



Paper: 43

Investigating the Effects of Individual and City Tier Characteristics on Motorized Two-Wheeler Usage Behaviour: A Multilevel Modelling Approach



Mr. Aitichya Chandra Ms. Hemanthini Allirani Prof. Ashish Verma IISc Sustainable Transportation Lab. (IST Lab.) Department of Civil Engineering, Indian Institute of Science (IISc) Bangalore, Karnataka, India





Background & Introduction

Aim & Research Questions

Data Description

Framework: k-means Clustering

Framework: Multilevel Modelling

Results and Discussions

Implications and Conclusions



Background & Introduction



- Two-wheeler in India -
 - Most popular transport mode across India
 - Maximum share of all motorized vehicles in production (79.8%) and sales (80.8%)
 - 49.7% of the households in India
 - Highest households ownership in Punjab (75.6%), Rajasthan (66.4%, and Tamil Nadu (63.9%)
- Reasons for Popularity -
 - Easy navigation capability through city traffic
 - Low purchase and maintenance cost
 - Convenience, less space requirement, and reduced travel time
 - Suitability to heterogeneous traffic conditions
 - Preferred choice for female and/or middle-income groups
 - Increased urban sprawl
 - Lack of Public Transit services







Background & Introduction

IISc Sustainable Transportation Lab (IST Lab)



- Share of trips is high in tier-2 cities compared to tier-1 cities
- Average distance travelled higher in tier-1 and tier-2 cities
- Parking for a short duration in tier-2 cities
- Traditional Two-wheeler Usage Policies -
 - Common for all tiers
 - No account of two-wheeler usage variation across city tiers
 - Tier classification based on population composition
- Understand:
 - The motorized two-wheeler usage aspects of individuals
 - Variation among city-tiers







Aim & Research Questions





To investigate the effect of individual and city tier characteristics on motorized two-wheeler



What are the different tiers of cities and can we re-define them?



Is there a variation in motorized two-wheeler usage characteristics among the different tiers of cities?



How much is the variation, and which variables are causing the variation?



What are the resulting policy implications?



Data Description







Data Description



Motorized Two-Wheeler Usage Characteristics **Parking Characteristics Travel Characteristics** Helmet Usage Years of Usage in % **Travel Distance in %** Parking Type > 10 km 32.8 >10 37.3 Always 65.9% Off-Street Parking*^b 39.8% 6-10 km 22.6 5-10 24.1 Sometimes or **On-Street Parking** 60.2% 34.1% 2-6 km 33.6 2-5 20.6 Never*^b **Parking Hour** < 2 km*b 11 < 2*b 18.2 PUC Check Less than 1 hour 47.1% Frequency of Usage At least twice **Travel Time in %** 38.1% Otherwise*b 52.9% a year Everyday 62.9% > 45 mins 20.3 Otherwise*^b 37.1% 30-45 mins 29.6 **Parking Cost** Less than 61.9% twice a year*b 15-30 mins 31.8 **Annual Maintenance Cost** No Parking Charge*^b 55.9% < 15 mins*b 18.3 > Rs. 8,000 5.8 Share with Family Less than Rs. 10 14.3% Rs. 6,000-8,000 🔳 10 **Co-Passenger** 26.1% Rs. 10-30 Rs. 5,000-6,000 23.8 78.7% Yes 31.8% Yes < Rs. 5,000*b 60.4 No*^b No^{*b} More than Rs. 30 21.3% 3.8% 68.2%

*b Variables are 2000 sidered as a base in the models



Framework: k-means Clustering



- Idea -
 - Minimize the total within-cluster sum of squares (WSS) defined as:

 $WSS = \sum_{k=1}^{k} \sum_{x_i \in C_k} (x_i - \mu_k)^2$

- x_i is the observation belonging to cluster C_k
- μ_k is the cluster centroid
- Variables -
 - Population Density (PD)
 - Literacy Rate (LR)
 - Gender Ratio (GR)



WSS vs k Plot

Centroids	Cluster-1	Cluster-2	Cluster-3	Cluster-4	Cluster-5
PD	16822	27710	2784	10663	6611
LR	87.12	86.77	85.17	85.03	88.09
GR	967	905	959	910	966



Framework: Multilevel Modelling



• Idea-

- Nested data structure violates the independence of residuals assumption
- Individuals in same cluster perform similarly than individuals in different clusters
- Disentangle the within-cluster effects from betweencluster effects
- Models-
 - Empty Model- Odds of outcome variables vary from one cluster to another
 - Constrained intermediate model (CIM)- Overall effect of level-1 and Level-2 variables
 - Augmented intermediate model (AIM)- CIM+city tierbased variation of the effect of Level-1 variables
 - If $AIM \gg CIM \rightarrow$ Tier level variation in Level-1 variable effects





- Empty Model Results -
 - ICC- Quantifies the extent to which log-odds of outcome vary between city tiers, *ICC* ∈ [0,1]
 - No between-tier variation in
 - Parking Hour
 - PUC Check
 - Co-passenger
 - Family Share
 - Small between-tier variation in
 - Helmet Usage
 - Parking type
 - Frequency of Usage

Results and Discussions



$$Logit (odds) = B_{00} + u_{0j}$$
$$ICC = \frac{var(u_{0j})}{var(u_{0j}) + (\frac{\pi^2}{3})}$$

- B_{00} is the fixed intercept
- *u*_{0j} is the level-2 residual signifying the deviance of the cluster-specific intercept from the fixed intercept

Denerate (Wenishler	Parameters					
Dependent variables	B_{00}	μ_{0j}	Log Likelihood	ICC		
Helmet Usage	0.717***	0.388	-254.5	3.35		
Parking Hour		0	-275.9	0		
Parking Type	-0.411***	9.15e-08	-268.3	2.54e-15		
PUC Check	-0.485***	0	-265.1	0		
Frequency of Usage	0.539***	0.146	-263.0	0.01		
Co-passenger	-0.761***	0	-249.6	0		
Family Share	1.306***	0	-206.7	0		
*** $p - value \leq$						
0.004						

0.001



CIM Results- Helmet Usage



Odds of Regular Helmet Usage

- Maintenance costs
- Income
- Gender Ratio
- Literacy Rate
- Population Density



- Years of usage
- Age

Results and Discussions



- X_{ij} is the Level-1 predictor variable
- *C_i* is the Level-2 predictor variable
- B_{10} is the fixed slope of X_{ij} (overall effect of Level-1 variable)
- B_{01} is the fixed slope of C_j (overall effect of Level-2 variable)



- Education
- Marital Status
 - Parking Cost



CIM Results- Parking Hours



Odds of Parking Hours



- Maintenance costs
- Age
- Education
- Gender Ratio
- Population Density



- Years of usage
- Travel Time

Results and Discussions



- X_{ij} is the Level-1 predictor variable
- *C_i* is the Level-2 predictor variable
- B_{10} is the fixed slope of X_{ij} (overall effect of Level-1 variable)
- B_{01} is the fixed slope of C_j (overall effect of Level-2 variable)





CIM Results- Parking Type



Odds of On-street Parking



- Years of Usage
- Travel Distance
- Travel Time
- Gender Ratio
- Population Density



Maintenance CostAge

Results and Discussions



- X_{ij} is the Level-1 predictor variable
- C_i is the Level-2 predictor variable
- B_{10} is the fixed slope of X_{ij} (overall effect of Level-1 variable)
- B_{01} is the fixed slope of C_j (overall effect of Level-2 variable)



- Income
- Occupation
- Parking Cost
- Education



CIM Results- PUC Check



Odds of Regular PUC Check



- Travel Time
- Travel Distance
- Maintenance Cost
- Education



Gender Ratio

Results and Discussions



- X_{ij} is the Level-1 predictor variable
- C_i is the Level-2 predictor variable
- B_{10} is the fixed slope of X_{ij} (overall effect of Level-1 variable)
- B_{01} is the fixed slope of C_j (overall effect of Level-2 variable)



- Income
- Age
- Parking Cost
- Population Density



CIM Results- Frequency of Usage

- Years of Usage
- Travel Distance
- Age
- Education
- Gender Ratio
- Literacy Rate



• Income

Results and Discussions



- X_{ij} is the Level-1 predictor variable
- C_j is the Level-2 predictor variable
- B_{10} is the fixed slope of X_{ij} (overall effect of Level-1 variable)
- B_{01} is the fixed slope of C_j (overall effect of Level-2 variable)





CIM Results- Co Passenger



Odds of accompanying a Passenger

- Education
- Population Density



- Years of Usage
- Age
- Gender Ratio
- Literacy Rate

Results and Discussions



- X_{ij} is the Level-1 predictor variable
- C_i is the Level-2 predictor variable
- B_{10} is the fixed slope of X_{ij} (overall effect of Level-1 variable)
- B_{01} is the fixed slope of C_j (overall effect of Level-2 variable)



- Maintenance Cost
 - Parking Cost
- Travel Distance



CIM Results- Family Share



Odds of sharing with Family



- Years of Usage
- Income
- Education
- Female



- Travel Time
- Travel Distance
- Parking Cost
- Maintenance Cost

Results and Discussions



- X_{ij} is the Level-1 predictor variable
- C_j is the Level-2 predictor variable
- B_{10} is the fixed slope of X_{ij} (overall effect of Level-1 variable)
- B_{01} is the fixed slope of C_j (overall effect of Level-2 variable)



- Population Density
- Literacy Rate
- Gender Ratio



• AIM Results -

- Only for variables showing nonnegative ICC
- *AIM* << *CIM*
- City tier-based variation of the effect of Level-1 variables does not improve model results significantly

$LR \chi^2(1) = Deviance (CIM) - Deviance (AIM)$

Results and Discussions



- X_{ij} is the Level-1 predictor variable
- *C_j* is the Level-2 predictor variable
- B_{10} is the fixed slope of X_{ij} (overall effect of Level-1 variable)
- B_{01} is the fixed slope of C_j (overall effect of Level-2 variable)
- u_{1j} is the Level-1 residual signifying the deviance of the cluster-specific slope from the fixed slope

Dependent Variable	Model	AIC	BIC	Log Likelihood	Deviance	χ^2	df	p- value
Helmet	CIM	528.8	672.40	-228.4	456.8			
Usage	AIM	750.9	1357.3	-223.4	446.9	9.84	116	1
Parking Type	CIM	554.3	697.97	-241.2	482.3			
	AIM	771.1	1377.4	-233.5	467.1	15.2	116	1
Frequency of	CIM	508.8	652.41	-218.4	436.8			
Usage	AIM	710.7	1317.1	-203.4	406.7	30.0	116	1



Implications and Conclusions











Helmet Usage, Parking type, and Frequency of Usage behaviour vary between different tiers of cities Common policy measures may not provide consistent results across all tiers of cities Customized policies based on city tier may ensure better enforcement Re-organization of transport infrastructure to induce push from twowheelers and pull towards sustainable alternatives



Low-income older individuals should be the target group for promoting adherence to helmet usage in densely populated cities



Policymakers must focus on low- and middle-income males to ensure regular PUC checks



Implications and Conclusions





Cities with high population density and gender ratio can modify parking charges to disincentivize people from using parking facilities for long durations



Push: People belonging to higher and middle-income groups can be chosen as the target population to administer the disincentive policies



Public transit system might be the key to motivating individuals relying on family-owned twowheelers to shift to public transit



Pull: High-income and middleincome individuals using twowheelers for longer period need more incentive to use public transit



Implications and Conclusions





Current data set cannot bring out a significant variance in usage behaviour between city-tiers and needs more introspection



Need for customization of regulatory policies based on the population density, literacy, and gender ratio of the city remains one of the key highlights



Overall results bring out new findings that extend the existing literature on two-wheeler usage behaviour



Further research is encouraged to improve the model accuracy and perform policy analysis based on the observations







Thank You and Stay Sustainable