limit setting features, and employee interventions have all shown some merits in  
439 reducing harms (Ladouceur et al., 2017; Tanner et al., 2017).

#### 440 *Limitations and Future Work*

441 The current research examined a large sample of individuals who gambled, and all were  
442 residents of the same country. The findings may not be generalizable to other cultures, and the  
443 findings do not provide intuition about how non-gamblers may convert to casino gambling.  
444 Similar work in other regions and with other goods would add to our broader understanding of  
445 access convenience and public health risks. Future work should also consider the policy  
446 implications of our findings more thoroughly. For example, research examining how to manage  
447 risks by developing interventions that reduce access convenience for high frequency groups, or  
448 research that assesses differentiated impacts to specific demographic groups.

449 There is ongoing opportunity to explore with greater detail how access convenience is  
450 implicated in local risk. For example, it remains unclear if population-wide adaptation is related  
451 to spontaneous recovery by individuals, or whether specific policies and programs meaningfully  
452 contribute to protection efforts. Like many jurisdictions, regulated Canadian casino, lottery, and



453 online operators spend between 0.4% and 1.6% of their gross revenue from gambling on  
454 responsible gambling programs (Canadian Partnership For Responsible Gambling & Responsible  
455 Gambling Council, 2018), but it is unknown how this leads to adaptive effects, if at all  
456 (Ladouceur et al., 2017).

457         There may be proximity related effects that are unrelated to access convenience. For  
458 instance, increased saliency of the gambling product may occur if the casino venue is in the  
459 individual's neighbourhood, which may prompt increase visitation that is not directly related to  
460 ease of travel. Future studies could accommodate these effects with instruments that capture  
461 salience-related interventions, such as mail advertisements that are sent to specific postal codes.

#### 462 ***Conclusion***

463         Our findings about the mechanism implicated in problem gambling risk provide clarity to  
464 an area of the literature that has received substantial debate, and that has important policy  
465 implications for casino stakeholders. Casinos can have substantial positive impacts on local  
466 economies by way of tourism, but as managers, policymakers, and other stakeholders consider  
467 access in their communities, they should examine how residents' travel times will change,  
468 particularly among those that may have other risk-factors. To offset potential harms from  
469 increases in gambling frequency, policymakers should consider interventions, including  
470 prevention and treatment programs, if casinos are placed in close proximity to their constituents.

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