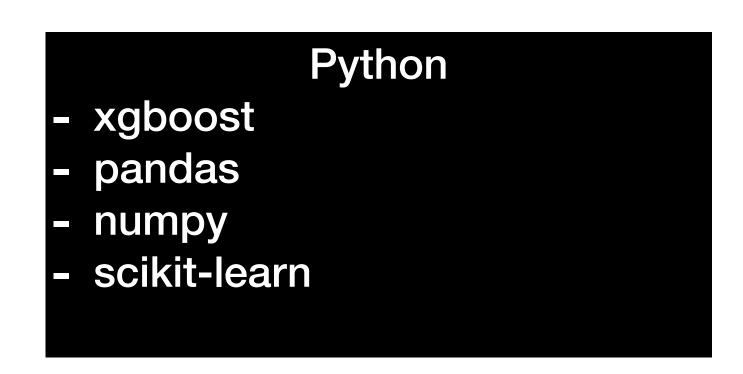
# Predicting COVID rates in municipalities

## Roadmap A

- Setup
- Overview of XGBoost
- Impact of COVID-19
- Format/transform data
- Build and test model

- Need to have python installed, or RStudio installed
- Need to have an internet connection to download packages and data
- All code, media, and data are available on GitHub





## Questions?

## Machine Learning

- Programming computers to make predictions about an input data set
- More efficient and accurate than rules and filters (usually)
- Requires lots of data to be accurate



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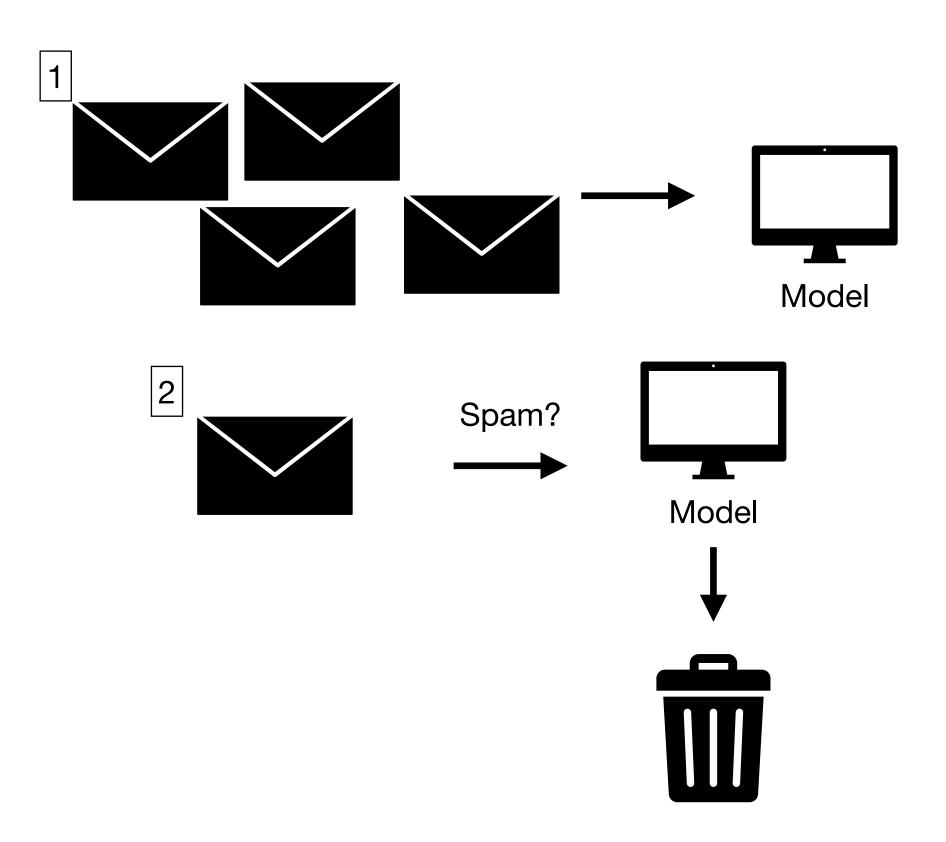
 If header contains 'GREAT DEAL'

- If body contains 'SAVINGS'
- If body contains 'act now'



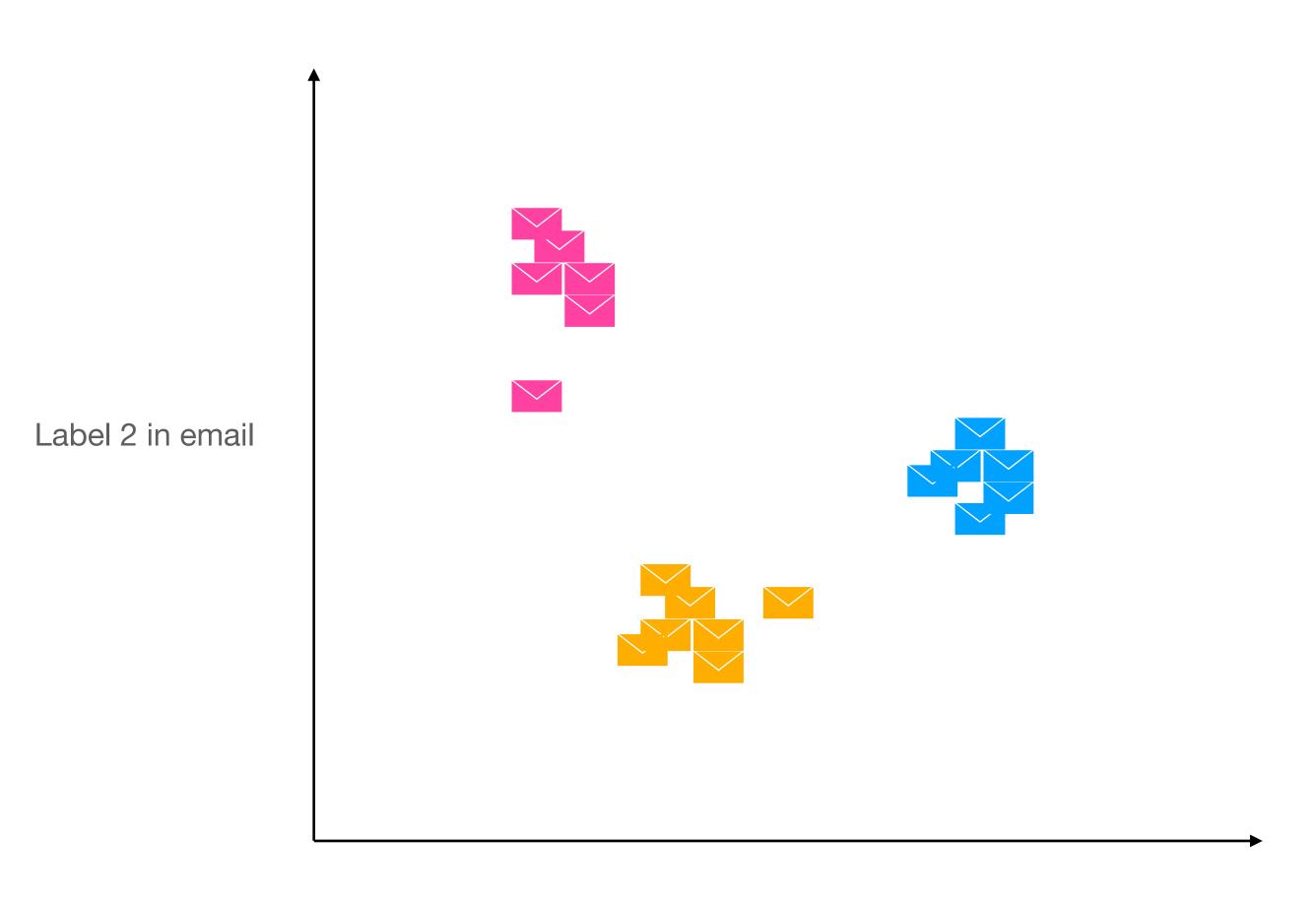
## Machine Learning

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#### Unsupervised Learning: k means

Among all emails, find 3 clusters with similar characteristics



Label 1 in email

#### Supervised Learning: Decision Tree Ensemble

Randomly select features and split data, then score the splits. Repeat to find the tree with the best prediction Contain string 'Sincerely' **Not Spam** Spam Contain string 'dealz' Does not Does not

## Key Terms

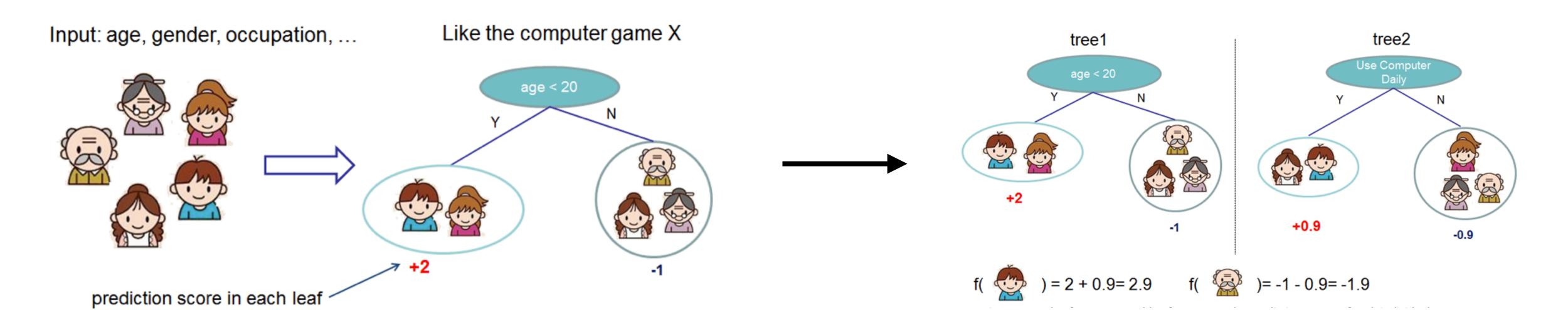
#### Train model with best parameters $\Theta$ to fit training data x and labels y

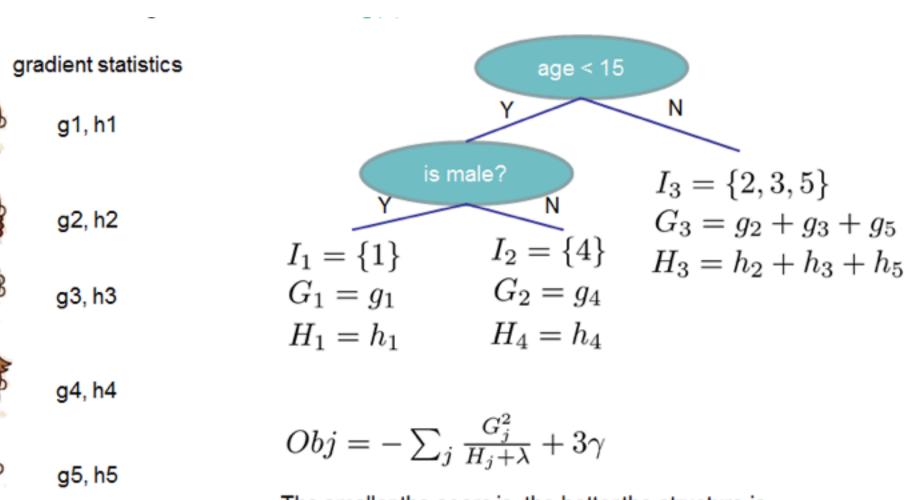
- objective function: measures how well the model fits the training data
- training loss: measure of how predictive the model is with respect to the training data, or how accurate the model's prediction was on data x with y
- regularization term: a penalty for overfitting the model to data, or too closely fitting the model to x

#### XGBoost: Extreme Gradient Boosting

Instance index

Example: Will a person like computer game x?



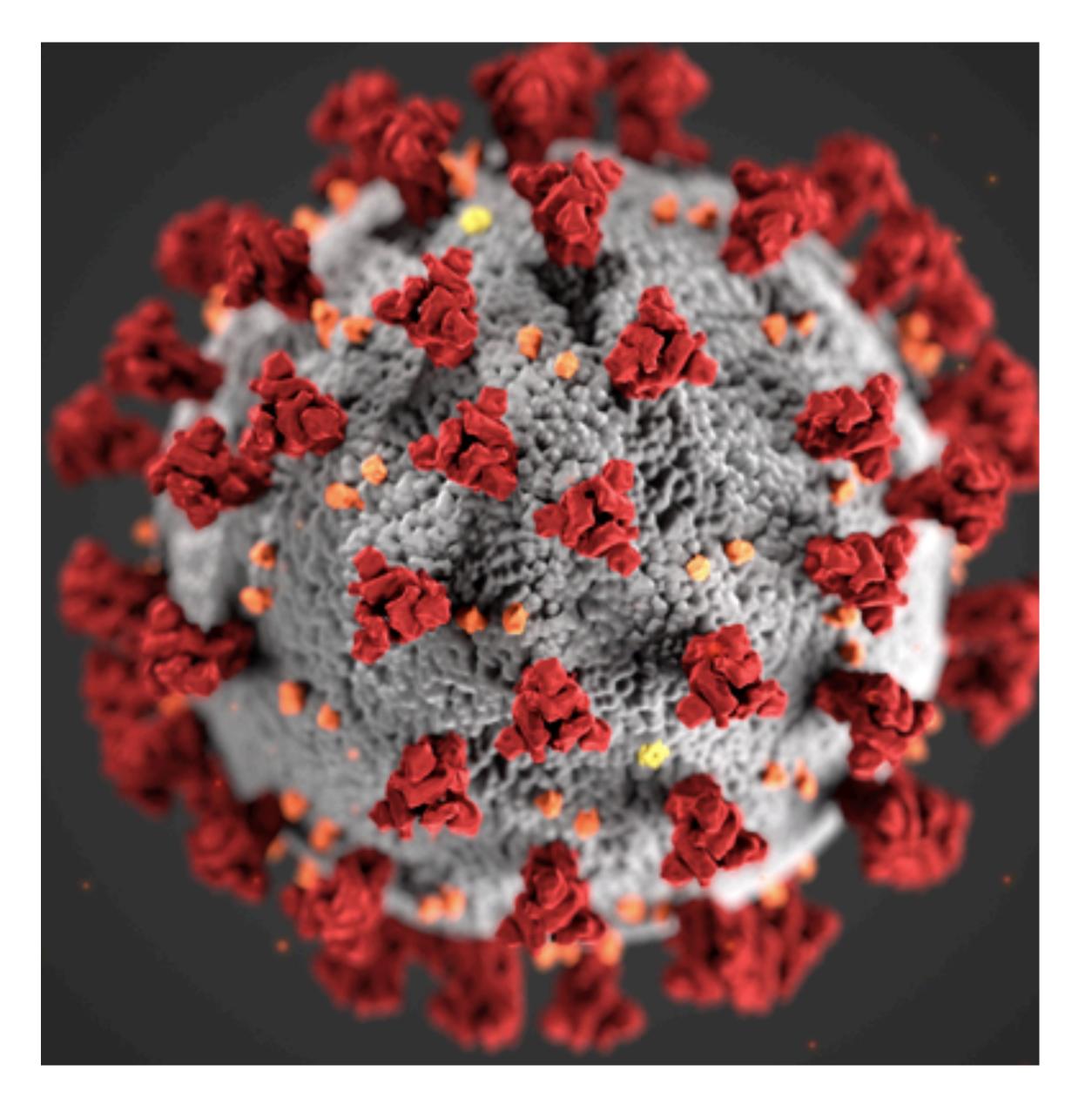


The smaller the score is, the better the structure is

## Questions?

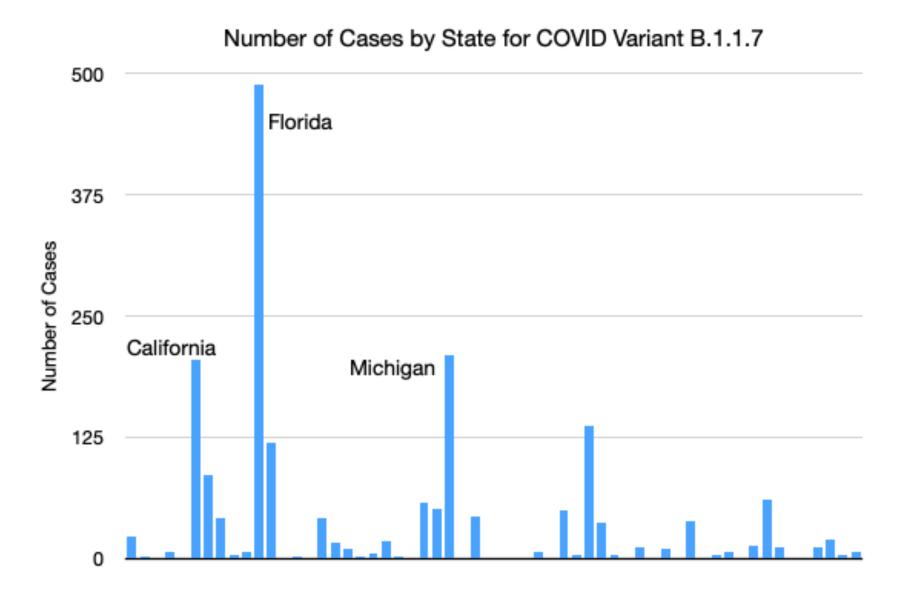
### COVID-19

- Pandemic affecting the world
- Caused by SARS-CoV-2 virus
- 28,828,370 cases reported (as of 02/23/2021)
- 495,270 deaths (as of 02/23/2021)



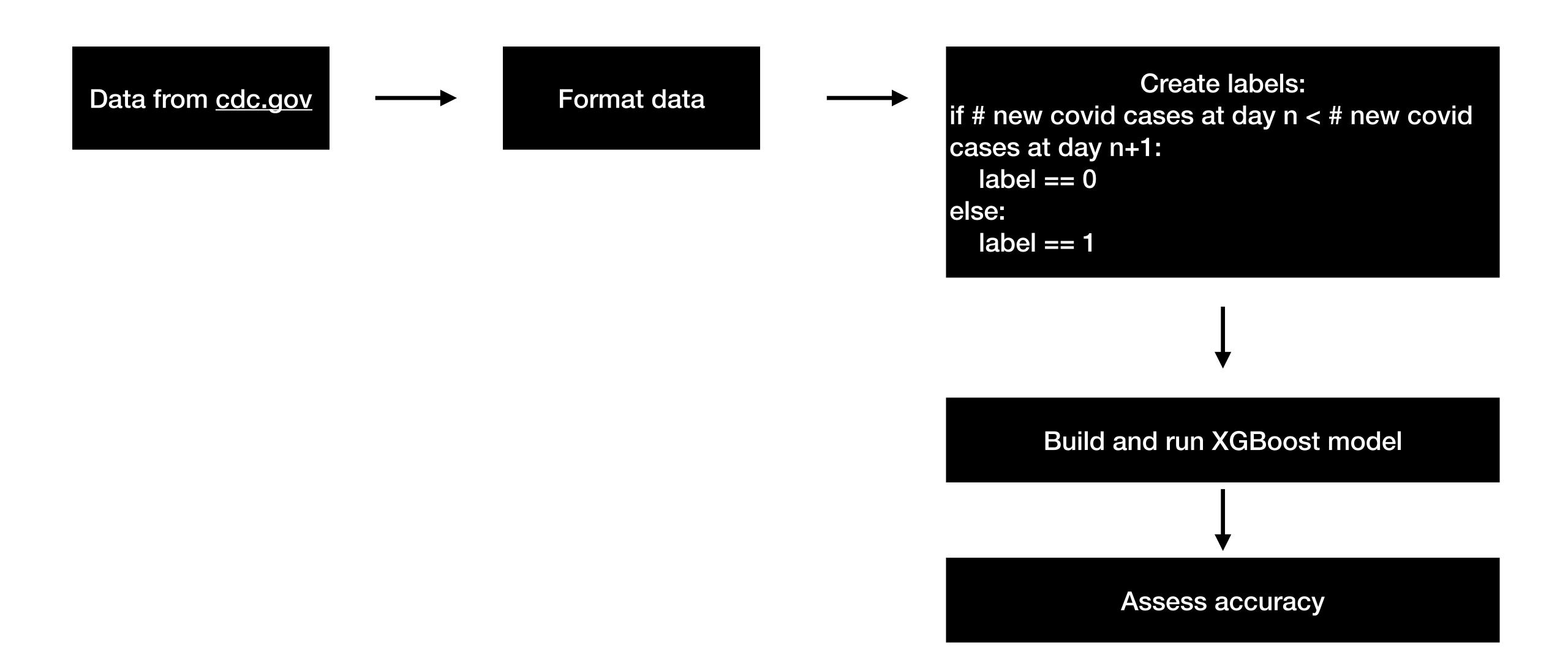
## Mask up!

- Vaccination efforts are underway
- Variants are emerging
- Transmission prevention efforts are critical as well as vaccination



- Is there a way to predict new cases?
  - Yes, several models and predictors exist. The models use numerous parameters, such as how well prevention guideline like social distancing and masking are followed
  - https://www.cdc.gov/coronavirus/2019-ncov/cases-updates/forecastscases.html
- Can we use existing data to build a simplified model to predict if there will be an increase in COVID cases the next day?

The number of hospitalizations and covid deaths can predict whether there will be an increase in COVID-19 cases the following day



- Download dataset from <a href="https://data.dhsgis.wi.gov/datasets/covid-19-historical-data-by-state/data?">https://data.dhsgis.wi.gov/datasets/covid-19-historical-data-by-state/data?</a>
   orderBy=DATE
- Descriptions on the columns is available here:
  - <a href="https://www.dhs.wisconsin.gov/publications/p02677.pdf">https://www.dhs.wisconsin.gov/publications/p02677.pdf</a>
- Import into Python pandas, RStudio (or Excel)
- Dataset is also available at GitHub

- Import csv data
- Order by date
- Create columns with counts of new hospitalizations, positive male cases, positive female cases, negative cases
- Create label column, then trim first 20 rows

## Import data, sort by date

#### Import data, sort by date

Python

df = pd.read\_csv('COVID-19\_Historical\_Data\_by\_State.orig.csv')

df = df.sort\_values('DATE')
df = df.reset\_index(drop=True)

#### **RStudio**

df = read.csv('COVID-19\_Historical\_Data\_by\_State.orig.csv')
df = df[order(df\$DATE),]
row.names(df) <- NULL</pre>

#### Create formatted columns and labels

#### Create formatted columns and labels

```
# repeat for the values "POS_FEM_NEW", "POS_MALE_NEW", "HOSP_UNK_NEW",
df['HOSP_NEW'] = df['HOSP_YES'].diff()

# setup labels
df['INCREASE'] = -1
df.loc[df['SCAN'] < 0, ['INCREASE']] = 1
df.loc[df['SCAN'] >= 0, ['INCREASE']] = 0

df = df.iloc[20:]
df_filtered.drop(df.tail(1).index, inplace=True)
```

```
# add HOSP_NEW values as difference between current day and prev day df = df %>% mutate(HOSP_NEW = HOSP_YES - lag(HOSP_YES, default = HOSP_YES[1]))

# repeat for the values: HOSP_NO, HOSP_UNK, POS_FEM, POS_MALE

df = df %>% mutate(SCAN = POS_NEW - lead(POS_NEW, default = POS_NEW))

df$INCREASE = ifelse(df$SCAN < 0, 1.0, 0.0)
```

#### **Build and Run Model!**