# An exploratory analysis of changes in trip-making frequency by mode during the COVID-19 emergency in Bangladesh

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# Abstract

The COVID-19 pandemic has had a profound impact on mobility in every country and region around the world. Recent research provides illuminates the nature and magnitude of the changes in mobility, but the evidence is still scant in developing countries. The objective of this paper is to present an exploratory analysis of the changes in the frequency of trip-making by mode during the COVID-19 emergency in Bangladesh. The results confirm an overall loss of mobility in the form of reduced trip-making frequency by all modes, but the changes are not uniform across modes.

## Research Questions and Hypotheses

The spread of the COVID-19 pandemic has led to limitations to movement in many countries and regions, either because of lock-down policies or self-censoring by segments of the public. The magnitude of changes in mobility has been studied by recent research, including DeWeese et al. (2020) and Molloy et al. (2020). While the evidence available indicates that overall there was a reduction in mobility in much of the world, the changes were uneven depending on the mode of transportation or the purpose of the trip (see Lock, 2020; Paez, 2020). Alas, with few exceptions evidence remains more spotty for developing countries, most of which have large populations segments that are less able to absorb losses in mobility (e.g., Astroza et al., 2020; Huynh, 2020; Saha et al., 2020).

The objective of this paper is to investigate changes in the trip-making frequency by different modes of transportation during the COVID-19 emergency in Bangladesh. Using data from a recent survey that asked respondents to report mobility levels before and during the pandemic, we pose the following questions:

- Was there a reduction of mobility in Bangladesh during COVID-19?
- And if so, what forms of transportation were more affected?

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This paper is a reproducible research document (see Brunsdon and Comber, 2020); the code and data necessary to replicate the tables and figures are available in a public repository<sup>1</sup>

### Methods and Data

Data used for this paper come from a survey conducted during **DATES** in **regions/cities** in Bangladesh. **BRIEFLY ADD DETAILS ABOUT SURVEY**.

The survey asked respondents to self-report their trip-making frequency by eight modes of transportation, namely **car**, **ridesharing** (e.g., Uber, Pathao), **rickshaw**, **cng auto-rickshaw** (a rickshaw-like vehicle powered by compressed natural gas), **bus**, motorcycle/scooter (hereafter just **motorcycle**), **walking**, and **bicycle** (there was an additional catch-all category **other** which we ignore here). Participants in the survey used the following levels to report their frequency of traveling by each mode both before and during COVID-19: *Never*, *Rarely*, *Once a week*, 2-3 trips per week, 4-5 trips per week, Almost daily. There are n = 800 responses in the data set.

To describe changes in the frequency of travel by mode in the transition to the pandemic, we use well-established exploratory data analysis (EDA) techniques.

# **Findings**

Figure 1 shows the number of responses (out of 800) in each trip-making frequency class by mode of transportation. The white bars and gray bars are for travel before and during the pandemic, respectively. Considering travel before the pandemic, travel by rickshaw, bus were relatively common for many respondents (few respondents reported never using these modes). The mode most commonly used on a quotidian basis was walk. In contrast, respondents reported less frequent travel by car, rideshare services, cng auto-rickshaw, motorcycle, and bicycle. During the pandemic we see that while there were reductions in mobility by car, motorcycle, and bicycle (with more respondents reporting never traveling by these modes), the changes were relatively minor. The frequency of trip-making by other modes changed more noticeably: the frequency of travel by rideshare services, rickshaw, cng auto-rickshaw, and bus collapsed, with vastly more respondents reporting never using these modes during the pandemic than before. The frequency of walking trips also decreased (fewer respondents report walking almost daily), but the reductions in mobility were not so heavily concentrated at the bottom of the scale.

Table 1 is a cross-tabulation of the number of cases in each trip-making frequency class before and during the pandemic. If no changes had occurred, all values would be concentrated on the main diagonal of the matrices. Values in the lower triangular matrix represent a *loss* of mobility (lower travel frequency),

 $<sup>^{1}</sup>$  place.holder

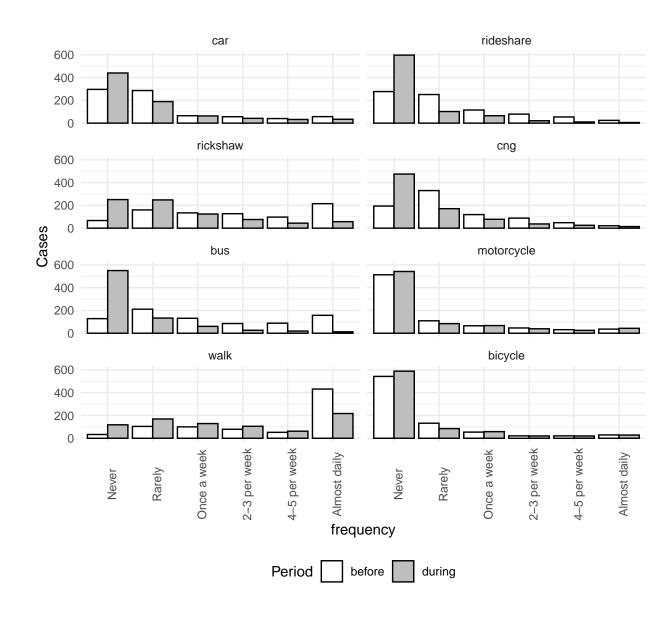


Figure 1: Number of responses by trip-making frequency class and mode, before and during COVID-19  $\,$ 

whereas values in the upper triangular matrix are *gains* (higher travel frequency). The further away a value is from the main diagonal, the greater the loss or gain.

Despite across-the-board losses of mobility, there appears to have been some adaptation that varied by mode of transportation. To illustrate, 103 respondents, or 65.61% of those who traveled by bus almost daily before, reported never using it during the pandemic. In contrast, only 12 respondents, or 1.5% of those who never used buses before started doing so during the pandemic. By way of comparison, 24.14% of respondents who cycled almost daily before the pandemic stopped doing so - but 4.88% who never cycled before started doing so during the pandemic.

To more easily understand the transitions towards different trip-making frequencies, including possible adaptations by mode, we convert the cross-tabulations to probability transition matrices, which we then visualize using circular plots.

Figures 2 to 5 present these plots. Each of the trip-making frequency sectors on the left hemisphere of the circle represent 100% of responses before the pandemic. The size of the links is proportional to the probability  $p_{ij}$  of transitioning from frequency class i before to the right hemisphere are proportional to the transition probabilities to each frequency during the pandemic. There are three transparency levels for the links: solid colors are for  $p_{ij} > 2/3$ , intermediate transparency is for  $1/3 < p_{ij} \le 2/3$ , and the more transparent links are for  $p_{ij} \le 1/3$ .

From Figure 2 we see that the probability of traveling less frequently for those who frequently traveled by car before is high, but their probability of not using this mode at all during the pandemic is quite small. The probability of traveling more frequently by car for those who originally never or rarely used this mode remained is low. In contrast, we see that the probabilities of never ridesharing during the pandemic are high irrespective of the initial level of use of this mode of transportation.

The probabilities of change in trip frequency by rickshaw and and cng autorickshaw are similar (see Figure 3), although the probabilities of being less mobile by cng autorickshaw are greater: between 1/3 and 2/3 of respondents who used this mode almost daily, stopped using it during the pandemic. Very rarely there was an increase in mobility by these modes.

After ridesharers, bus users (Figure 4) were the most likely to stop using it, irrespective of their initial level of use of this mode, while the probabilities of using it *more* frequently are extremely small.

# References

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 $Table \ 1: \quad Cross-tabulation \ of \ counts \ of \ travel \ frequency \ by \ mode, \ before \ and \ during \ COVID-19$ 

	Never	Rarely	Once a week	2-3 per week	4-5 per week	Almost daily
car						
Never	267	20	5	1	1	5
Rarely	142	99	21	12	9	
Once a week	11	23	19	7	2	;
2-3 per week	10	23	7	6	8	4
4-5 per week	4	9	5	6	12	
Almost daily	6	15	6	10	0	20
rideshare						
Never	254	13	6	4	0	
Rarely	192	44	12	1	1	
Once a week	60	20	31	3	1	
2-3 per week	49	12	9	7	2	
4-5 per week	28	10	3	5	6	
Almost daily	14	3	4	1	0	
rickshaw						
Never	51	9	6	0	1	
Rarely	64	70	18	4	3	
Once a week	36	38	45	9	3	
2-3 per week	34	36	18	29	6	
4-5 per week	21	30	7	16	15	
Almost daily	45	65	30	18	16	4
v		0.0	30	10	10	4
cng auto-ricks						
Never	176	14	3	1	0	
Rarely	213	95	14	4	1	
Once a week	36	31	38	9	2	
2-3 per week	29	18	16	14	8	
4-5 per week	13	6	5	9	14	
Almost daily	8	7	2	0	0	
ous						
Never	116	9	3	0	0	
Rarely	155	46	9	0	0	
Once a week	73	32	22	4	0	
2-3 per week	43	19	11	7	5	
4-5 per week	60	6	5	4	12	
Almost daily	103	21	10	11	2	1
motorcycle						
Never	467	22	12	3	1	
Rarely	44	38	13	10	0	
Once a week	11	11	34	3	3	
2-3 per week	13	6	6	15	4	
4-5 per week	6	4	0	4	14	
Almost daily	1	3	2	4	3	2
	1	9	2	4	3	2
walk		_			0	
Never	17	7	3	4	0	
Rarely	26	53	13	6	2	
Once a week	12	19	54	9	4	
2-3 per week	10	14	12	26	13	
4-5 per week	7	3	7	11	16	4.0
Almost daily	46	73	40	49	27	19
oicycle						
Never	504	22	7	4	2	
Rarely	59	43	20	4	2	
Once a week	14	13	21	3	2	
2-3 per week	3	4	55	7	2	
4-5 per week	$^{2}$	2	4	1	9	
Almost daily	7	1	1	1	3	1

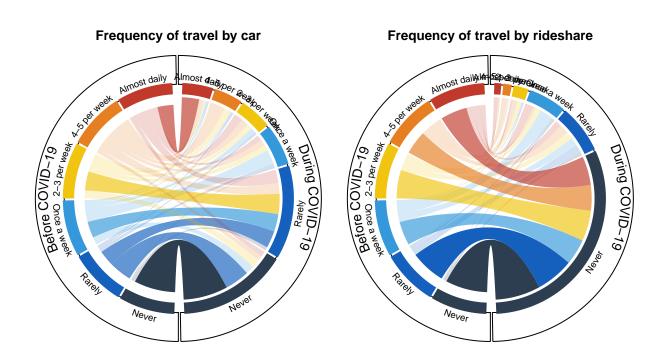


Figure 2: Transition probabilities in trip frequency from before to during COVID-19: car and rideshare

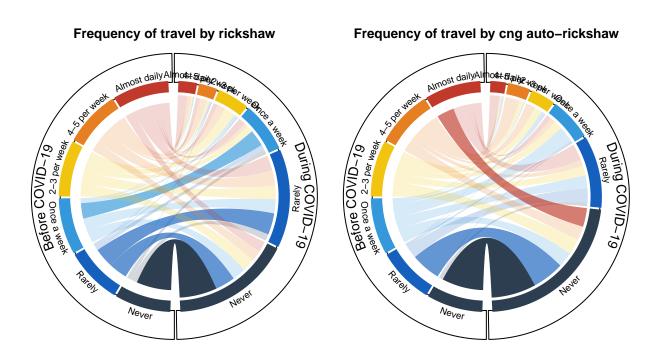


Figure 3: Transition probabilities in trip frequency from before to during COVID-19: rickshaw and cng auto-rickshaw

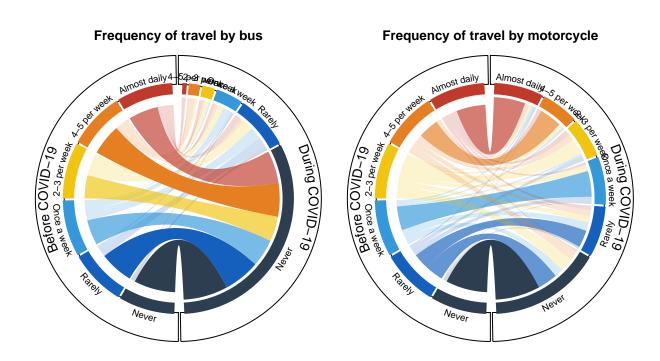


Figure 4: Transition probabilities in trip frequency from before to during COVID-19: motorcycle and bus

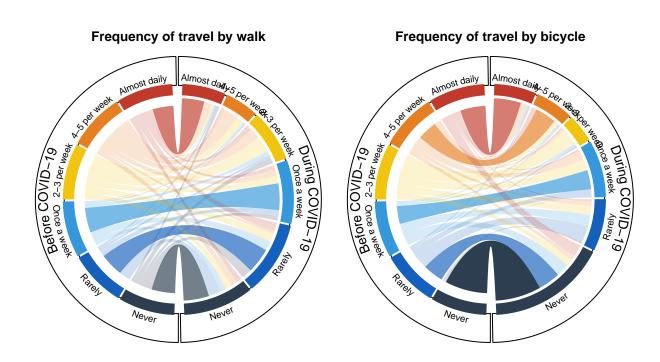


Figure 5: Transition probabilities in trip frequency from before to during COVID-19: bicycle and walk

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