

MiWay Transit Access for UTM Students: District-Level Analysis

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Abstract

Mississauga's MiWay transit system plays a major role in how University of Toronto Mississauga (UTM) students commute to campus. In this project, we use the City of Mississauga's MiWay Bus Stops dataset to compare transit accessibility across districts, with a focus on direct UTM service and GO Train connectivity. We analyze bus stops across the city and summarize district-level differences in accessibility, shelter coverage, route diversity, direct UTM access, and GO connectivity. For each research question, we compute conditional proportions, create visual summaries, and conduct chi-square tests of association to assess whether these characteristics vary by district. We find strong evidence that accessibility, shelter availability, UTM service, and GO connectivity all differ substantially across Mississauga. Districts along Dundas Street and Mississauga Road provide the strongest direct access to UTM, while districts containing GO stations have more GO-serving bus routes. These findings highlight important transit inequalities that affect where UTM students may choose to live.

Introduction

Public transit availability can strongly influence how easily University of Toronto Mississauga (UTM) students can travel between home, campus, and regional transportation hubs. MiWay provides local and express routes across the city, including several routes that serve UTM directly and others that connect to GO Train stations. Understanding how these routes and stops are distributed geographically is useful for students deciding where to live.

This project uses the **City of Mississauga MiWay Bus Stops dataset**, which includes one row per bus stop and information on district, accessibility, shelter availability, and routes. We cleaned the dataset by removing missing or "UNKNOWN" district labels (primarily GO platforms or boundary stops) and separated the route list into a long format.

We address four research questions:

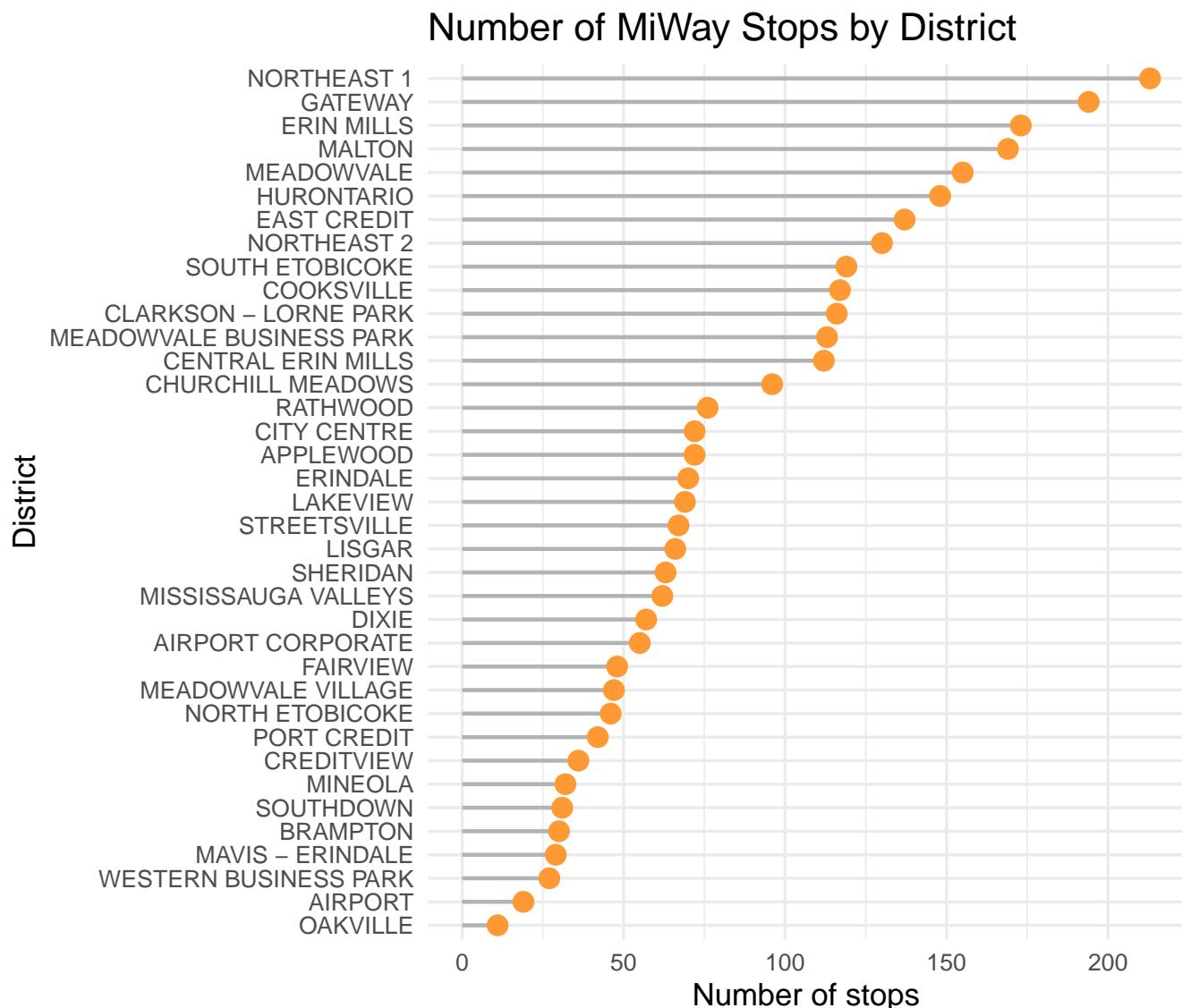
1. How does MiWay stop coverage vary across districts?
2. Does accessibility and shelter coverage differ across districts?
3. Which districts have the strongest direct connections to UTM?
4. Which districts have the best connectivity to GO Train stations?

Each question is answered using summary statistics, bar charts, and chi-square tests.

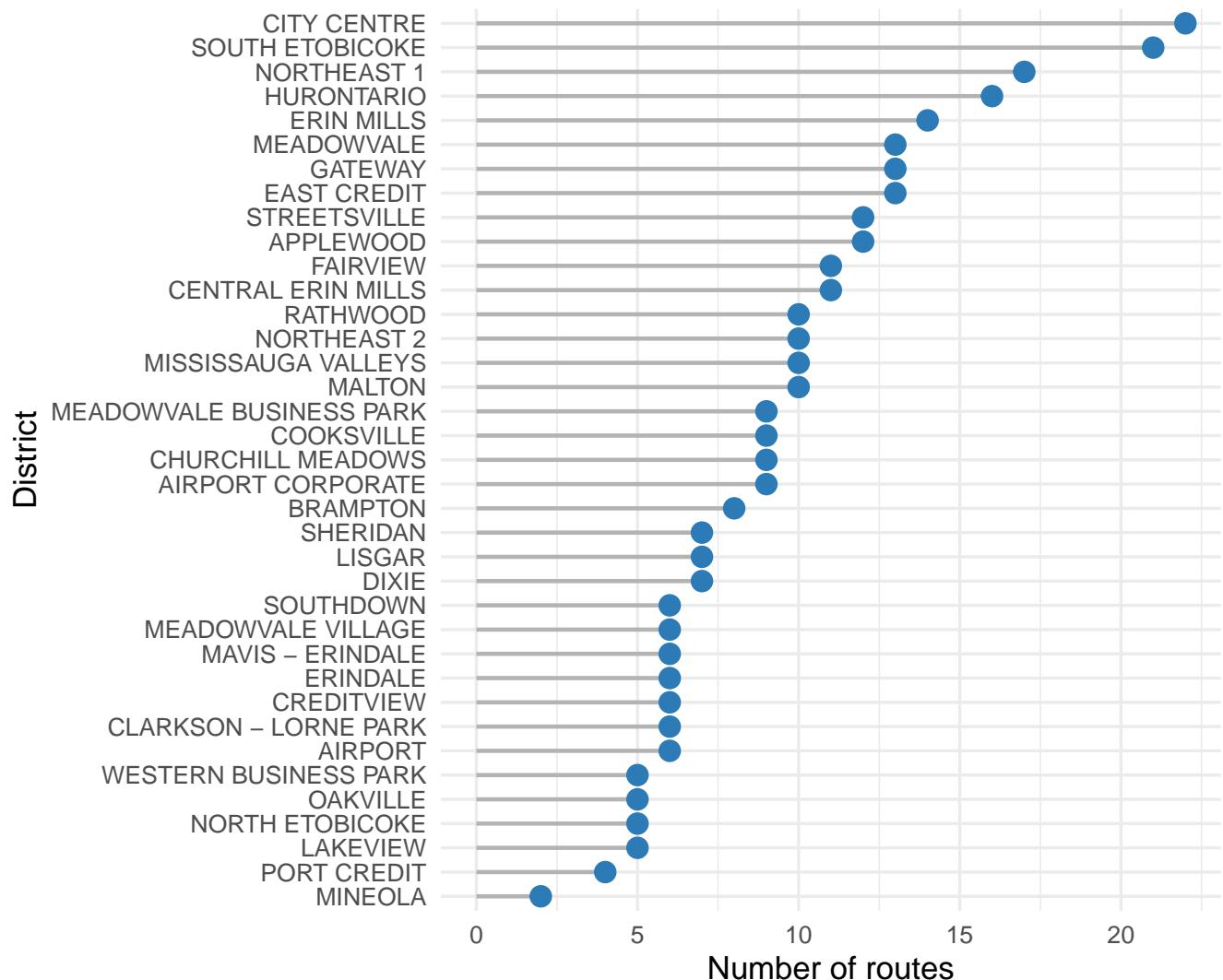
Results

1. MiWay Coverage Across Districts

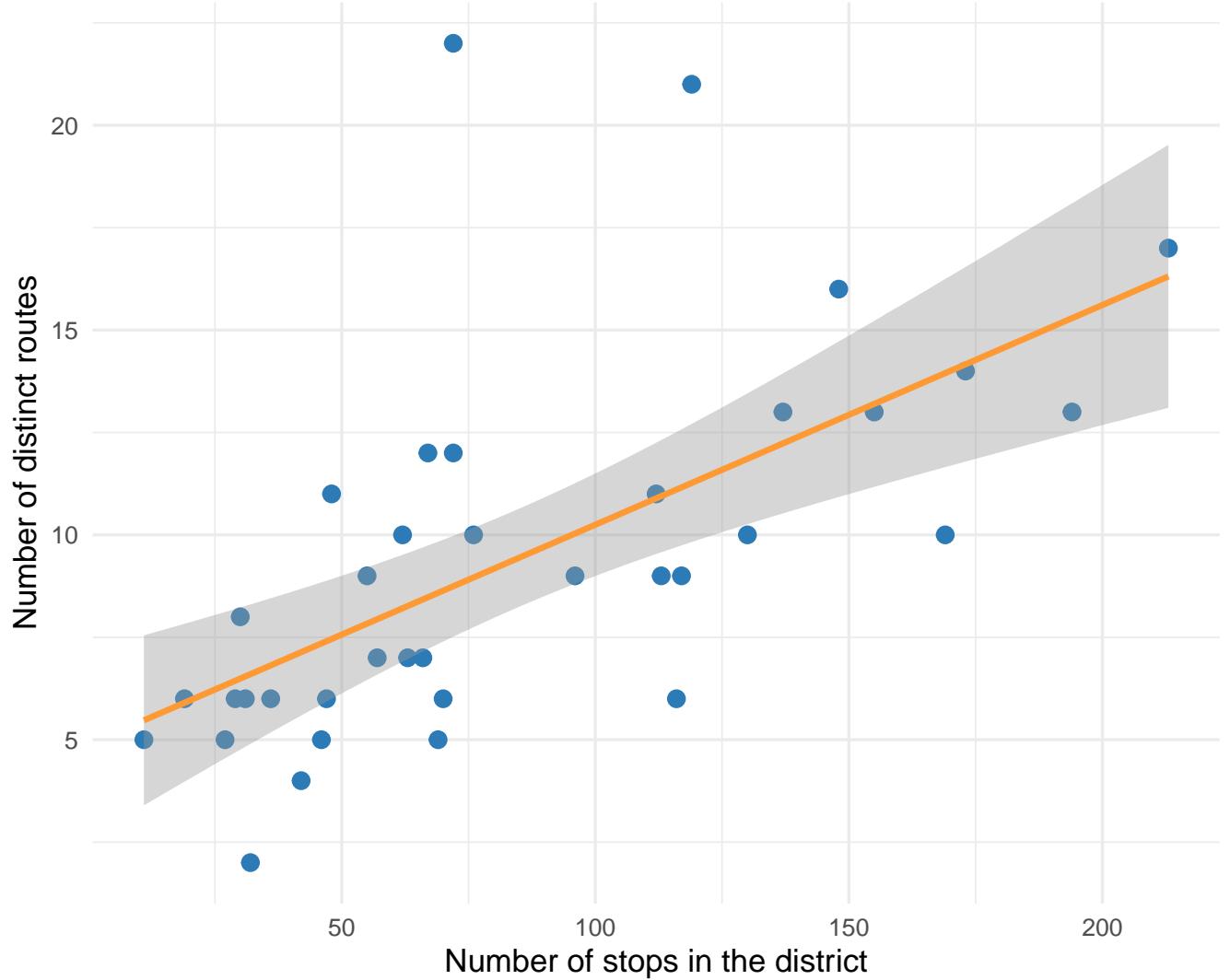
Districts differ widely in the number of MiWay bus stops they contain. Some districts along major corridors have significantly more stops, while others have fewer. Route diversity also varies, with some districts having many distinct routes and others offering fewer options.



Number of MiWay Routes by District



Relationship Between Number of Stops and Routes by District

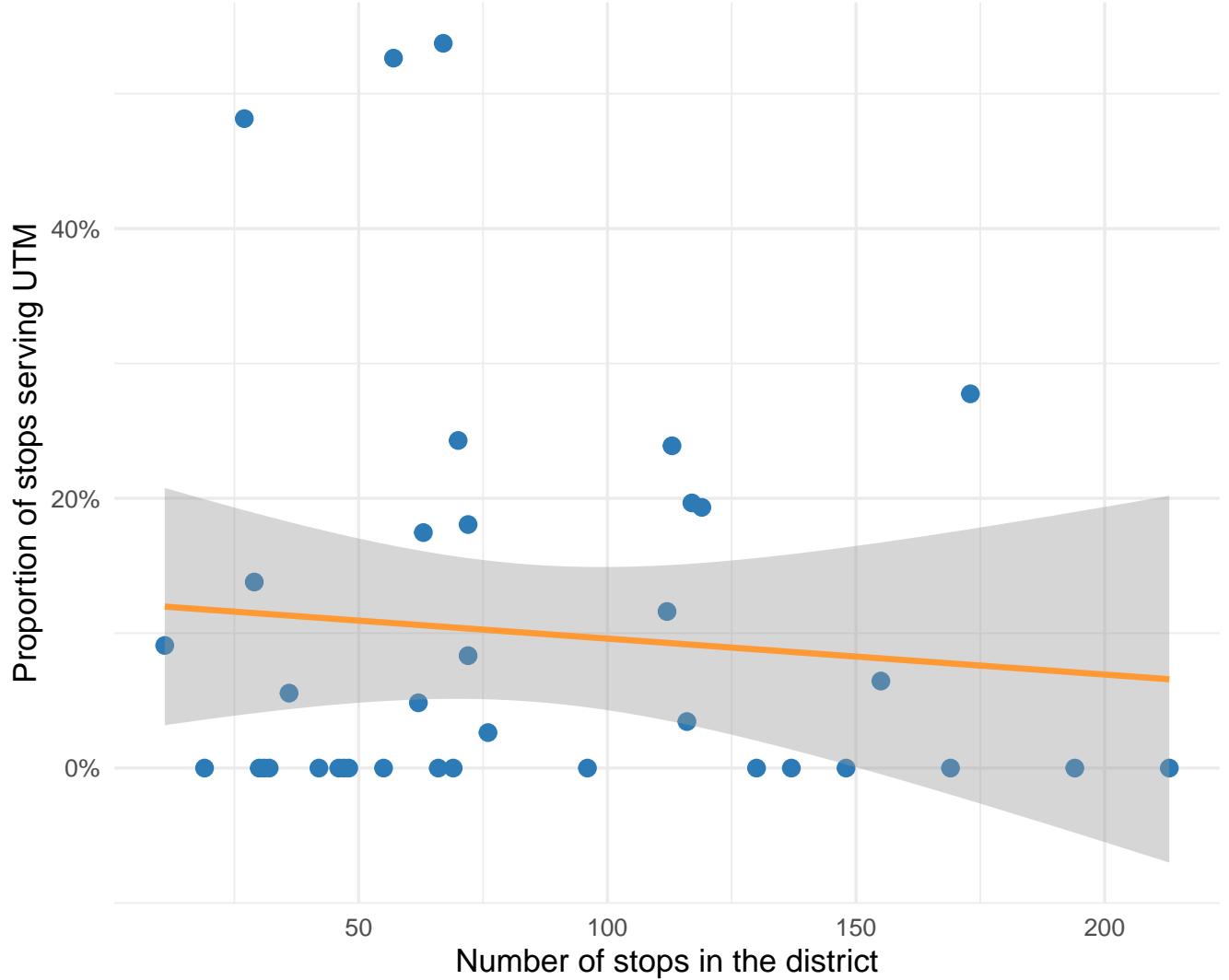


```
##
## Call:
## lm(formula = n_routes ~ n, data = coverage_summary)
##
## Residuals:
##    Min     1Q   Median     3Q    Max 
## -5.1061 -1.9451 -0.5463  1.0397 13.2543 
## 
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 4.88331   1.12679   4.334 0.000118 ***
## n          0.05364   0.01139   4.711 3.83e-05 ***
## ---        
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 
## Residual standard error: 3.59 on 35 degrees of freedom
## Multiple R-squared:  0.388, Adjusted R-squared:  0.3706
```

```
## F-statistic: 22.19 on 1 and 35 DF, p-value: 3.829e-05

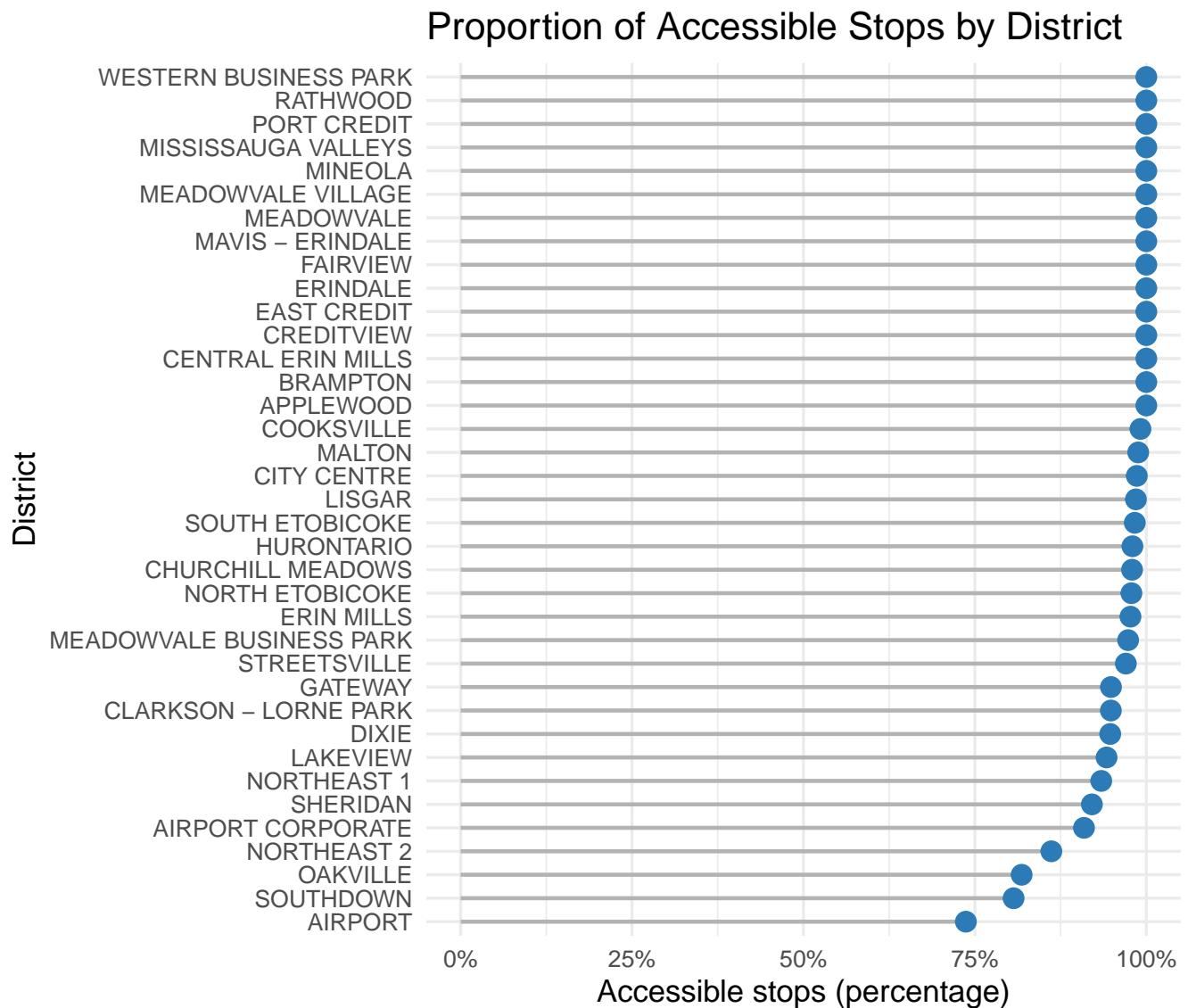
##
## Call:
## lm(formula = utm_rate ~ n, data = utm_with_counts)
##
## Residuals:
##      Min       1Q   Median       3Q      Max 
## -0.11759 -0.10506 -0.06589  0.06874  0.43252 
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 0.1226501 0.0477669  2.568  0.0147 *  
## n          -0.0002665 0.0004827 -0.552  0.5844    
## ---        
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 
##
## Residual standard error: 0.1522 on 35 degrees of freedom
## Multiple R-squared:  0.008632, Adjusted R-squared: -0.01969 
## F-statistic: 0.3048 on 1 and 35 DF, p-value: 0.5844
```

Direct UTM Access versus Number of Stops by District



2. Accessibility and Shelter Coverage

Most MiWay stops are designed so people with mobility challenges can board the bus more easily. Because of this city-wide design standard, almost every district has very high accessibility rates.

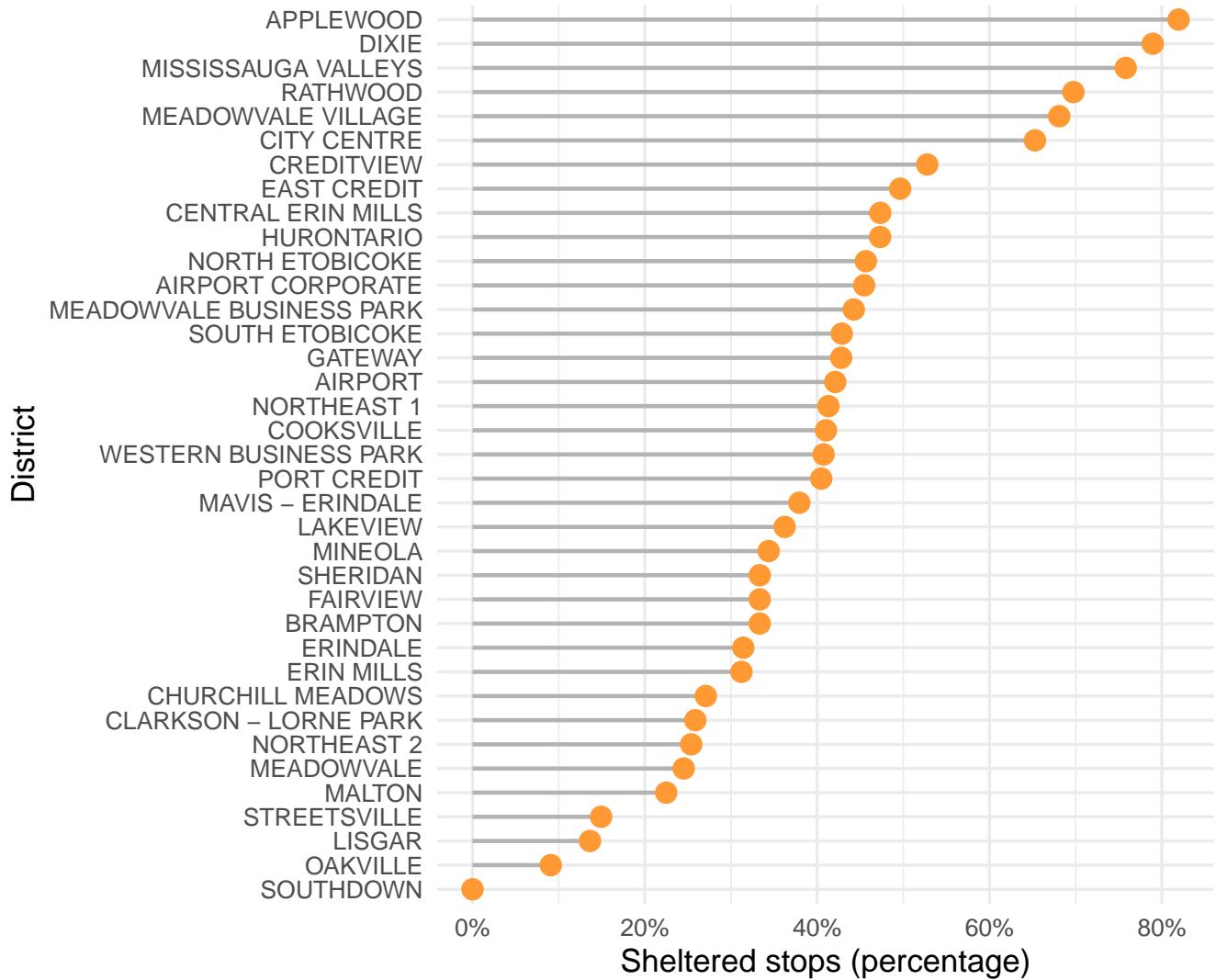


```
##  
## Pearson's Chi-squared test  
##  
## data: tab_access  
## X-squared = 179.32, df = 36, p-value < 2.2e-16
```

The chi-square test shows that accessibility is not evenly distributed across districts ($p < 0.001$). Although most stops in Mississauga are accessible, some districts have slightly higher or lower proportions than others, meaning accessibility is associated with district location.

Shelter coverage varies more widely:

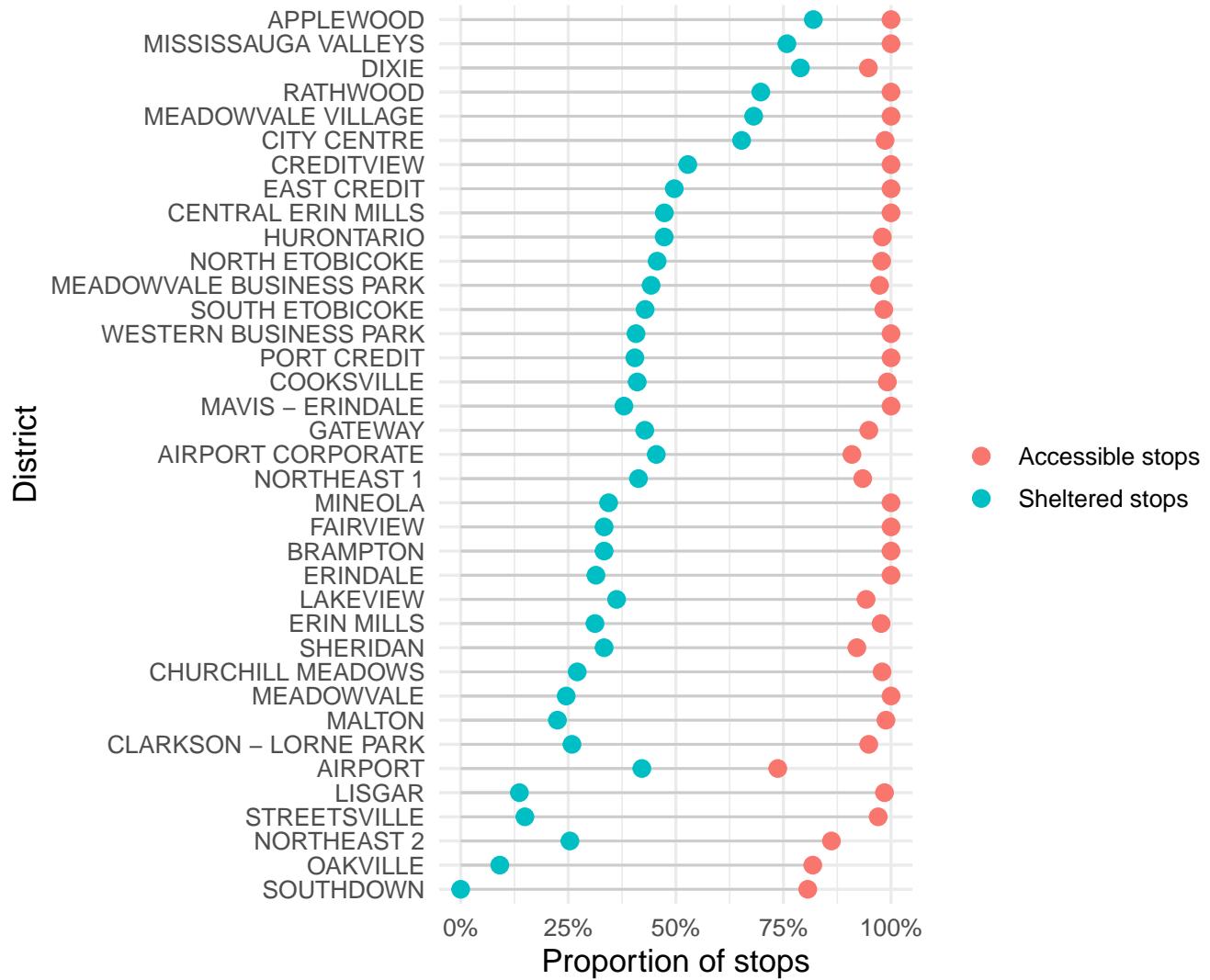
Proportion of Stops with Shelters by District



```
##
##  Pearson's Chi-squared test
##
## data: tab_shelter
## X-squared = 339.18, df = 36, p-value < 2.2e-16
```

The chi-square test indicates that shelter coverage differs significantly across districts ($p < 0.001$). Some districts provide many more sheltered stops than others, showing that weather protection is not evenly available across the city.

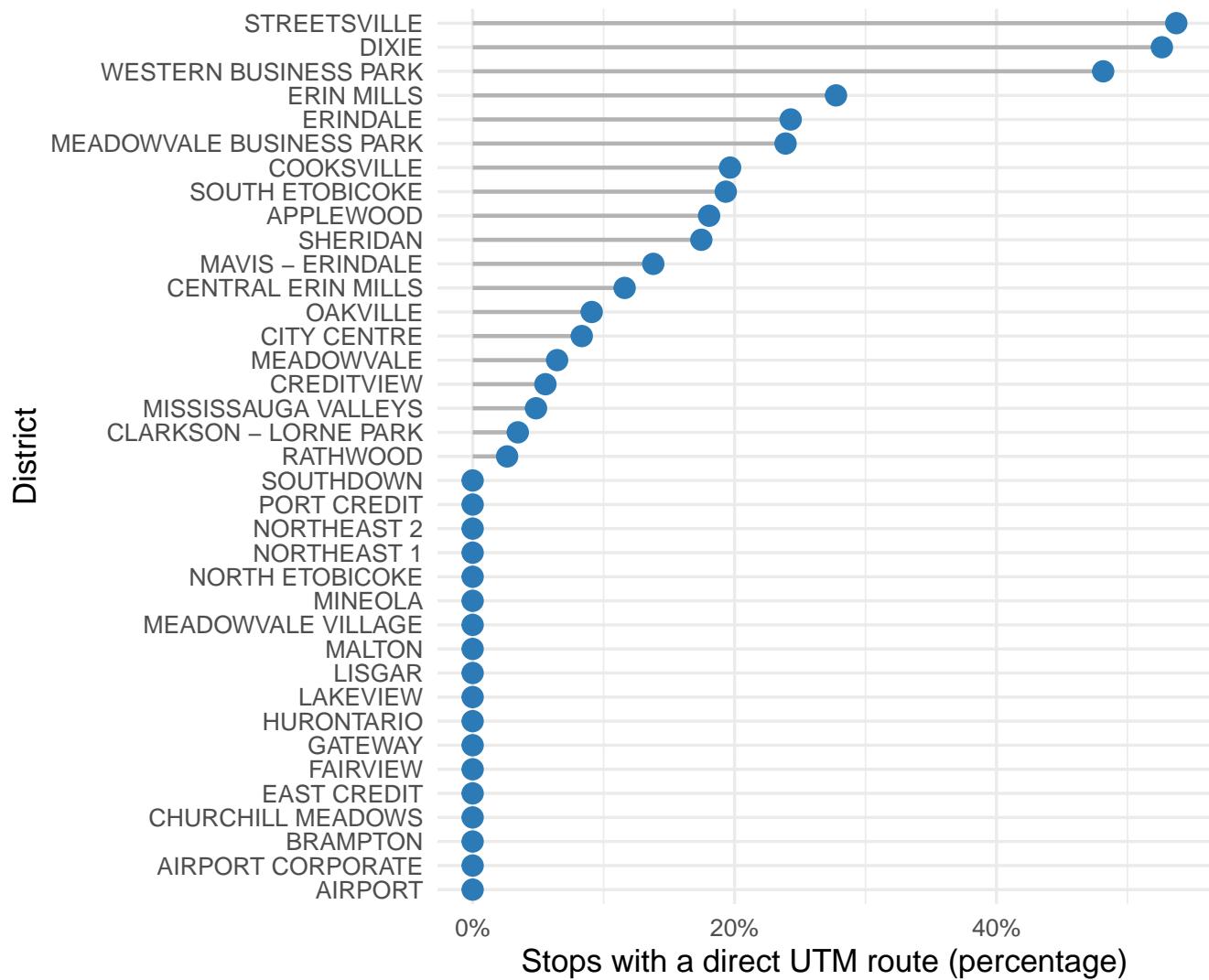
Accessibility and Shelter Coverage by District



3. Direct UTM Access

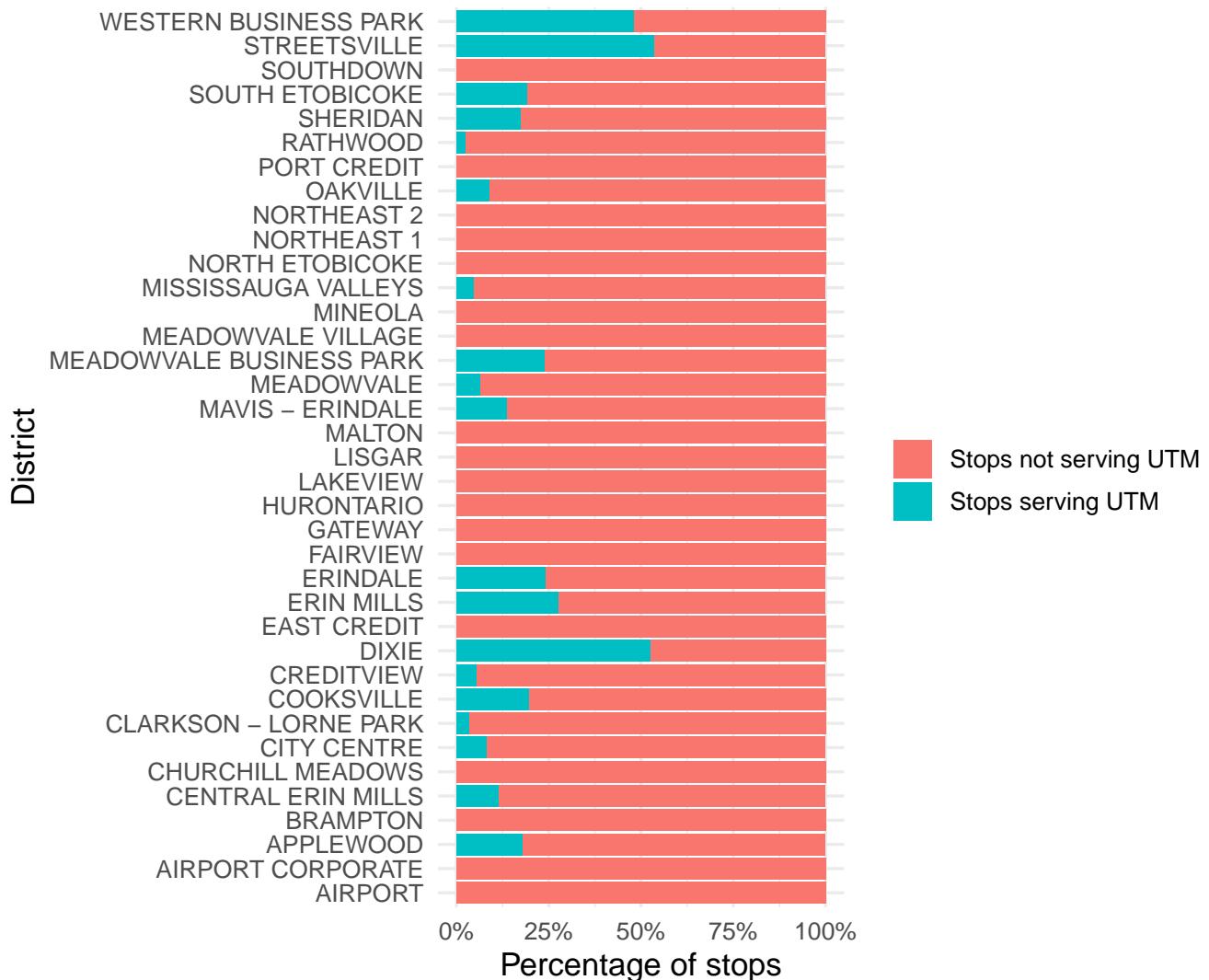
Direct UTM routes (1, 44, 101, 110, 126) concentrate in districts along Dundas and Mississauga Road.

Proportion of Stops Serving UTM by District



Normalized rates provide a fairer comparison:

Share of Stops Serving UTM by District



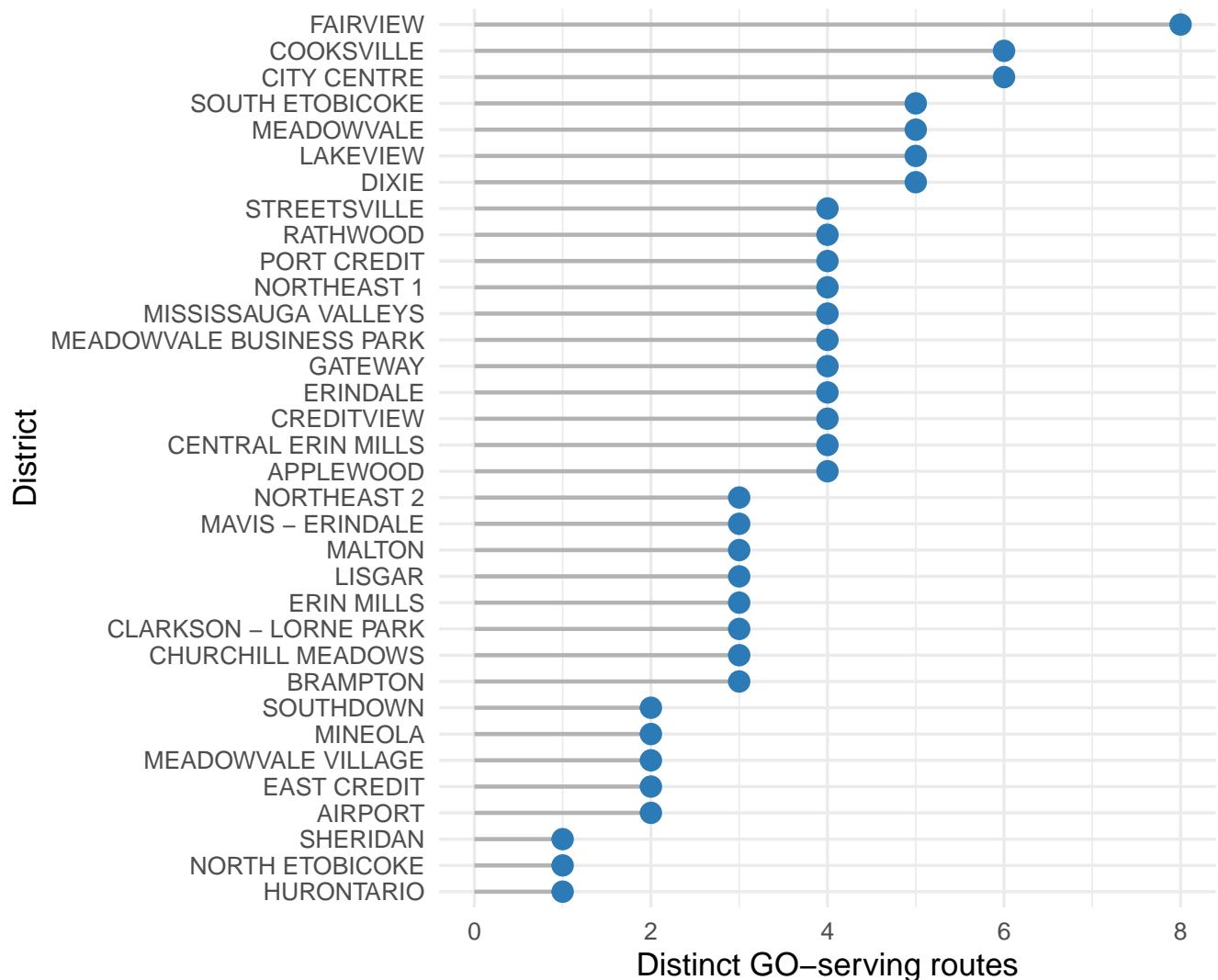
```
##
## Pearson's Chi-squared test
##
## data: tab_utm
## X-squared = 672.88, df = 36, p-value < 2.2e-16
```

The chi-square test shows a strong association between district and whether a stop serves a UTM route ($p < 0.001$). Districts along Dundas Street and Mississauga Road have far more UTM-serving stops than other districts, meaning direct access to campus varies considerably by location.

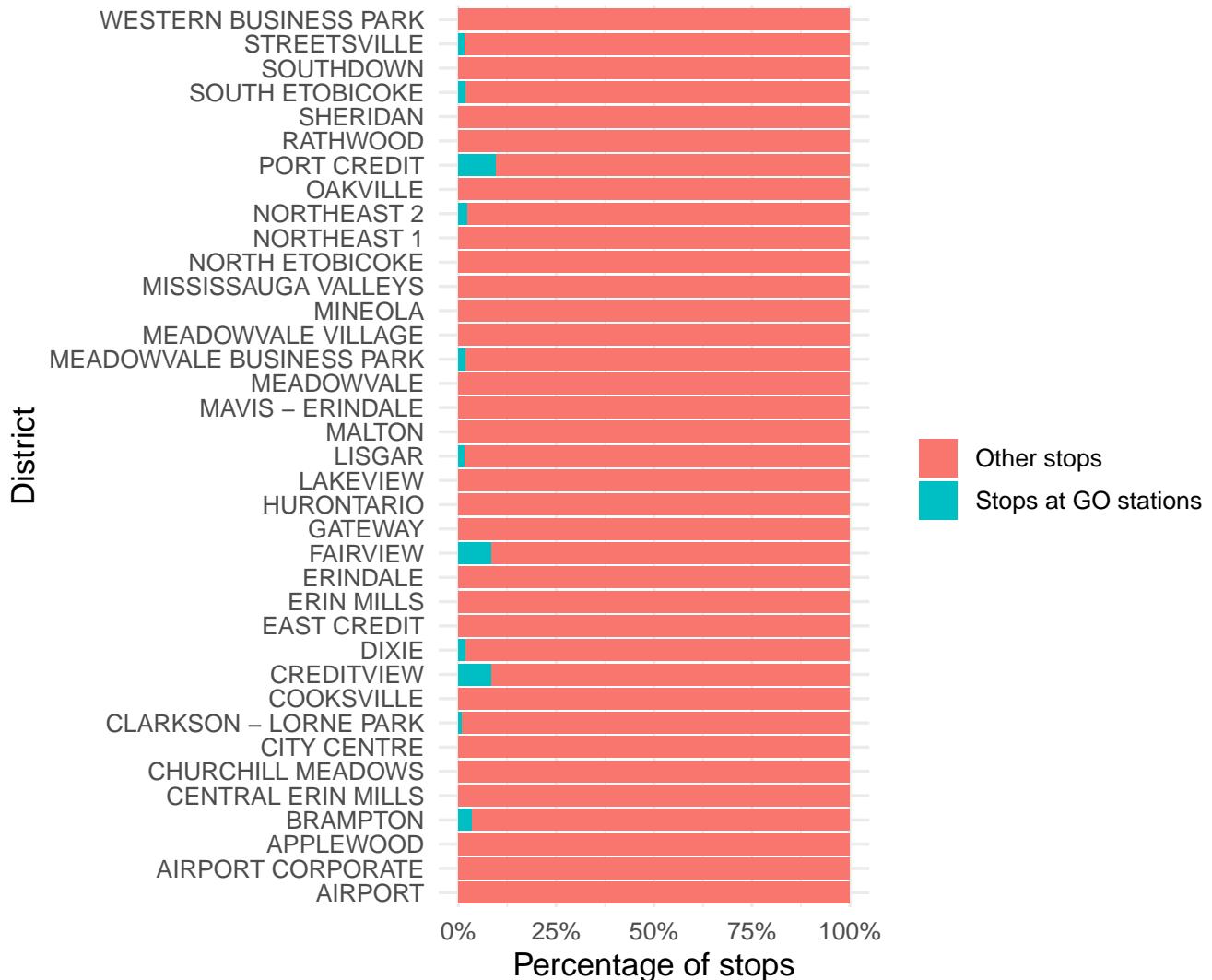
4. GO Train Connectivity

Districts near Cooksville, Port Credit, and Meadowvale GO stations have more GO-serving routes.

Number of MiWay Routes Connecting to GO ↗



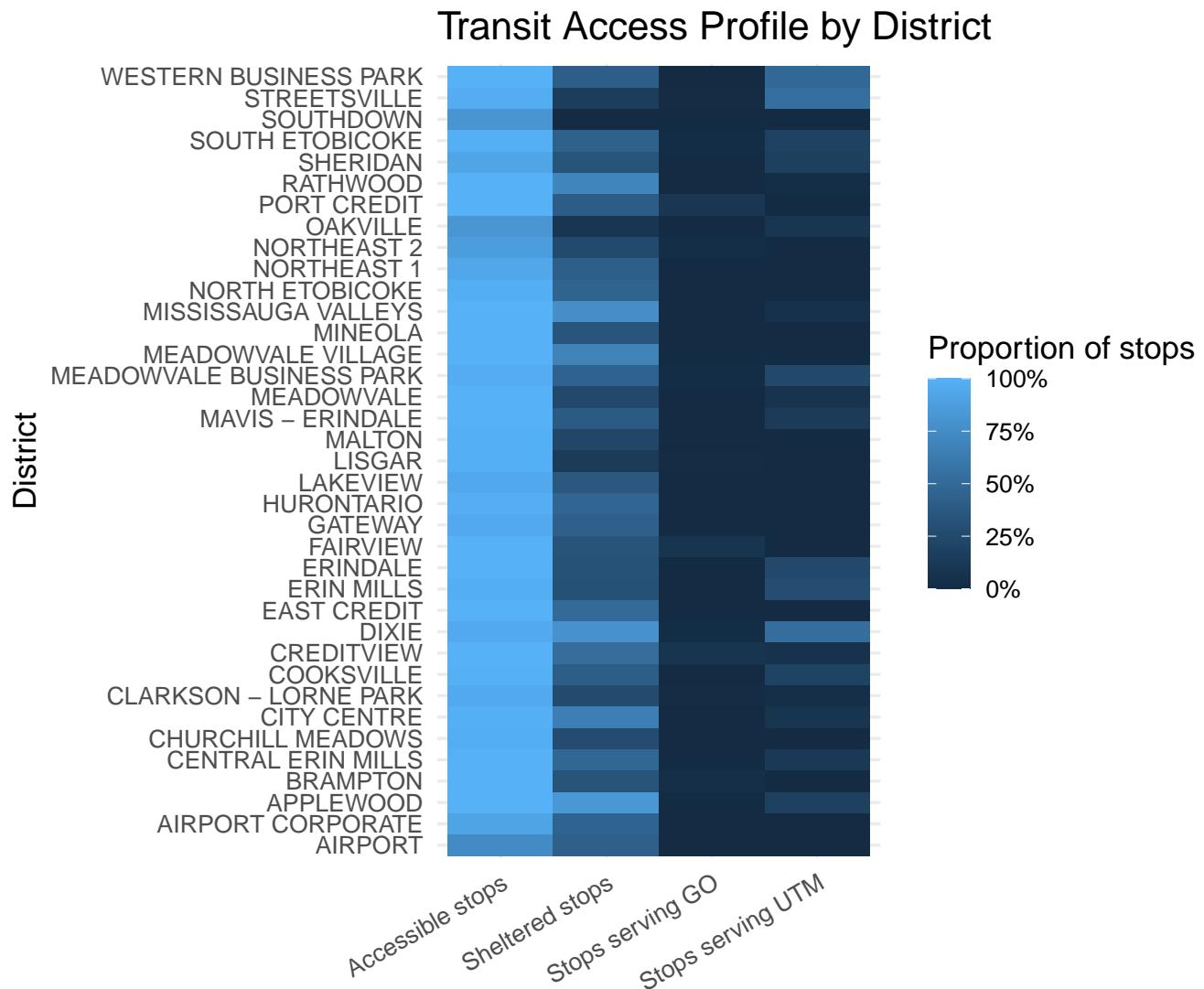
Share of Stops at GO Stations by District



```
##
##  Pearson's Chi-squared test
##
## data: tab_go
## X-squared = 139.69, df = 36, p-value = 3.843e-14
```

The chi-square test finds a significant relationship between district and whether a stop is located at a GO station ($p < 0.001$). Only a few districts contain GO stations, so GO-connected routes are concentrated in those districts rather than being evenly distributed.

5. Overall Transit Profile by District



Conclusion

Our analysis shows clear and statistically significant differences in MiWay transit access across Mississauga's districts. The chi-square tests for accessibility, shelter coverage, direct UTM service, and GO connectivity all provided very small p-values, indicating that these features are not evenly distributed across the city. Some districts, especially those along Dundas Street and Mississauga Road, have more stops that directly serve UTM, while districts containing GO stations have a greater number of routes that connect to regional rail. In contrast, other districts have fewer sheltered stops or fewer direct UTM and GO connections, which may make commuting less convenient or comfortable for students living there.

These patterns matter in practice because UTM students often choose where to live based on a mix of rent, safety, and transit access. Our results suggest that students who live in districts with strong UTM and GO connectivity may have more reliable and flexible commute options, while those in less-served areas may face longer or less convenient trips. At the same time, accessibility proportions are high across most districts, reflecting AODA requirements and an aging population, so the main differences lie in comfort (shelters) and connectivity (UTM and GO routes) rather than basic physical access.

There are several important limitations. First, our analysis uses only stop locations and route presence, not schedule or frequency information, so we cannot distinguish between a route that runs every few minutes and one that runs infrequently. Second, we do not incorporate travel time or crowding, which also affect how attractive a commute is. Finally, we work with district-level summaries, which may hide variation within districts.

Future work could address these gaps by incorporating GTFS schedule data to study service frequency and expected travel times, constructing a composite “transit access score” that combines shelters, accessibility, route diversity, UTM direct service, and GO connectivity, and linking MiWay data with student survey data on actual commute experiences. Together, these extensions would provide an even clearer picture of how transit infrastructure shapes the daily lives and choices of UTM students.

Appendix

```
# packages

library(dplyr)
library(ggplot2)
library(tidyr)
library(scales)

# read data

df <- read.csv("./Datasets/MiWay_Bus_Stops_-4490514804018385747.csv")

# clean districts

df <- df %>%
  mutate(District = trimws(District)) %>%
  filter(District != "", District != "UNKNOWN")

# UTM routes (MiWay routes that go to UTM)

utm_routes <- c("1", "44", "101", "110", "126")

# flag GO station stops

df$is_GO_stop <- grepl("GO STATION", df$Stop_Description, ignore.case = TRUE)

# one row per stop-route

routes_long <- df %>%
  separate_rows(Routes_Service, sep = "/") %>%
  mutate(Routes_Service = trimws(Routes_Service)) %>%
  filter(Routes_Service != "")

# stop-level data: whether each stop has a UTM route

stop_level <- routes_long %>%
  group_by(Stop_Number, District) %>%
  summarise(
    has_utm = any(Routes_Service %in% utm_routes),
    .groups = "drop"
  )

# routes that touch a GO station at least once

go_routes <- df %>%
  filter(is_GO_stop) %>%
  separate_rows(Routes_Service, sep = "/") %>%
  mutate(Routes_Service = trimws(Routes_Service)) %>%
  filter(Routes_Service != "") %>%
  pull(Routes_Service) %>%
  unique()
```

```

# summaries by district

counts <- df %>%
group_by(District) %>%
summarise(n = n()) %>%
arrange(desc(n))

access_summary <- df %>%
group_by(District) %>%
summarise(
accessible = sum(Accessible),
total      = n(),
proportion = accessible / total
) %>%
arrange(desc(proportion))

shelter_summary <- df %>%
group_by(District) %>%
summarise(
shelters          = sum(Shelter),
total            = n(),
proportion_sheltered = shelters / total
) %>%
arrange(desc(proportion_sheltered))

routes_by_district <- routes_long %>%
group_by(District) %>%
summarise(n_routes = n_distinct(Routes_Service)) %>%
arrange(desc(n_routes))

utm_long <- routes_long %>%
filter(Routes_Service %in% utm_routes)

district_utm <- utm_long %>%
group_by(District) %>%
summarise(
utm_stops        = n_distinct(Stop_Number),
utm_bus_occurrences = n(),
utm_distinct_routes = n_distinct(Routes_Service),
.groups = "drop"
) %>%
arrange(desc(utm_stops))

utm_normalized <- stop_level %>%
group_by(District) %>%
summarise(
total_stops = n(),
utm_stops   = sum(has_utm),
utm_rate    = utm_stops / total_stops,
.groups = "drop"
) %>%
arrange(desc(utm_rate))

```

```

go_by_district <- routes_long %>%
  filter(Routes_Service %in% go_routes) %>%
  group_by(District) %>%
  summarise(
    distinct_go_routes = n_distinct(Routes_Service),
    total_go_stops     = n_distinct(Stop_Number),
    .groups = "drop"
  ) %>%
  arrange(desc(distinct_go_routes))

utm_share <- stop_level %>%
  group_by(District) %>%
  summarise(
    utm      = sum(has_utm),
    non_utm  = sum(!has_utm),
    .groups = "drop"
  ) %>%
  pivot_longer(
    cols = c(utm, non_utm),
    names_to = "utm_type",
    values_to = "n"
  ) %>%
  group_by(District) %>%
  mutate(prop = n / sum(n)) %>%
  ungroup() %>%
  mutate(
    utm_type = recode(
      utm_type,
      utm = "Stops serving UTM",
      non_utm = "Stops not serving UTM"
    )
  )

go_share <- df %>%
  mutate(
    GO_type = if_else(is_GO_stop, "Stops at GO stations", "Other stops")
  ) %>%
  group_by(District, GO_type) %>%
  summarise(n = n(), .groups = "drop") %>%
  group_by(District) %>%
  mutate(prop = n / sum(n)) %>%
  ungroup()

district_profile <- df %>%
  group_by(District) %>%
  summarise(
    prop_accessible = mean(Accessible, na.rm = TRUE),
    prop_shelter    = mean(Shelter,    na.rm = TRUE),
    prop_utm        = mean(Stop_Number %in% stop_level$Stop_Number[stop_level$has_utm], na.rm = TRUE),
    prop_go         = mean(is_GO_stop, na.rm = TRUE),
    .groups = "drop"
  )

```

```

district_profile_long <- district_profile %>%
pivot_longer(
cols = starts_with("prop_"),
names_to = "feature",
values_to = "value"
) %>%
mutate(
feature = recode(
feature,
prop_accessible = "Accessible stops",
prop_shelter = "Sheltered stops",
prop_utm = "Stops serving UTM",
prop_go = "Stops serving GO"
)
)

tab_access <- table(df$District, df$Accessible)
chisq_access <- chisq.test(tab_access)

tab_shelter <- table(df$District, df$Shelter)
chisq_shelter <- chisq.test(tab_shelter)

tab_utm <- table(stop_level$District, stop_level$has_utm)
chisq_utm <- chisq.test(tab_utm)

tab_go <- table(df$District, df$is_GO_stop)
chisq_go <- chisq.test(tab_go)

chisq_access

## 
## Pearson's Chi-squared test
##
## data: tab_access
## X-squared = 179.32, df = 36, p-value < 2.2e-16

chisq_shelter

## 
## Pearson's Chi-squared test
##
## data: tab_shelter
## X-squared = 339.18, df = 36, p-value < 2.2e-16

chisq_utm

## 
## Pearson's Chi-squared test
##
## data: tab_utm
## X-squared = 672.88, df = 36, p-value < 2.2e-16

```

```
chisq_go
```

```
##  
## Pearson's Chi-squared test  
##  
## data: tab_go  
## X-squared = 139.69, df = 36, p-value = 3.843e-14
```