

```
library(readr)
```

Warning: package 'readr' was built under R version 4.4.3

```
library(dplyr)
library(lubridate)
library(ggplot2)
```

Warning: package 'ggplot2' was built under R version 4.4.3

```
library(here)
library(tidyr)
```

Warning: package 'tidyr' was built under R version 4.4.3

```
library(scales)
library(sf)
```

Warning: package 'sf' was built under R version 4.4.3

```
here:::i_am("hypothesis2.qmd")
```

```
# Load data and define the analysis window.
taxi <- read_csv(here("data", "taxidata_clean.csv"), show_col_types = FALSE) %>%
  mutate(
    year = year(trip_start_timestamp),
    pickup_community_area = as.integer(pickup_community_area)
  )

late_hours <- c(23, 0, 1, 2, 3, 4)

taxi_clean <- taxi %>%
  filter(year %in% c(2014, 2019), !is.na(pickup_community_area)) %>%
  mutate(
    hour = hour(trip_start_timestamp),
    late = hour %in% late_hours,
    dow = wday(trip_start_timestamp),
    month = month(trip_start_timestamp)
  )
```

## Per-area proportion tests with BH and BY correction

```
# Official City of Chicago community area names via Socrata open-data API.
# This avoids hard-coding names and ensures alignment with the city's numbering.
ca_lookup <- read_sf("https://data.cityofchicago.org/resource/igwz-8jzy.geojson") %>%
  st_drop_geometry() %>%
  transmute(
    pickup_community_area = as.integer(area_num_1),
    area_name = tools::toTitleCase(tolower(community)))
) %>%
  arrange(pickup_community_area)

# Area-year summary table.
area_year <- taxi_clean %>%
  group_by(pickup_community_area, year) %>%
  summarise(
    n_total = n(),
    n_late = sum(late),
    .groups = "drop"
  )

area_wide <- area_year %>%
  pivot_wider(
    names_from = year,
    values_from = c(n_total, n_late),
    values_fill = list(n_total = 0, n_late = 0)
  )

# One proportion test per area, then BH and BY correction.
test_results <- area_wide %>%
  filter(n_total_2014 >= 200, n_total_2019 >= 200) %>%
  rowwise() %>%
  mutate(
    p_value = prop.test(
      c(n_late_2014, n_late_2019),
      c(n_total_2014, n_total_2019),
      correct = FALSE
    )$p.value
  ) %>%
  ungroup() %>%
  left_join(ca_lookup, by = "pickup_community_area") %>%
  mutate(
```

```

late_share_2014 = n_late_2014 / n_total_2014,
late_share_2019 = n_late_2019 / n_total_2019,
share_change = late_share_2019 - late_share_2014,
p_adj_bh = p.adjust(p_value, method = "BH"),
p_adj_by = p.adjust(p_value, method = "BY"),
bh_reject = p_adj_bh < 0.05,
by_reject = p_adj_by < 0.05
) %>%
arrange(p_adj_bh)

test_results %>%
select(
  pickup_community_area, area_name,
  n_total_2014, n_total_2019,
  late_share_2014, late_share_2019, share_change,
  p_value, p_adj_bh, p_adj_by, bh_reject
) %>%
slice_head(n = 15)

# A tibble: 15 x 11
  pickup_community_area area_name      n_total_2014 n_total_2019 late_share_2014
              <int> <chr>                  <int>          <int>        <dbl>
1                      8 Near North S~     64253         64355       0.224
2                      32 Loop            40253         58501       0.0911
3                      28 Near West Si~    17825         23118       0.153
4                      7 Lincoln Park    13440         4384        0.338
5                      6 Lake View        16430         6751        0.345
6                      76 Ohare           9373          17374       0.146
7                      33 Near South S~   5217          5927        0.0947
8                      3 Uptown           3757          1922        0.259
9                      77 Edgewater        2199          1556        0.225
10                     4 Lincoln Squa~   1095          665         0.327
11                     24 West Town        9447         1967        0.381
12                     5 North Center     1760          428         0.313
13                     22 Logan Square     3270          732         0.454
14                     41 Hyde Park        1155          705         0.0632
15                     21 Avondale          627          259         0.319
# i 6 more variables: late_share_2019 <dbl>, share_change <dbl>, p_value <dbl>,
#   p_adj_bh <dbl>, p_adj_by <dbl>, bh_reject <lgl>

```

## Supplementary global-null tests

```
m <- nrow(test_results)

fisher_stat <- -2 * sum(log(test_results$p_value))
fisher_p <- pchisq(fisher_stat, df = 2 * m, lower.tail = FALSE)

p_sorted <- sort(test_results$p_value)
simes_p <- min(1, min(p_sorted * m / seq_len(m)))

tibble(
  test      = c("Fisher combination", "Simes"),
  p_value = c(fisher_p, simes_p)
)
```

```
# A tibble: 2 x 2
  test          p_value
  <chr>        <dbl>
1 Fisher combination    0
2 Simes           0
```

## Discovery counts

```
tibble(
  method      = c("BH (FDR control)", "BY (FDR under dependence)"),
  discoveries = c(sum(test_results$bh_reject), sum(test_results$by_reject)),
  tested      = c(m, m)
)
```

```
# A tibble: 2 x 3
  method          discoveries tested
  <chr>            <int>    <int>
1 BH (FDR control)       20     24
2 BY (FDR under dependence) 18     24
```

```
# Verify the direction of all BH discoveries.
n_decline <- sum(test_results$bh_reject & test_results$share_change < 0)
n_increase <- sum(test_results$bh_reject & test_results$share_change > 0)
tibble(
```

```

direction = c("Decline (share_change < 0)", "Increase (share_change > 0)" ),
n_discoveries = c(n_decline, n_increase)
)

```

```

# A tibble: 2 x 2
  direction          n_discoveries
  <chr>                  <int>
1 Decline (share_change < 0)      20
2 Increase (share_change > 0)       0

```

## Visualization

```

plot_df <- test_results %>%
  slice_max(order_by = abs(share_change), n = 20) %>%
  mutate(area_label = paste0(pickup_community_area, " - ", area_name))

ggplot(plot_df, aes(x = reorder(area_label, share_change), y = share_change, fill = bh_reject)
  geom_col() +
  coord_flip() +
  scale_y_continuous(labels = label_percent(accuracy = 0.1)) +
  scale_fill_manual(
    values = c(FALSE = "grey70", TRUE = "#1f78b4"),
    labels = c(FALSE = "Not rejected", TRUE = "Rejected")
  ) +
  labs(
    title = "Largest Changes in Late-Night Taxi Share",
    subtitle = "Blue = BH-significant at FDR 5%",
    x = NULL,
    y = "Change in late-night share (pp)",
    fill = "BH (alpha = 0.05)"
  ) +
  theme_minimal(base_size = 11) +
  theme(legend.position = "bottom")

```

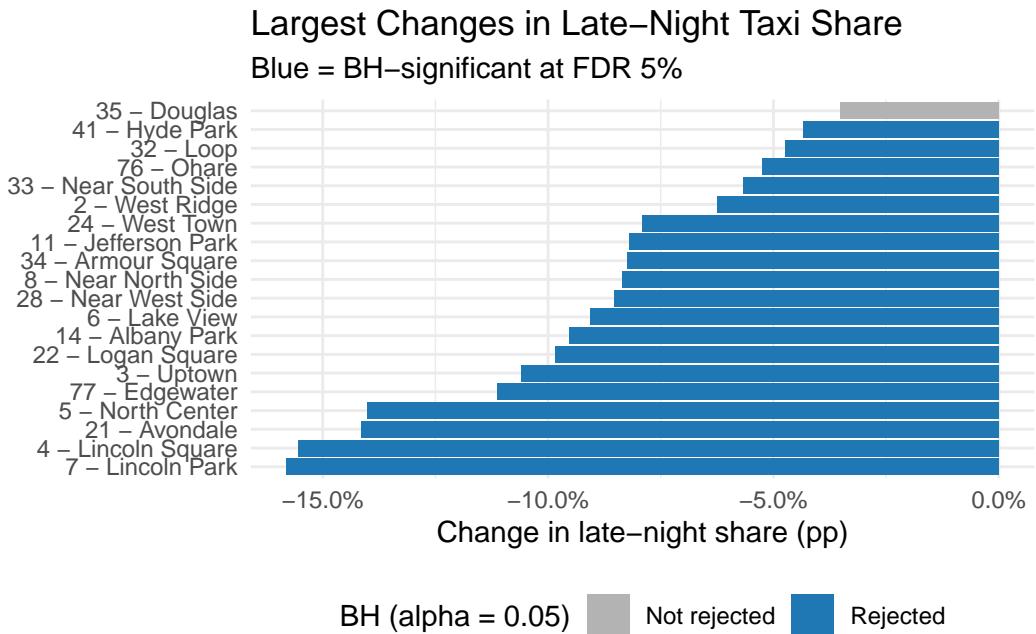


Figure 1: Change in late-night taxi share by community area (2019 minus 2014). Blue bars are BH-significant at FDR 5%; grey bars are not.

#### Grouped comparison: pre-specified area sets

```

nightlife_residential <- c(6, 7, 8, 22, 24)
airport_downtown_business <- c(28, 32, 33, 76)

group_year <- area_year %>%
  mutate(
    area_group = case_when(
      pickup_community_area %in% nightlife_residential ~ "Nightlife/Residential",
      pickup_community_area %in% airport_downtown_business ~ "Airport/Downtown/Business",
      TRUE ~ "Other"
    )
  ) %>%
  filter(area_group != "Other") %>%
  group_by(area_group, year) %>%
  summarise(
    n_total = sum(n_total),
    n_late = sum(n_late),
    .groups = "drop"
  )
  
```

```

)
group_results <- group_year %>%
  pivot_wider(
    names_from = year,
    values_from = c(n_total, n_late)
  ) %>%
  mutate(
    late_share_2014 = n_late_2014 / n_total_2014,
    late_share_2019 = n_late_2019 / n_total_2019,
    share_change     = late_share_2019 - late_share_2014
  ) %>%
  rowwise() %>%
  mutate(
    p_value = prop.test(
      c(n_late_2014, n_late_2019),
      c(n_total_2014, n_total_2019),
      correct = FALSE
    )$p.value
  ) %>%
  ungroup()

group_results %>%
  select(
    area_group, n_total_2014, n_total_2019,
    late_share_2014, late_share_2019, share_change, p_value
  )

# A tibble: 2 x 7
area_group      n_total_2014 n_total_2019 late_share_2014 late_share_2019
<chr>            <int>        <int>       <dbl>        <dbl>
1 Airport/Downtown/Business          72668        104920      0.114       0.0569
2 Nightlife/Residential             106840        78189       0.278       0.158
# i 2 more variables: share_change <dbl>, p_value <dbl>

# Difference-in-differences: year x group interaction test.
interaction_data <- taxi_clean %>%
  filter(
    pickup_community_area %in% c(nightlife_residential, airport_downtown_business)
  ) %>%
  mutate(
    nightlife = as.integer(pickup_community_area %in% nightlife_residential),

```

```

    post      = as.integer(year == 2019)
  )

fit <- glm(late ~ post * nightlife, data = interaction_data, family = binomial)
summary(fit)

```

Call:

```
glm(formula = late ~ post * nightlife, family = binomial, data = interaction_data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )							
(Intercept)	-2.05409	0.01169	-175.74	<2e-16 ***							
post	-0.75360	0.01773	-42.52	<2e-16 ***							
nightlife	1.09854	0.01354	81.14	<2e-16 ***							
post:nightlife	0.03953	0.02137	1.85	0.0644 .							
---											
Signif. codes:	0	'***'	0.001	'**'	0.01	'*'	0.05	'. '	0.1	' '	1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 313092 on 362616 degrees of freedom  
 Residual deviance: 291890 on 362613 degrees of freedom  
 AIC: 291898

Number of Fisher Scoring iterations: 5

```

# Extract and interpret the interaction coefficient.
beta3 <- coef(fit)["post:nightlife"]
se3 <- summary(fit)$coefficients["post:nightlife", "Std. Error"]
p3 <- summary(fit)$coefficients["post:nightlife", "Pr(>|z|)"]
tibble(
  term      = "year2019 x nightlife (interaction)",
  estimate  = beta3,
  std_error = se3,
  p_value   = p3
)

```

```

# A tibble: 1 x 4
term                  estimate std_error p_value
<chr>                <dbl>     <dbl>    <dbl>
1 year2019 x nightlife (interaction) 0.0395   0.0214  0.0644

```

## Robustness: stratified permutation test

```
set.seed(42)
B <- 2000

# Observed late-night share difference per area.
obs_diff <- taxi_clean %>%
  group_by(pickup_community_area, year) %>%
  summarise(late_share = mean(late), n = n(), .groups = "drop") %>%
  pivot_wider(
    names_from = year, values_from = c(late_share, n),
    values_fill = list(late_share = NA, n = 0)
  ) %>%
  filter(n_2014 >= 200, n_2019 >= 200) %>%
  mutate(obs_stat = late_share_2019 - late_share_2014)

# Keep only areas passing the sample-size filter.
areas_to_test <- obs_diff$pickup_community_area

perm_data <- taxi_clean %>%
  filter(pickup_community_area %in% areas_to_test) %>%
  mutate(stratum = interaction(month, dow, drop = TRUE)) %>%
  select(pickup_community_area, year, late, stratum)

# Function: one permutation round for all areas.
# We shuffle year labels within each (area, stratum) cell using base-R split-apply
# to avoid the dplyr grouped-mutate size constraint.
# Note: sample(x) when length(x)==1 is interpreted as sample.int(x,1), so we use
# x[sample.int(length(x))] instead, which is safe for any length.
one_perm <- function(df) {
  idx <- split(seq_len(nrow(df)), list(df$pickup_community_area, df$stratum))
  year_perm <- df$year # start as a copy; overwrite in place
  for (i in idx) {
    n_i <- length(i)
    if (n_i > 1L) {
      year_perm[i] <- df$year[i][sample.int(n_i)]
    }
    # n_i <= 1: nothing to permute
  }
  df$year_perm <- year_perm
}

df %>%
```

```

group_by(pickup_community_area, year_perm) %>%
  summarise(late_share = mean(late), .groups = "drop") %>%
  pivot_wider(names_from = year_perm, values_from = late_share) %>%
  mutate(perm_stat = `2019` - `2014`) %>%
  select(pickup_community_area, perm_stat)
}

# Run B permutations.
perm_stats <- bind_rows(lapply(seq_len(B), function(b) {
  one_perm(perm_data) %>% mutate(b = b)
}))

# Compute two-sided permutation p-values.
perm_pvals <- perm_stats %>%
  inner_join(obs_diff %>% select(pickup_community_area, obs_stat), by = "pickup_community_area") %>%
  group_by(pickup_community_area) %>%
  summarise(
    perm_p = (sum(abs(perm_stat) >= abs(obs_stat)) + 1) / (n() + 1),
    .groups = "drop"
  )

# Merge with test_results and apply BH.
perm_results <- obs_diff %>%
  select(pickup_community_area, obs_stat) %>%
  inner_join(perm_pvals, by = "pickup_community_area") %>%
  mutate(
    perm_p_bh     = p.adjust(perm_p, method = "BH"),
    perm_reject  = perm_p_bh < 0.05
  ) %>%
  arrange(perm_p_bh)

perm_results <- perm_results %>%
  left_join(ca_lookup, by = "pickup_community_area")

perm_results %>%
  select(pickup_community_area, area_name, obs_stat, perm_p, perm_p_bh, perm_reject) %>%
  slice_head(n = 15)

# A tibble: 15 x 6
  pickup_community_area area_name      obs_stat    perm_p perm_p_bh perm_reject
  <int> <chr>          <dbl>      <dbl>      <dbl>      <lgl>
1           3 Uptown      -0.106   0.000500  0.000750 TRUE

```

```

2          4 Lincoln Square    -0.156  0.000500  0.000750 TRUE
3          5 North Center     -0.140  0.000500  0.000750 TRUE
4          6 Lake View        -0.0906 0.000500  0.000750 TRUE
5          7 Lincoln Park      -0.158  0.000500  0.000750 TRUE
6          8 Near North Side   -0.0835 0.000500  0.000750 TRUE
7          21 Avondale         -0.141  0.000500  0.000750 TRUE
8          22 Logan Square     -0.0983 0.000500  0.000750 TRUE
9          24 West Town        -0.0792 0.000500  0.000750 TRUE
10         28 Near West Side   -0.0853 0.000500  0.000750 TRUE
11         32 Loop             -0.0474 0.000500  0.000750 TRUE
12         33 Near South Side  -0.0567 0.000500  0.000750 TRUE
13         41 Hyde Park        -0.0433 0.000500  0.000750 TRUE
14         56 Garfield Ridge   -0.0277 0.000500  0.000750 TRUE
15         76 Ohare            -0.0525 0.000500  0.000750 TRUE

```

```
cat("Permutation-based BH discoveries:", sum(perm_results$perm_reject), "of", nrow(perm_resu
```

Permutation-based BH discoveries: 20 of 24

```
cat("Parametric BH discoveries:      ", sum(test_results$bh_reject), "of", nrow(test_resul
```

Parametric BH discoveries: 20 of 24