

## Class 17: Vaccination rate mini project

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#Getting Started

```
# Import vaccination data
vax <- read.csv("covid19vaccinesbyzipcode_test.csv")
head(vax)
```

```
##   as_of_date zip_code_tabulation_area local_health_jurisdiction   county
## 1 2021-01-05                92804                Orange    Orange
## 2 2021-01-05                92626                Orange    Orange
## 3 2021-01-05                92250                Imperial  Imperial
## 4 2021-01-05                92637                Orange    Orange
## 5 2021-01-05                92155                San Diego  San Diego
## 6 2021-01-05                92259                Imperial  Imperial
##   vaccine_equity_metric_quartile          vem_source
## 1                               2 Healthy Places Index Score
## 2                               3 Healthy Places Index Score
## 3                               1 Healthy Places Index Score
## 4                               3 Healthy Places Index Score
## 5                               NA                No VEM Assigned
## 6                               1      CDPH-Derived ZCTA Score
##   age12_plus_population age5_plus_population persons_fully_vaccinated
## 1                76455.9                84200                19
## 2                44238.8                47883                NA
## 3                 7098.5                8026                NA
## 4                16027.4                16053                NA
## 5                 456.0                456                NA
## 6                 119.0                121                NA
##   persons_partially_vaccinated percent_of_population_fully_vaccinated
## 1                        1282                        0.000226
## 2                         NA                        NA
## 3                         NA                        NA
## 4                         NA                        NA
## 5                         NA                        NA
## 6                         NA                        NA
##   percent_of_population_partially_vaccinated
## 1                        0.015226
## 2                         NA
## 3                         NA
## 4                         NA
## 5                         NA
## 6                         NA
```

```
## percent_of_population_with_1_plus_dose
## 1 0.015452
## 2 NA
## 3 NA
## 4 NA
## 5 NA
## 6 NA
## redacted
## 1 No
## 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?

- The 9th column: “persons\_fully\_vaccinated”

Q2. What column details the Zip code tabulation area?

- The 2nd column

Q3. What is the earliest date in this dataset?

```
head(vax$as_of_date)
```

```
## [1] "2021-01-05" "2021-01-05" "2021-01-05" "2021-01-05" "2021-01-05"
## [6] "2021-01-05"
```

- 2021-01-05 (January 5th, 2021)

Q4. What is the latest date in this dataset?

```
tail(vax$as_of_date)
```

```
## [1] "2021-11-16" "2021-11-16" "2021-11-16" "2021-11-16" "2021-11-16"
## [6] "2021-11-16"
```

- 2021-11-16 (November 16th, 2021)

As we have done previously, let’s call the `skim()` function from the `skimr` package to get a quick overview of this dataset:

```
skimr::skim(vax)
```

Table 1: Data summary

Name	vax
Number of rows	81144
Number of columns	14
Column type frequency:	
character	5
numeric	9
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	46	0
local_health_jurisdiction	0	1	0	15	230	62	0
county	0	1	0	15	230	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	hist
zip_code_tabulation_area0	1.00	93665.11	17.39	0.0001	192257.75	3658.53	380.37	635.0		
vaccine_equity_metric_4002	0.95	2.44	1.11	1	1.00	2.00	3.00	4.0		
age12_plus_population	0	1.00	18895.04	993.94	0	1346.93	3685.11	756.82	556.7	
age5_plus_population	0	1.00	20875.21	106.05	0	1460.50	5364.00	1877.00	1902.0	
persons_fully_vaccinated	8256	0.90	9456.49	1498.25	1	506.00	4105.00	5859.00	1078.0	
persons_partially_vaccinated	8256	0.90	1900.62	113.07	1	200.00	1271.00	2893.00	20185.0	
percent_of_population_fully_vaccinated	8256	0.90	0.42	0.27	0	0.19	0.44	0.62	1.0	
percent_of_population_partially_vaccinated	8256	0.90	0.10	0.10	0	0.06	0.07	0.11	1.0	
percent_of_population_with_1_plus_doses	8256	0.90	0.50	0.26	0	0.30	0.53	0.70	1.0	

Q5. How many numeric columns are in this dataset?

- 9 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] 8256
```

- 8256 NA values

Q7. What percent of persons\_fully\_vaccinated values are missing (to 2 significant figures)?

```
per_vax <- sum(is.na(vax$persons_fully_vaccinated)) / sum(vax$persons_fully_vaccinated, na.rm=TRUE)
signif((per_vax *100), 2)
```

```
## [1] 0.0012
```

- 0.0012%

Q8. [Optional]: Why might this data be missing?

- The areas don't report their vaccination. For example, the military has their own health systems and does not participate in this record keeping.

#Working with dates

##Ensure that the date column is useful

We will use the **lubridate** package to make life a lot easier when dealing with date and times

```
#install lubridate
#install.packages("lubridate")

library(lubridate)
```

```
## Warning: package 'lubridate' was built under R version 4.1.2
```

```
##
```

```
## Attaching package: 'lubridate'
```

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

```
##
```

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

```
today()
```

```
## [1] "2021-11-23"
```

Here we make our ~as-of-date' column lubridate format

```
# Specify that we are using the Year-month-day format
vax$as_of_date <- ymd(vax$as_of_date)
```

Now we can do math with dates. For example: How many days have passed since the first vaccination reported in this dataset?

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

```
## Time difference of 322 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 315 days
```

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

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

```
## Time difference of 7 days
```

- 7 days

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

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

```
## [1] 46
```

-There are 46 unique dates in the dataset

#Working with ZIP codes

In R we can use the zipcodeR package to make working with these codes easier.

```
#install.packages("zipcodeR")  
library(zipcodeR)
```

```
## Warning: package 'zipcodeR' was built under R version 4.1.2
```

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

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

Calculate the distance between the centroids of any two ZIP codes in miles, e.g.

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

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

More usefully, we can pull census data about ZIP code areas (including median household income etc.). For example:

```
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
## # ... with 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 use this `reverse_zipcode()` to pull census data later on for any or all ZIP code areas we might be interested in.

```
# Pull data for all ZIP codes in the dataset
#zipdata <- reverse_zipcode( vax$zip_code_tabulation_area )
```

*#Focus on the San Diego area*

Let's now focus in on the San Diego County area by restricting ourselves first to `vax$county == "San Diego"` entries. We have two main choices on how to do this. The first using base R the second using the `dplyr` package:

Using base R, the code would look like this:

```
# Subset to San Diego county only areas
sd <- vax[ vax$county == "San Diego" , ]
```

Using `dplyr` the code would look like this:

```
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] 4922
```

Using dplyr is often more convenient when we are subsetting across multiple criteria (more advanced sub-setting) - for example all San Diego county areas with a population of over 10,000.

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

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

```
length(unique(sd$zip_code_tabulation_area))
```

```
## [1] 107
```

- There are 107 distinct zip codes

Q12. What San Diego County Zip code area has the largest 12 + Population in this dataset?

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

```
## [1] 23
```

```
sd[23,]
```

```
##   as_of_date zip_code_tabulation_area local_health_jurisdiction   county  
## 23 2021-01-05                92154                San Diego San Diego  
##   vaccine_equity_metric_quartile                vem_source  
## 23                        2 Healthy Places Index Score  
##   age12_plus_population age5_plus_population persons_fully_vaccinated  
## 23                76365.2                82971                32  
##   persons_partially_vaccinated percent_of_population_fully_vaccinated  
## 23                1336                0.000386  
##   percent_of_population_partially_vaccinated  
## 23                0.016102  
##   percent_of_population_with_1_plus_dose redacted  
## 23                0.016488                No
```

- zip code: 92154

Using dplyr select all San Diego “county” entries on “as\_of\_date” “2021-11-16” and use this for the following questions.

Q13. What is the overall average “Percent of Population Fully Vaccinated” value for all San Diego “County” as of “2021-11-09”?

```
library(dplyr)
```

```
sd.now <- filter(sd, as_of_date == "2021-11-09")  
(mean(sd.now$percent_of_population_fully_vaccinated, na.rm = TRUE))*100
```

```
## [1] 67.27567
```

- 67.28%

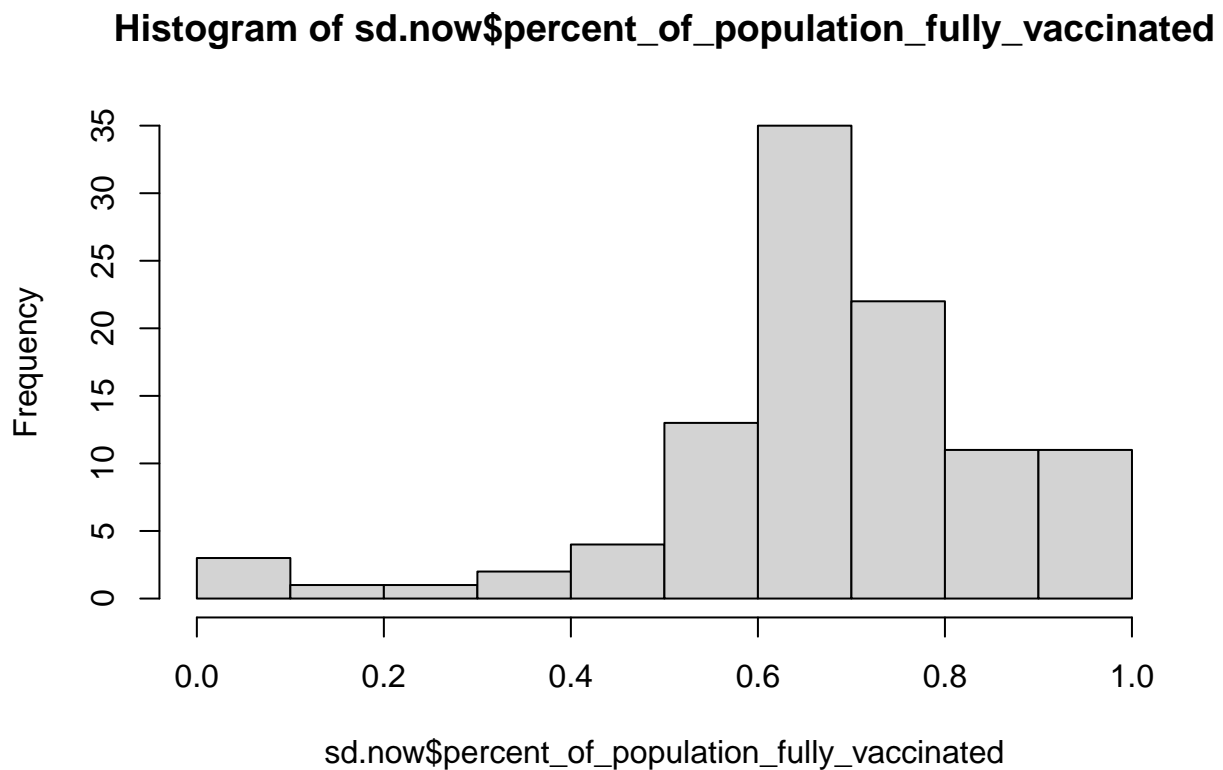
We can also look at the 6-number summary

```
summary(sd.now$percent_of_population_fully_vaccinated)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.     NA's  
## 0.01017 0.60776 0.67700 0.67276 0.76164 1.00000         4
```

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 “2021-11-09”?

```
hist(sd.now$percent_of_population_fully_vaccinated)
```

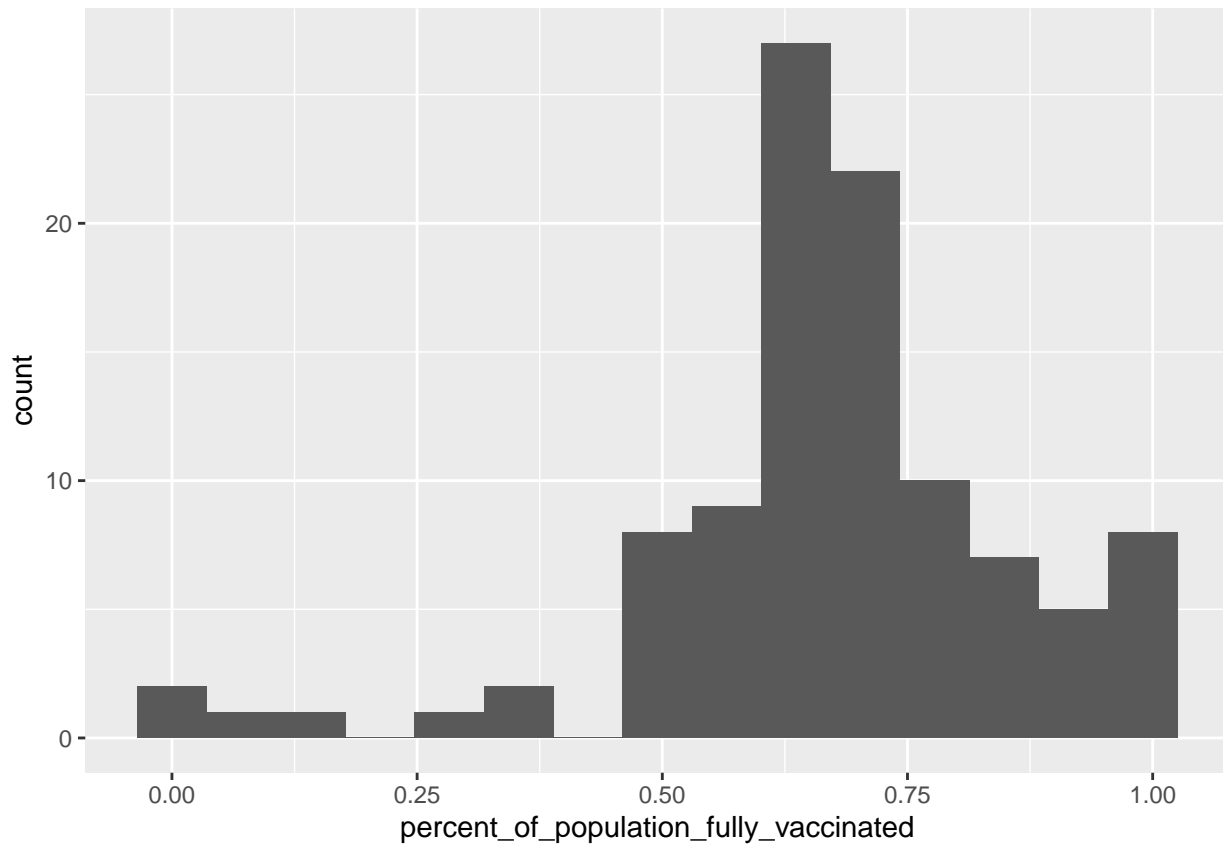


or

```
library(ggplot2)  
  
ggplot(sd.now) +  
  aes(percent_of_population_fully_vaccinated)+  
  geom_histogram(bins=15)
```



```
## Warning: Removed 4 rows containing non-finite values (stat_bin).
```



```
#Focus on UCSD/La Jolla
```

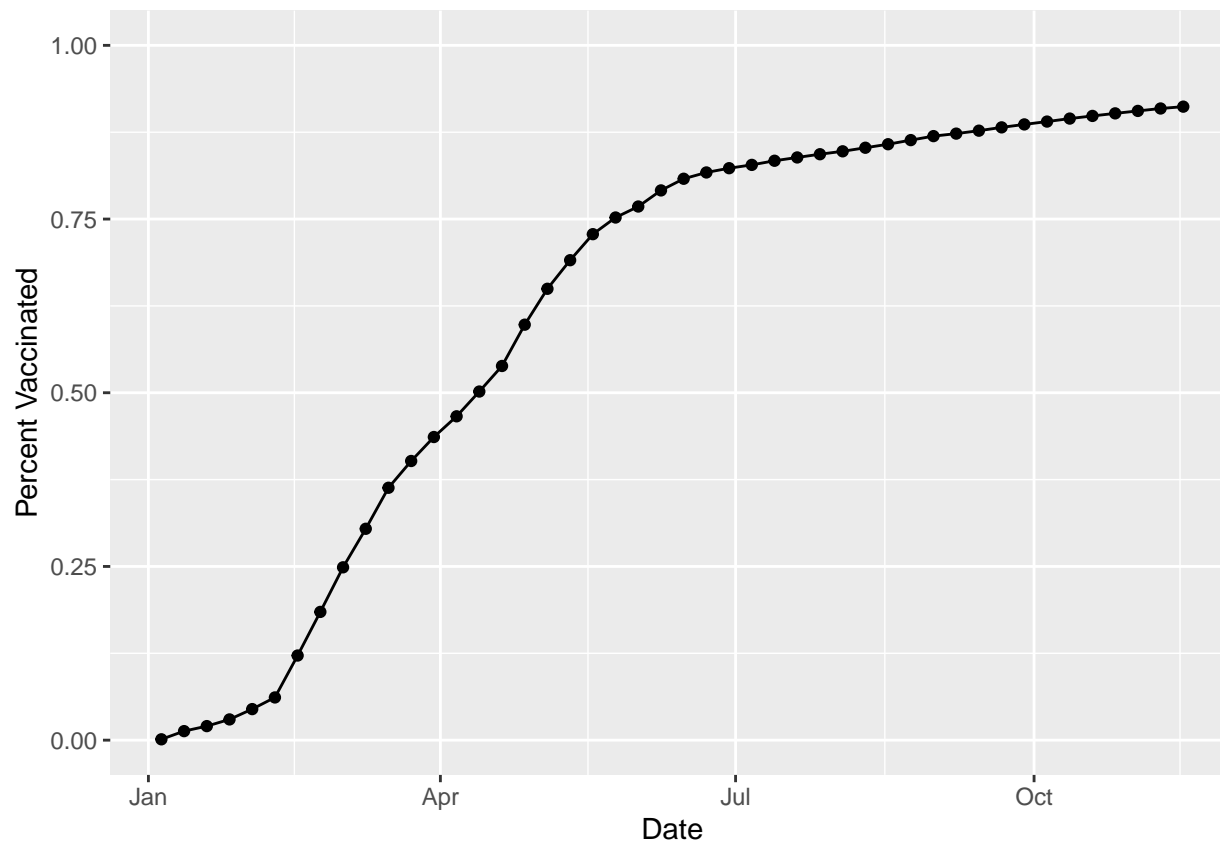
UC San Diego resides in the 92037 ZIP code area and is listed with an age 5+ population size of 36,144.

```
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:

```
library(ggplot2)
ggplot(ucsd) +
  aes(as_of_date, percent_of_population_fully_vaccinated) +
  geom_point() +
  geom_line(group=1) +
  ylim(c(0,1)) +
  labs(x="Date", y="Percent Vaccinated")
```



#Comparing 92037 to other similar sized areas?

*# Subset to all CA areas with a population as large as 92037*

```
vax.36 <- filter(vax, age5_plus_population > 36144 &
  as_of_date == "2021-11-16")
```

```
head(vax.36)
```

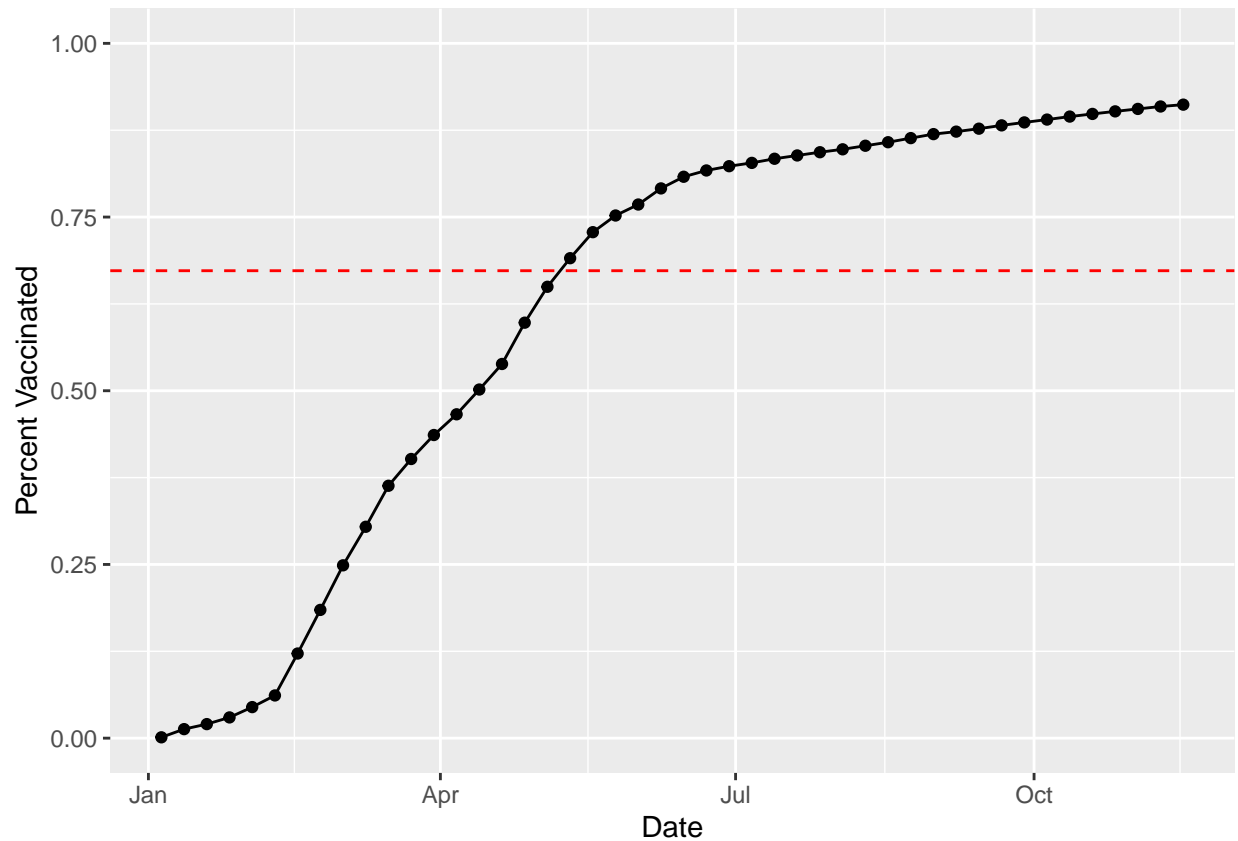
```
##   as_of_date zip_code_tabulation_area local_health_jurisdiction      county
## 1 2021-11-16          92833                Orange            Orange
## 2 2021-11-16          92234                Riverside          Riverside
## 3 2021-11-16          92507                Riverside          Riverside
## 4 2021-11-16          92555                Riverside          Riverside
## 5 2021-11-16          92345            San Bernardino San Bernardino
## 6 2021-11-16          91306                Los Angeles        Los Angeles
##   vaccine_equity_metric_quartile      vem_source
## 1                             3 Healthy Places Index Score
## 2                             1 Healthy Places Index Score
## 3                             1 Healthy Places Index Score
## 4                             2 Healthy Places Index Score
## 5                             1 Healthy Places Index Score
## 6                             2 Healthy Places Index Score
##   age12_plus_population age5_plus_population persons_fully_vaccinated
## 1             43985.4             48623             34668
## 2             46401.1             51202             34191
## 3             51432.5             55253             31704
```

```
## 4          36725.7          41446          23776
## 5          66047.5          75539          35332
## 6          42671.1          46573          31858
##  persons_partially_vaccinated percent_of_population_fully_vaccinated
## 1              3377              0.712996
## 2              3966              0.667767
## 3              3434              0.573797
## 4              2424              0.573662
## 5              4428              0.467732
## 6              3372              0.684044
##  percent_of_population_partially_vaccinated
## 1              0.069453
## 2              0.077458
## 3              0.062150
## 4              0.058486
## 5              0.058619
## 6              0.072402
##  percent_of_population_with_1_plus_dose redacted
## 1              0.782449          No
## 2              0.745225          No
## 3              0.635947          No
## 4              0.632148          No
## 5              0.526351          No
## 6              0.756446          No
```

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 “2021-11-16”. Add this as a straight horizontal line to your plot from above with the geom\_hline() function?

```
avg.full.vax <- mean(vax.36$percent_of_population_fully_vaccinated, na.rm=TRUE)

library(ggplot2)
ggplot(ucsd) +
  aes(as_of_date, percent_of_population_fully_vaccinated) +
  geom_point() +
  geom_line(group=1) +
  ylim(c(0,1)) +
  labs(x="Date", y="Percent Vaccinated") +
  geom_hline(yintercept = 0.6727567, linetype="dashed", col = "red")
```



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 “2021-11-16”?

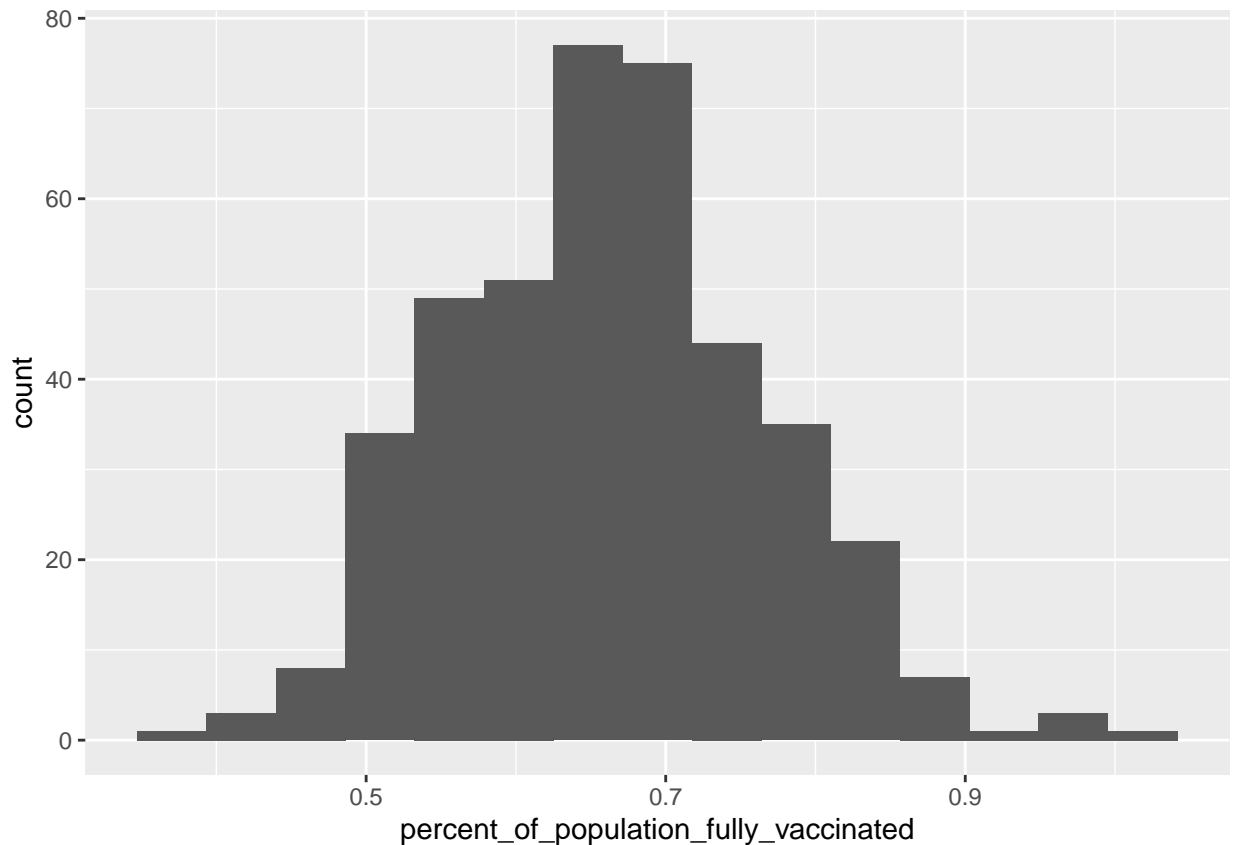
```
summary(vax.36$percent_of_population_fully_vaccinated)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.3519  0.5891  0.6649  0.6630  0.7286  1.0000
```

Q18. Using ggplot generate a histogram of this data.

```
library(ggplot2)

ggplot(vax.36) +
  aes(percent_of_population_fully_vaccinated)+
  geom_histogram(bins=15)
```



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 == "2021-11-16") %>%
  filter(zip_code_tabulation_area=="92109") %>%
  select(percent_of_population_fully_vaccinated)
```

```
## percent_of_population_fully_vaccinated
## 1 0.687763
```

```
vax %>% filter(as_of_date == "2021-11-16") %>%
  filter(zip_code_tabulation_area=="92040") %>%
  select(percent_of_population_fully_vaccinated)
```

```
## percent_of_population_fully_vaccinated
## 1 0.520463
```

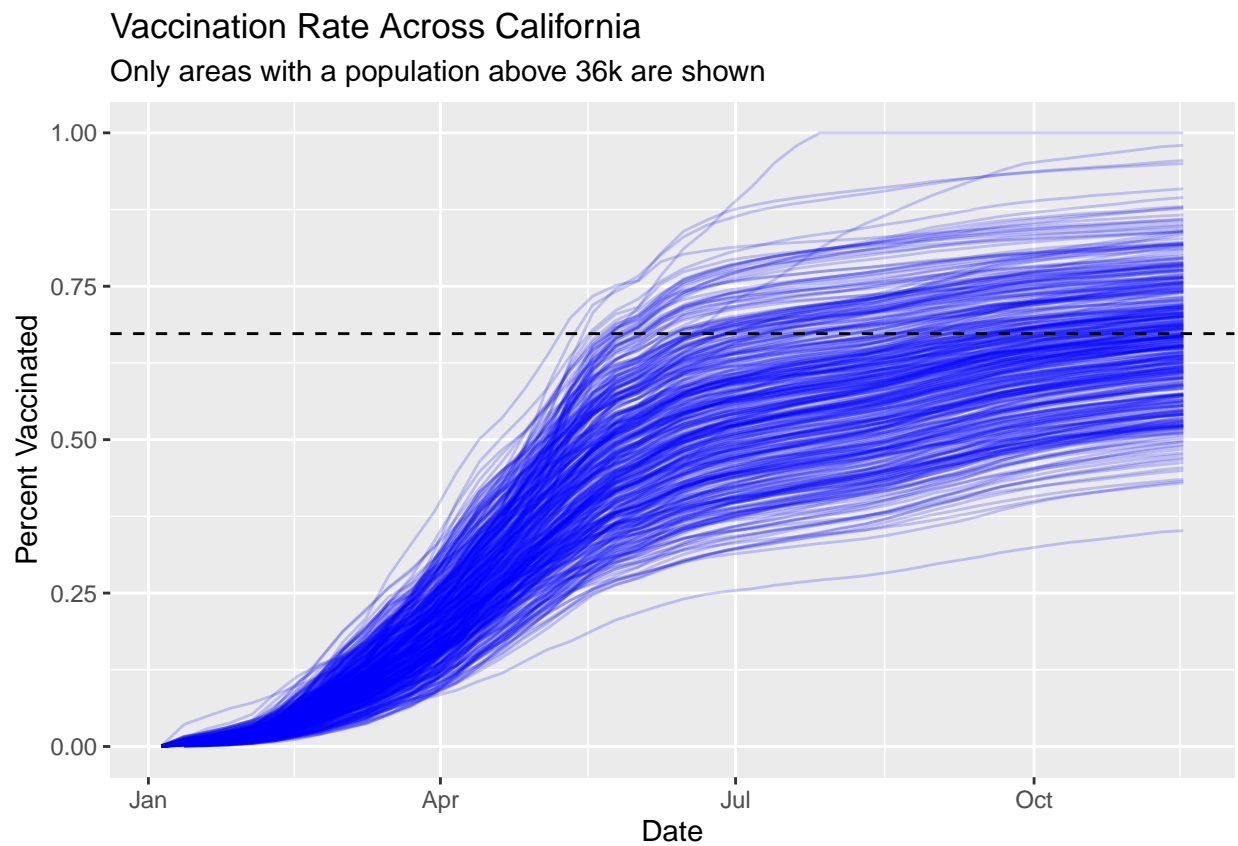
- Zip Code 92109 is above average at 68.78% vax where as zip code 92040 is below average at 52.05% vax.

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) +
  geom_line(alpha=0.2, color="blue") +
  ylim(c(0,1)) +
  labs(x="Date", y="Percent Vaccinated",
       title="Vaccination Rate Across California",
       subtitle="Only areas with a population above 36k are shown") +
  geom_hline(yintercept = 0.6727567, linetype="dashed")
```

## Warning: Removed 180 row(s) containing missing values (geom\_path).



Q21. How do you feel about traveling for Thanksgiving and meeting for in-person class next Week?