## The relationship between health and active travel

Mathijs de Haas
The Netherlands Institute for Transport Policy Analysis (KiM)

## Introduction

, Research shows that physical activity positively influences health

- Reduces the risk of diabetes, cardiovascular diseases, depression etc.
, Worldwide around a third of all adults do not reach public health guidelines for recommended levels of physical activity
- In The Netherlands this amounts to almost 50\%
, Promoting active travel could potentially be effective in increasing physical activity levels
- And thereby contributing to physical and mental health


## Research questions

> How are subjective health and body-mass index (BMI) related to mode use in the Netherlands?
, To what extent does active travel contribute to reaching physical activity recommendations?
, Does BMI influence mode use, or does mode use influence BMI?

## Data

, MPN wave 5, 6 and 7 (2017-2019)
, MPN has two indicators of health
$-\mathrm{BMI}=$ weight $(\mathrm{kg}) /(\text { length }(\mathrm{m}))^{2}$

- Healthy weight $=\mathrm{BMI}<25$
- Overweight $=\mathrm{BMI} \geq 25 \& \mathrm{BMI}<30$
- Obese $=\mathrm{BMI} \geq 30$

| Class | Share |
| :--- | :--- |
| Healthy weight $(<25)$ | $48,3 \%$ |
| Overweight $(25-30)$ | $35,0 \%$ |
| Obese $(30+)$ | $16,7 \%$ |

- 'Subjective' health = how healthy people find themselves
, Physical activity: the time people are cycling or walking


## Methods

> Relation between subjective health and BMI and mode use

- Multivariate regression models
> Active travel and physical activity guidelines
- Latent class analysis
> Assessing whether health influences mode use or vice versa
- Random Intercept Cross-Lagged Panel Model


## Bicycle and e-bike ownership per BMI class

Bicycle ownership per age- and BMI class


E-bike ownership per age- and BMI class


## Health and mode use

, Multivariate linear regression models to assess whether health and the use of individual travel modes are related (cross-sectional)
> Control for counfounding variables:

- Gender
- Work status
- Level of education
- Age
- Income
- Country of origin of respondent


## Health and mode use－results

|  | Trips（\＃per three days） |  |  |  | Distance（kilometres） |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Overweight （ref． healthy weight） | Obese（ref． healthy weight） | Good subj． health（ref． bad subj． health） | Average trips | Overweight （ref． healthy weight） | Obese（ref． healthy weight） | Good subj． health（ref． bad subj． health） | Average distance |
| Car as driver | 0.56 （0．00） | 0.75 （0．00） | 0.29 （0．07） | 3.20 | 9.45 （0．00） | 9.34 （0．01） | 8.67 （0．02） | 55.8 |
| Car as passenge | －0．06（0．23） | －0．19（0．00） | －0．05（0．42） | 0.79 | －0．62（0．69） | －4．20（0．03） | －1．04（0．60） | 16.3 |
| Train | 0.00 （0．92） | 0.00 （0．88） | 0.00 （0．94） | 0.23 | －0．46（0．77） | －0．07（0．97） | 0.80 （0．70） | 12.4 |
| BTM | －0．01（0．65） | 0.00 （0．97） | －0．03（0．26） | 0.17 | －0．47（0．27） | －0．14（0．80） | －1．36（0．02） | 2.5 |
| Bicycle | －0．32（0．00） | －0．50（0．00） | 0.71 （0．00） | 1.62 | －1．25（0．01） | －2．37（0．00） | 2.87 （0．00） | 5.3 |
| E－bike | 0.01 （0．92） | 0.17 （0．02） | $0.08(0.28)$ | 0.52 | 0.09 （0．77） | 0.07 （0．86） | 0.67 （0．10） | 2.3 |
| Walking | －0．07（0．37） | －0．35（0．00） | 0.17 （0．11） | 1.48 | －0．50（0．00） | －0．94（0．00） | 0.46 （0．01） | 2.0 |
| Total | 0.08 （0．65） | －0．01（0．97） | 0.88 （0．00） | 8.30 | 6.08 （0．10） | 3.33 （0．48） | 10.43 （0．03） | 101.8 |

Most significant effects are found for car as driver，bicycle and walking

## Active travel as physical activity

, WHO recommends 150 minutes of physical activity per week

- Cycling (regular bicycle and e-bike) and walking are also considered physical activity
> MPN has a three day travel diary $\rightarrow$ minimum 64 minutes of active travel to meet the guideline
, How many people meet this guideline with their daily mobility?


## Physical activity per travel pattern

, Latent class analysis
, Five travel modes:

- Car
- Public Transport
- Bicycle
- E-bike
- Walking



## Seven different travel patterns

- Seven different travel patterns
- Not only differences in travel behaviour, also differences in background characteristics
- For example:
- Car users are generally employed men with relatively good incomes
- Low mobility class have low incomes and low level of education

|  | Pattern |  | Pattern | Pattern | Pattern |  | Pattern | Pattern |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | Pattern

## Physical activity per travel pattern

- In total, $\pm 33 \%$ of people meet the physical activity recommendation with their daily mobility
- Large differences in meeting the physical activity recommendation per travel pattern
- Only 7\% of strict car users meet the recommendation with their daily mobility
- $80 \%$ of people who primarily use the bicycle do so
- Of course, people can meet the recommendation with other physical activities besides cycling or walking
$\left.\begin{array}{lcccccccc}\hline & \text { Pattern } & \text { Pattern } & \text { Pattern } & \text { Pattern } & \text { Pattern } & \text { Pattern } & \text { Pattern } \\ & \mathbf{1} & \mathbf{2} & \mathbf{3} & \mathbf{4} & \mathbf{5} & \mathbf{6} & \mathbf{7} \\ \hline & \text { Car } & \text { Low } & \text { Car }+ & \text { Car }+ & \text { Bobility } & \text { bicycle } & \text { walking } & \text { E-bike }\end{array} \begin{array}{c}\text { Public } \\ \text { Transport }\end{array}\right]$

Causal relation between health and active travel?
> MPN allows for studying the relationship between health and active travel over time
, Random Intercept Cross-Lagged Panel Model (RI-CLPM)


## RI-CLPM results - trips and BMI

## , Preliminary results!

, No significant effects ( $\mathrm{p}<0.05$ ).
, At $p<0.10$ effects of BMI on trips in the 'Bicycle total' and 'Total active modes' model

| Mode | Cross-lagged relation | Parameter (p-value) |
| :---: | :---: | :---: |
| Bicycle | Trips -> BMI | * |
|  | BMI -> trips | * |
| E-bike | Trips -> BMI | * |
|  | BMI -> trips | * |
| Bicycle total | Trips -> BMI | * |
|  | BMI -> trips | * |
| Walking | Trips -> BMI | * |
|  | BMI -> trips | * |
| Total active modes | Trips -> BMI | * |
|  | BMI -> trips | * |

## 迷 <br> RI-CLPM multigroup results - trips and BMI

## , Preliminary results!

> Significant effects in non-obese group of BMI on trips
> No significant effects in obese group

| Model | Cross-lagged relation | Parameter (p-value) $(\mathrm{BMI}<30)$ | $\begin{aligned} & \text { Parameter }(p- \\ & \text { value) }(B M I=>30) \end{aligned}$ |
| :---: | :---: | :---: | :---: |
| Bicycle | Trips -> BMI | * | * |
|  | BMI -> trips | * | * |
| E-bike | Trips -> BMI | * | * |
|  | BMI -> trips | * | * |
| Bicycle total | Trips -> BMI | * | * |
|  | BMI -> trips | * | * |
| Walking | Trips -> BMI | * | * |
|  | BMI -> trips | * | * |
| Total active modes | Trips -> BMI | * | * |
|  | BMI -> trips | * | * |

## Conclusions (1)

, Clear differences in mode use between people from different BMI classes and with different subjective health levels

- Higher BMI: less cycling and walking, more car use
- Obese: more e-bike!
- Better subjective health: Higher total mobility. More car use, cycling and walking.
- Active travel is an important factor in meeting the physical activity guideline of 150 minutes per week
- $80 \%$ of people with a cycling-only travel pattern meet the guideline
- Only 7\% of strict car users does


## Conclusions (2)

, Does BMI influence mode use, or does mode use influence BMI?

- Number of trips with active modes does not seem to influence BMI in a following year
- There is a significant negative effect of BMI on active mode use for non-obese people
- So, an increase in BMI leads to lower active mode use, but active mode use does not lead to a lower BMI
- Relation between active travel and travelled distances and subjective health over time not yet assessed


## Questions?

