

Non-work/school activity participation in a flexible work future: A pre/post-pandemic comparative study



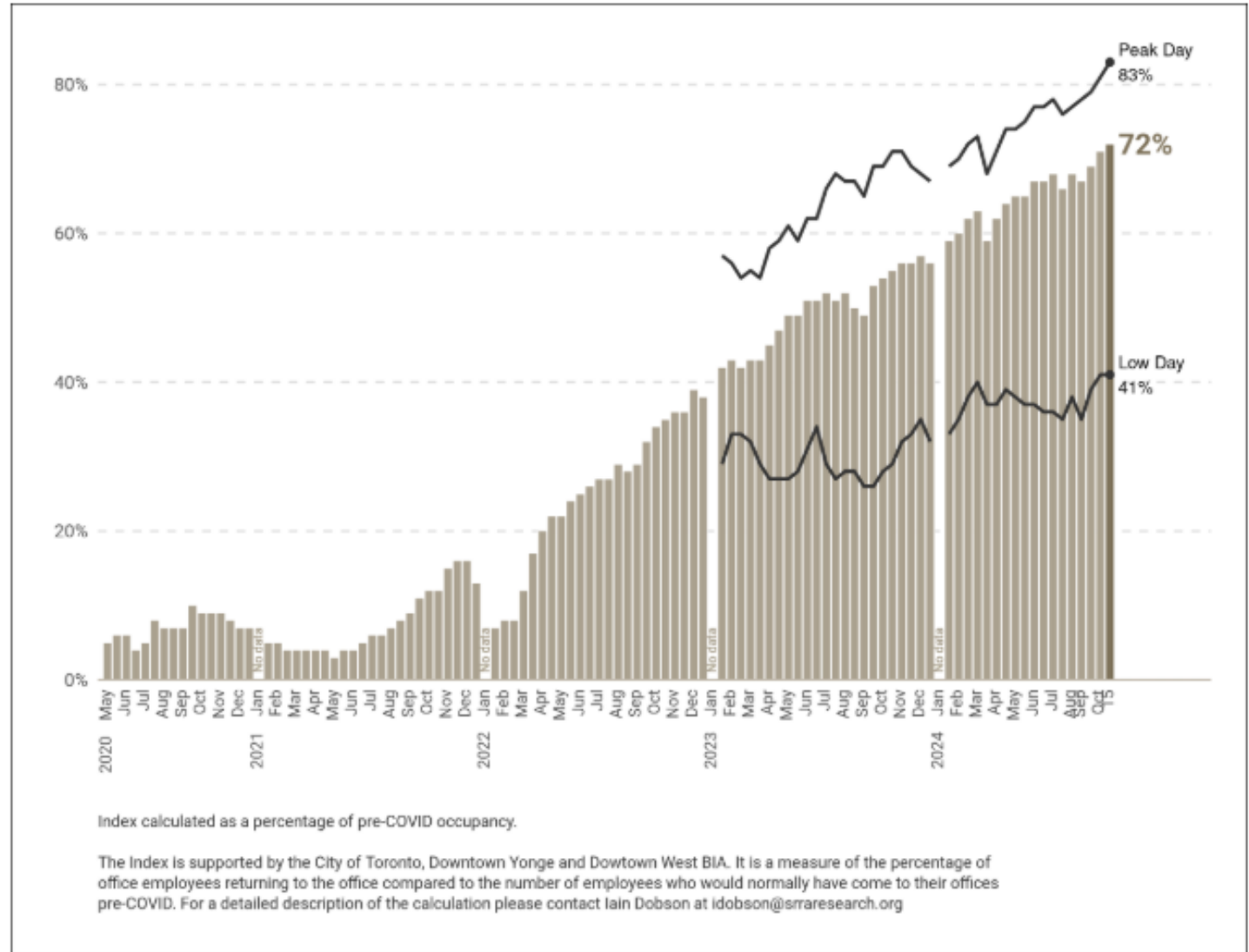
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City of Toronto Downtown Occupancy Index

- A hybrid work environment is here to stay.
- This raises many challenges for activity-based travel demand modelling.
- In this study we look at the relationship of NWS activity participation with work “modality”.



Worker “Modality”



- Onsite worker: Has an out-of-home workplace and must work onsite.
- Remote worker: Does not have an out-of-home workplace and always works at home.
- Hybrid worker: Has an out-of-home workplace; sometimes works onsite, sometimes works from home (WfH).

Non-work/school activities investigated

- Shopping.
- Leisure/personal business.



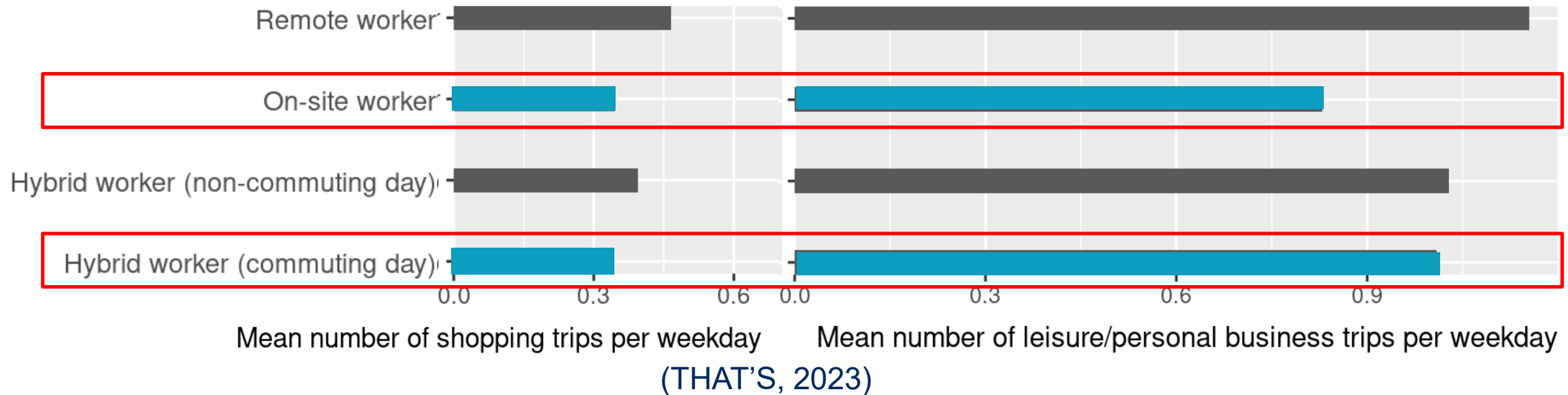
Data

- Toronto Tomorrow Survey (TTS) Fall, 2016
 - 5% sample
 - Primarily web-based
 - One weekday
- Toronto Household Activity-Travel Survey (THATS) Spring, 2023
 - Small sample
 - Smartphone app + web
 - Week-long (7-day)

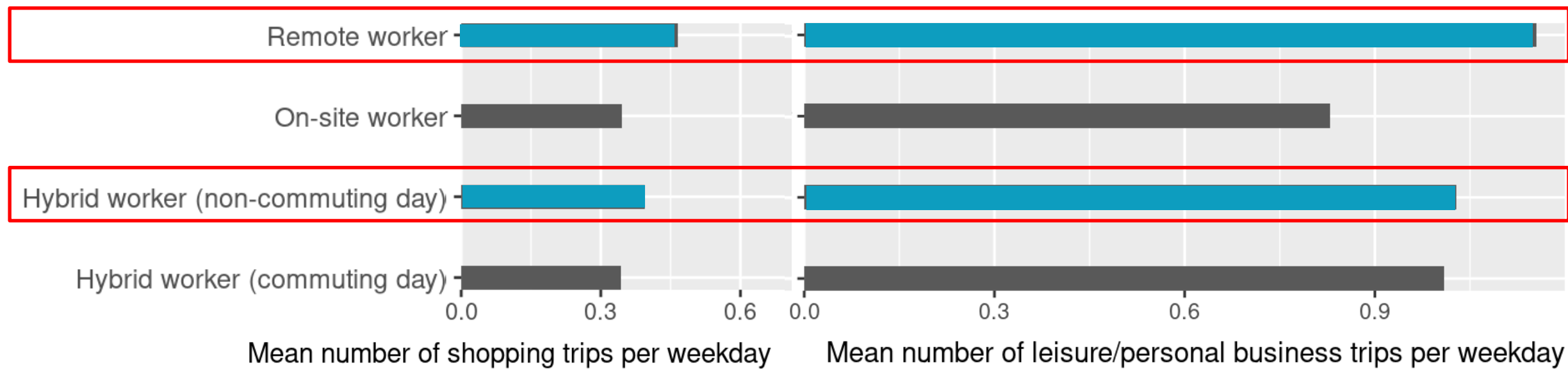
| Attribute | Category | TTS 2016 | THATS 2023 |
|------------------------------|----------------|----------|------------|
| Gender | Female | 0.4772 | 0.4716 |
| | Male | 0.5228 | 0.5157 |
| | Others | N/A | 0.0128 |
| Age | 17 or less | 0.0144 | 0.0063 |
| | 18 to 24 | 0.0680 | 0.0314 |
| | 25 to 34 | 0.1823 | 0.2954 |
| | 35 to 49 | 0.3442 | 0.4460 |
| | 50 to 64 | 0.3344 | 0.1930 |
| | 65 or more | 0.0568 | 0.0278 |
| Work modality | On-site worker | 0.9411 | 0.3323 |
| | Remote worker | 0.0589 | 0.1943 |
| | Hybrid worker | N/A | 0.4734 |
| Work status | Full time | 0.8330 | 0.8935 |
| | Part time | 0.1670 | 0.1065 |
| Sample size (# of trip-days) | | 172244 | 10975 |

*All numbers are proportions except for sample sizes.

Hypothesis (1): Activity participation of **hybrid workers on commuting days** is similar to that of **on-site workers**.



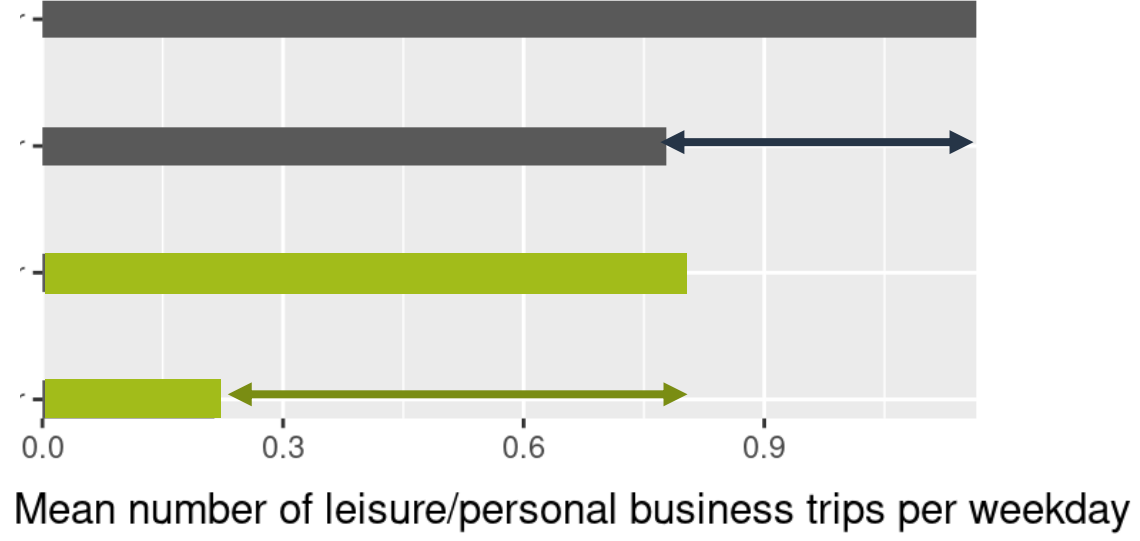
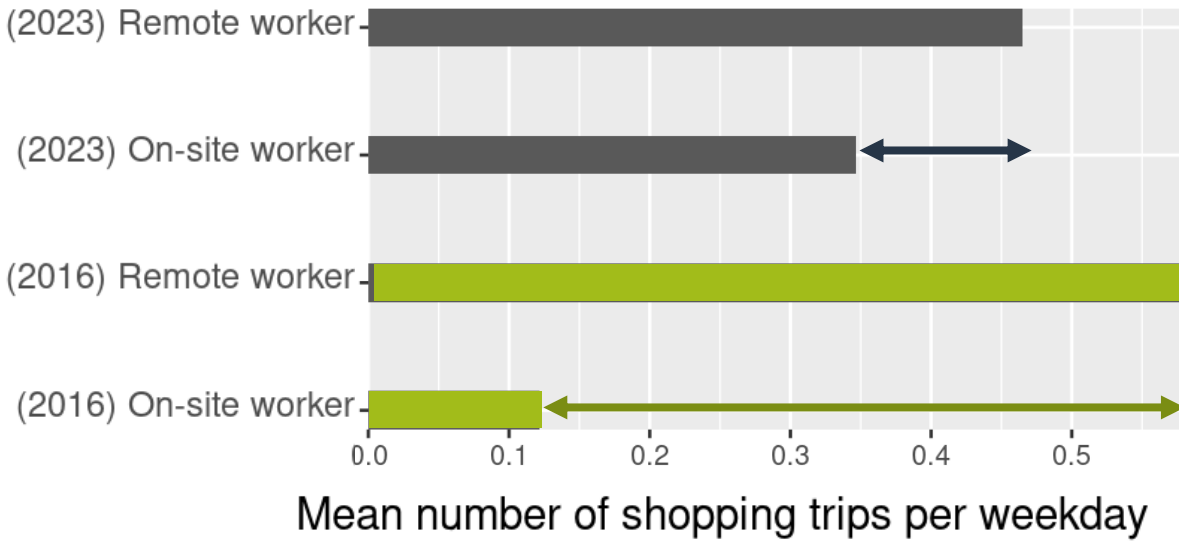
Hypothesis (2): Activity participation of **hybrid workers on non-commuting days** is similar to that of **remote workers**.



(THAT'S, 2023)

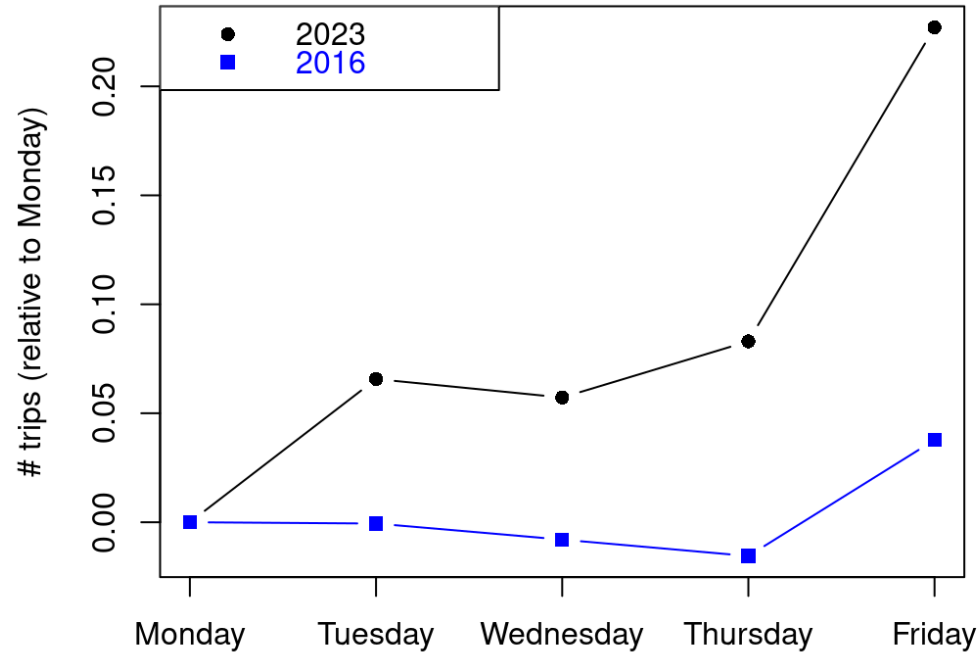


Hypothesis (3): **Post-pandemic** activity participation behaviour differs in nature from **pre-pandemic** behaviour.

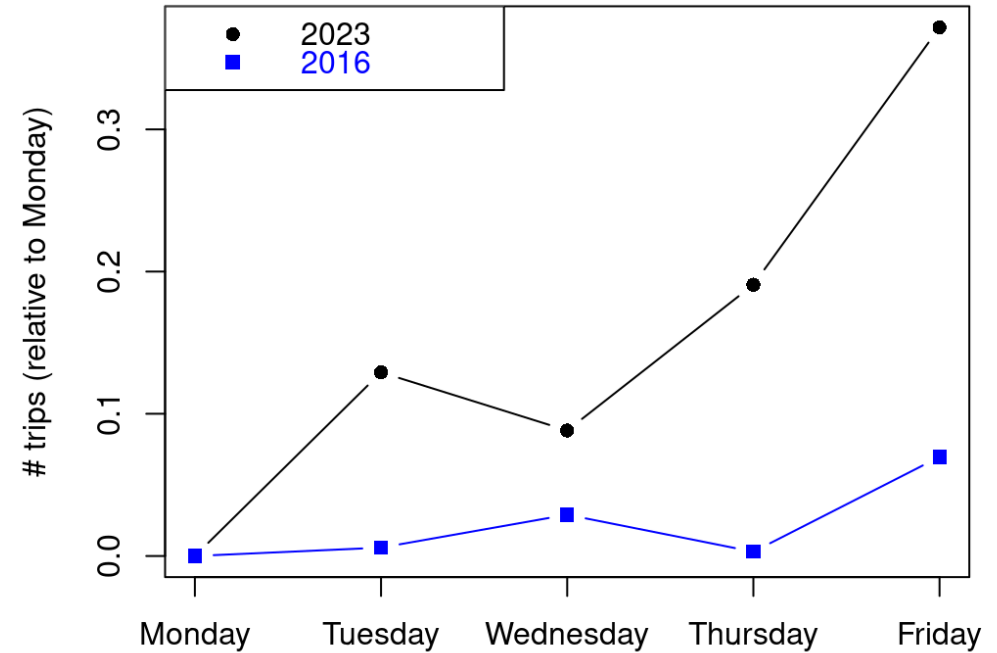


Hypothesis (3): Post-pandemic activity participation behaviour differs in nature from pre-pandemic behaviour.

Mean number of shopping trips

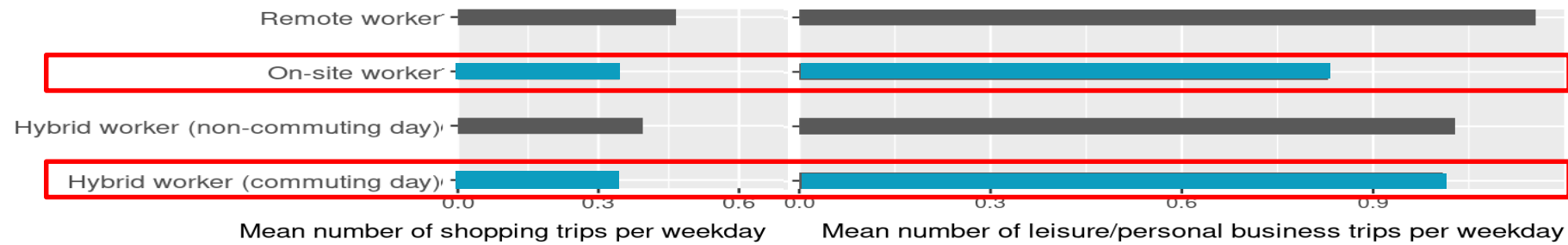


Mean number of leisure/personal business trips



Models

- Dependent variable: daily weekday person trips by activity type (by year).
- Models:
 - Negative binomial.
 - Hurdle count:
 - Stage 1: Binomial logit: make 1 or more trips; do not make any trips.
 - Stage 2: Number of daily trips if travelling this day; truncated discrete regression (e.g., Poisson regression; **negative binomial**).
 - For 2023: Models extended to incorporate mixed-effects available in the week-long dataset.
 - Random intercept.
 - Additional fixed effects available in THATS:
 - Prior day trip indicator.
 - Trip made 2 days prior indicator.



Summary of Results: Hypothesis 1

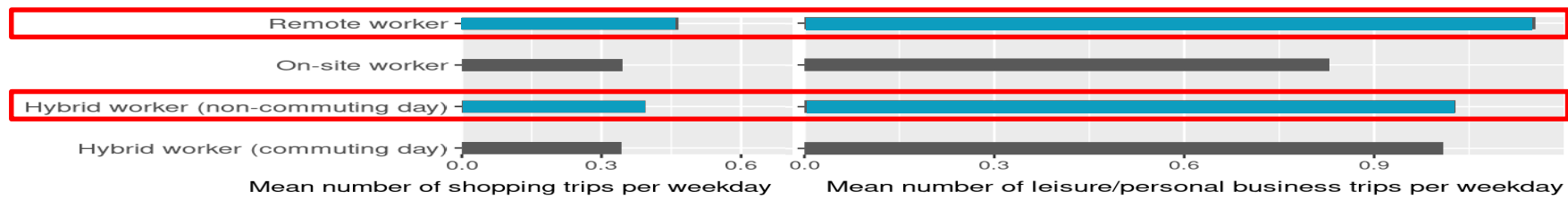
Table 4: Hypothesis 1: Activity participation of hybrid workers on commuting days is similar to that of on-site workers.

Null Hypothesis: coefficient (Hybrid) + coefficient (Hybrid & Commute) = 0

| | Shopping participation | | | Leisure/Personal business participation | | |
|---------|------------------------|-------------|--------------|---|-------------|--------------|
| | Neg. Binomial | Hurdle-Zero | Hurdle-Count | Neg. Binomial | Hurdle-Zero | Hurdle-Count |
| Value | 0.0321 | 0.0275 | 0.1807 | 0.2837 | 0.2888 | 0.1795 |
| Z-score | 0.2906 | 0.2201 | 1.0226 | 3.5612 | 2.6930 | 1.8485 |
| P-value | 0.7713 | 0.8258 | 0.3065 | 0.0004*** | 0.0071** | 0.0645 |

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

- Hybrid workers on commuting days & onsite workers do not have statistically significant weekday shopping participation rate differences.
- Hybrid workers on commuting days do, however, generate significantly more leisure/personal business trips than onsite workers.



Summary of Results: Hypothesis 2

Table 5: Hypothesis 2: Activity participation of hybrid workers on non-commuting days is similar to that of remote workers.

Null Hypothesis: coefficient (Hybrid & Non-commute) – coefficient (Home) = 0

| | Shopping participation | | | Leisure/Personal business participation | | |
|---------|------------------------|-------------|--------------|---|-------------|--------------|
| | Neg. Binomial | Hurdle-Zero | Hurdle-Count | Neg. Binomial | Hurdle-Zero | Hurdle-Count |
| Value | -0.1410 | -0.1054 | -0.1462 | -0.1215 | -0.1898 | -0.0536 |
| Z-score | -1.1965 | -0.7891 | -0.7924 | -1.2928 | -1.6407 | -0.4785 |
| P-value | 0.2315 | 0.4300 | 0.4281 | 0.1961 | 0.1009 | 0.6323 |

Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

- Remote workers on average make more shopping & leisure/PB trips than hybrid non-commuters, but the differences are not significant at a 90% confidence level.
- The leisure/PB difference in making or not-making at least one weekday trip is marginally significant.

Summary of Results: Hypothesis 3 (1)



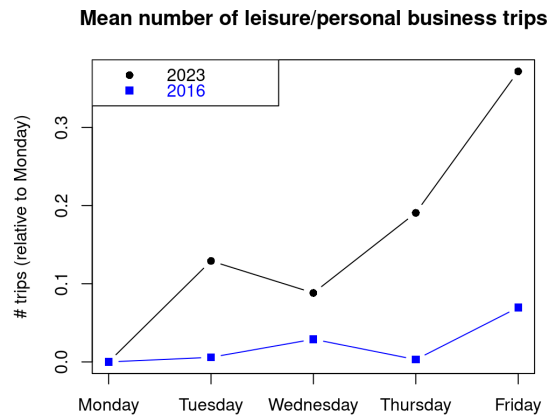
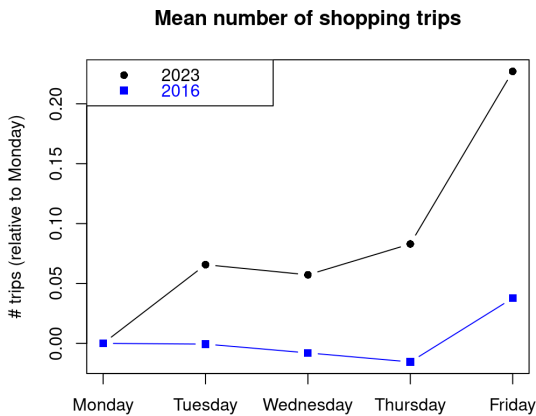
- Remote workers are more likely to engage in both shopping & leisure/PB activities in both years.
- BUT the gap has narrowed significantly:
 - Remote workers are making fewer trips & onsite workers are making more trips.

Table 6: Negative Binomial model parameter estimation differences: 2023 vs. 2016

| variables | Shopping | | Leisure/Personal Business | |
|-------------------------|----------------------|----------|---------------------------|---------|
| | b_diff | b_ratio | b_diff | b_ratio |
| (Intercept) | 1.5706***(4.1699) | 0.5911 | 0.6238*(2.0428) | 0.7083 |
| age | -0.0334*(-1.9823) | 0.5162 | 0.0065(0.4681) | 1.4764 |
| driver_licTRUE | -0.3733***(-3.9762) | -0.2629 | -0.2089**(-2.6863) | 0.4575 |
| emp_modalityHome | -1.1162***(-12.435) | 0.2203 | -0.7997***(-11.0576) | 0.3476 |
| emp_statusP | 0.1085(1.1854) | 2.2309 | 0.0231(0.2991) | 1.3385 |
| HHincome_above_100kTRUE | -0.1764**(-2.7168) | -1.0617 | -0.1958***(-3.6684) | 0.2501 |
| HHincome_below_40kTRUE | -0.0168(-0.1207) | 1.5736 | 0.3057**(2.7114) | -1.1294 |
| I(age^2) | 0.0003(1.3652) | 0.5877 | -0.0002(-1.2098) | 2.6213 |
| n_person | 0.1213*** (3.3935) | -0.2322 | 0.1456*** (4.9092) | 0.0596 |
| n_student | -0.0223(-0.5073) | 1.7481 | -0.1663***(-4.5736) | -1.0460 |
| n_vehicle | 0.0418(1.0028) | 4.3116 | 0.0122(0.3539) | 1.2636 |
| onsite_NonCommuteTRUE | -1.3289***(-14.7748) | 0.0752 | -1.0522***(-15.96) | 0.1182 |
| PD1TRUE | 0.0503(0.5744) | 0.7146 | -0.1572*(-2.3419) | -0.4219 |
| Region_TorontoTRUE | 0.0102(0.1605) | 0.8714 | 0.0804(1.5229) | 2.6796 |
| Remote_CommuteTRUE | 0.4747.(1.9308) | 0.5444 | 0.75*** (5.1115) | 0.1711 |
| sexF | -0.359***(-6.4885) | -0.4782 | -0.1933***(-4.2967) | -1.4903 |
| trip_day2 | 0.1738*(2.1039) | 5.2657 | 0.1035.(1.9596) | 2.8132 |
| trip_day3 | 0.1894*(2.2707) | -12.5191 | 0.0606(1.1445) | 1.5480 |
| trip_day4 | 0.2858*** (3.4771) | -13.5799 | 0.203*** (3.8906) | 3.9995 |
| trip_day5 | 0.3853*** (4.8983) | 3.0876 | 0.2148*** (4.2668) | 1.9493 |

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Summary of Results: Hypothesis 3 (2)



- Both shopping & leisure/PB activity participation has increased Tuesday-Friday relative to a Monday base.
- Thursday NWS activity is notably increased.

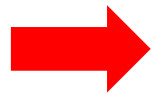
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Summary

- Hybrid workers' shopping trips on commuting days are similar to on-site workers', and similar to remote workers' on non-commuting days.
- Regarding out-of-home leisure and personal business activity participation, hybrid workers tend to **squeeze trips on commuting days** and **favor staying at home on non-commuting days**.
- Compared to the pre-pandemic era, remote workers engage in fewer out-of-home shopping and leisure/personal business trips, potentially influenced by attractive home-based alternatives like **online shopping** or a **sedentary lifestyle** adopted during lockdown. On non-commuting days, there is a higher barrier to making trips, but once trips are initiated, their frequency exceeds that of commuting days.
- Friday traditionally experiences the highest volume of out-of-home NWS trips compared to other weekdays. In the post-pandemic era, Thursday has begun to rival Friday, indicating a shift with "**Thursday is the new Friday!**" sentiment, although the clientele is different.
- Reinforces the case for multi-day (week-long) models.
- Limitation: survey bias.
- Future work: in-home vs out-of-home activity participation



Linkages clearly exist between work modality (scheduling) and NWS activity participation that may affect activity scheduling model structures.

Thank you.

Questions?



Appendix 1: Model Equations

The regression model for negative binomial models is;

$$y_i \sim NB(\mu_i, \mu_i[1 + \alpha\mu_i]), \quad (1)$$

$$\log(\mu_i) = \beta_0 + \mathbf{x}'_{ij}\boldsymbol{\beta} + \varepsilon_i, \quad (2)$$

and for the mixed-effect negative binomial models:

$$y_{ij} \sim NB(\mu_{ij}, \mu_{ij}[1 + \alpha\mu_{ij}]), \quad (3)$$

$$\log(\mu_{ij}) = (\beta_0 + b_i) + \mathbf{x}'_{ij}\boldsymbol{\beta} + \varepsilon_{ij}, \quad (4)$$

where the number of trips of a given activity type made by individual i in observation j is denoted as y_i in NB models and y_{ij} in NB mixed model, $\boldsymbol{\beta}$ is the vector of fixed effects including work modalities, household attributes, and the additional dummy variables for trip making history for 2023 models, b_i is the individual-specific random effect added for 2023 mixed models, α is the dispersion parameter that captures the extent of data over-dispersion.

Hurdle count models represent the two stages of activity participation explicitly. The first stage of the process, whether to participate or not, is modeled by its zero-hurdle component, f_{zero} ; and the second stage of the process, the frequency of participation given the state of participating, $\{Y|Y > 0\}$, is modelled by its positive count component, f_{count} . A general structure of hurdle models is as follows:

$$P(Y = y) = \begin{cases} f_{zero}(0) & y = 0, \\ (1 - f_{zero}(0)) \frac{f_{count}(y)}{1 - f_{count}(0)} & y > 0. \end{cases} \quad (5)$$

In this study, we employ a logistic regression model for f_{zero} , and negative binomial regression models for f_{count} , with the same set of parameters. Similarly, for the 2023 data, the logistic regression and NB models are extended to mixed models by incorporating individual-specific random effects, similar to Equation (4). The models are fitted using maximum likelihood estimation with Laplace approximation for mixed models. The likelihood function for hurdle models is simply the multiplication of the likelihoods of two stages.

