Pedestrian Modeling: Assessing Pedestrian Impacts of Future Land Use and Transportation Scenarios

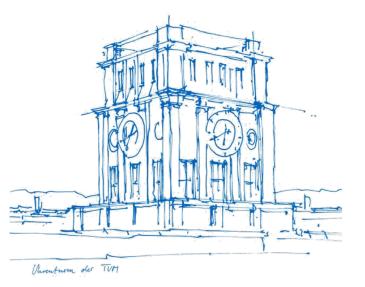
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MobilLAB final event, 14 Sep 2020





WaaS: Walk as a Solution?



CARS

ELECTRIC CARS



AUTONOMOUS CARS



UBER/LYFT CARS









From Twitter



Background

Among 48 large Metropolitan planning organizations (MPOs) in US¹:

- 38 % did not estimate walking
- 33 % estimated non motorized (walking + bicycling) travel
- 29 % estimated walking

 \rightarrow Pedestrian travel demands are less represented in the existing transport models. Transport planning and decision-making have often overlooked them.

¹Singleton et al. 2018. "Making Strides: State of the Practice of Pedestrian Forecasting in Regional Travel Models."

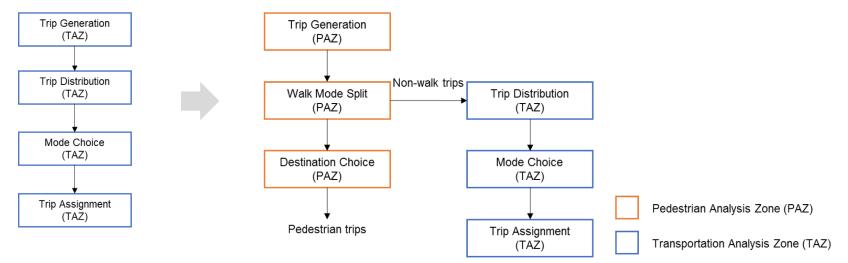
Objective

Incorporate pedestrian modelling with the existing transport model.

- Pedestrian modeling at small spatial analysis scale
- Better represent pedestrian behavior (pedestrian-related variables)

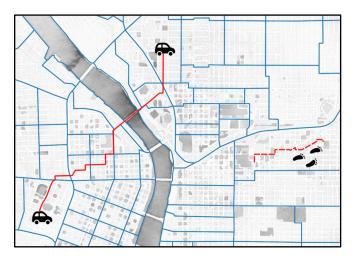
Existing transport model

New modelling framework



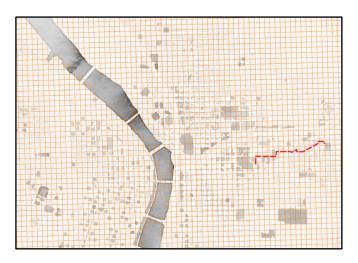


Pedestrian Analysis Zone (PAZ)



TAZ – transport analysis zone

- + Homogeneous traffic activities
- Not sensitive to walking trips
- Too many intrazonal trips



PAZ – 80 * 80 m grid cell

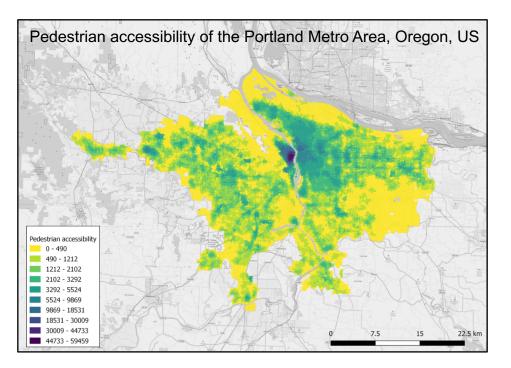
- + Sensitive to walking trips
- Millions of analysis zones
- Need of fine socio-economic data



Pedestrian Accessibility

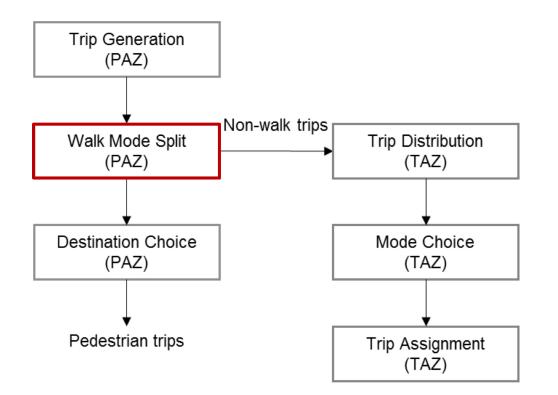
Activity density within an 800-meter pedestrian catchment area:

- Activity density = number of nonindustrial jobs + number of population
- Pedestrian catchment area is calculated based on network distance





Pedestrian modelling framework





Walk Mode Split | Methodology

- Data source: Oregon Household Activity Survey data 2011 (90% sample)
- Method: binary logit model

 $P_i(walk) = f(person characteristic, built environment attribute)$

Walk Mode Split | Model results

	Home-based purposes			Non-home-b		
	Estimate	Pr(> z)		Estimate	Pr(> z)	
(intercept)	-8.392	0.000	***	-7.411	0.000	***
Income category 2				-0.205	0.261	
Income category 3				0.222	0.046	*
Income category 4				0.448	0.000	***
Number of vehicle (0)	1.001	0.000	***	1.375	0.000	***
Number of vehicle (2)	-0.226	0.002	**	-0.898	0.000	***
Number of vehicle (2+)	-0.394	0.000	***	-0.963	0.000	***
Number of children (1)	-0.554	0.000	***			
Number of children (2)	-0.574	0.000	***			
Number of children (2+)	-0.718	0.000	***			
Child (Yes)			***	-0.162	0.039	*
log(pedestrian accessibility)	0.754	0.000	***	0.686	0.000	***
HBShop	1.029	0.000	***			
HBOther	1.046	0.000	***			
HBRecreation	1.566	0.000	***			
NHBWork				-0.362	0.000	***
	4400			2624		
Log-Likelihood:	-4189			-2624		
McFadden R^2:	13.5%			22.8%		

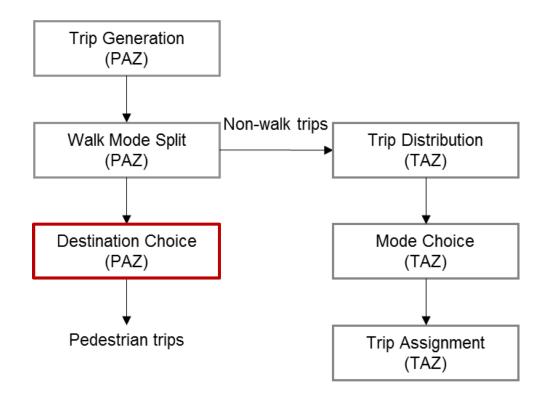


Walk Mode Split | Sensitivity analysis

Sensitivity of activity density for home-based-shop trips 27,5% 25,0% 22,5% 20,0% 20,0% 17,5% 15,0% 12,5% 10,0% 7,5° 7,5% 5,0% 2,5% 0,0% 200 0 100 300 400 500 600 700 800 900 1000 Number of non-industrial jobs

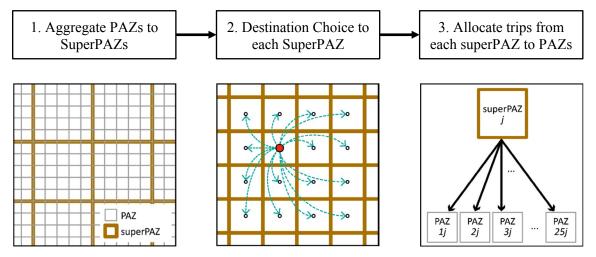


Pedestrian modelling framework



Walk Trip Destination Choice | Methodology

- Data source: Oregon Household Activity Survey data 2011 (90% sample)
- Method: Multinomial logit model
- Spatial unit: superPAZ and PAZ



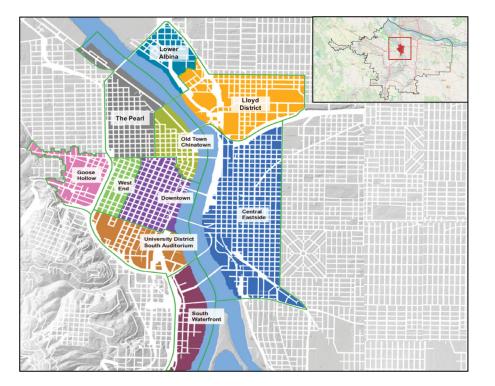
Prob(dest.) = f (network distance, pedestrian supports and barriers)



Walk Trip Destination Choice | Model results

	HBWork		HBShop			HB	HBRecreation		
	Estimat e	sd	Pr(> z)	Estimate	sd	Pr(> z)	Estimate	sd	Pr(> z)
Intrazonal (Yes)	0.71	0.19	0.00	-0.95	0.16	0.00	0.18	0.12	0.14
Distance on major road	-1.58	0.10	0.00	-2.26	0.15	0.00	-2.26	0.10	0.00
Distance on residential road	-1.48	0.13	0.00	-2.37	0.12	0.00	-1.99	0.11	0.00
Size term (In)									
Service jobs (#)	0.41	0.05	0.00				0.10	0.03	0.00
Retail jobs (#)				0.95	0.04	0.00			
Finance jobs (#)	0.32	0.05	0.00				-		-
Government jobs (#)									
All other non-industrial jobs (#)									
Households (#)							-0.04	0.04	0.33
Share of industrial jobs	-1.04	0.55	0.06	-1.36	0.43	0.00			
Mean slope	-0.19	0.07	0.01	-0.38	0.07	0.00	-0.15	0.04	0.00
Freeway in zone (Yes)				-0.32	0.18	0.08	-0.36	0.22	0.09
Park (Yes)							0.68	0.11	0.00
Null model Log-Likelihood:	-1618			-3574			-3412		
Final model Log-Likelihood:	-936			-1564			-1883		
Mc'Fadden pseudo R2	40.47%			55.93%			44.50%		





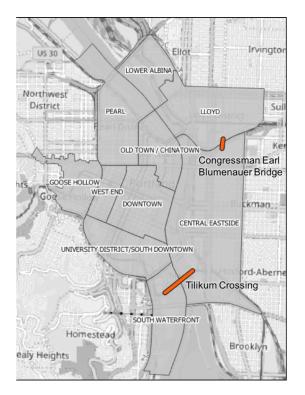
Household

23,100 → 60,300 (+161%)

Employment

123,800 → 174,000 (+41%)











	A: 2010	B: 2035 with growth	C: 2035 with growth and bridges	
District	Total walk trips	Total walk trips	difference to B (% Change)	
CENTRAL EASTSIDE	3,290	16,043 (+388%)	+1.1%	
DOWNTOWN	24,085	48,140 (+100%)	0.0%	
GOOSE HOLLOW	6,745	10,077 (+49%)	0.0%	
LLOYD	4,717	23,650 (+401%)	+0.2%	
LOWER ALBINA	158	701 (+343%)	0.0%	
OLD TOWN / CHINATOWN	5,540	8,931 (+61%)	0.0%	
PEARL	10,794	25,103 (+133%)	0.0%	
SOUTH WATERFRONT	704	8,594 (+1121%)	+0.4%	
UNIVERSITY DISTRICT/ SOUTH DOWNTOWN	7,593	17,551 (+131%)	+0.1%	HousingCommercial
WEST END	10,657	21,018 (+97%)	0.0%	Education
Sum	n 74,283	179,807 <mark>(142%)</mark>	+0.2%	□ Industrial



	A: 2010	B: 2035 with growth	C: 2035 with growth	
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Conclusion

- Contribution
 - One of the first practical planning tool for pedestrian demand estimation at fine spatial scale
 - Better understanding of the effectiveness of various demographic policies and infrastructure planning on promoting more walk trips.
- On progress further works
 - Link the pedestrian modelling to health model for health impact analysis, crash risk exposure, pollutant exposure
 - Feedback health impacts of walking into land use model for estimating life expectancy



Thank you for your listening!

Qin Zhang MobilLAB final event, 14 Sep 2020





District			Total hous	eholds	Total employment			
		2010	2035	Change in %	2010	2035	Change in %	
Central Eastside		900	7900	+778%	17000	25000	+47%	
Downtown		1600	4600	+188%	48200	55200	+15%	
Goose Hollow		3900	4900	+26%	5300	7300	+38%	
Lloyd		1000	9000	+800%	16800	25800	+54%	
Lower Albina		100	300	+200%	2100	2300	+10%	
Old Town		1900	3900	+105%	5200	8200	+58%	
Pearl		5600	11600	+107%	10700	14700	+37%	
South Waterfront		1100	5100	+364%	1200	11200	+833%	
University District		3200	6200	+94%	10400	14400	+38%	
West End		3800	6800	+79%	6900	9900	+43%	
	Sum	23100	60300	+161%	123800	174000	+41%	