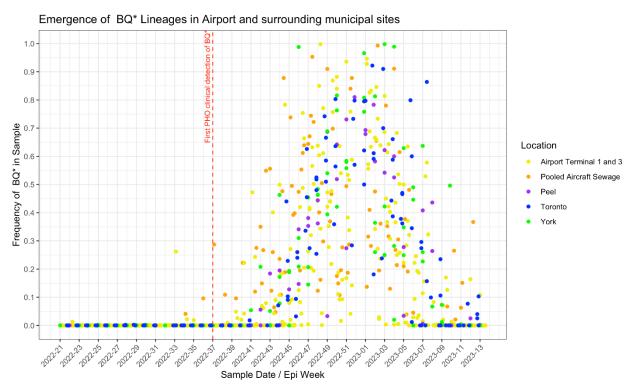
# Genomic Surveillance of Canadian Airport Wastewater Samples Allows Early Detection of Emerging SARS-CoV-2 Lineages

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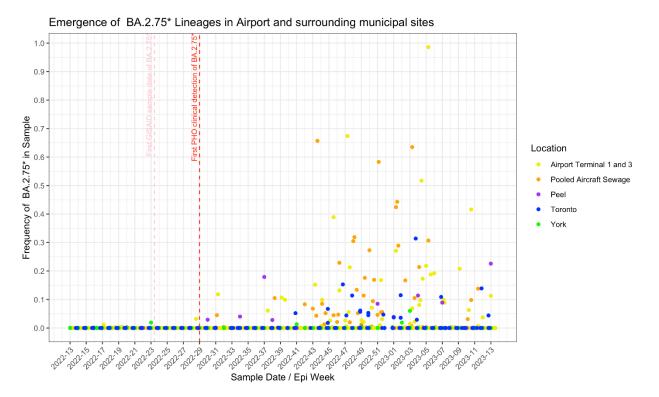
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<sup>5</sup>Regional Municipality of Peel
<sup>6</sup>Toronto Public Health
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# Supplementary Figures S1-S7, Supplementary table S3 and Supplementary methods

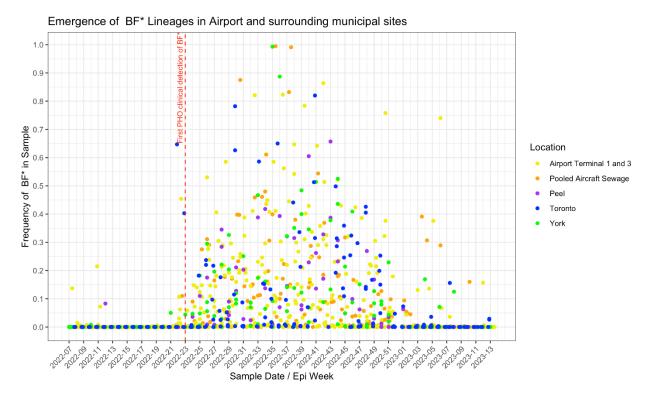


## Supplementary Figures and Tables

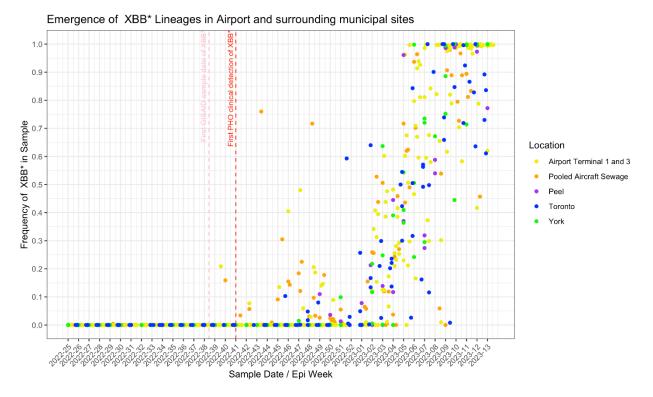
Supplementary Figure S1. Frequency of BQ\* lineages in WW samples from Toronto Pearson and surrounding municipal sites in Toronto, York, and Peel regions. Each point represents a single WW sample. Red dashed line represents the date of the first clinical sequence for BQ\* in Ontario from PHO reports.



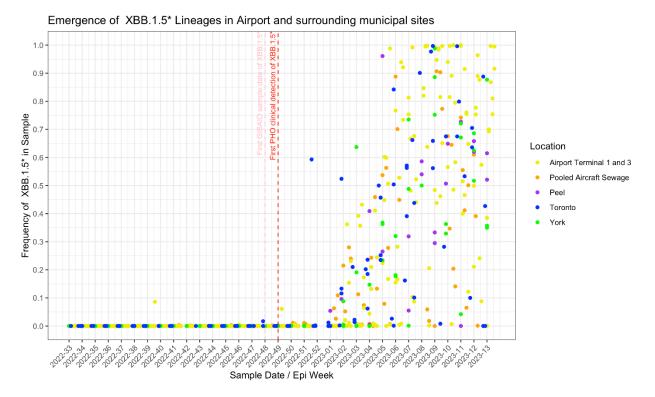
Supplementary Figure S2. Frequency of BA.2.75\* lineages in WW samples from Toronto Pearson and surrounding municipal sites in Toronto, York, and Peel regions. Each point represents a single WW sample. Red dashed line represents the date of the first clinical sequence for BA.2.75\* in Ontario from PHO reports. Pink dashed line represents the date of the first clinical sequence for BA.2.75 available in GISAID.



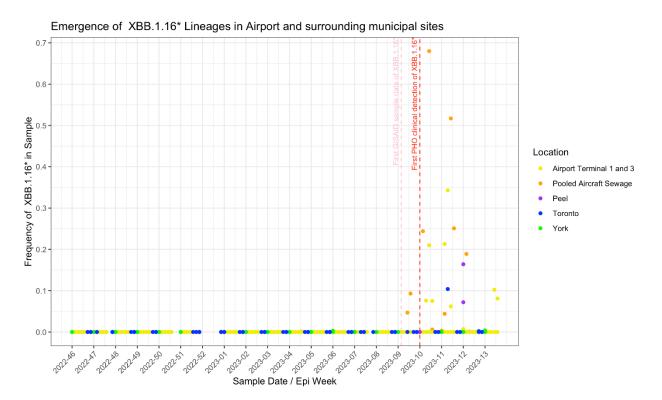
Supplementary Figure S3. Frequency of BF\* lineages in WW samples from Toronto Pearson and surrounding municipal sites in Toronto, York, and Peel regions. Each point represents a single WW sample. Red dashed line represents the date of the first clinical sequence of BF\* in Ontario from PHO reports.



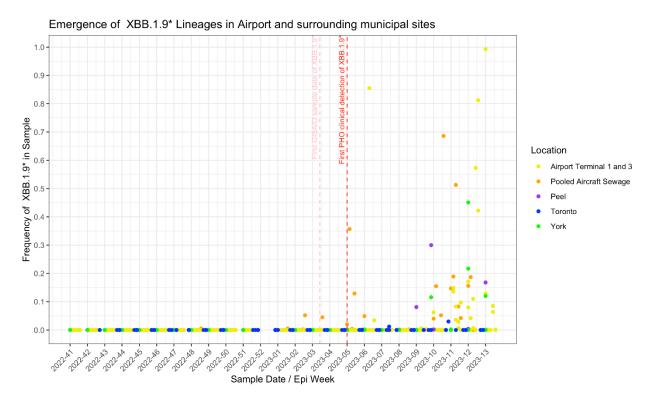
Supplementary Figure S4. Frequency of XBB\* lineages in WW samples from Toronto Pearson and surrounding municipal sites in Toronto, York, and Peel regions. Each point represents a single WW sample. Red dashed line represents the date of the first clinical sequence of XBB\* in Ontario from PHO reports. Pink dashed line represents the date of the first clinical sequence for XBB\* available in GISAID.



Supplementary Figure S5. Frequency of XBB.1.5\* lineages in WW samples from Toronto Pearson and surrounding municipal sites in Toronto, York, and Peel regions. Each point represents a single WW sample. Red dashed line represents the date of the first clinical sequence of XBB.1.5\* in Ontario from PHO reports. Pink dashed line represents the date of the first clinical sequence for XBB.1.5\* available in GISAID.



Supplementary Figure S6. Frequency of XBB.1.16\* lineages in WW samples from Toronto Pearson and surrounding municipal sites in Toronto, York, and Peel regions. Each point represents a single WW sample. Red dashed line represents the date of the first clinical sequence of XBB.1.16\* in Ontario from PHO reports. Pink dashed line represents the date of the first clinical sequence for XBB.1.16\* available in GISAID.



Supplementary Figure S7. Frequency of XBB.1.9\* lineages in WW samples from Toronto Pearson and surrounding municipal sites in Toronto, York, and Peel regions. Each point represents a single WW sample. Red dashed line represents the date of the first clinical sequence of XBB.1.9\* in Ontario from PHO reports. Pink dashed line represents the date of the first clinical sequence for XBB.1.9\* available in GISAID.

Supplementary Tables S1 and S2 are attached as spreadsheets

Sample	SequencingPartner	AverageBOC	Median BOC	st dev
		Region		•
Terminal 1 and 3	Waterloo	76.55423729	82	18.38826839
Pooled Aircraft	Guelph	80.87124464	96	27.17438024
York	Waterloo	71.46892655	76	20.65701603
Peel	Waterloo	77.07142857	83.5	19.06012982
Toronto	Western	66.75618375	76	25.45294117
		Terminal		•
Terminal 1		75.35932203	80	18.03216635
Terminal 3		77.74915254	84	18.66124755
	Pooled A	ircraft Sample Method		•
Aircraft passive sampler		73.07801418	90	30.11492665
Aircraft auto sampler		92.81521739	98	15.63714179
	Munic	ipal Sample Method	•	
Municipal Grab		74.45132743	79	19.93041
Municipal Composite 24hr		69.42917548	78	23.75998187
	Se	quencing Partner	-	-
	Waterloo	75.61659193	81	19.07687053
	Guelph	80.87124464	96	27.17438024
	Western	66.75618375	76	25.45294117

#### Supplementary Table S3: Breadth of Coverage Analysis

Table showing average, median and st dev of BOC > 10 data broken into different sample groups as indicated by sub-headers for comparison. All sequenced samples were included even if they were discarded from lineage calling analysis.

### Supplementary Methods

#### Sum Frequencies for Lineages - Python

```
import pandas as pd
import json
```

# Load the alias key
with open('/Users/jennk/Documents/Data\_Chales\_Lab/paper\_airport/alias\_key.json', 'r') as f:
 alias\_key = json.load(f)

# Load the CSV file csv\_file\_path = '/Users/jennk/Documents/Data\_Chales\_Lab/paper\_airport/all\_samples\_lineages.csv' df = pd.read\_csv(csv\_file\_path)

```
# Long form names for specified lineages
lineages_of_interest = {
  'BQ': 'B.1.1.529.5.3.1.1.1.1',
  'BA.2.75': 'B.1.1.529.2.75',
  'BF': 'B.1.1.529.5.2.1',
  'XBB': 'XBB', # Special handling required
  'XBB.1.5': 'XBB.1.5',
  'XBB.1.16': 'XBB.1.16',
  'XBB.1.9': 'XBB.1.9'
}
# Function to get all child lineages for a given lineage
def get_subset_keys(dictionary, start_string):
  return {key for key, value in dictionary.items() if isinstance(value, str) and
value.startswith(start_string) or isinstance(value, list) and any(isinstance(item, str) and
item.startswith(start_string) for item in value)}
lineage_results = {}
for lineage, start string in lineages of interest.items():
  lineage_keys = get_subset_keys(alias_key, start_string)
  lineage_keys.add(lineage) # Include the lineage itself
  lineage_results[lineage] = lineage_keys
# Exclude XBF from BF lineage
lineage_results['BF'] = {key for key in lineage_results['BF'] if not key.startswith('XBF')}
# Initialize dictionary to store lineage columns
lineage_columns = {lineage: set() for lineage in lineage_results.keys()}
# Identify columns matching each lineage and its children
for lineage, child keys in lineage results.items():
  for col in df.columns:
    col lineages = col.split(' or ')
    if all(any(child_key in lineage_part for child_key in child_keys) for lineage_part in col_lineages):
       lineage_columns[lineage].add(col)
# Debugging: print identified columns for each lineage
for lineage, columns in lineage_columns.items():
  print(f"{lineage}: {columns}")
# Function to sum columns for each lineage and round to 3 decimal places
def sum lineage columns(df, lineage, columns):
  if columns:
    df[f'{lineage} summary'] = df[list(columns)].sum(axis=1).round(3)
```

else: df[f'{lineage}\_summary'] = 0

# Sum the columns and create summary columns for lineage, columns in lineage\_columns.items(): sum\_lineage\_columns(df, lineage, columns)

```
# Reorder columns to insert summary columns after 'BreadthOfCoverage'
summary_columns = [f'{lineage}_summary' for lineage in lineages_of_interest]
cols = list(df.columns)
for col in summary_columns:
    cols.insert(3, cols.pop(cols.index(col)))
```

df = df[cols]

```
# Save the modified DataFrame back to a CSV file
output_csv_file_path =
'/Users/jennk/Documents/Data_Chales_Lab/paper_airport/summary_all_samples_lineages.csv'
df.to_csv(output_csv_file_path, index=False)
```

### Data Preparation for Plots

```
suppressPackageStartupMessages({
```

```
library(readxl)
library(dplyr)
library(tidyr)
library(ggplot2)
theme_set(theme_bw())
library(lubridate)
```

})

### NOT RUN # Processing script to remove unused columns and # give columns standardized names.

# This was run once then copied here for posterity.

```
air <- air %>%
mutate(Location = gsub("(20\\d+)|(-\\d+)", "", Sample_Name)) %>%
mutate(Location = ifelse(
```

```
Location %in% c("Ashbridges", "AshbridgesBay", "AshbridgeBay"),
    yes = "Ashbridges",
    no = ifelse(
      Location %in% c("HighlandCreek", "Highland", "HighlandCR"),
      yes = "HighlandCreek",
      no = Location)
  )) %>%
  mutate(Group = case when(
    Location %in% c("Humber", "Ashbridges", "HighlandCreek", "NorthToronto") ~ "Toronto",
    Location %in% c("As", "At") ~ "Pooled Aircraft Sewage",
    Location %in% c("A1", "A3") ~ "Airport Terminal 1 and 3",
    Location %in% c("P1", "P2") ~ "Peel",
    Location %in% c("Y1", "Y5", "Y6") ~ "York",
    TRUE ~ "Devan Missed One"
  ))
air <- select(air, -Sample_Name) %>%
  pivot longer(cols = !c(Group, Location, Sample Date, Breadth of Coverage),
    values to = "Frequency", names to = "Lineage") %>%
  mutate(Sample Date = ymd(Sample Date),
    Frequency = ifelse(Frequency > 1, 1, Frequency))
write.csv(air, file = "summary_all_samples clean.csv")
rm(air1)
### END NOT RUN
air <- read.csv("summary_all_samples_clean.csv") %>%
  filter(Breadth_of_Coverage >= 20)
gisaid <- data.frame(
  Lineage = c("BQ*", "BA.2.75*", "BF*", "XBB*",
    "XBB.1.5*", "XBB.1.16*", "XBB.1.9*"),
  Gisaid_Sample = ymd(c("", "2022-06-09", "", "2022-09-22",
      "2022-11-28", "2023-02-28", "2023-01-19")),
  Gisaid_Report = ymd(c("", "2022-06-21", "", "2022-10-03",
      "2022-12-12", "2023-04-24", ""))
)
pho <- data.frame(
  Lineage = c("BQ*", "BA.2.75*", "BF*", "XBB*",
    "XBB.1.5*", "XBB.1.16*", "XBB.1.9*"),
  Pho_Week = c(37, 29, 23, 41,
    49, 52 + 10, 52 + 5) + 52
)
date_seq <- seq(ymd("2022-01-01"), ymd("2024-12-31"), 1)
epiweeks <- data.frame(</pre>
  EpiWeek = epiweek(date_seq) + 52 * (year(date_seq) - 2021),
```

date = date\_seq)
epiweeks <- epiweeks[epiweeks\$EpiWeek %in% pho\$Pho\_Week, ]</pre>

```
air2 <- left_join(air, gisaid, by = "Lineage") %>%
mutate(Lead_Sample = as.numeric(ymd(Sample_Date) - ymd(Gisaid_Sample)),
Lead_Report = as.numeric(ymd(Sample_Date) - ymd(Gisaid_Report)),
Epi_Week = 52 * (year(Sample_Date) - 2021) + epiweek(Sample_Date)) %>%
left_join(pho, by = "Lineage") %>%
mutate(Lead_Epi = Epi_Week - Pho_Week) %>%
mutate(Detection = Frequency >= 0.01) %>%
mutate(Group = ordered(Group, levels = levels(factor(Group))[c(1,3,4,5,2)]))
```

```
air5 <- air2 %>%
filter(Lead_Epi > -17) %>%
group_by(Lineage, Group) %>%
arrange(Sample_Date) %>%
mutate(`Freq >= 0.01` = cumsum(Frequency > 0.01),
`Freq >= 0.05` = cumsum(Frequency > 0.05)) %>%
pivot_longer(cols = c(`Freq >= 0.01`, `Freq >= 0.05`))
# https://stackoverflow.com/questions/54438495/shift-legend-into-empty-facets-of-a-faceted-plot-in-
ggplot2
shift_legend3 <- function(p) {
    pnls <- cowplot::plot_to_gtable(p) %>% gtable::gtable_filter("panel") %>%
    with(setNames(grobs, layout$name)) %>%
    purrr::keep(~identical(.x, zeroGrob()))
```

```
if (length(pnls) == 0) stop("No empty facets in the plot")
```

```
lemon::reposition_legend(p, "center",
    panel = names(pnls))
```

```
}
```

#### Frequency Plots (log scale)

```
g <- ggplot(filter(air2, Lead_Epi > -17, Lead_Epi < 10, Frequency > 0)) + theme_bw() +
```

```
aes(x = Lead_Epi, y = Frequency, colour = Group) +
geom_point(size = 1, mapping = aes(alpha = Detection)) +
scale_alpha_manual(values = c(0.4, 1)) +
facet_wrap(~ Lineage, ncol = 2) +
geom_vline(xintercept = 0) +
geom_hline(yintercept = c(0.01), col = "grey", linetype = 2) +
geom_hline(yintercept = c(0.05), col = "grey", linetype = 2) +
labs(x = "Epiweeks to/from First Clinical Case",
    y = "Relative Demixing Frequency of Lineage from Alcov (log scale)",
    title = "", colour = NULL) +
guides(alpha = "none") +
scale_y_log10() +
scale_colour_brewer(palette = "Dark2")
shift_legend3(g)
```

#### **Cumulative Plots**

```
g <- filter(air4, Lead_Epi > -20, Lead_Epi < 10) %>%
  mutate(name = ordered(name, levels = rev(levels(factor(name))))) %>%
  ggplot() +
  theme_bw() +
  geom_vline(xintercept = 0, colour = "darkgreen", linewidth = 1.25) +
  aes(x = Lead_Epi, y = value, colour = Group, linetype = name) +
  geom_step() +
  facet_wrap(~Lineage, ncol = 2) +
  coord_cartesian(ylim = c(0, 10)) +
  labs(x = "Epiweeks Since First Clinical Case",
    y = "Cumulative Detections",
    linetype = "Detection Threshold",
    colour = "Location") +
  scale_y_continuous(minor_breaks = 0:10, breaks = seq(0, 10, 2)) +
  scale_x_continuous(minor_breaks = -20:20, breaks = seq(-20, 20, 2)) +
  scale colour brewer(palette = "Dark2") +
  theme(legend.box = "horizontal")
```

shift\_legend3(g)

#### **Summary Statistics**

```
air4 %>%
filter(Lead_Epi > -20, value == 1) %>%
mutate(Detection_Threshold = case_when(
    Detection ~ 0.01,
    TRUE ~ 0.05
)) %>%
group_by(Group, Detection_Threshold) %>%
summarise(
    earliest_first_date = min(Lead_Epi),
```

```
mean_first_date = mean(Lead_Epi),
    sd_first_date = sd(Lead_Epi)
) %>%
arrange(Group, Detection_Threshold) %>%
knitr::kable()
```

#### Summary Statistics GISAID

```
air5 %>%
filter(Lead_Epi > -20, value == 1) %>%
mutate(Detection_Threshold = case_when(
    Detection ~ 0.01,
    TRUE ~ 0.05
)) %>%
group_by(Group, Detection_Threshold) %>%
summarise(
    earliest_first_date = min(Lead_Sample, na.rm = TRUE),
    mean_first_date = mean(Lead_Sample, na.rm = TRUE),
    sd_first_date = sd(Lead_Sample, na.rm = TRUE)
) %>%
arrange(Group, Detection_Threshold) %>%
knitr::kable()
```

#### Supplementary Scatter Plots - R

```
suppressPackageStartupMessages({
    library(readxl)
    library(dplyr)
    library(magrittr)
    library(tidyr)
    library(ggplot2)
    theme_set(theme_bw())
    library(lubridate)
    library(epitools)
})
```

```
air <- read.csv("summary_all_samples_clean.csv") %>%
filter(Breadth_of_Coverage >= 20)
```

```
)
```

```
pho <- data.frame(
 Lineage = c("BQ*", "BA.2.75*", "BF*", "XBB*",
       "XBB.1.5*", "XBB.1.16*", "XBB.1.9*"),
 Pho Week = c(37, 29, 23, 41,
        49, 10, 5),
Year = c(2022, 2022, 2022, 2022, 2022, 2023, 2023)
)
# Convert Pho Week and Year to dates
pho <- pho %>%
mutate(Pho_Date = as.Date(paste(Year, Pho_Week, 1, sep = "-"), "%Y-%U-%u"))
# Ensure Sample Date is a Date object and add Epiweek and Year columns
air <- air %>%
mutate(Sample_Date = as.Date(Sample_Date),
    Epiweek = as.numeric(as.week(Sample_Date)$week),
    Year = year(Sample Date),
    Epiweek Year = paste(Year, sprintf("%02d", as.numeric(Epiweek)), sep = "-"))
# Function to calculate start date 16 weeks before a given date
start_date <- function(date) {</pre>
date - weeks(16)
}
# Define the end date
end_date <- ymd("2023-03-31")
# Unique lineages
lineages <- unique(air$Lineage)
# Loop through each lineage and create a scatter plot
for (lineage in lineages) {
print(paste("Processing lineage:", lineage)) # Print the lineage being processed
air_lineage <- air %>%
 filter(Lineage == lineage)
 # Get the first PHO clinical detection date for the current lineage
 first_pho_detection_date <- pho %>%
  filter(Lineage == lineage) %>%
  pull(Pho Date)
 # Get the first GISAID clinical sample date for the current lineage
 first_gisaid_sample_date <- gisaid %>%
 filter(Lineage == lineage) %>%
  pull(Gisaid_Sample)
```

```
# Calculate the start date 16 weeks before the first clinical detection date
 start date lineage <- start date(first pho detection date)
 # Filter air lineage based on the calculated start date and end date
 air lineage <- air lineage %>%
  filter(Sample Date >= start date lineage & Sample Date <= end date)
 # Create breaks at weekly intervals
 breaks <- seq(min(air_lineage$Sample_Date), max(air_lineage$Sample_Date), by = "week")
 # Create labels for the breaks
 labels <- paste(year(breaks), sprintf("%02d", as.numeric(as.week(breaks)$week)), sep = "-")
 # Adjust breaks and labels for specific lineages
 if (lineage %in% c("BQ*", "BA.2.75*", "BF*")) {
  breaks <- breaks[seq(1, length(breaks), by = 2)]</pre>
  labels <- labels[seq(1, length(labels), by = 2)]</pre>
 }
 # Define the order of legend labels
 legend order <- c("Airport Terminal 1 and 3", "Pooled Aircraft Sewage", "Peel", "Toronto", "York")
 p <- ggplot(air lineage,
       aes(x = Sample_Date, y = Frequency, color = Group)) +
  geom point() +
  labs(title = paste("Emergence of ", lineage, "Lineages in Airport and surrounding municipal sites"),
     x = "Sample Date / Epi Week",
     y = paste("Frequency of ", lineage, "in Sample"),
     color = "Location") +
  scale color manual(values = c("Toronto" = "blue", "Peel" = "purple", "York" = "green",
                   "Pooled Aircraft Sewage" = "orange", "Airport Terminal 1 and 3" = "yellow2"),
             limits = legend_order) +
  scale x date(breaks = breaks, labels = labels) +
  geom_vline(xintercept = first_pho_detection_date, linetype = "dashed", color = "red") +
  annotate("text", x = first pho detection date, y = Inf, label = paste("First PHO clinical detection of",
lineage),
       angle = 90, viust = -0.5, hiust = 1, color = "red", size = 3) +
  geom vline(xintercept = first gisaid sample date, linetype = "dashed", color = "pink") +
  annotate("text", x = first gisaid sample date, y = Inf, label = paste("First GISAID sample date of",
lineage),
       angle = 90, vjust = -0.5, hjust = 1, color = "pink", size = 3) +
  theme(axis.text.x = element text(angle = 45, hjust = 1)) +
  scale y continuous(breaks = seq(0, max(air$Frequency, na.rm = TRUE), by = 0.1)) # Adjust y-axis
breaks
 # Display plot
 print(p)
```

```
# Save plot
ggsave(filename = paste0("scatter_plot_", lineage, ".png"), plot = p, width = 10, height = 6)
}
```

#### S:R346T Plot - R

```
suppressPackageStartupMessages({
library(readxl)
library(dplyr)
library(magrittr)
library(tidyr)
library(ggplot2)
theme_set(theme_bw())
library(lubridate)
library(epitools)
})
airS1 <- read.csv("all_samples_SR346T_mutations.csv")
airS <- airS1[, 1:4]
names(airS) <- c("Sample_Name", "Sample_Date", "Breadth_of_Coverage",
       "S:R346T")
airS <- airS %>%
mutate(Location = gsub("(20\\d+)|(-\\d+)", "", Sample_Name)) %>%
mutate(Location = ifelse(
 Location %in% c("Ashbridges", "AshbridgesBay", "AshbridgeBay"),
 yes = "Ashbridges",
 no = ifelse(
  Location %in% c("HighlandCreek", "Highland", "HighlandCR"),
  yes = "HighlandCreek",
  no = Location)
)) %>%
mutate(Group = case_when(
 Location %in% c("Humber", "Ashbridges", "HighlandCreek", "NorthToronto") ~ "Toronto",
 Location %in% c("As", "At") ~ "Pooled Aircraft Sewage",
 Location %in% c("A1", "A3") ~ "Airport Terminal 1 and 3",
 Location %in% c("P1", "P2") ~ "Peel",
 Location %in% c("Y1", "Y5", "Y6") ~ "York",
 TRUE ~ "Devan Missed One"
))
airS <- select(airS, -Sample_Name) %>%
pivot_longer(cols = !c(Group, Location, Sample_Date, Breadth_of_Coverage),
      values_to = "Frequency", names_to = "Mutation") %>%
mutate(Sample_Date = ymd(Sample_Date),
    Frequency = ifelse(Frequency > 1, 1, Frequency))
write.csv(airS, file = "summary_all_samples_SR346T_clean.csv")
```

rm(airS1)

```
airS <- read.csv("summary_all_samples_clean.csv") %>%
# Remove samples with less than 20% BOC and no coverage of S:R346T (Frequency = -1)
filter(Breadth_of_Coverage >= 20 & Frequency >= 0)
# Ensure Sample_Date is a Date object and add Epiweek and Year columns
airS <- airS %>%
mutate(Sample_Date = as.Date(Sample_Date),
    Epiweek = as.numeric(as.week(Sample_Date)$week),
    Year = year(Sample_Date),
    Epiweek_Year = paste(Year, sprintf("%02d", as.numeric(Epiweek)), sep = "-"))
# Define the start and end dates
start_date <- ymd("2022-04-01")
end_date <- ymd("2023-03-31")
# Filter airS based on the calculated start date and end date
airS <- airS %>%
filter(Sample_Date >= start_date & Sample_Date <= end_date)
# Create breaks at weekly intervals
breaks <- seq(min(airS$Sample_Date), max(airS$Sample_Date), by = "week")
# Create labels for the breaks
labels <- paste(year(breaks), sprintf("%02d", as.numeric(as.week(breaks)$week)), sep = "-")
# Adjust breaks and labels for readability
breaks <- breaks[seq(1, length(breaks), by = 2)]</pre>
labels <- labels[seq(1, length(labels), by = 2)]
# Define the order of legend labels
legend_order <- c("Airport Terminal 1 and 3", "Pooled Aircraft Sewage", "Peel", "Toronto", "York")
p <- ggplot(airS,
     aes(x = Sample_Date, y = Frequency, color = Group)) +
geom_point() +
labs(title = paste("Emergence of S:R346T mutation in Airport and surrounding municipal sites"),
   x = "Sample Date / Epi Week",
   y = paste("Frequency of S:R346T in Sample"),
   color = "Location") +
scale_color_manual(values = c("Toronto" = "blue", "Peel" = "purple", "York" = "green",
               "Pooled Aircraft Sewage" = "orange", "Airport Terminal 1 and 3" = "yellow2"),
         limits = legend_order) +
scale_x_date(breaks = breaks, labels = labels) +
theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
```

scale\_y\_continuous(breaks = seq(0, max(airS\$Frequency, na.rm = TRUE), by = 0.1)) # Adjust y-axis breaks

# Display plot print(p)

# Save plot

ggsave(filename = paste0("scatter\_plot\_SR346T.png"), plot = p, width = 10, height = 6)