

A BETTER CYCLING DEMAND MODEL



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Heathcote Valley

Mount Pleasant

erry Rd

Westmorland



Halswell

MSD

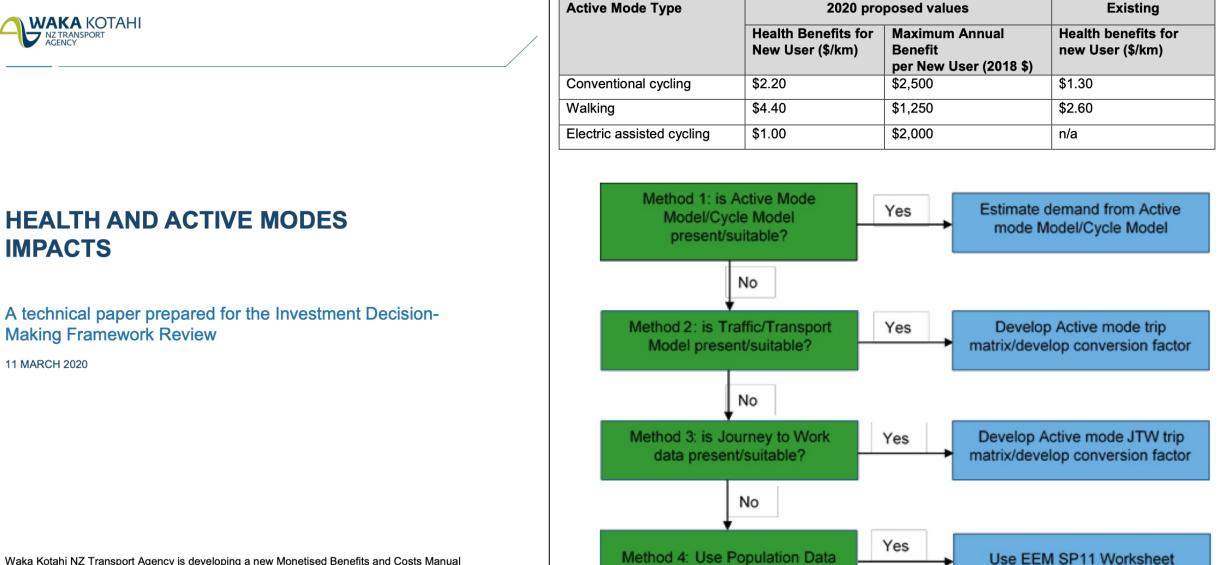
Tauhinu-Korok Scenic

Problem

- Forecasts of cycling use are needed to develop appropriate designs, inform funding applications, and support decision making and public interest.
- The current Simplified Procedure 11 worksheet 7 is known to substantially over-estimate demand based on population density and census journey to work inputs only.
- The current procedure is based on data taken from just two sites in the United States



Background



Waka Kotahi NZ Transport Agency is developing a new Monetised Benefits and Costs Manual (MBCM) to replace the existing Economic Evaluation Manual (EEM). Parameter values for health and active modes, including the use of e-bikes, have been updated.

Types of cycling demand models

- a. Comparison studies
- b. Aggregate behaviour studies
- c. Sketch planning method
- d. Discrete choice models
- e. Traditional demand models
- f. GIS-based approaches
- g. Combination of the above approaches

Source: Health and active mode impacts report (Weerappulige & Khoo, 2020)

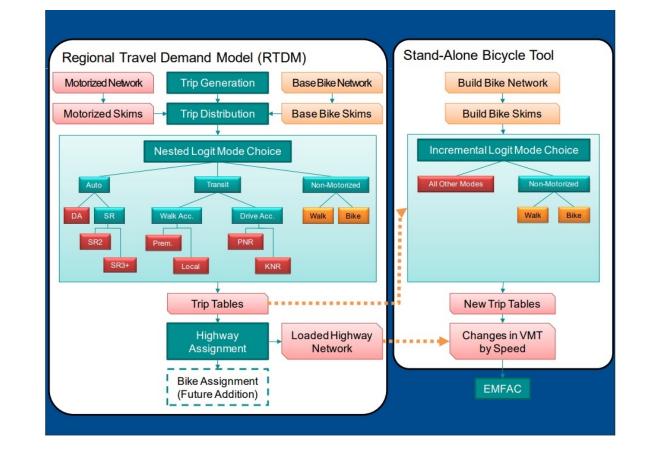
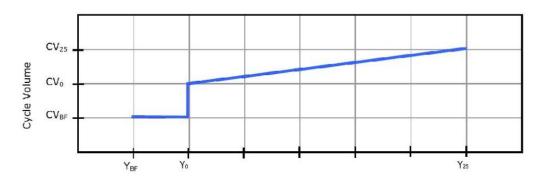


Figure 5-6: AMBAG Bike Model from NCHRP 08-36 Task 141 Evaluation of Walk and Bicycle Demand Modeling Practice (RAND, 2019)



Available non-proprietary models

Research Report 340



 $\therefore NC = 1.6 \times \sqrt{MS \times MV} + 0.5 \times PCV_{BF}$

Equation 22

Off-road paths

MS = the mode share (percentage of people that travel to work by bike) from Census data. Values for MS are listed in Appendix G. MV = the motor vehicle volume on the road parallel to the off-road path. PCV_{BF} = the existing cycle volume on the road parallel to the off-road path before the installation of the path.

On-road lanes

 $NC = CV_{BF} \times 0.2$

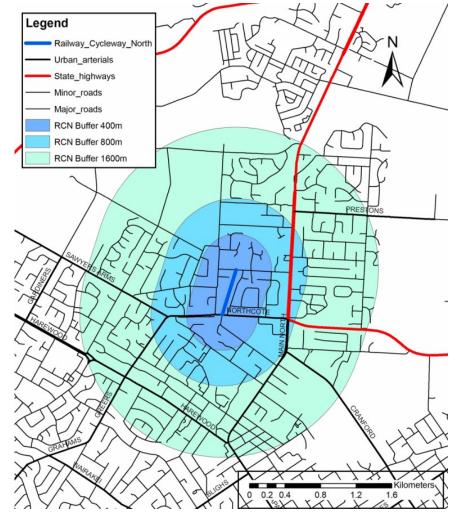
 CV_{BF} = the existing cycle volume before installation of a cycle facility.

 $CGR = \left(BG + 8\%\right)/2$

Equation 4

Equation 8







Propensity to cycle tool

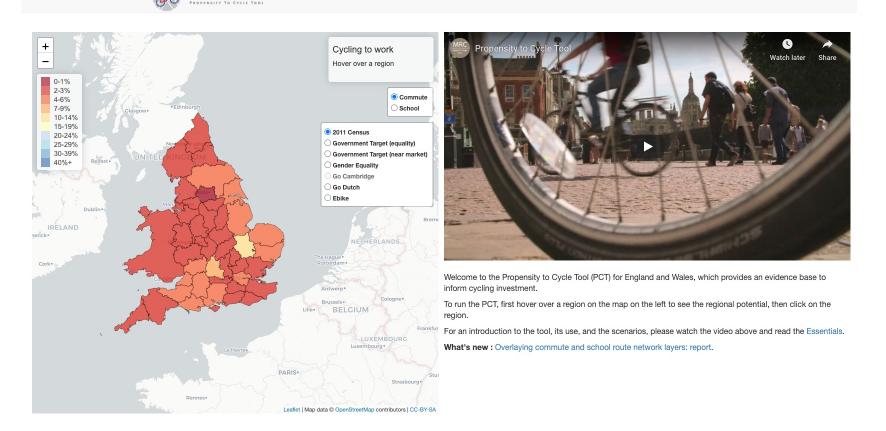
Blog

Manual

About

WWW.PCT.BIKE

- Based on hypothetical national scenarios of cycling uptake
- Does not provide estimates of cycling resulting from a given intervention
- Deterministic (not probabilistic) route choice









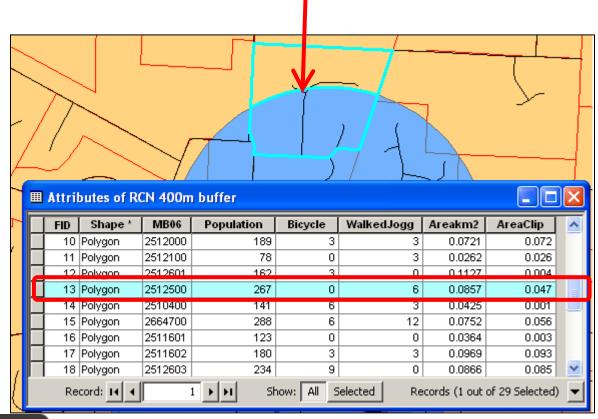






What is SP11?

- AADT estimation from population buffers 400, 800, 1600
- Cut the meshblocks to obtain population within them



Population_{400m} = 0.047 / 0.0857 * 267 = 146



SP11: AADT estimation from census data

New	and existing cyclists						
Buffe	rs (km)	<0.4	0.4 to <0.8	0.8 to ≤ 1.6			
Area	(km²)	0.93	2.24	8.50			
Densi	ity per square kilometre	2,264	2,450	1,787			
Popul	lation in each buffer (3) = (1) \times (2)	1,993	5,488	15,193			
Total	population in all buffers (Sum of (3))		22,674				
Comn	nute share (single value for all)		6.6				
Likelil	hood of new cyclist multiplier	1.04	0.54	0.21			
Row ((7) = (3) × (6)	2,073	2,964	3,190			
Sum	of row (7)		8,227				
Cyclis	st rate (9) = ((5) × 0.96) + 0.32		6.66				
) Total	existing daily cyclists (10) = (4) \times (9)		1,510				
I Total	new daily cyclists (11) = (8) × (9)	548					



SP 11 limitations

- Based on two sites in the USA quite old
- Only based on residential population
 - does not consider trip attractors e.g. workplaces
- Considers facility in isolation
 - no allowance for connecting to existing facilities / strategic network
- Relative attractiveness of different cycle facility types only affect travel time calculations, not cycle volumes
- Cycling likelihood multiplier based on census data for territorial authority
- Generally, substantially over-estimates
 - Or, will under-estimate if low surrounding residential population



Build a database

Overview of the project							Usage (before			Usage (aj	fter)					
	RCA	Project cluster no.	FACILITY	LOC REFORE	VETER	USAGE BEFORE	CYC BEFORE	W&C BEFORE	USE AFTER	CYC AFTER	W&C AFTER	BEST PRACTICE	CONTEXT	DESTINATIONS	CONTACTS	VARIABLES
What is the name of the project?			[-	•		T	-	T	-	-		T	T



Key Figures Map Last Year

Whole Domain





Categorisation schema for long list variables

Туре	Type 2	Value	Description					
	Catagoriaal	Nominal	named categories, no implicit order					
	Categorical	Ordinal	categories with an implied order assigned by modeller					
Data format	Numerical	Discrete	only particular values are possible (either by the nature of the variable, or as assigned ranges by mode developers), can be counted but not measured					
		Continuous	any numerical value along a scale is possible, can be measured					
		High	very important to accuracy of output					
Relevance		Medium	reasonably important to accuracy of output					
		Low	a "nice to have"; a modifier					
		High	already provided in existing data source					
Likely availability		Medium	available from external source or modeler's interpretation					
		Low	may require further investigation to obtain					



Variables that:

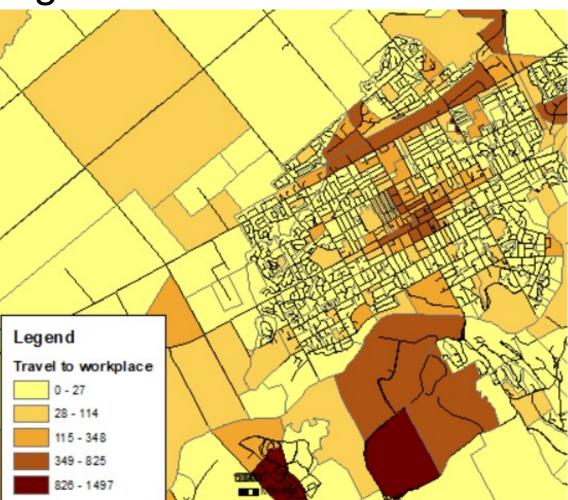
- We can change
- Constants of the urban area e.g. underlying environmental / urban area / cultural variables captured in HTS/JTW)
- We are mostly interested in new trips (latent demand) because the biggest benefits accrue to new users
 - Trips diverted from other routes are less important to the economics from a social benefit
 - Those variables that will have a bigger impact on attracting new users rather than re-routing existing users



Variables – measures of existing cycling demand

Possible calibration coefficients?

- Proportion of cycle/e-scooters using similar facilities
- Current cycle volumes
- Cycling trip to work mode share





Variables – environmental characteristics

- Relative size of city / town
- Location of facility
- Local road default speed limit
- Motor traffic volume on corridor
- Length of facility
- Proximity to residences
- Proximity to school students age 10+

- Degree of connectedness with trip generators / cycle network
- Avg gradient / hilliness
- Max gradient / hilliness
- Destination elevation
- Directness of facility



Variables – cross-elasticities of other modes

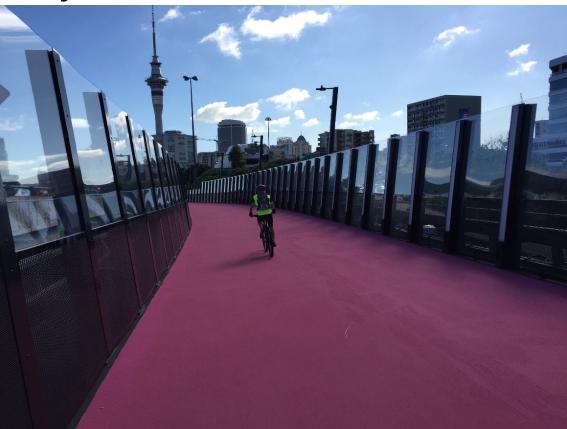
- Network level variables influence all routes in the area
 - -Degree of motor vehicle congestion
 - -Availability / price of parking at destinations
 - -Availability / LOS of PT provision
 - -Level of integration with PT
 - -Availability of bike (or e-scooter) share schemes





Variables – facility design

- Type of facility
- Number and type of intersections / crossings
- Number of driveways crossing facility
- Alignment with best practice
- Expected (change in) LoS (QoS)





Variables – network planning

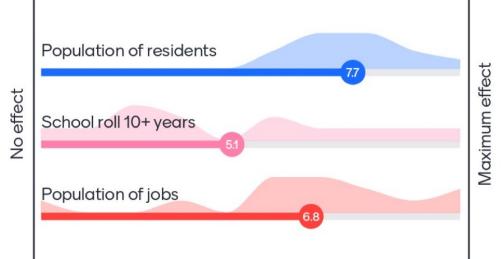
- Type of user(s) anticipated
- Aesthetic attractiveness of surroundings
- Network impact of facility
- Degree of e-bike uptake





Menti results

Transport demand variables



Takeup of cycling (latent demand)



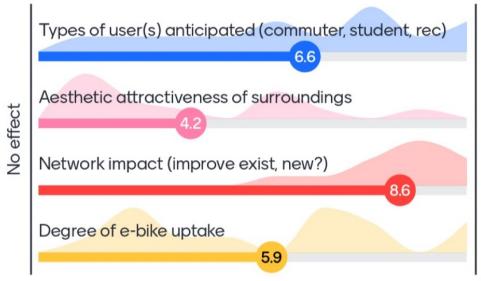


Menti results

Cross elasticities



Network planning variables



Maximum effect



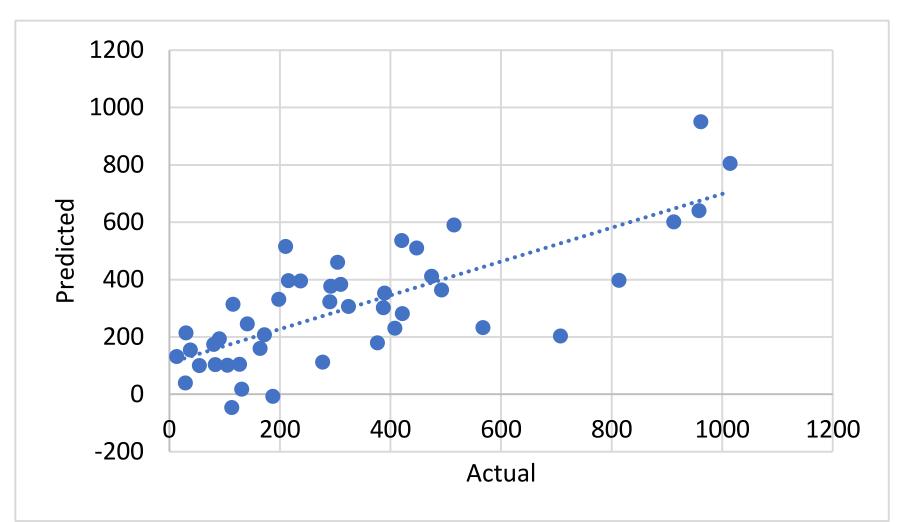
Variables tested in the modelling

- Rd is the route density score (0= viable alternate routes, 1= alt routes available but subject route preferred, 2= no other routes)
- QoS is the corridor average Quality of Service Score (where 4 is best, and 0 is worst - a transformation of the Auckland Transport QoS method that includes facility type, gradient, intersections, etc)
- Pop and Job are the census population and jobs respectively within 400, 800 and 1600 m buffers of the corridor, weighted as per previous SP11 method
- TTR is the travel time ratio (worst peak period travel time by car / best off peak travel time by car) – a measure of congestion. People are more likely to cycle when the alternative is unattractive in terms of the generalised cost of travel.
- Parking is the degree to which parking is abundant and low cost



Model 8

Model 8 includes directness and is slightly more accurate, Model 10 does not Count = e^(-3.608+(.789 * QoS)+(0.00001682 * Jobs) + (7.387 * Directness)





Accuracy comparison

	Model 8	Model 10	Model 8c	SP11
	negative	negative	pooled	
	binomial	binomial		
Model coefficients				
Directness	7.387	n/a	1745.4	-
QoS	.789	.867	252.8	-
Jobs	0.00001682	0.00001661	0.0061	-
Predicted vs. actual count (absolute error %)				
estimation before	205%	242%	147%	3979%
estimation after	47%	46%	31%	647%
estimation before & after	126%	144%	89%	2313%
validation after	52%	62%	51%	383%
estimation after & validation after	48%	51%	38%	547%

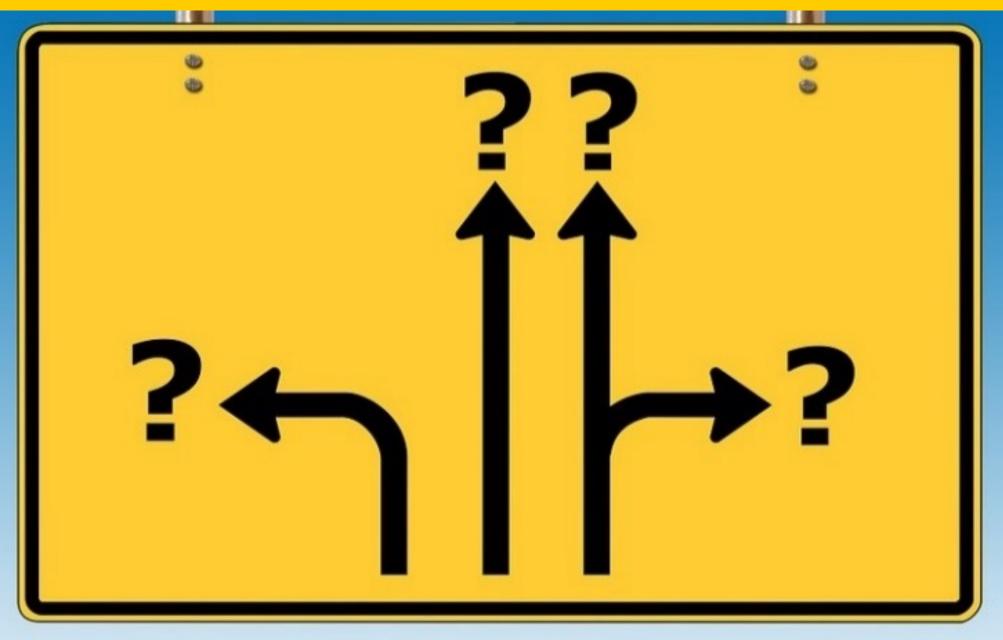


Model results

- Based on the 22 sites considered in the modelling, a half-point improvement in QoS score results in an increase of approximately 130 riders per day.
- For the 22 sites, the average change in ridership after implementation was a +81% increase in daily cycling numbers.
- The models tend to over-predict if the actual count is low, and to under-predict if the actual count is high. This is typical of models that that are subject to measurement error. For future work, it is critical to obtain larger samples of "before" implementation count data (i.e., longer duration counts).



Thank you, are there any questions?



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