

# Changing the scope and scale of regional travel models to better estimate pedestrian activity: Applications for public health

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Moving Active Transportation to Higher Ground

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Portland State  
UNIVERSITY



TREC

*Transportation Insight for Vibrant Communities*



Metro

# Outline

- **Background**
- **Project & data**
- **Method & results**
  - Trip generation
  - Walk mode split
  - Destination choice
- **Summary, conclusions, & future work**



Adapted from: <http://www.flickr.com/photos/takomabibelot/3223617185>

## Why model pedestrian travel?



plan for pedestrian investments  
& non-motorized facilities



mode shifts



health & safety



greenhouse  
gas emissions



new data

## How can estimates of walking be used for health impact assessment?



# walk trips, distances, locations

- Pedestrian crash risk & safety assessment
- Minutes walked → levels of physical activity
- Health impacts of projects / scenarios

## How do travel models estimate walking?

- Among 48 large MPOs:
  - 38% did not estimate walking
  - 33% estimated non-motorized (walking + bicycling) travel
  - 29% estimated walking
- Few used many BE measures or small spatial units

### Trip-based model sequence

1. Generation
2. Distribution
3. Mode choice
4. Assignment

Source: Singleton, P. A., & Clifton, K. J. (2013). Pedestrians in regional travel demand forecasting models: State-of-the-practice.

## What are some challenges?

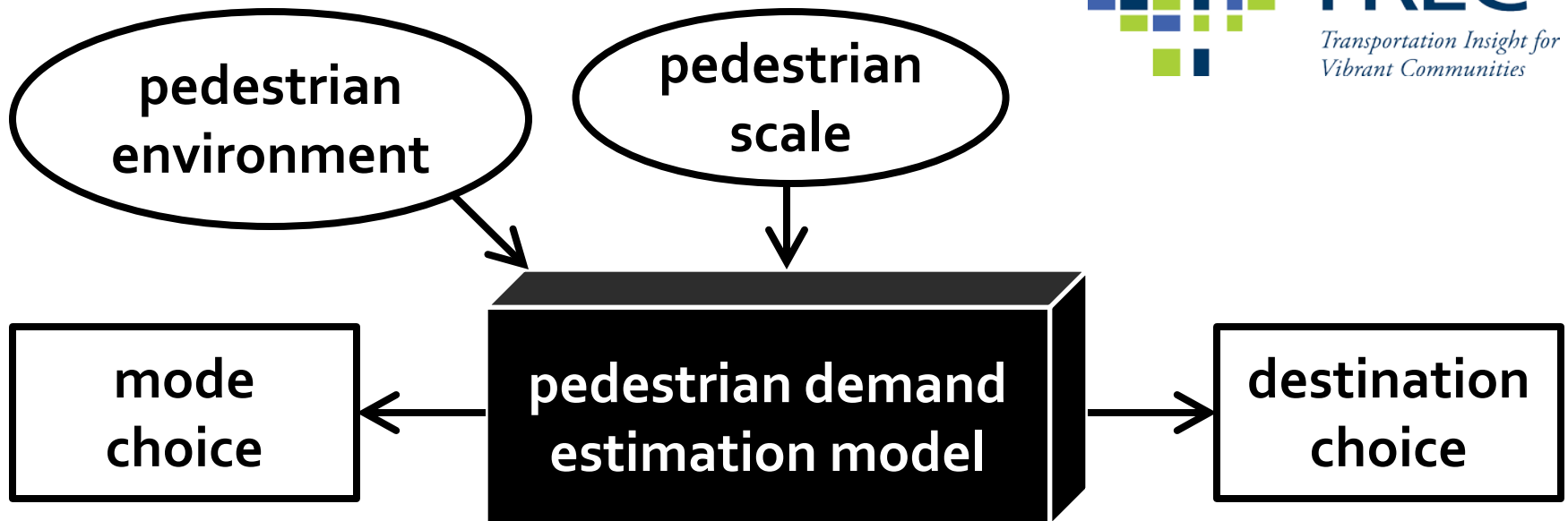
- ***Walking behavior data:***
  - difficult to obtain
- ***Built environment data:***
  - inconsistent/incomplete info on sidewalks, ...
- ***Travel demand models:***
  - large TAZs & coarse networks obscure BE variation
- ***Walking behavior research:***
  - determinants of walking were lacking

## What are some opportunities?

- ***Walking behavior data:***
  - improved travel surveys, ped. count data collection
- ***Built environment data:***
  - archived spatial datasets, GIS processing
- ***Travel demand models:***
  - smaller zones, complete networks, computer power
- ***Walking behavior research:***
  - more knowledge and studies

# Project overview

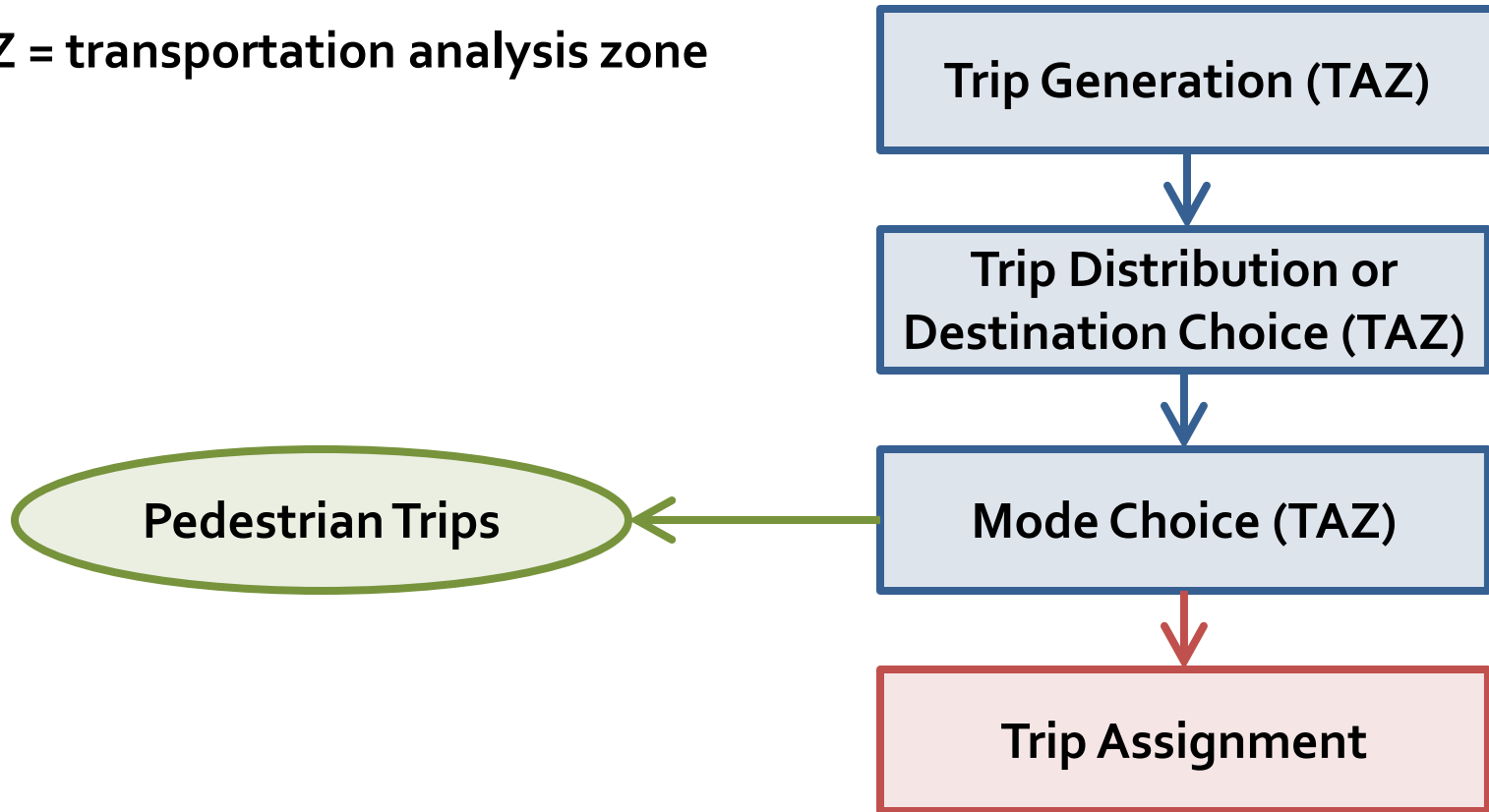
- **Metro:** metropolitan planning organization for Portland, OR
- **Two research projects**





# Current method

TAZ = transportation analysis zone

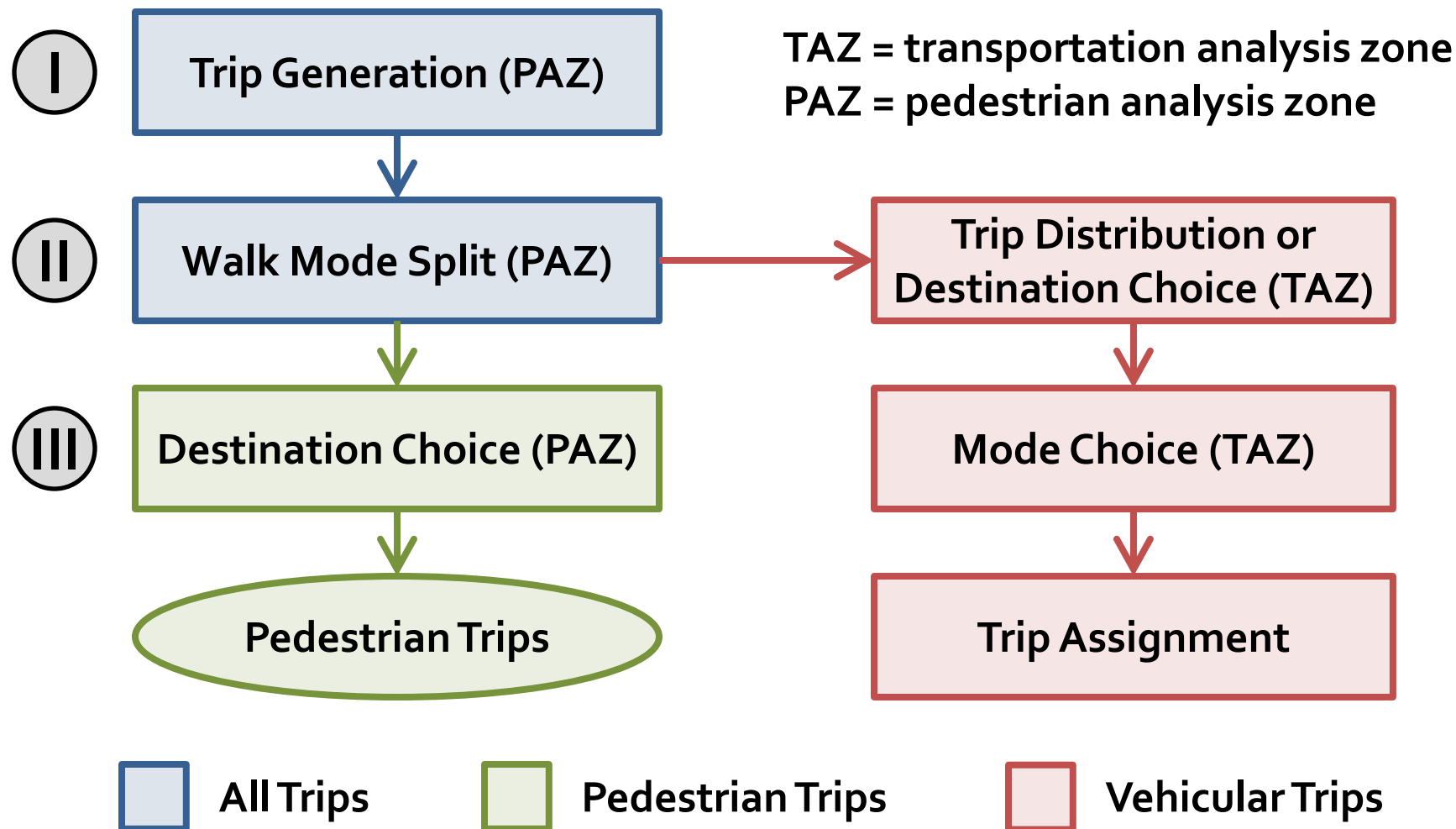


 All Trips

 Pedestrian Trips

 Vehicular Trips

# New method

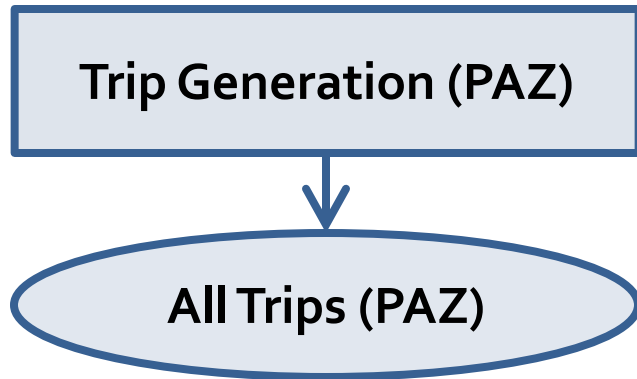


# Travel survey data

- Oregon Household Activity Survey (OHAS)
- Portland region dataset (2011)
  - 6,100 households
  - 13,400 people
  - 56,000 trips ÷ 4,500 walk trips ≈ 8% walk
  - 90% sample\* for estimation;  
10% for validation

Note: \* The 90% estimation sample was stratified by trip purpose and walk vs. not-walk.

# ① Trip generation



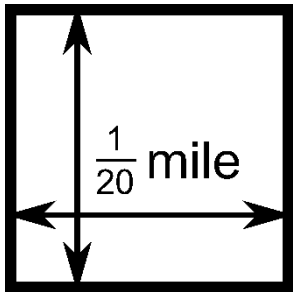
**# trips = *function of*...**

- traveler characteristics
- average trip rates

- ***Data:*** 2010 US Census pop. est.
- ***Method:*** cross-classification model\*
- ***Spatial unit:*** pedestrian analysis zone

Note: \* We did not estimate new cross-classification models; instead, we applied Metro's existing models to PAZ-level data.

# Pedestrian analysis zones



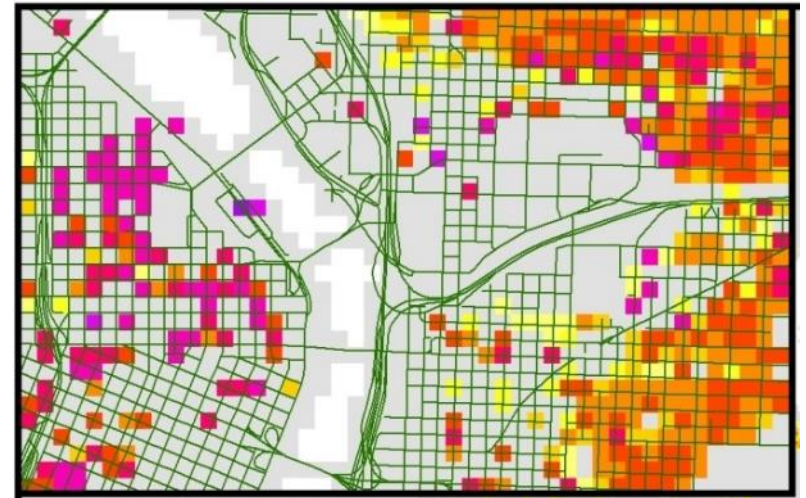
$\frac{1}{20}$  mile = 264 feet  $\approx$  1 minute walk

Metro:  $\sim$ 2,000 TAZs  $\rightarrow$   $\sim$ 1.5 million PAZs

TAZs

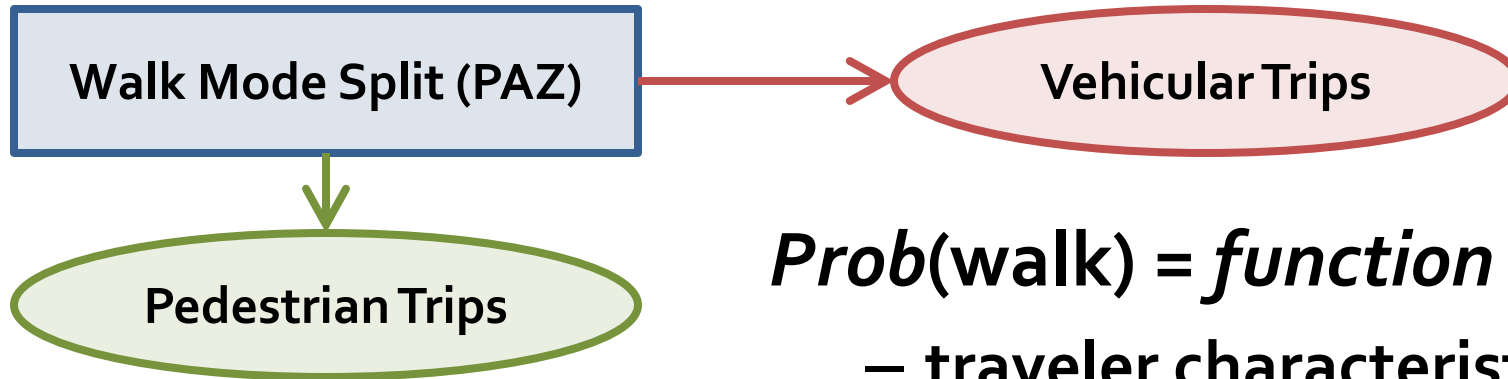


PAZs



Home-based work trip productions

# II Walk mode split



*Prob(walk) = function of...*

- traveler characteristics
- pedestrian environment

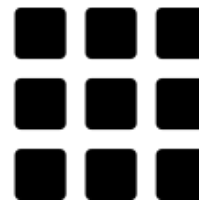
- *Data:* 2011 OHAS
- *Method:* binary logit model
- *Spatial unit:* pedestrian analysis zone

## Pedestrian Index of the Environment (PIE)

20–100 score = calibrated *sum*(6 dimensions)



People and job  
density



Block size



Transit access



Sidewalk  
extent



Urban living  
infrastructure

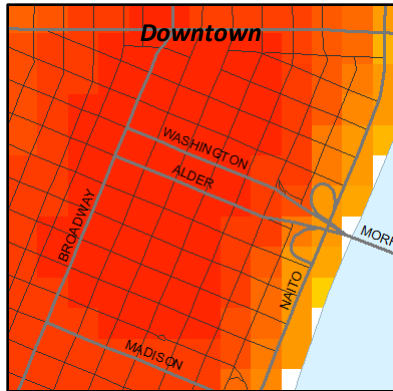


Comfortable  
facilities

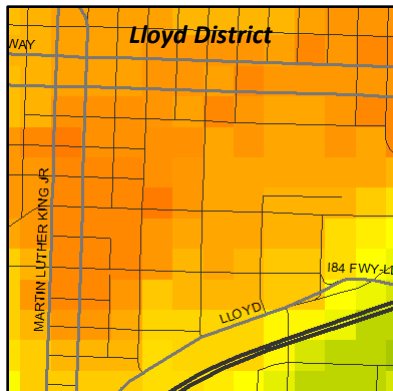
ULI = Urban Living Infrastructure: pedestrian-friendly shopping and service destinations used in daily life.

# Visualizing PIE

## 100 – Downtown core



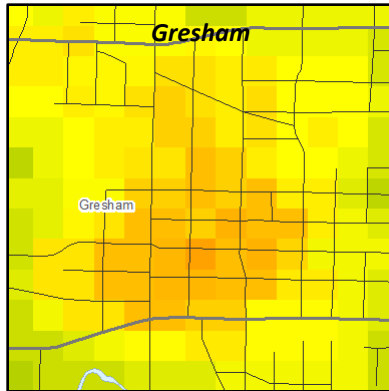
## 80 – Major neighborhood centers



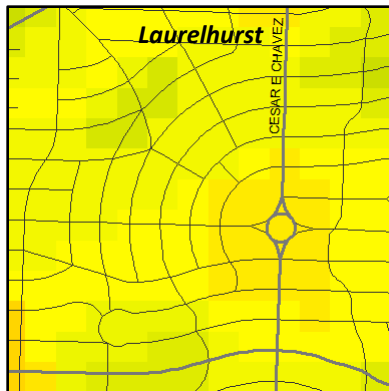


# Visualizing PIE

## 70 – Suburban downtowns

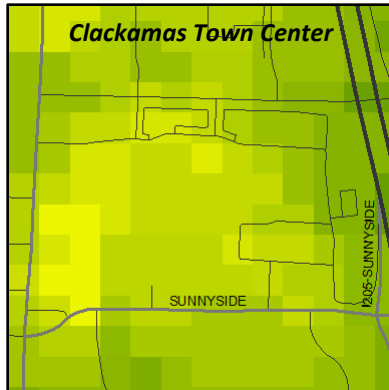


## 60 – Residential inner-city neighborhoods

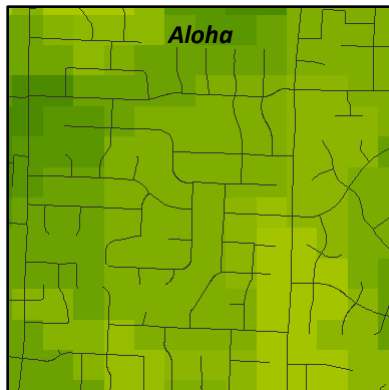


# Visualizing PIE

## 50 – Suburban shopping malls

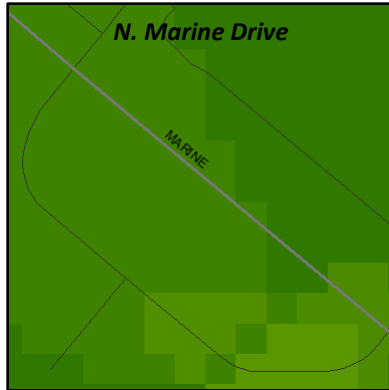


## 40 – Suburban neighborhoods/subdivisions

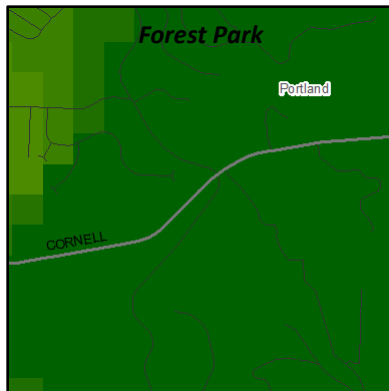


# Visualizing PIE

## 30 – Isolated business and light industry



## 20 – Rural, undeveloped, forested





# Walk model results

## • Traveler characteristics

*+ positively related to walking*

*- negatively related to walking*

number of children in HH

age of household head

HH vehicle ownership

## • Pedestrian environment

*+ positively related to walking*

*Δ odds of choosing to walk*

**+ 10 points PIE  
associated with:**

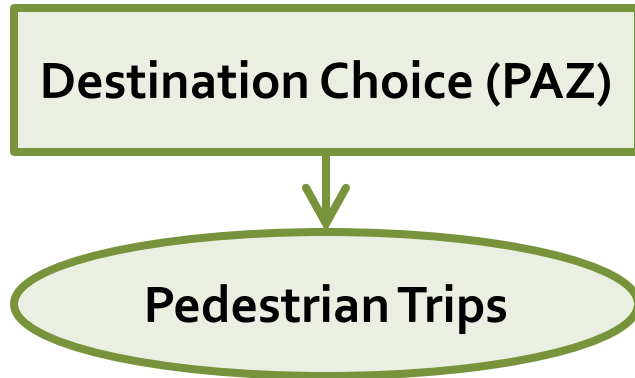
**43% increase (HBW)**

**54% increase (HBO)**

**67% increase (NHB)**

Pseudo R<sup>2</sup>

**0.137 (HBO) – 0.253 (NHB)**



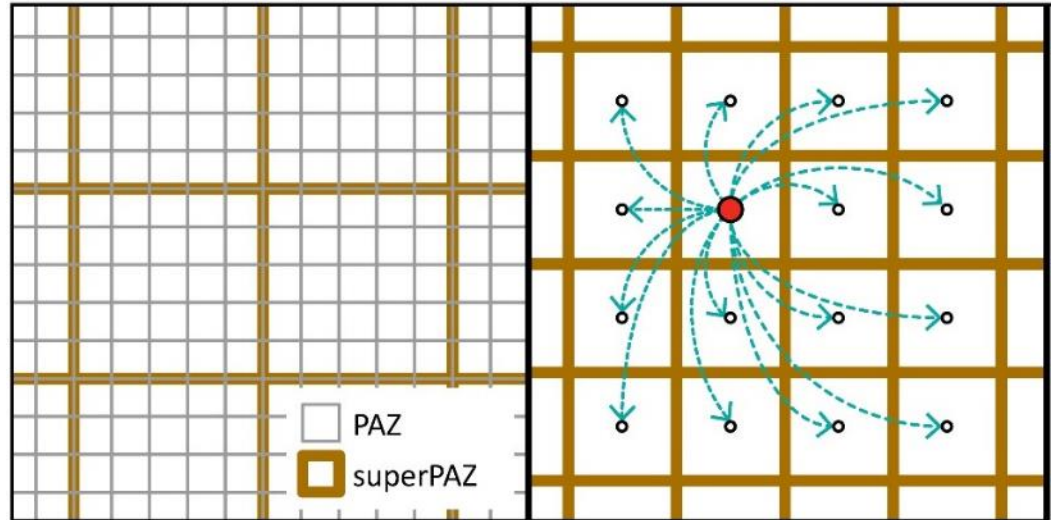
*Prob(dest.) = function of...*

- network distance
- size / # of destinations
- pedestrian environment
- traveler characteristics

- *Data:* 2011 OHAS
- *Method:* multinomial logit model
- *Spatial unit:* super-pedestrian analysis zone

# Destination choice

- **superPAZ:**
  - a grid of  
 $5 \times 5 = 25$   
PAZs



- **Choice set generation:**
  - Random sample of 10 superPAZs within 3 miles  
(99% of OHAS walk trips < 3 miles)



# Dest. choice results

## Preliminary results:

	<u><i>Δ odds of walking to destination</i></u>
+ 1 mile of distance	<b>72–85% decrease</b>
2 × # destinations	<b>32–39% increase</b> 4% increase (HBrec) 92% increase (HBshop)
+ 10 points PIE	<b>15–46% increase*</b>
Pseudo R <sup>2</sup>	<b>0.417 (HBrec) – 0.668 (HBshop)</b>

Note: \* Except for HBshop.

# Summary of results

## II Walk mode split model:

+ 10 point PIE → +45% to +65% odds of walking

## III Destination choice model (preliminary):

+ 1 mile distance → -75% odds of walking to dest.

2 × #destinations → +33% odds of walking to dest.

+ 10 points PIE → +15% to +45% odds of walk to dest.

+ 10 points PIE ≈ -450 to -1600 ft (-0.08 to -0.30 mi)



- **Walking demand impacts of projects, policies, programs, or suites of scenarios**
  - $\Delta$  sidewalk/off-street path network, or  $\Delta$  land use density/diversity  $\rightarrow \Delta$  PIE  $\rightarrow \Delta$  pedestrian travel demand
- **Crash analysis / safety assessment**
  - # walk trips across / along a street, or distances walked in an area  $\rightarrow$  denominator in a crash rate calculation

- **Health impact assessment / model**
  - distances walked by location/neighborhood
    - minutes walked by location/neighborhood
    - levels of physical activity
- **Health & transportation equity analysis**
  - # walk trips & distances walked:
    - by neighborhood,
    - by categories of age, income, etc.

- **Continue destination choice modeling**
- **Refine and verify PIE**
  - Compare PIE to other walkability measures (e.g., WalkScore)
  - Construct with widely available data sources (e.g., EPA's Smart Location Database)

- **Test method in other region(s)**
  - **Examine relationships in contexts beyond those in Portland, OR**
  - **Assess PIE's transferability (whether other regions prefer different "flavors" of PIE)**
  - **Construct full pedestrian modeling tool**
  - **Provide agency guidance for making pedestrian enhancements to urban travel demand models**

# Questions?

## Project info & reports:

<http://trec.pdx.edu/research/project/510>

<http://trec.pdx.edu/research/project/677>



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