

Class 17 Mini-Project: COVID-19 Vaccination Rates

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Date: 2023/06/12

The goal of this hands-on mini-project is to examine and compare the Covid-19 vaccination rates around San Diego.

Getting Started

First, we need to read the .csv that includes data about statewide COVID-19 vaccines administered by ZIP code:

```
vax <- read.csv("covid19vaccinesbyzipcode_test.csv")
head(vax)
```

```
  as_of_date zip_code_tabulation_area local_health_jurisdiction    county
1 2021-01-05           94579                Alameda    Alameda
2 2021-01-05           93726                Fresno    Fresno
3 2021-01-05           94305            Santa Clara Santa Clara
4 2021-01-05           93704                Fresno    Fresno
5 2021-01-05           94403            San Mateo    San Mateo
6 2021-01-05           93668                Fresno    Fresno
 vaccine_equity_metric_quartile      vem_source
1                3 Healthy Places Index Score
2                1 Healthy Places Index Score
3                4 Healthy Places Index Score
4                1 Healthy Places Index Score
5                4 Healthy Places Index Score
6                1      CDPH-Derived ZCTA Score
age12_plus_population age5_plus_population tot_population
```

1	19192.7	20872	21883
2	33707.7	39067	42824
3	15716.9	16015	16397
4	24803.5	27701	29740
5	37967.5	41530	44408
6	1013.4	1199	1219

persons_fully_vaccinated		persons_partially_vaccinated
1	NA	NA
2	NA	NA
3	NA	NA
4	NA	NA
5	NA	NA
6	NA	NA

percent_of_population_fully_vaccinated	
1	NA
2	NA
3	NA
4	NA
5	NA
6	NA

percent_of_population_partially_vaccinated	
1	NA
2	NA
3	NA
4	NA
5	NA
6	NA

percent_of_population_with_1_plus_dose		booster_recip_count
1	NA	NA
2	NA	NA
3	NA	NA
4	NA	NA
5	NA	NA
6	NA	NA

bivalent_dose_recip_count	eligible_recipient_count
1	4
2	2
3	8
4	5
5	7
6	0

eligible_bivalent_recipient_count	
1	4

2	2
3	8
4	5
5	7
6	0

redacted

1 Information redacted in accordance with CA state privacy requirements
 2 Information redacted in accordance with CA state privacy requirements
 3 Information redacted in accordance with CA state privacy requirements
 4 Information redacted in accordance with CA state privacy requirements
 5 Information redacted in accordance with CA state privacy requirements
 6 Information redacted in accordance with CA state privacy requirements

Q1: What column details the total number of people fully vaccinated? *Answer:* persons_fully_vaccinated

Q2: What column details the Zip code tabulation area? *Answer:* zip_code_tabulation_area

Q3: What is the earliest date in this dataset? *Answer:* 2021-01-05

Q4: What is the latest date in this dataset? *Answer:* 2023-05-23

Call the `skim()` function from the `skimr` package to get a quick overview of this dataset.

```
skimr::skim_without_charts(vax)
```

Table 1: Data summary

Name	vax
Number of rows	220500
Number of columns	19
Column type frequency:	
character	5
numeric	14
Group variables	None

Variable type: character

skim_variable	n_missing	complete_rate	min	max	empty	n_unique	whitespace
as_of_date	0	1	10	10	0	125	0

skim_variable	n_missing	complete_rate	min	max	empty	n_unique	whitespace
local_health_jurisdiction	0	1	0	15	625	62	0
county	0	1	0	15	625	59	0
vem_source	0	1	15	26	0	3	0
redacted	0	1	2	69	0	2	0

Variable type: numeric

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100
zip_code_tabulation_area	0	1.00	93665.11817	389000	192257.79	3658.50	5380.50	7635.0	
vaccine_equity_metric_q1	10871	0.95	2.44	1.11	1	1.00	2.00	3.00	4.0
age12_plus_population	0	1.00	18895.04	8993.87	0	1346.95	13685.10	1756.18	8556.7
age5_plus_population	0	1.00	20875.22	1105.97	0	1460.50	15364.00	4877.00	1902.0
tot_population	10750	0.95	23372.72	2628.50	12	2126.00	18714.00	8168.00	11165.0
persons_fully_vaccinated	17711	0.92	14272.72	25264.17	11	954.00	8990.00	23782.00	7724.0
persons_partially_vaccinated	17711	0.92	1711.05	2071.56	11	164.00	1203.00	2550.00	42259.0
percent_of_population_fully_vaccinated	17711	0.90	0.58	0.25	0	0.44	0.62	0.75	1.0
percent_of_population_partially_vaccinated	17711	0.90	0.08	0.09	0	0.05	0.06	0.08	1.0
percent_of_population_w23732_plus0.80se	17711	0.80	0.64	0.24	0	0.50	0.68	0.82	1.0
booster_recip_count	74388	0.66	6373.43	7751.70	11	328.00	3097.00	10274.00	60022.0
bivalent_dose_recip_count	159956	0.27	3407.91	4010.38	11	222.00	1832.00	5482.00	29484.0
eligible_recipient_count	0	1.00	13120.40	5126.17	0	534.00	6663.00	22517.25	7437.0
eligible_bivalent_recipient_count	0	1.00	13016.51	5199.08	0	266.00	6562.00	22513.00	7437.0

Q5: How many numeric columns are in this dataset? *Answer:* There are 14 numeric columns.

Q6: Note that there are “missing values” in the dataset. How many NA values there in the persons_fully_vaccinated column?

```
sum(is.na(vax$persons_fully_vaccinated))
```

```
[1] 17711
```

Answer: There are 17711 missing values in the persons_fully_vaccinated column.

Q7: What percent of persons_fully_vaccinated values are missing (to 2 significant figures)?

```
(1-0.9196780)*100
```

```
[1] 8.0322
```

Answer: 8.03% of persons_fully_vaccinated values are missing (used the value found in column complete_rate).

Q8: [Optional]: Why might this data be missing? *Answer:* The data could be missing because whatever method they are using to track every individual's data may not be efficient in collecting every peice of data in the table.

Working with dates

One of the “character” columns of the data is `as_of_date`, which contains dates in the Year-Month-Day format.

Using the **lubridate** package:

```
library(lubridate)
```

Attaching package: 'lubridate'

The following objects are masked from 'package:base':

```
date, intersect, setdiff, union
```

Today's date is:

```
today()
```

```
[1] "2023-06-12"
```

To make our `as_of_date` column usable for mathematical operations, we will convert our date data into a lubridate format.

```
vax$as_of_date <- ymd(vax$as_of_date)
```

How many days have passed since the first vaccination reported in this dataset?

```
today() - vax$as_of_date[1]
```

Time difference of 888 days

Using the last and the first date value we can now determine how many days the dataset span.

```
vax$as_of_date[nrow(vax)] - vax$as_of_date[1]
```

Time difference of 868 days

Q9: How many days have passed since the last update of the dataset?

```
today() - vax$as_of_date[220500]
```

Time difference of 20 days

Answer: 20 days have passed since the last update of the dataset (different from the answer ‘7’ in the lab because our today dates are different).

Q10: How many unique dates are in the dataset (i.e. how many different dates are detailed)?

```
length(unique(vax$as_of_date))
```

```
[1] 125
```

Answer: There are 125 unique dates in the dataset.

Working with ZIP codes

One of the numeric columns in the dataset (namely `vax$zip_code_tabulation_area`) are actually ZIP codes - a postal code used by the United States Postal Service (USPS).

Using the **zipcodeR** package to make working with the zip codes easier.

```
library(zipcodeR)
```

The legacy packages `mapproj`, `rgdal`, and `rgeos`, underpinning this package will retire shortly. Please refer to R-spatial evolution reports on <https://r-spatial.org/r/2023/05/15/evolution4.html> for details. This package is now running under evolution status 0

Finding the centroid of the La Jolla 92037 ZIP code area:

```
geocode_zip('92037')
```

```
# A tibble: 1 x 3
  zipcode lat lng
  <chr>   <dbl> <dbl>
1 92037   32.8 -117.
```

Calculating the distance between the centroids of any two ZIP codes in miles:

```
zip_distance('92037','92109')
```

```
  zipcode_a zipcode_b distance
1     92037     92109      2.33
```

Pulling census data about ZIP code areas:

```
reverse_zipcode(c('92037', "92109"))
```

```
# A tibble: 2 x 24
  zipcode zipcode_type major_city post_office_city common_city_list county state
  <chr>   <chr>         <chr>      <chr>                <blob> <chr>  <chr>
1 92037   Standard      La Jolla   La Jolla, CA          <raw 20 B> San D~ CA
2 92109   Standard      San Diego San Diego, CA          <raw 21 B> San D~ CA
# i 17 more variables: lat <dbl>, lng <dbl>, timezone <chr>,
#   radius_in_miles <dbl>, area_code_list <blob>, population <int>,
#   population_density <dbl>, land_area_in_sqmi <dbl>,
#   water_area_in_sqmi <dbl>, housing_units <int>,
#   occupied_housing_units <int>, median_home_value <int>,
#   median_household_income <int>, bounds_west <dbl>, bounds_east <dbl>,
#   bounds_north <dbl>, bounds_south <dbl>
```

We can also use `reverse_zipcode()` to pull census data later on for any or all ZIP code areas we might be interested in:

```
zipdata <- reverse_zipcode(vax$zip_code_tabulation_area)
```

Focus on the San Diego area

Focus in on the San Diego County area by restricting ourselves first to `vax$county == "San Diego"` entries.

We can do this by using the **dplyr** package.

```
library(dplyr)
```

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':

```
filter, lag
```

The following objects are masked from 'package:base':

```
intersect, setdiff, setequal, union
```

```
sd <- filter(vax, county == "San Diego")  
nrow(sd)
```

```
[1] 13375
```

Using `**dplyr*` is more convenient when subsetting across multiple criteria.

```
sd.10 <- filter(vax, county == "San Diego" & age5_plus_population > 10000)
```

Q11: How many distinct zip codes are listed for San Diego County?

```
sd %>% group_by(zip_code_tabulation_area) %>% summarise()
```

```
# A tibble: 107 x 1  
  zip_code_tabulation_area  
  <int>  
1             91901  
2             91902
```



```

3          91905
4          91906
5          91910
6          91911
7          91913
8          91914
9          91915
10         91916
# i 97 more rows

```

Answer: There are 107 distinct zip codes listed for San Diego County.

Q12: What San Diego County Zip code area has the largest population in this dataset?

```
which.max(sd$age12_plus_population)
```

```
[1] 87
```

```
sd[87, ]
```

```

      as_of_date zip_code_tabulation_area local_health_jurisdiction    county
87 2021-01-05          92154                San Diego San Diego
      vaccine_equity_metric_quartile          vem_source
87                2 Healthy Places Index Score
      age12_plus_population age5_plus_population tot_population
87          76365.2          82971          88979
      persons_fully_vaccinated persons_partially_vaccinated
87                18                1403
      percent_of_population_fully_vaccinated
87                0.000202
      percent_of_population_partially_vaccinated
87                0.015768
      percent_of_population_with_1_plus_dose booster_recip_count
87                0.01597                NA
      bivalent_dose_recip_count eligible_recipient_count
87                NA                18
      eligible_bivalent_recipient_count
87                18
                                                    redacted
87 Information redacted in accordance with CA state privacy requirements

```

Answer: 92154

Q13: What is the overall average (with 2 decimal numbers) “Percent of Population Fully Vaccinated” value for all San Diego “County” as of “2023-05-23”?

```
df_23.05.23 <- sd %>% filter(as_of_date == "2022-11-15")

df_23.05.23_clean <- df_23.05.23 %>% filter(!is.na(percent_of_population_fully_vaccinated))

mean(df_23.05.23_clean$percent_of_population_fully_vaccinated)
```

```
[1] 0.7392451
```

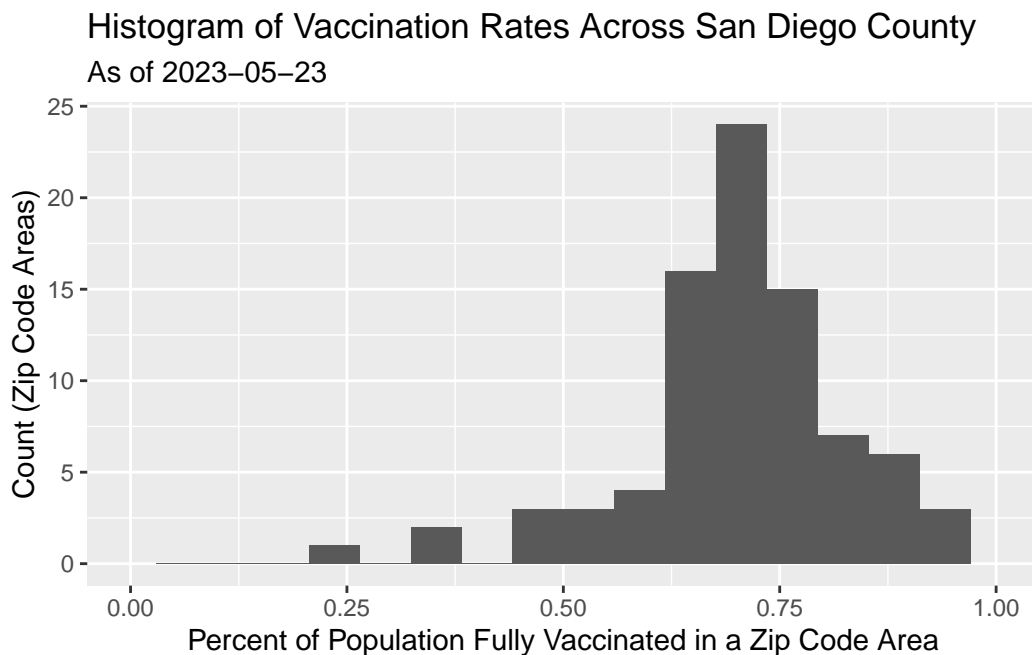
Answer: 0.74 is the overall average.

Q14: Using either ggplot or base R graphics make a summary figure that shows the distribution of Percent of Population Fully Vaccinated values as of “2023-05-23”?

```
library(ggplot2)

ggplot(df_23.05.23_clean) + aes(x = percent_of_population_fully_vaccinated) + geom_histogram
```

Warning: Removed 2 rows containing missing values (`geom_bar()`).



Focus on UCSD/La Jolla

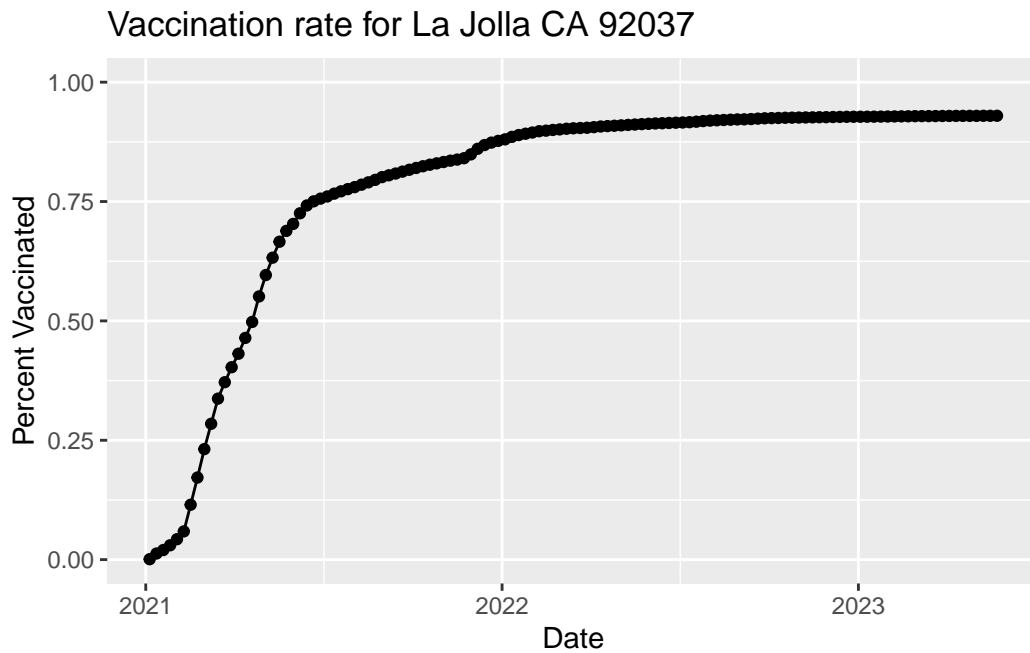
```
ucsd <- filter(sd, zip_code_tabulation_area=="92037")
ucsd[1,]$age5_plus_population
```

```
[1] 36144
```

Q15: Using ggplot make a graph of the vaccination rate time course for the 92037 ZIP code area:

```
ggplot(ucsd) + aes(x = as_of_date, y = ucsd$percent_of_population_fully_vaccinated) + geom
```

Warning: Use of `ucsd\$percent_of_population_fully_vaccinated` is discouraged.
i Use `percent_of_population_fully_vaccinated` instead.
Use of `ucsd\$percent_of_population_fully_vaccinated` is discouraged.
i Use `percent_of_population_fully_vaccinated` instead.



Comparing to similar sized areas

Return to the full dataset and look across every zip code area with a population at least as large as that of 92037 on as_of_date “2023-05-23”.

```
vax.36 <- filter(vax, age5_plus_population > 36144 & as_of_date == "2023-05-23")
```

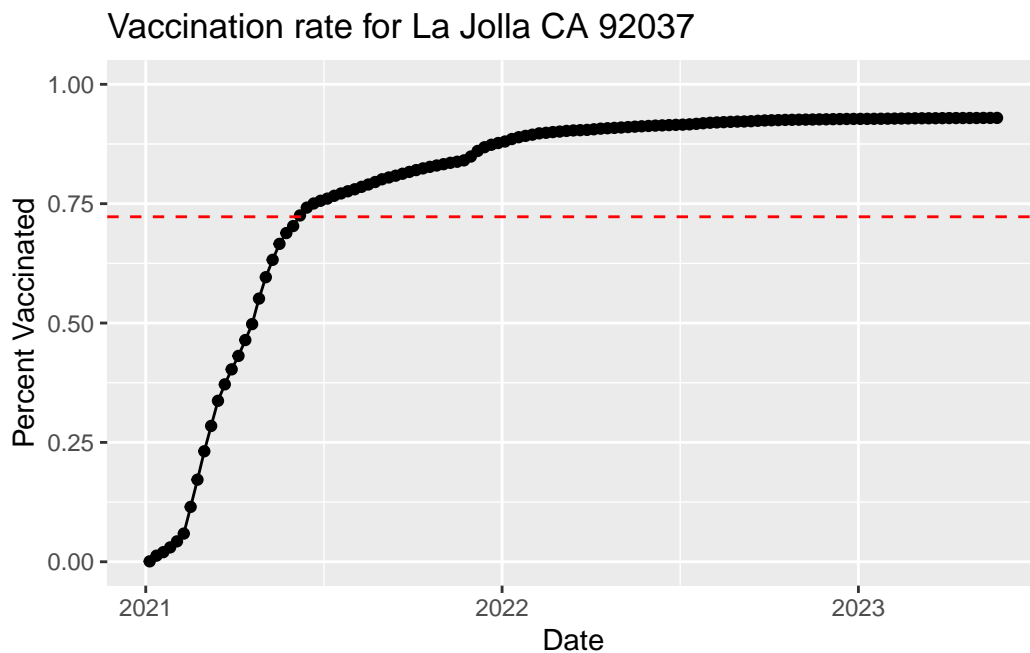
Q16: Calculate the mean “Percent of Population Fully Vaccinated” for ZIP code areas with a population as large as 92037 (La Jolla) as_of_date “2023-05-23”. Add this as a straight horizontal line to your plot from above with the `geom_hline()` function:

```
mean(vax.36$percent_of_population_fully_vaccinated)
```

```
[1] 0.7225892
```

```
ggplot(ucsd) + aes(x = as_of_date, y = ucsd$percent_of_population_fully_vaccinated) + geom
```

Warning: Use of `ucsd\$percent_of_population_fully_vaccinated` is discouraged.
i Use `percent_of_population_fully_vaccinated` instead.
Use of `ucsd\$percent_of_population_fully_vaccinated` is discouraged.
i Use `percent_of_population_fully_vaccinated` instead.



Q17: What is the 6 number summary (Min, 1st Qu., Median, Mean, 3rd Qu., and Max) of the “Percent of Population Fully Vaccinated” values for ZIP code areas with a population as large as 92037 (La Jolla) as_of_date “2023-05-23”?

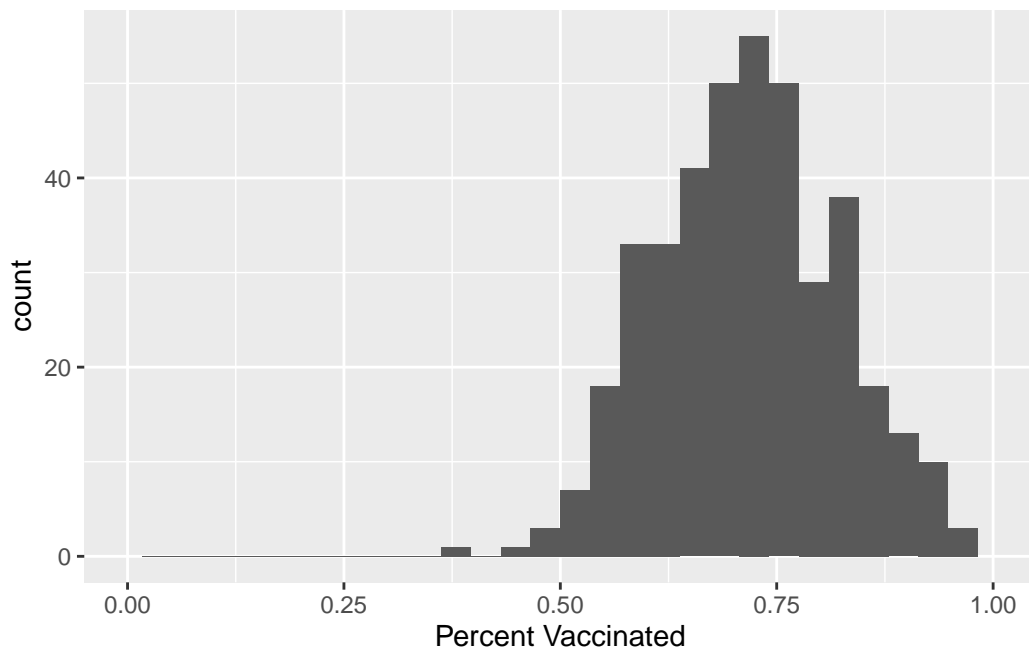
```
summary(vax.36$percent_of_population_fully_vaccinated)
```

```
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.3816  0.6469  0.7207  0.7226  0.7924  1.0000
```

Q18: Using ggplot generate a histogram of this data.

```
ggplot(vax.36) + aes(x = percent_of_population_fully_vaccinated) + geom_histogram(bins = 30)
```

Warning: Removed 2 rows containing missing values (`geom_bar()`).



Q19: Is the 92109 and 92040 ZIP code areas above or below the average value you calculated for all these above?

```
vax %>% filter(as_of_date == "2023-05-23") %>% filter(zip_code_tabulation_area=="92040") %>% summarise(mean_vax = mean(percent_of_population_fully_vaccinated))
```

```
percent_of_population_fully_vaccinated
1 0.552434
```

```
vax %>% filter(as_of_date == "2023-05-23") %>% filter(zip_code_tabulation_area=="92109") %>%
```

```
percent_of_population_fully_vaccinated
1 0.69487
```

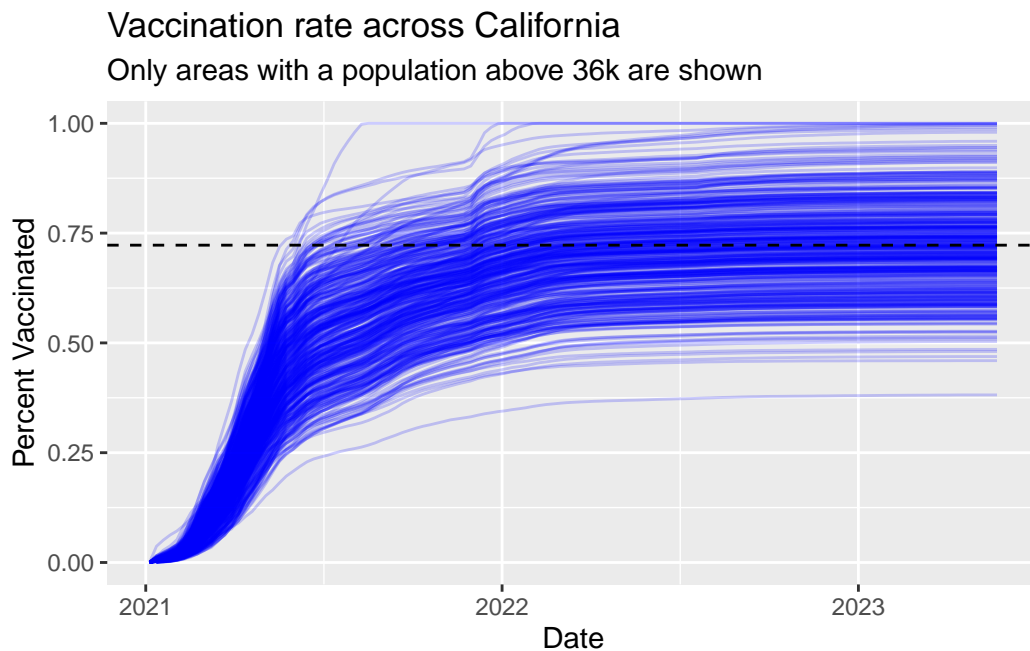
Answer: The percent of population fully vaccinated for 92040 is 0.55 which is below the average 0.7226. For 92109, the percent of population fully vaccinated in 0.69 which is also below the average 0.7226.

Q20: Finally make a time course plot of vaccination progress for all areas in the full dataset with a `age5_plus_population > 36144`.

```
vax.36.all <- filter(vax, age5_plus_population > 36144)

ggplot(vax.36.all) + aes(as_of_date, percent_of_population_fully_vaccinated, group = zip_code_tabulation_area)
```

Warning: Removed 185 rows containing missing values (``geom_line()``).



About this document

```
sessionInfo()
```

```
R version 4.2.3 (2023-03-15 ucrt)
Platform: x86_64-w64-mingw32/x64 (64-bit)
Running under: Windows 10 x64 (build 22621)
```

```
Matrix products: default
```

```
locale:
```

```
[1] LC_COLLATE=English_United States.utf8
[2] LC_CTYPE=English_United States.utf8
[3] LC_MONETARY=English_United States.utf8
[4] LC_NUMERIC=C
[5] LC_TIME=English_United States.utf8
```

```
attached base packages:
```

```
[1] stats      graphics  grDevices  utils      datasets  methods    base
```

```
other attached packages:
```

```
[1] ggplot2_3.4.2  dplyr_1.1.2    zipcodeR_0.3.5  lubridate_1.9.2
```

```
loaded via a namespace (and not attached):
```

```
[1] Rcpp_1.0.10      lattice_0.21-8   tidyr_1.3.0      class_7.3-21
[5] digest_0.6.31    utf8_1.2.3       R6_2.5.1         repr_1.1.6
[9] RSQLite_2.3.1    evaluate_0.21    e1071_1.7-13     httr_1.4.6
[13] pillar_1.9.0     rlang_1.1.0      curl_5.0.1       uuid_1.1-0
[17] rstudioapi_0.14  raster_3.6-20    blob_1.2.4       rmarkdown_2.22
[21] labeling_0.4.2   readr_2.1.4      stringr_1.5.0    munsell_0.5.0
[25] bit_4.0.5        proxy_0.4-27     compiler_4.2.3   xfun_0.39
[29] pkgconfig_2.0.3  tigris_2.0.3     base64enc_0.1-3  htmltools_0.5.5
[33] tidyselect_1.2.0 tibble_3.2.1     codetools_0.2-19 fansi_1.0.4
[37] crayon_1.5.2     tzdb_0.4.0       withr_2.5.0      sf_1.0-13
[41] tidycensus_1.4.1 rappdirs_0.3.3   grid_4.2.3       gtable_0.3.3
[45] jsonlite_1.8.5   lifecycle_1.0.3  DBI_1.1.3        magrittr_2.0.3
[49] scales_1.2.1     units_0.8-2      KernSmooth_2.23-20 cli_3.6.1
[53] stringi_1.7.12   cachem_1.0.8     farver_2.1.1     sp_1.6-1
[57] skimr_2.1.5      xml2_1.3.4       generics_0.1.3   vctrs_0.6.2
[61] tools_4.2.3      bit64_4.0.5      glue_1.6.2       purrr_1.0.1
[65] hms_1.1.3        fastmap_1.1.1    yaml_2.3.7       colorspace_2.1-0
```

[69]	timechange_0.2.0	terra_1.7-29	classInt_0.4-9	rvest_1.0.3
[73]	memoise_2.0.1	knitr_1.43		