

Northwestern

# Public Health Informatics

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# About Me

## Training

- **PhD**, Epidemiology  
@University of Pittsburgh Graduate School of Public Health

## Business as Usual:

- **Assistant Professor**, Department of Preventive Medicine, Division(s) of Biomedical Informatics & Epidemiology @Northwestern Feinberg School of Medicine
- **Director, Data Science Services** @NUCATS Center for Data Science & Informatics

## These Days:

- **Chief Public Health Informatics Officer and Deputy Commissioner, Bureau of Informatics and IT** @Chicago Department of Public Health

# Objectives

- Understand and discuss public health informatics
- Understand public health surveillance
- Identify outbreaks in time-series data
- Discuss clustering, classification, and outbreak detection approaches

# Informatics

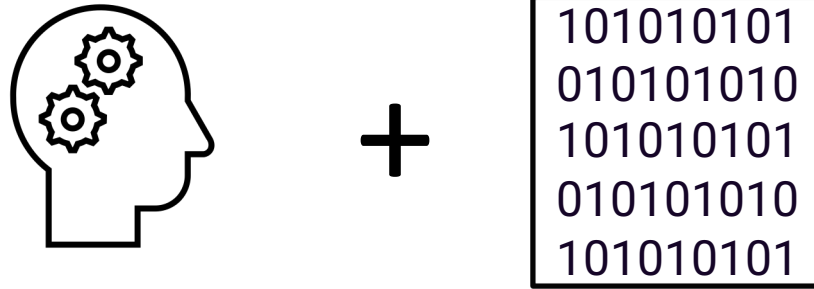
# What is Informatics?

The science of how we collect, organize and use information.

The kind of informatics depends on the domain user.

# A Single Unit of Informatics

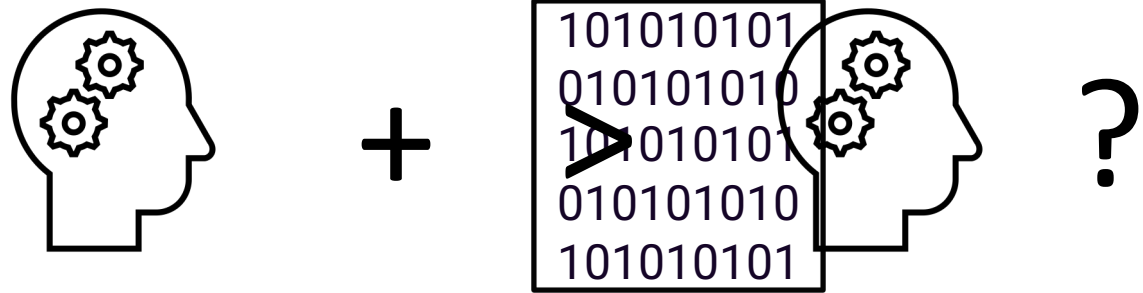
*People Using Information*



From: Friedman CP. A “fundamental theorem” of biomedical informatics. Journal of the American Medical Informatics Association. 2009 Mar 1;16(2):169-70.

# Asking a Good Informatics Question

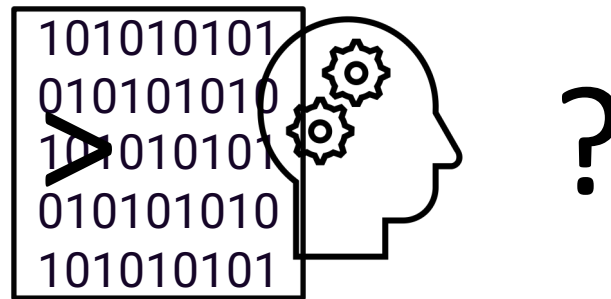
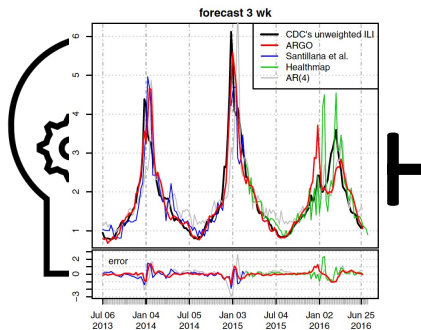
*Is a user, working in concert with an information resource, measurably better at achieving their objectives than without the resource?*



From: Friedman CP. A “fundamental theorem” of biomedical informatics. Journal of the American Medical Informatics Association. 2009 Mar 1;16(2):169-70.

# Information for Action

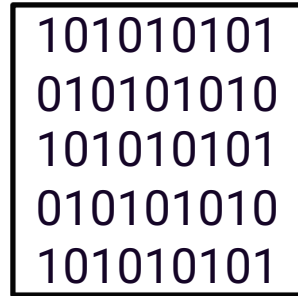
*Public Health Information for Public Health Actions*



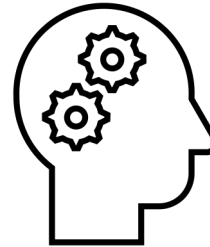
From: Friedman CP. A “fundamental theorem” of biomedical informatics. Journal of the American Medical Informatics Association. 2009 Mar 1;16(2):169-70.



# Informatics – What it isn't



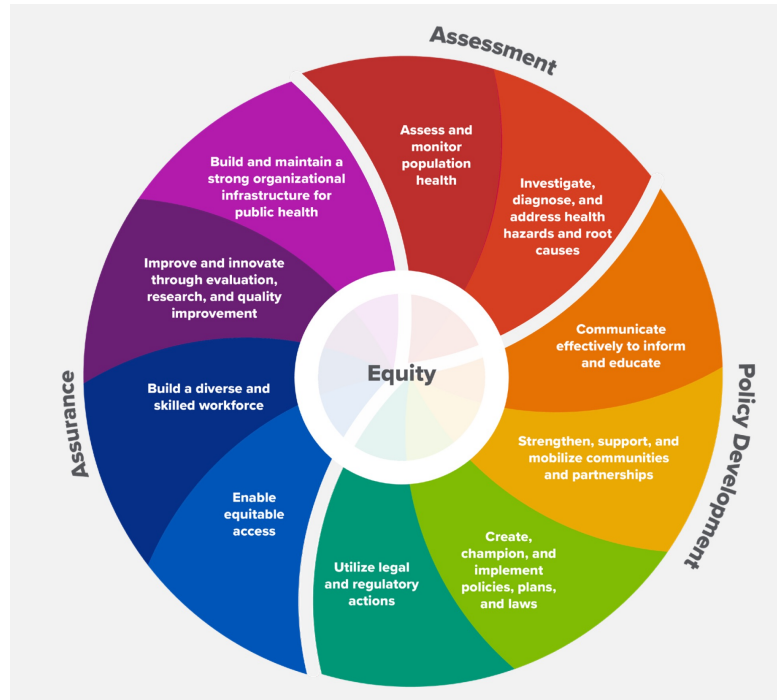
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From: Friedman CP. A “fundamental theorem” of biomedical informatics. Journal of the American Medical Informatics Association. 2009 Mar 1;16(2):169-70.

# Public Health

# Essential Public Health Services



Framework to protect and promote the health of all people in all communities.

CDC. 10 Essential Public Health Services. <https://www.cdc.gov/publichealthgateway/publichealthservices/essentialhealthservices.html>

# Chicago Public Health

[Home](#) / [Departments](#) / [Public Health](#)

The Chicago Department of Public Health provides guidance, services, and strategies that make Chicago a healthier and safer city.

## Featured Services and Programs



### HEALTHY COMMUNITIES

Environment, Healthy Chicago 2.0, Healthy Chicago 2025, PlayStreets, Vector Control



### HEALTH DATA AND REPORTS

Epi. Data Request Forms, HIV/STI Surveillance, IRB



### HEALTHY HOMES

Bed Bugs, Lead



### HEALTH INSPECTIONS

Food Protection Services, Food Service Sanitation Certificate



### HEALTHY LIVING

Adolescent Health, Cannabis, HIV, Obesity, Pharmaceutical Representative License, Tobacco, Substance Use, Violence Prevention



### HEALTHY MOTHERS AND BABIES

Family Connects, WIC



### HEALTH PROTECTION AND RESPONSE

Communicable Disease, Emergency Preparedness, Immunization, Tuberculosis



### HEALTH SERVICES

Breast Health, HIV Testing, Mental Health, Primary Care, WIC, Office of Compliance

## Department Main Office

### Public Health

Phone: 312.747.9884  
TTY: 312.747.2374

### Contact Us

333 S. State Street  
Room 200  
Chicago, IL 60604 (For 24-hour assistance or to report a public health issue, call 311.)

[Get Directions](#)

### Leadership

Allison Arwady, M.D.  
Commissioner, Chicago Department of Public Health

## Department Facts

### About Us:

Mission

Leadership

Our Structure

## Related Links

[Senior Safety Ordinance](#)

[Healthy Chicago 2025](#)

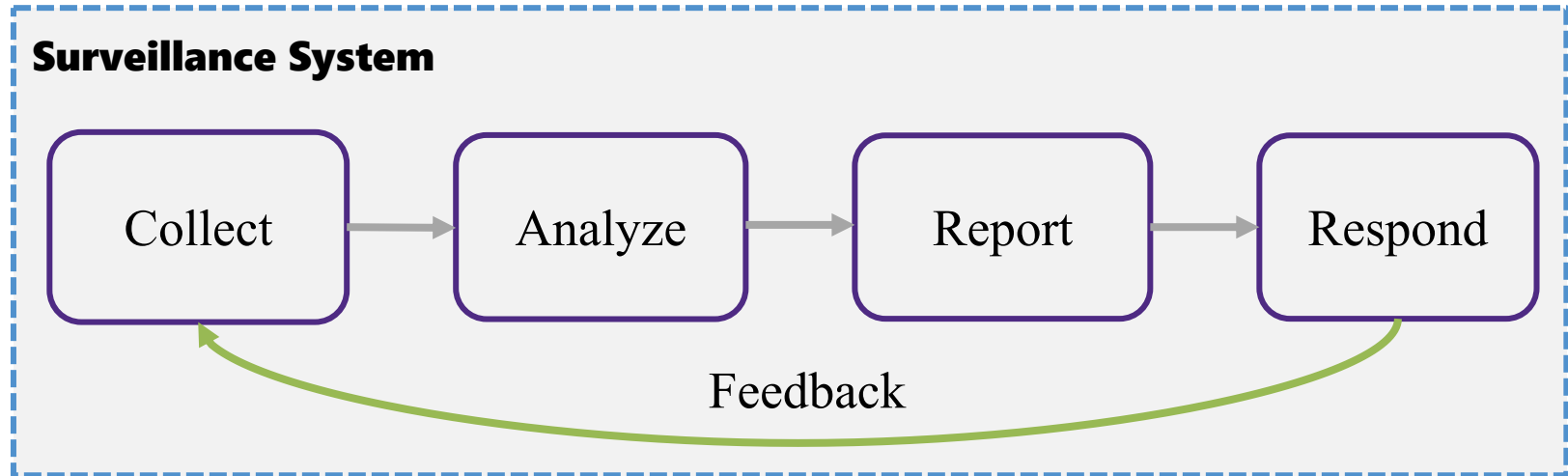
[Coronavirus Disease 2019 \(COVID-19\)](#)

[Tobacco Products Sign \(pdf\)](#)

# Public Health Surveillance

# Public Health Workflows

*Traditional public health practice use of information reflects many informatics use cases.*



Thacker, Stephen B., and Ruth L. Berkelman. "Public health surveillance in the United States." *Epidemiologic reviews* 10.1 (1988): 164-190.

# Clusters

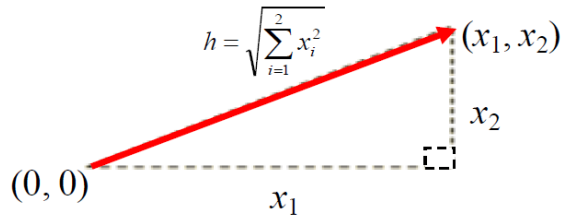
# Measuring Distances

PROC DISTANCE METHOD=EUCLID    PROC DISTANCE METHOD=CITYBLOCK

## Euclidean Distance

$$D_E = \|\mathbf{x} - \mathbf{w}\| = \sqrt{\sum_{i=1}^k (x_i - w_i)^2}$$

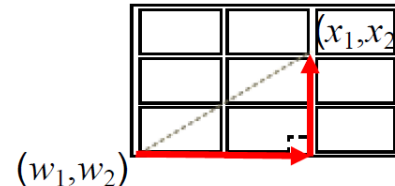
- Euclidean distance gives the linear distance between any two points in  $n$ -dimensional space.
- It is a generalization of the Pythagorean theorem.



## City Block Distance

$$D_{M_1} = \sum_{i=1}^d |x_i - w_i|$$

- The distance between two points is measured along the sides of a right triangle.
- It is the distance that you would travel if you had to walk along the streets of a right-angled city.

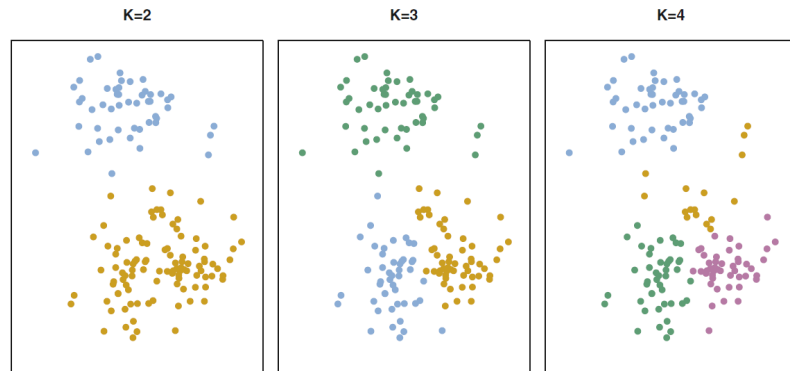


[https://documentation.sas.com/?docsetId=statug&docsetTarget=statug\\_introclus\\_sect004.htm&docsetVersion=14.3&locale=en](https://documentation.sas.com/?docsetId=statug&docsetTarget=statug_introclus_sect004.htm&docsetVersion=14.3&locale=en)



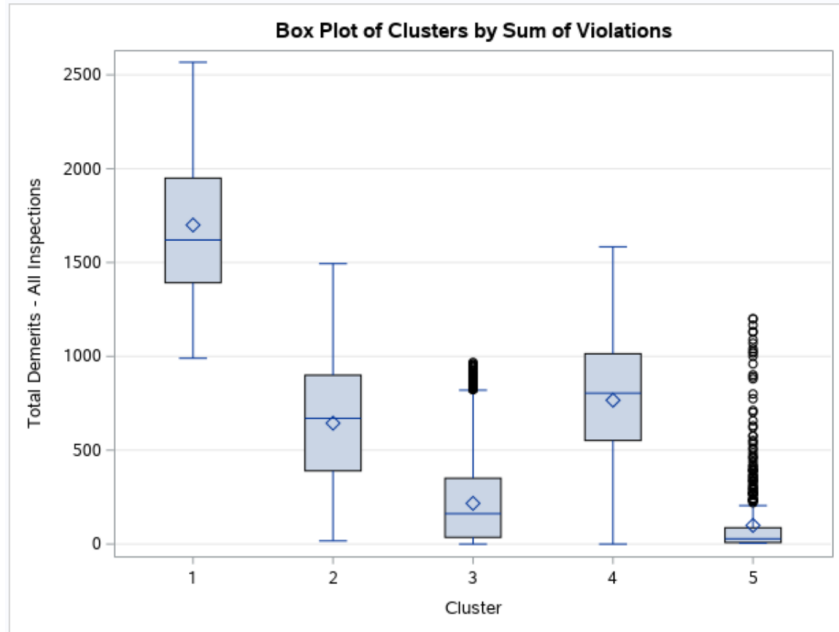
# K-means Clustering

- “Partive” clustering
- Specify number of clusters
- K-means algorithm assigns exactly each observation to exactly one cluster



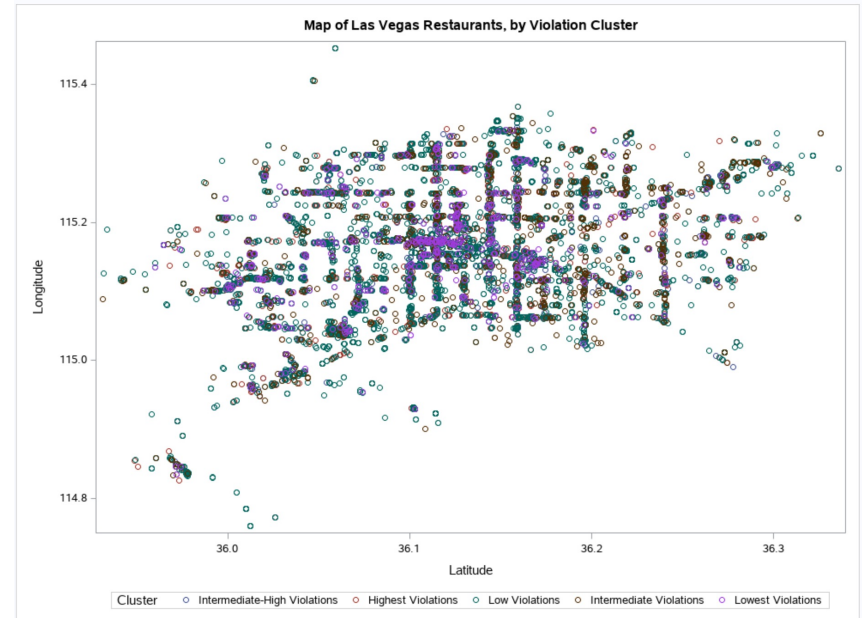
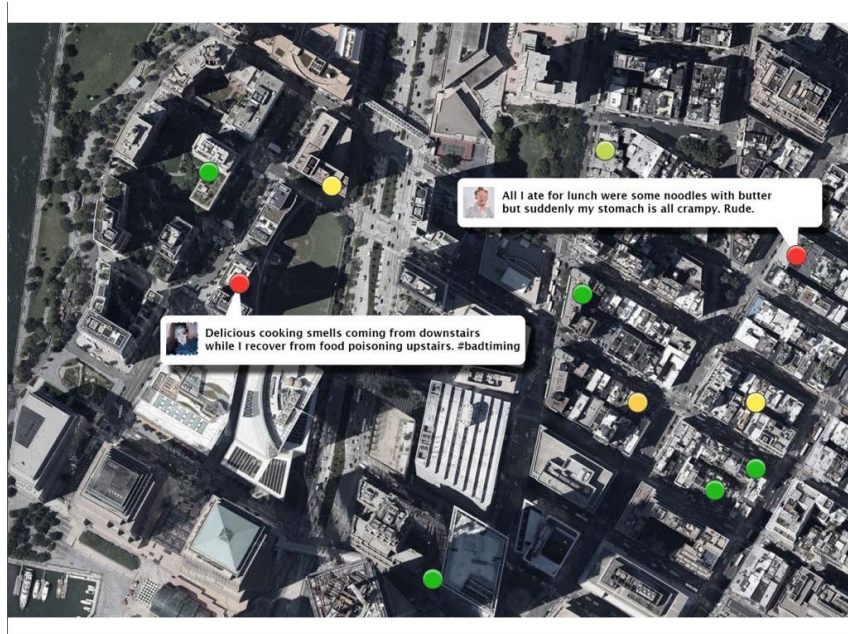
May be the only practical choice when the data set is large.

# Plot by Coordinates



See: Course shared files > Clustering > sas > lv\_inspections\_clustering\_190522\_1434.sas

# Refined Social Listening



See: Course shared files > Clustering > sas > lv\_inspections\_clustering\_190522\_1434.sas

# Classification

# Surveillance Case Definitions

A surveillance case definition is a set of uniform criteria used to define a disease for public health surveillance. Surveillance case definitions:

- Enable public health officials to **classify and count** cases consistently across reporting jurisdictions.  
**Not intended to be used by healthcare providers for making a clinical diagnosis** or determining how to meet an individual patient's health needs.
- The Council of State and Territorial Epidemiologists recommends that state health departments **report cases** of selected diseases to CDC's National Notifiable Diseases Surveillance System (NNDSS).
- CSTE issues Position Statements, providing **uniform criteria** of national notifiable infectious and non-infectious conditions for reporting purposes.

CDC. "Surveillance Case Definitions for Current and Historical Conditions." <https://ndc.services.cdc.gov/>

# Terms Used in Case Classification

- **Confirmed case:** a case that is classified as confirmed for reporting purposes.
- **Probable case:** a case that is classified as probable for reporting purposes.
- **Suspected case:** a case that is classified as suspected for reporting purposes.

From: CDC. Case definitions for infectious conditions under public health surveillance. MMWR

# Terms Used in Case Classification

- **Clinically compatible case:** a clinical syndrome generally compatible with the disease, as described in the clinical description.
- **Epidemiologically linked case:** a case in which a) the patient has had **contact** with one or more persons who either have/had the disease or have been **exposed** to a point source of infection (i.e., a single source of infection, foodborne-disease outbreak) and b) transmission of the agent by the usual modes of transmission is **plausible**. A case may be considered epidemiologically linked to a laboratory confirmed case if at least one case in the chain of transmission is laboratory confirmed.

From: CDC. Case definitions for infectious conditions under public health surveillance. MMWR

# Terms Used in Case Classification

- **Laboratory-confirmed case:** a case that is confirmed by one or more of the laboratory methods listed in the case definition under Laboratory Criteria for Diagnosis. Although other laboratory methods can be used in clinical diagnosis, only those listed are accepted as laboratory confirmation for national reporting purposes.
- **Supportive or presumptive laboratory results:** specified laboratory results that are **consistent** with the diagnosis, yet do not meet the criteria for laboratory confirmation.

From: CDC. Case definitions for infectious conditions under public health surveillance. MMWR



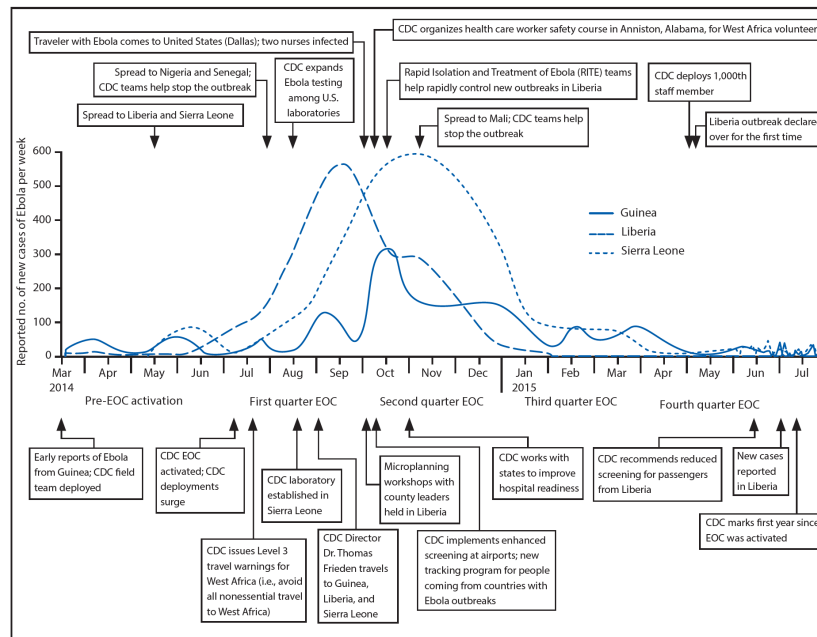
Ebola

# Ebola Virus Disease (EVD)

- Rare and deadly disease in people and nonhuman primates.
- The viruses that cause EVD are located mainly in sub-Saharan Africa.
- People can get EVD through direct contact with an infected animal (bat or nonhuman primate) or a sick or dead person infected with Ebola virus.

# 2014–2016 Ebola Epidemic

FIGURE 1. CDC's response to the Ebola epidemic, from the first reported cases through the first year after CDC's EOC was activated, and approximate number of reported new cases of Ebola per week — Guinea, Liberia, and Sierra Leone, March 2014–July 2015



Abbreviations: Ebola = Ebola virus disease; EOC = Emergency Operations Center.

Bell BP, Damon IK, Jernigan DB, Kenyon TA, Nichol ST, O'Connor JP, Tappero JW. Overview, Control Strategies, and Lessons Learned in the CDC Response to the 2014-2016 Ebola Epidemic. MMWR Suppl. 2016 Jul 8;65(3):4-11. doi: 10.15585/mmwr.su6503a2. PMID: 27389903.

# WHO Ebola '14 Case Definition

## Standard

**Suspected:** Illness with onset of fever and no response to treatment for usual causes of fever in the area, and at least one of the following signs: bloody diarrhoea, bleeding from gums, bleeding into skin (purpura), bleeding into eyes and urine.

**Confirmed:** A suspected case with laboratory confirmation (positive IgM antibody, positive PCR or viral isolation)

World Health Organization. "Case definition recommendations for Ebola or Marburg Virus Diseases."  
<https://www.who.int/csr/resources/publications/ebola/ebola-case-definition-contact-en.pdf?ua=1>

# Ebola Outbreak West Africa '14



WHO and partners respond to the outbreak of Ebola virus disease in Guinea

“WHO and partners respond to the outbreak of Ebola virus disease in Guinea.” Posted 08/11/2014. <https://youtu.be/uDicsP3femI>

# Ebola Outbreak West Africa '14



- How do barriers to specimen collection affect case reporting?
- How do low resources affect laboratory confirmation?
- How does 'social mobilization' improve case detection in villages?

"WHO and partners respond to the outbreak of Ebola virus disease in Guinea." Posted 08/11/2014. <https://youtu.be/uDicsP3femI>

# Evaluating Definitions

# Evaluating Case Definitions

		Truth		
		Disease	No Disease	
Test Result	Positive	<b>A</b> True Positive	<b>B</b> False Positive	$T_{Test\ Positive}$
	Negative	<b>C</b> False Negative	<b>D</b> True Negative	
		$T_{disease}$	$T_{non-disease}$	Total

- Prevalence of disease
  - $T_{disease} / Total \times 100$
- Sensitivity - Probability that a test will indicate 'disease' among those with the disease
  - $A / (A + C) \times 100$
- Specificity - The fraction of those without disease who will have a negative test result
  - $D / (D + B) \times 100$

See: Doi SA, Williams GM, editors. Methods of clinical epidemiology. Berlin: Springer; 2013 Jun 1.



# Evaluating Case Definitions

		Truth		
		Disease	No Disease	
Test Result	Positive	<b>A</b> True Positive	<b>B</b> False Positive	$T_{Test\ Positive}$
	Negative	<b>C</b> False Negative	<b>D</b> True Negative	
		$T_{disease}$	$T_{non-disease}$	Total

- Positive Predictive Value: what is the chance that a person with a positive test truly has the disease?
  - $A/(A+B) \times 100$
- Negative Predictive Value: what is the chance that a person with a negative test truly does not have the disease?
  - $D/(D+C) \times 100$

See: Doi SA, Williams GM, editors. Methods of clinical epidemiology. Berlin: Springer; 2013 Jun 1.

# Classifiers in Practice

# Syndromic Surveillance

- Monitor emergency department visits to detect disease outbreaks **early**
- Collect **chief complaint** information transmitted electronically to the health department daily
- Analyze for **temporal and spatial aberrations**

Heffernan R et al. Syndromic surveillance in public health practice, New York City. Emerg Infect Dis. 2004 May;10(5):858-64. doi: 10.3201/eid1005.030646.

# Syndrome Coding

Table 1. Syndrome coding and hierarchy

Syndrome	Includes	Excludes
Common cold	Nasal drip, congestion, stuffiness	Chest congestion, sore throat
Sepsis	Sepsis, cardiac arrest, unresponsive, unconscious, dead on arrival	
Respiratory	Cough, shortness of breath, difficulty breathing, croup, dyspnea, bronchitis, pneumonia, hypoxia, upper respiratory illness, chest congestion	Cold
Diarrhea	Diarrhea, enteritis, gastroenteritis, stomach virus	
Fever	Fever, chills, flu, viral syndrome, body ache and pain, malaise	Hay fever
Rash	Vesicles, chicken pox, folliculitis, herpes, shingles	Thrush, diaper and genital rash
Asthma	Asthma, wheezing, reactive airway, chronic obstructive airway disease	
Vomiting	Vomiting, food poisoning	

Heffernan R et al. Syndromic surveillance in public health practice, New York City. *Emerg Infect Dis.* 2004 May;10(5):858-64. doi: 10.3201/eid1005.030646.

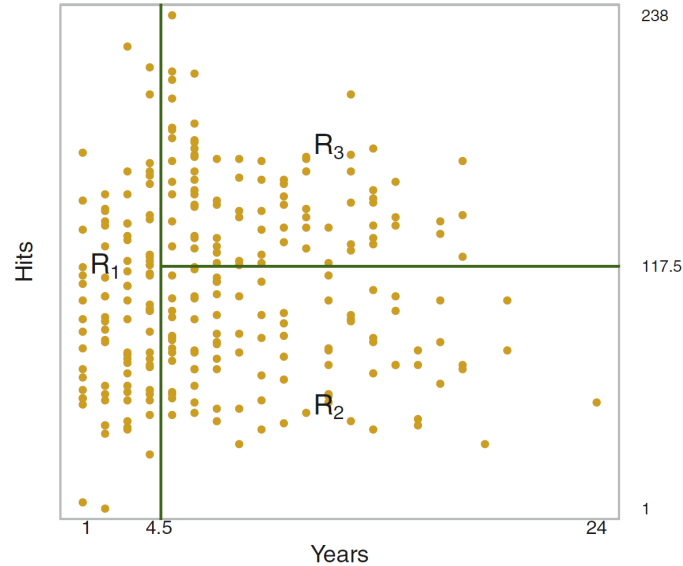
# Introduction to Trees

- Trees segment a predictor space into a number of **simple regions**.
- The set of splitting rules segmenting the predictor space can be summarized in a **tree**.
- This approach often called '**decision tree**' methods.
- Trees predicting numeric outcomes are **regression trees**.
- Trees predicting nominal outcomes are **classification trees**.
- Trees offer high interpretability
- Trees often **overfit**, resulting in high bias

# Partitioning

Tree for predicting the log salary of a baseball player, based on the number of years that he has played in the major leagues and the number of hits that he made in the previous year.

Three-region partition **regression** tree



See Figure 8.1, 8.2. Chapter 08: James, Gareth, et al. An introduction to statistical learning. Vol. 112. New York: springer, 2013.

# Tree terminology

**Terminal node**: final segment for an observation. Can also be called leaves.

**Leaves** are at the bottom of the **upside down** tree

**Internal nodes**: The points along the tree where the predictor space is split.

**Branches** emanate from an internal node split.

See Chapter 08: James, Gareth, et al. An introduction to statistical learning. Vol. 112. New York: springer, 2013.

# Classification Trees

- For a classification tree, we predict that each observation belongs to the **most commonly occurring class** of training observations in the region to which it belongs.
- Use **recursive binary splitting** to grow a classification tree.
- **Classification error rate**: Fraction of the training observations in that region that do not belong to the most common class.
- Classification error is **not** sufficiently sensitive for tree-growing.

See Chapter 08: James, Gareth, et al. An introduction to statistical learning. Vol. 112. New York: springer, 2013.



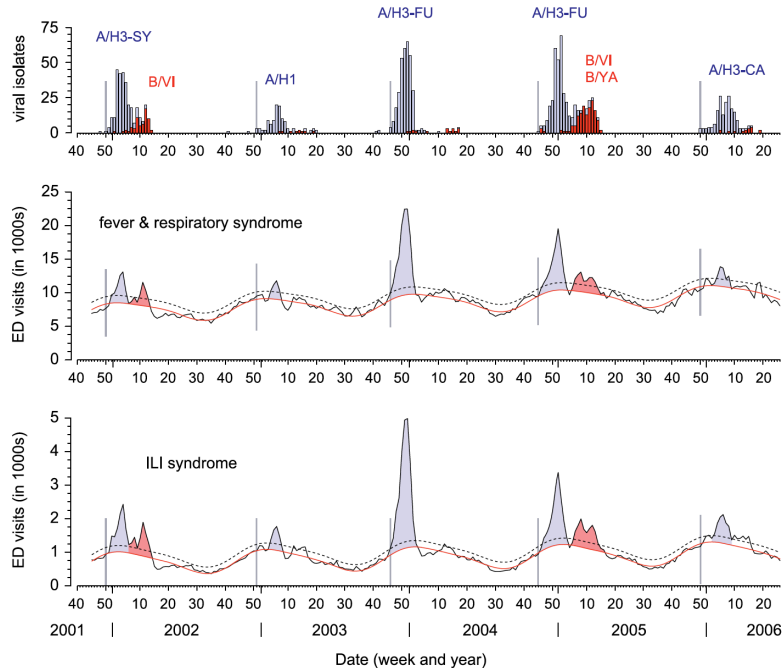
# Pruning a Classification Tree

- **Gini index** is a measure of purity; a small value indicates that a node contains predominantly observations from a single class.
- Gini measures total **variance** of the misclassification rate across all classes.
- The Gini index takes on a **small value** if all misclassification rates are close to zero or one.

See Chapter 08: James, Gareth, et al. An introduction to statistical learning. Vol. 112. New York: springer, 2013.

# Baseline Simulations

# Baselines

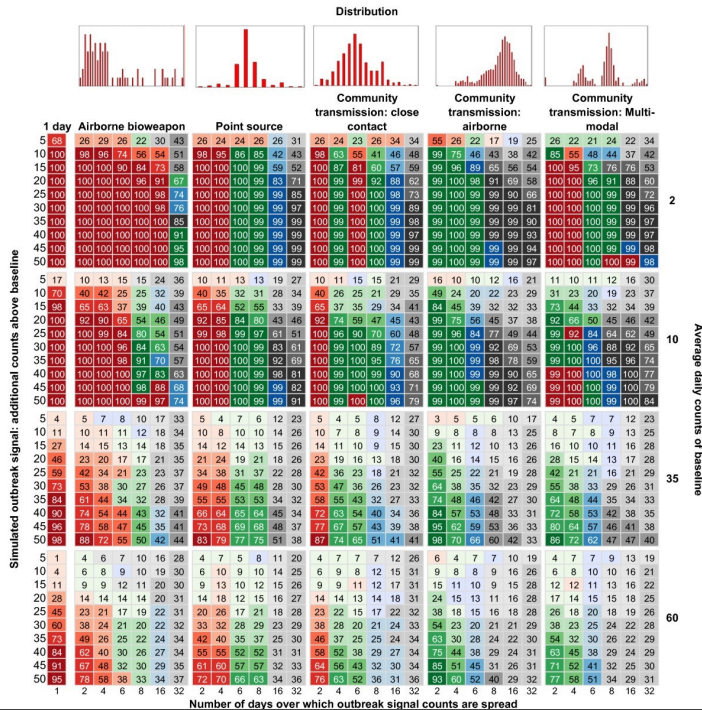


**Plot for trends** – seasonal trend, secular trend (regression)

- Describe **min, max, standard deviation**.
  - How strong is the seasonal affect?
  - Are there essentially no cases in the off-season?
- Does a **secondary data source** exist (e.g. google trends, CDC data) to correlate with your baseline?

Olson, Donald R., et al. "Monitoring the impact of influenza by age: emergency department fever and respiratory complaint surveillance in New York City." PLoS Medicine 4.8 (2007): e247.

# Determinants of Detectability



Visualize the effects of baseline and:

- outbreak temporal distribution,
- size, and
- duration

on relative algorithm performance

Figure 3. Jackson ML et al. A simulation study comparing aberration detection algorithms for syndromic surveillance. BMC Med Inform Decis Mak. 2007 Mar 1;7:6.

# Surveillance Dynamics

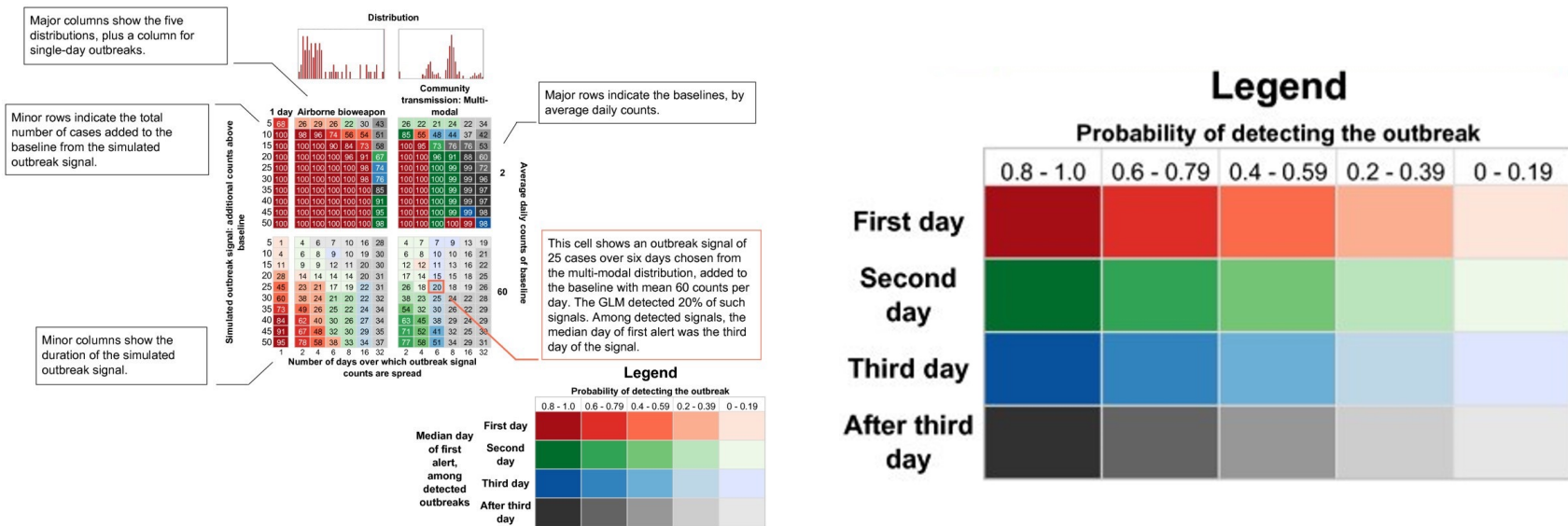
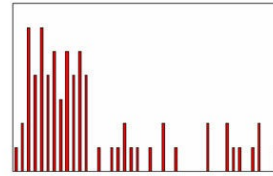


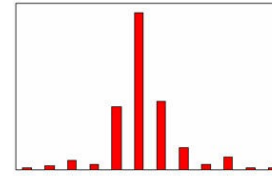
Figure 2. Jackson ML et al. A simulation study comparing aberration detection algorithms for syndromic surveillance. BMC Med Inform Decis Mak. 2007 Mar 1;7:6.

# What does it look like?

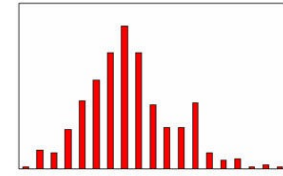
- Shape
- Spread
- Scenario



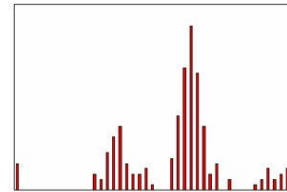
Airborne  
bioweapon  
(19)



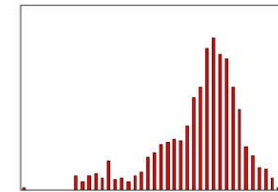
Point-source  
(20)



Community  
transmission:  
close contact  
(21)

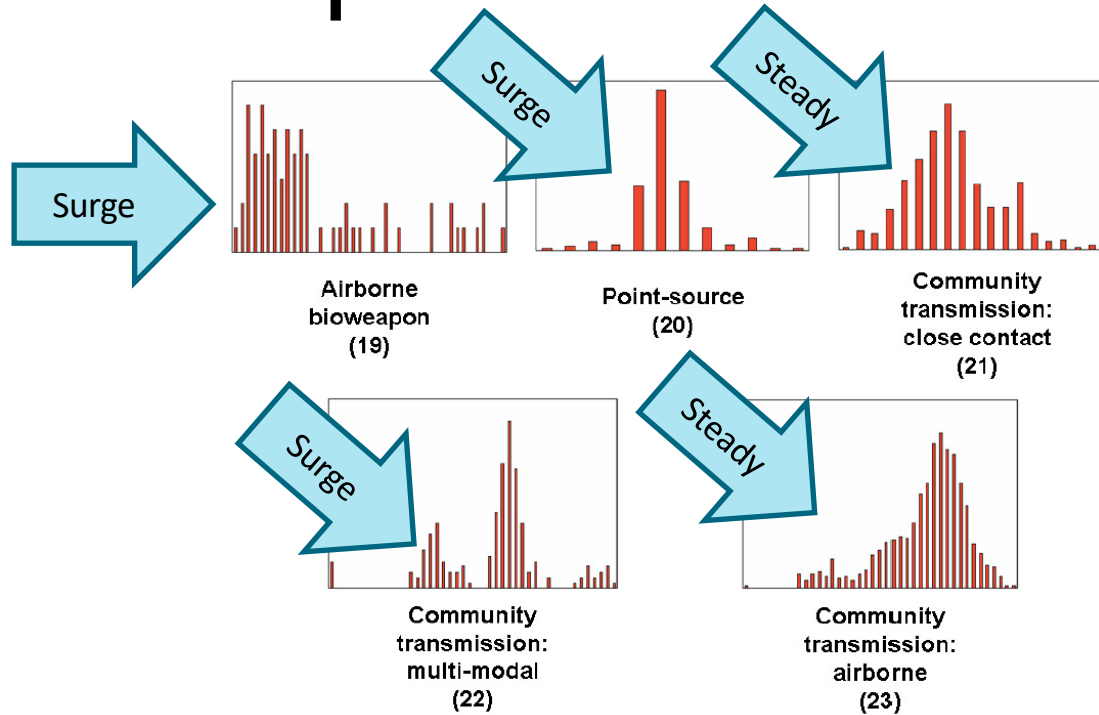


Community  
transmission:  
multi-modal  
(22)

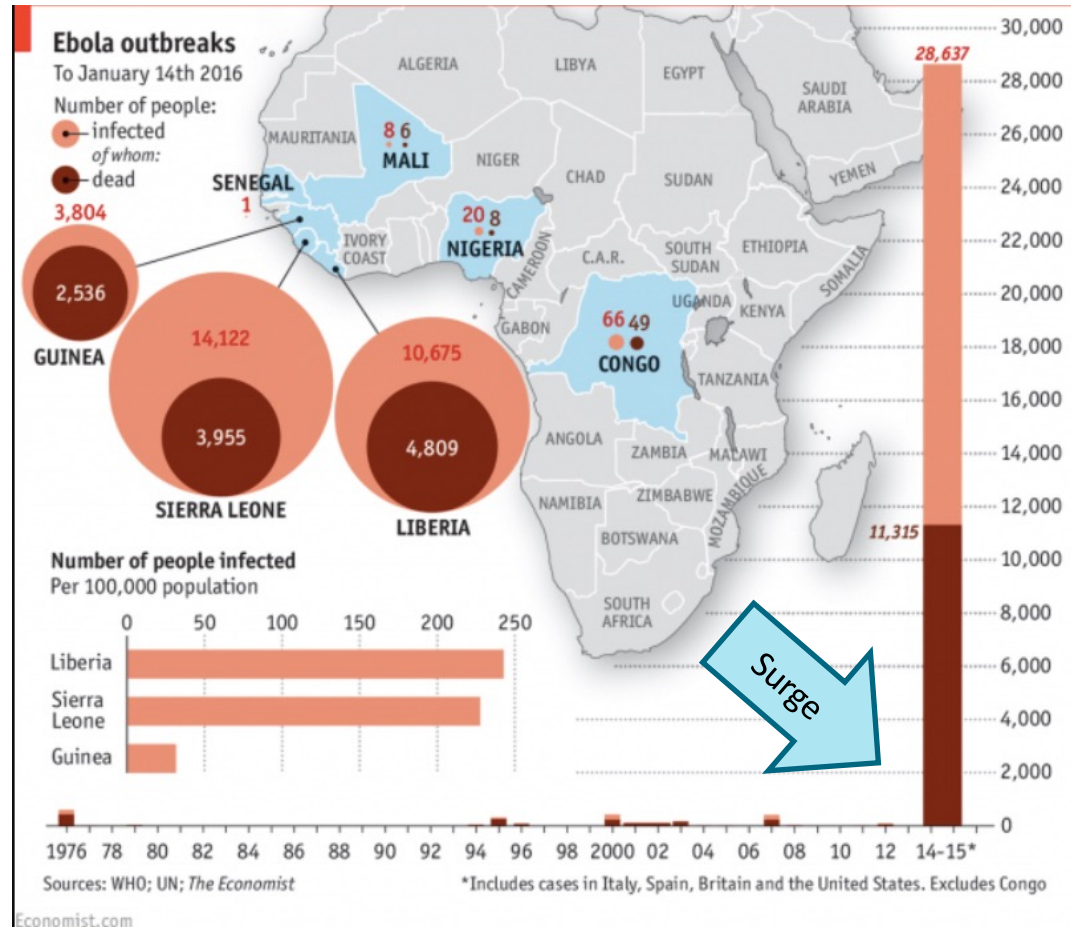
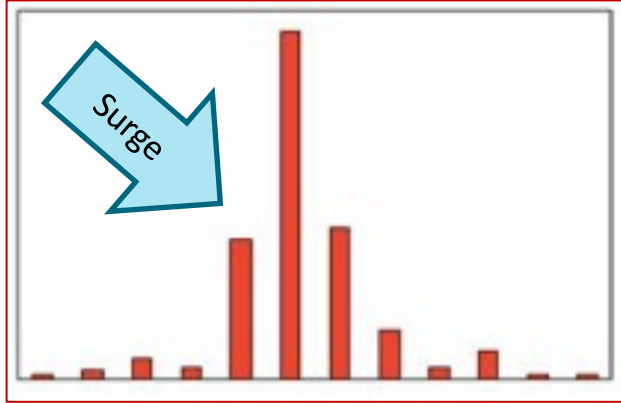


Community  
transmission:  
airborne  
(23)

# Epidemic Curves



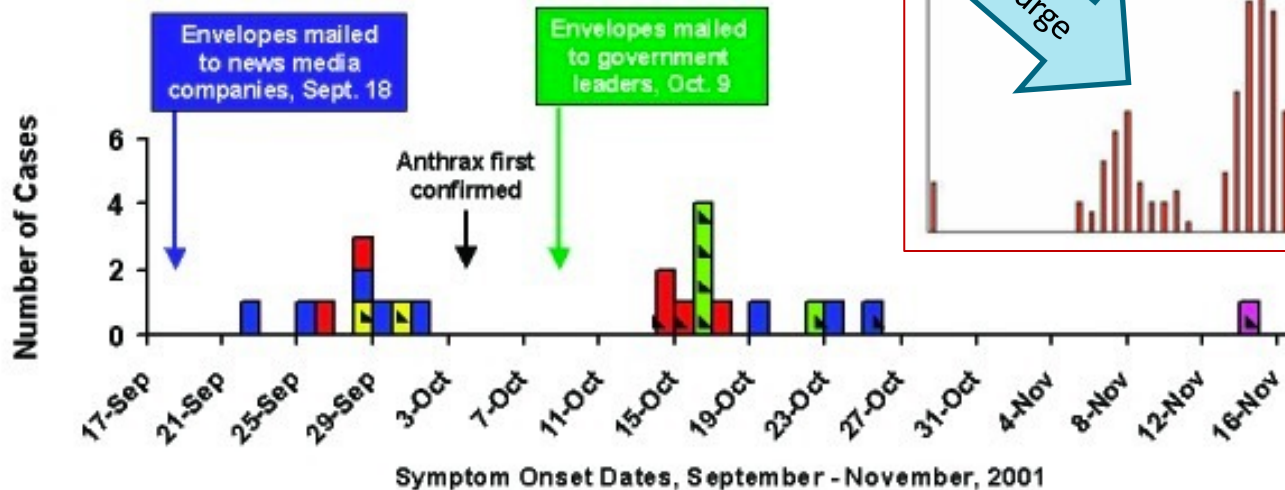
Jackson, Michael L., et al. "A simulation study comparing aberration detection algorithms for syndromic surveillance." *BMC Medical Informatics and Decision Making* 7.1 (2007): 6.



From the Economist: [Ebola in Africa: the end of a tragedy?](#)



# Epidemic Curve



Florida
  New York City
  New Jersey
  Dist. of Columbia\*
  Connecticut

▲ = inhalational anthrax case,

\* = Metropolitan District of Columbia Area, cases were residents of Maryland (3) and Virginia (2)

From: Jernigan, Daniel B., et al. "Investigation of bioterrorism-related anthrax, United States, 2001: epidemiologic findings." *Emerging infectious diseases* 8.10 (2002): 1019.

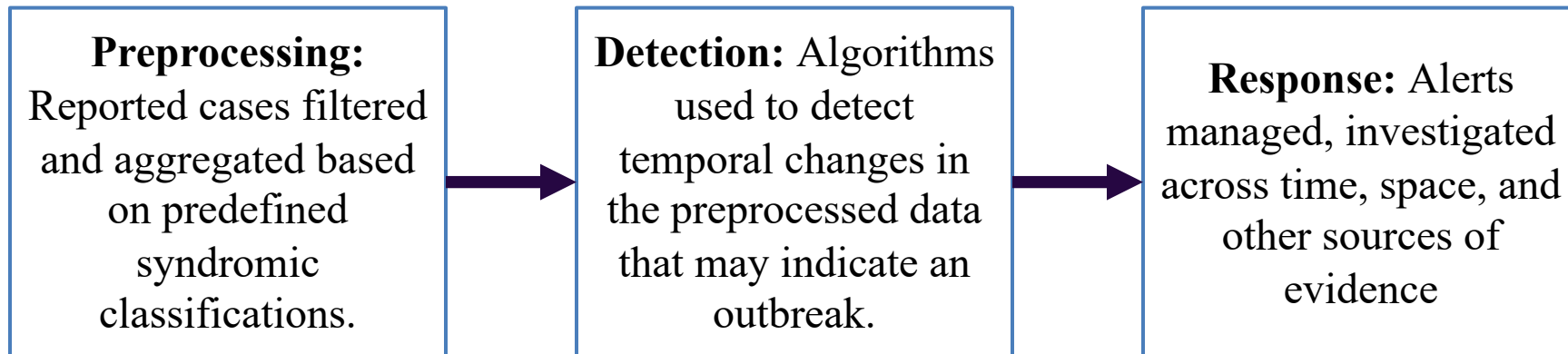
# Detecting Outbreaks

# What's An Outbreak

- **Epidemic:** refers to an increase, often sudden, in the number of cases of a disease above what is normally expected in that population in that area.
- **Outbreak** carries the same definition of epidemic, but is often used for a more limited geographic area.
- **Cluster** refers to an aggregation of cases grouped in place and time that are suspected to be greater than the number expected, even though the expected number may not be known.
- **Pandemic** refers to an epidemic that has spread over several countries or continents, usually affecting a large number of people.

Dicker, Richard, et al. "Principles of epidemiology in public health practice." Atlanta GA: US Department of Health and Human Services (2006).

# Algorithmic Surveillance



Murphy SP, Burkom H. Recombinant temporal aberration detection algorithms for enhanced biosurveillance. J Am Med Inform Assoc. 2008 Jan-Feb;15(1):77-86.

Which algorithm  
is most effective at detecting potential  
outbreaks?

# Challenging Factors

- **Size** with median from 0 to 100s
- **Cyclic behaviors** including day-of-week (DOW) effects and seasonal fluctuations
- **Pathogen**, its infectivity, the method of introduction to the susceptible population, and the population's underlying social network
- **Unknown shape** of the resulting time series

Murphy SP, Burkom H. Recombinant temporal aberration detection algorithms for enhanced biosurveillance. *J Am Med Inform Assoc.* 2008 Jan-Feb;15(1):77-86.

# Two steps

## ONE

Attempt to capture the expected behavior of the time series in the absence of an outbreak signal.

## TWO

Produce a test statistic based on the difference between the observed and expected data.

# Stages of Detection

*Table 1* ■ Decomposition of Several Existing Aberration Detection Algorithms from the Literature into Data Forecast and Anomaly Measure Stages

Algorithm	Data Forecasting	Anomaly Measure	Threshold
Fixed Data Threshold	Not applicable	Not applicable	Predetermined number of counts
C1 (EARS)	Moving average with 7-Day window	Z-Score	Three standard deviations above the mean (7-day)
C2 (EARS)	Moving average with 7-day window and 2-day guardband	Z-Score	Three standard deviations above the mean (7-day)
C3 (EARS)	Moving average with 7-day window and 2-day guardband	Sum of last three Z-Scores	Three standard deviations above the mean (7-day)
Reis [3]	Auto regressive moving average applied to trimmed seasonal model	7-day filter applied to residuals	Empirically derived based on desired sensitivity
GScan [5]	Spline fit to historic data	GLRT applied to sum of values in fixed-size moving window	Empirically derived based on simulation studies
Brillman [4]	Loglinear regression model with a fixed baseline	Page's test applied to the residuals in log space	Empirically derived

Murphy SP, Burkom H. Recombinant temporal aberration detection algorithms for enhanced biosurveillance. *J Am Med Inform Assoc.* 2008 Jan-Feb;15(1):77-86.



# Forecasting

- Simplest forecaster would simply use the previous day's value or the value from a week ago without further modification.
- The EARS family of algorithms (C1, C2, C3) all use a 7-day moving average to generate expected values.
- C1 uses the last 7 days of data
- C2 and C3 use a 2-day buffer data from 3 to 9 days in the past which ignores the current and two most recent days of data.
  - This guardband prevents outbreak effects from contaminating the predictions, due to presence of the early part of the outbreak signal in the baseline.

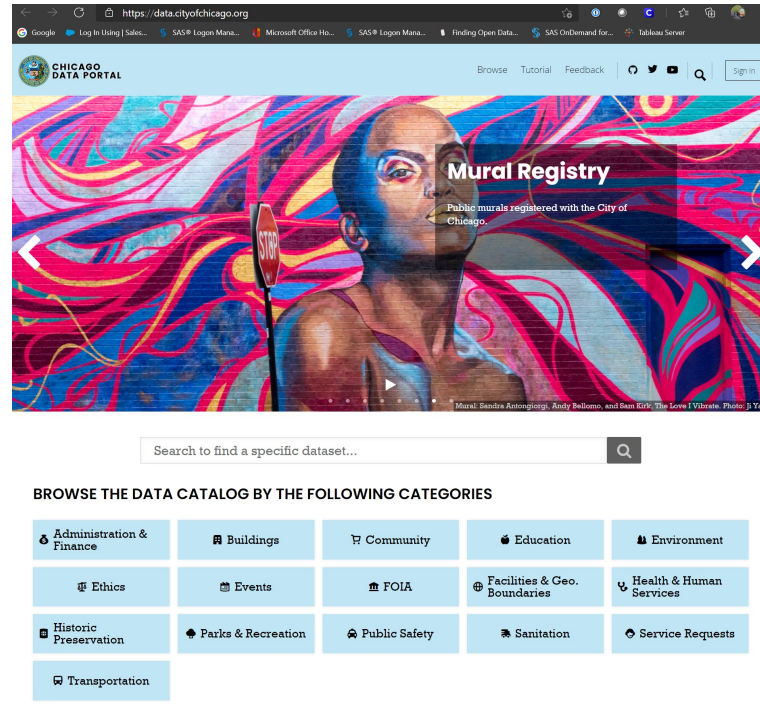
# Anomaly Detection

- Calculate a numeric value to quantify the positive **differences** between observation and prediction.
- Control **alert rates** for sensitivity to high values such overpredictions not tested for anomalies.
- Use calculated numeric value or **test statistic** to decide whether to investigate the cases causing the anomaly for evidence of a public health event.
- Calculate difference between observed and forecast, resulting in a time series of **residuals** that are then used to form a test statistic as the basis for an alerting decision.
- May **normalize** the residuals to account for natural variance in the baseline data.

Murphy SP, Burkom H. Recombinant temporal aberration detection algorithms for enhanced biosurveillance. J Am Med Inform Assoc. 2008 Jan-Feb;15(1):77-86.

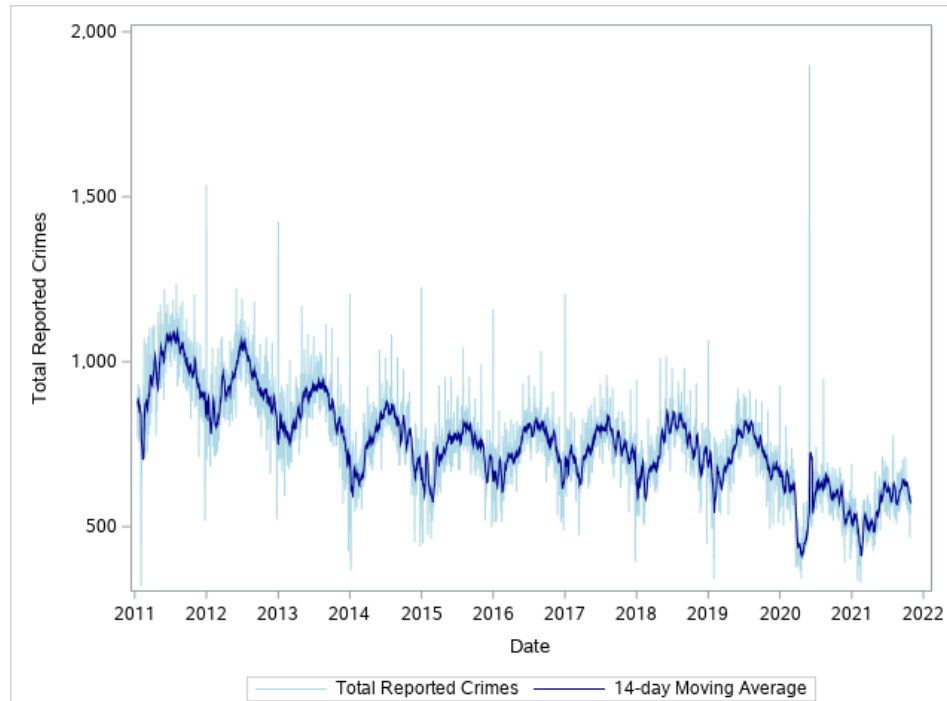
# Outbreak Example

# Chicago Open Data Portal



<https://data.cityofchicago.org/>

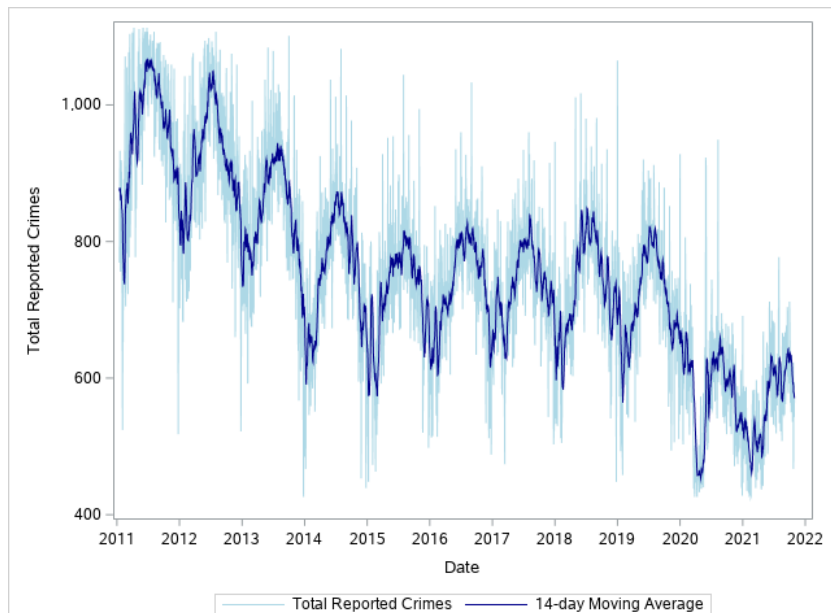
# Reported Crimes 2011 - Present



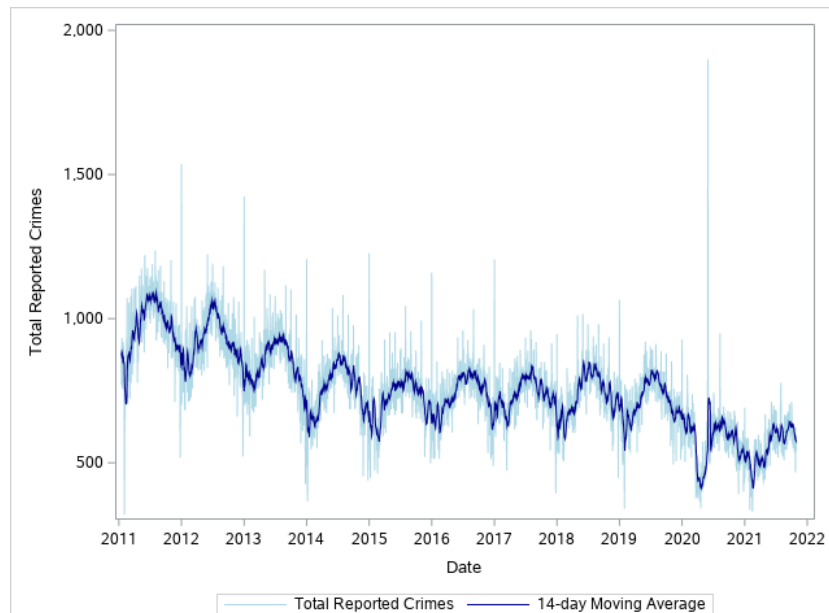
<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>

# Outliers

## Winsorized Observations – 1%

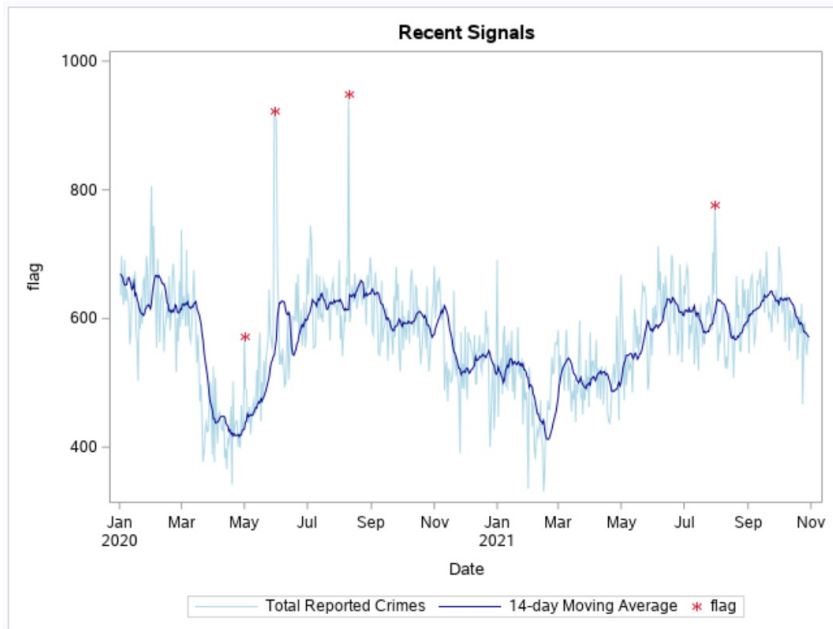


## All Data



<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>

# Recent Signals – All Crime



## Number of Flags

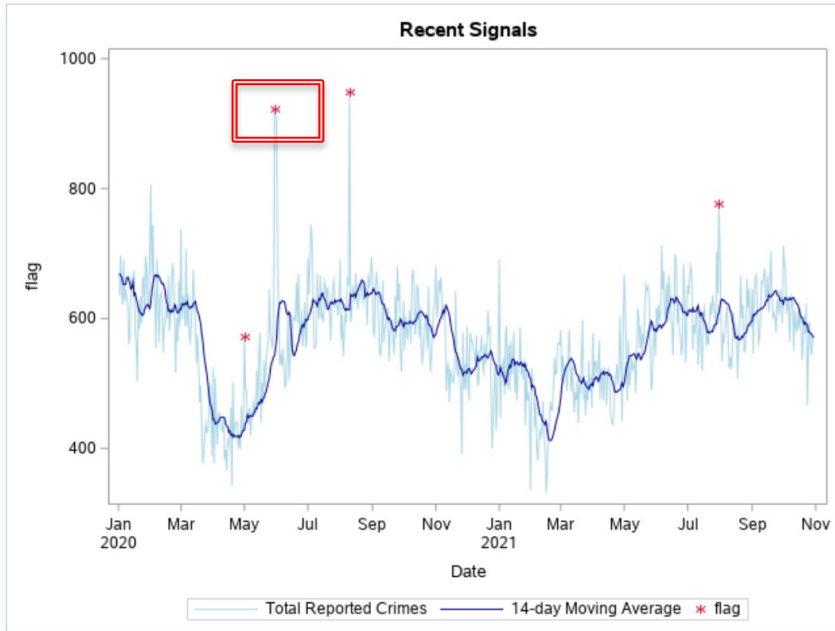
Total Intervals	Flags	Flags Per Hundred
3866	35	0.91

## Top 20 Recent Signals

Date	Total Reported Crimes	Upper CI
07/31/2021	777	748.18
08/10/2020	949	760.56
05/30/2020	923	738.59
05/01/2020	573	546.02
01/01/2020	928	894.83

<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>

# Recent Signals – All Crime



BREAKING NEWS NEWS

## Mayor imposes curfew after chaotic scenes unfold in Loop, Near North Side as protesters clash with police during demonstration over death of George Floyd in Minneapolis

By JEREMY GORNER, MADELINE BUCKLEY and PAIGE FRY  
CHICAGO TRIBUNE | MAY 31, 2020 AT 7:56 AM



Protests in Chicago in response to the killing of George Floyd

People in Chicago gathered this week to protest the killing of George Floyd by Minneapolis police officers. (Brian Cassella/Brian Cassella)

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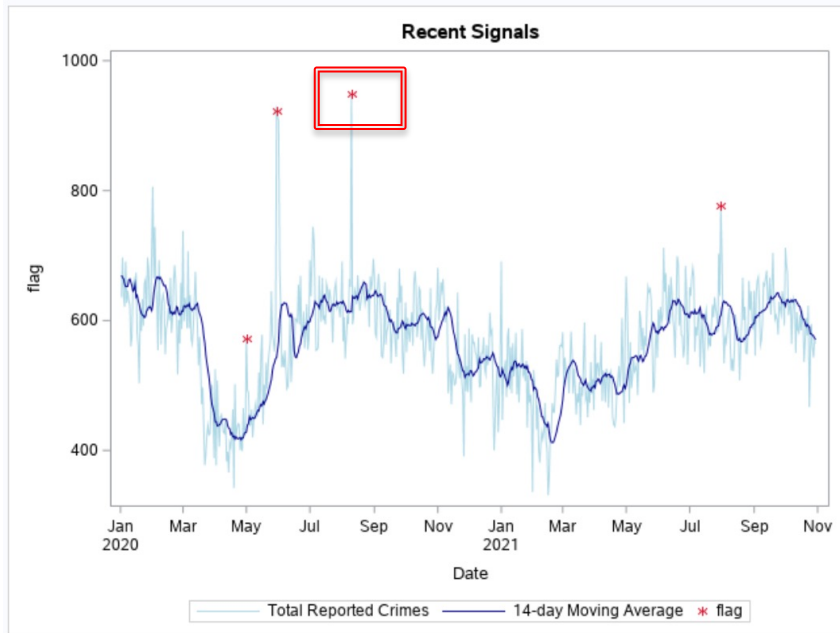
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<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>



# Recent Signals – All Crime



BREAKING NEWS NEWS

## Police shooting of Englewood man reignites political debate and looting as Mag Mile trashed, 13 cops injured, 2 people shot

By PAIGE FRY, JEREMY GORNER, PETER NICKEAS, GREGORY PRATT, MEGAN CREPEAU, STACY ST. CLAIR, CLAIRE HAO, WILLIAM LEE, DAN HINKEL, ANNIE SWEENEY, JOHN BYRNE and JAVONTE ANDERSON  
CHICAGO TRIBUNE | AUG 10, 2020 AT 8:55 PM



Damage at Zara  
Damaged windows at the Zara store following early morning looting in downtown Chicago on Aug. 10, 2020. (Antonio Perez / Chicago Tribune)

1 / 44

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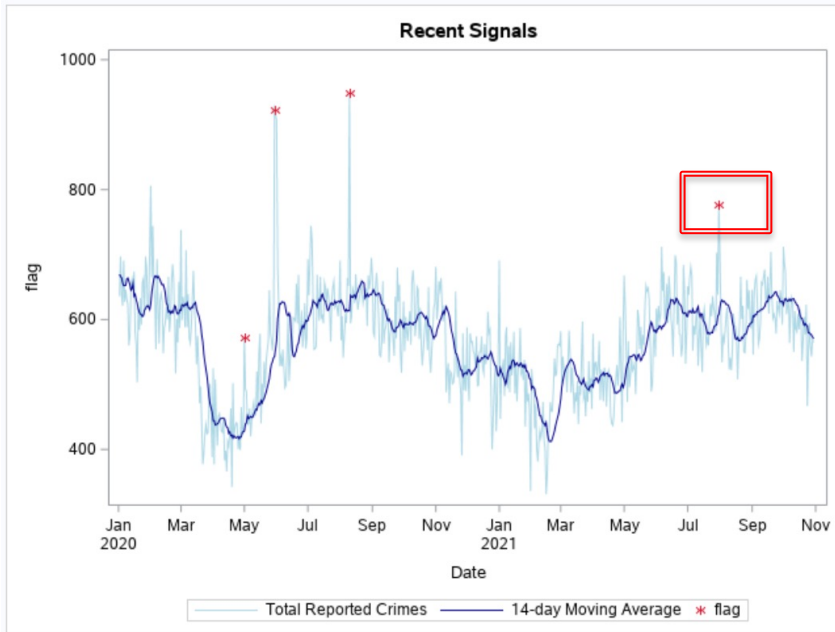
Google

Helping 654,000 Illinois businesses connect with customers

Learn more

<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>

# Recent Signals – All Crime



**THURSDAY** MILEY CYRUS • ILLENIUM • KAYTRANADA • PLAYBOI CARTI  
 BLACK PUMAS • STEVE AOKI • JIMMY EAT WORLD • SAINT JHN • ALL TIME LOW • TCHAMI • LUNAV  
 HIM PETRAS • CASH CASH • OLIVER TREE • ASHE • LP • OLIVIA O'BRIEN • ORVILLE PECK • FLO MILLI • DAVEYLOW • PARIS TEXAS • SAVVYNAMÉ • DOMBRESKY  
 ED MAVERICK • MAX • CLEVER • POST ANIMAL • JUSTIN JAY • ALMOST MONDAY • ANT CLEMONS • NIGHT LOVELL • ALV & AJ • BLOSSOM • SEBASTIAN PAUL  
 ABSOLFACTO • MOTHERFOLK • TAYLOR JANZEN • CHRISTIAN FRENCH • JULIAN LAMADRID • HOKO • MIGRANT MOTEL • TRISTAN SIMONE

**FRIDAY** TYLER, THE CREATOR • MARSHMELLO • RODDY RICCH • JACK HARLOW  
 POLO G • SUICIDEBOYS • LAUV • GIVEON • JAUZ • SUBTRONICS • OMAR APOLLO • ARIZONA ZERVAS • JACOB BANKS  
 WHITE REAPER • AG CLUB • BOY PABLO • ELOHIM • MICK JENKINS • GRANDSON • CAM • TOBI LOU • DABIN • RIOT TEN • WOOLI • EMOTIONAL ORANGES  
 BLACK PISTOL FIRE • GUS DAPPERTON • JAWNY • TAI VERDES • ROOKIE • LEVEL UP • MOB RICH • NJOMZA • RENFORSHORT • CONTRADASH  
 MOTHICA • KENNY MASON • OSTON • JAKE WESLEY ROGERS • SA-ROC • ELEPHANT HEART • CHOMPPA • BROWNIES & LEMONADE • ALL STARS

**SATURDAY** POST MALONE • JOURNEY • MEGAN THEE STALLION • LIMP BIZKIT  
 TRIPPIE REDD • SLANDER • YOUNG THE GIANT • FREDDIE GIBBS • ANGELS & AIRWAVES • MT. JOY • IANN DIOR  
 MARC REBILLET • WHITNEY • OLIVER HELDENS • TATE MCRAE • CAUTIOUS CLAY • BIA • TREVOR DANIEL • TIGHT • CAUETOWN • DRAMA • LOST KINGS  
 MONOPHONICS • PORCHES • JESSIA • CANNONS • VINTAGE CULTURE • CID • THE BACKSEAT LOVERS • MICHIGANDER • CLOVE • JOY OLADOKUN  
 VNSSA • LAUNDRY DAY • SOPHIE CATES • RENEE • KID QUILL • PHEM • SERENA ISIOMA • JAC ROSS • DANTE ELEPHANTE • JAKE DUBV

**SUNDAY** FOO FIGHTERS • YOUNG THUG • BROCKHAMPTON • MODEST MUSE  
 G HERBO • ALISON WONDERLAND • BRITTANY HOWARD • BAND OF HORSES • DERMOT HEARNEDY  
 SURFACES • THE FRONT BOTTOMS • BIG WILD • YELLOW CLAW • EARTHGANG • RICO NASTY • JPEGMAFIA • PEEKABOO • PRINCESS NOKIA  
 FLIPP DINERO • MIXMOTOON • TOOSII • NOGA EREZ • SULLIVAN KING • RMR • PEACH TREE RASCALS • ELDERBROOK • JXON • DR. FRESCH • SIR CHLOE  
 WILLIAM BLACK • LP GIOBBI • CHILID • NEZ • ANT SAUNDERS • RIZ LA VIE • SOFIA VALDES • SARAH BARRIOS  
 NEAL FRANCIS • MOORE KISMET • SHY CARTER • OTTO • THE AQUADOLLS • RADKEY

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<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>

# CCVI vs Domestic Crime

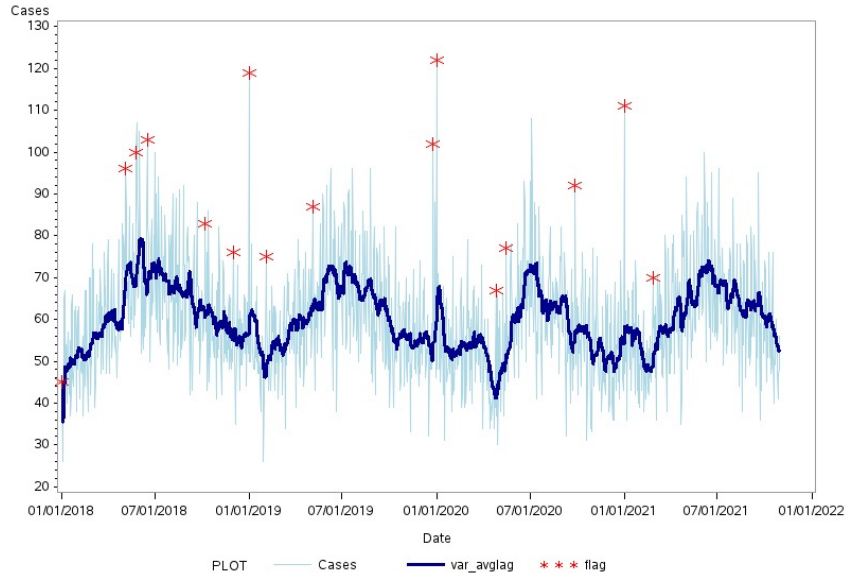
Domestic Crime	Frequency	High CCVI	Pct High CCVI	Medium CCVI	Pct Medium CCVI	Low CCVI	Pct Low CCVI
false	2,508,091	1,031,268	41.12%	636,132	25.36%	840,691	33.52%
true	476,328	244,876	51.41%	151,909	31.89%	79,543	16.70%

<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>

# High CCVI Domestic Crime

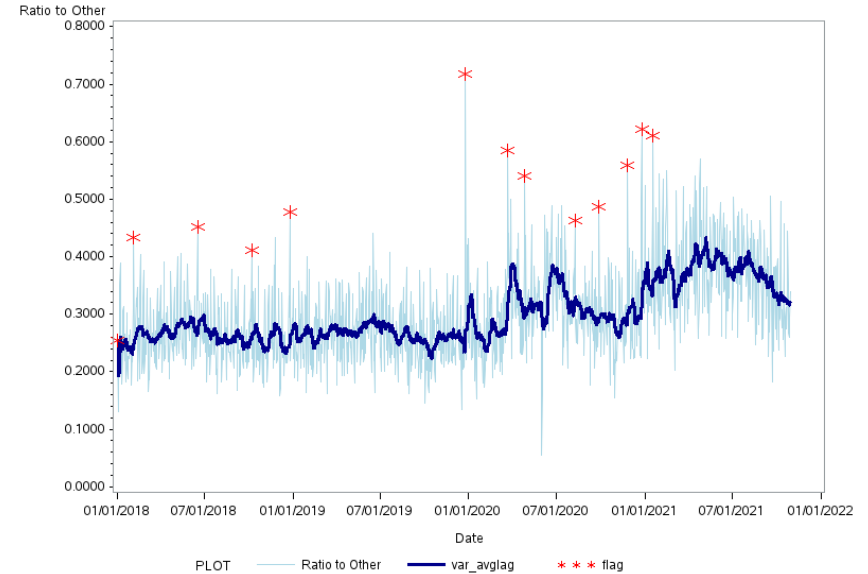
14-lag Moving Average (case\_sum) from 01/2018 to 10/2021

Source: chi\_crime\_daily



14-lag Moving Average (ratio\_to\_other) from 01/2018 to 10/2021

Source: chi\_crime\_daily



<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>

# High CCVI Domestic Crime

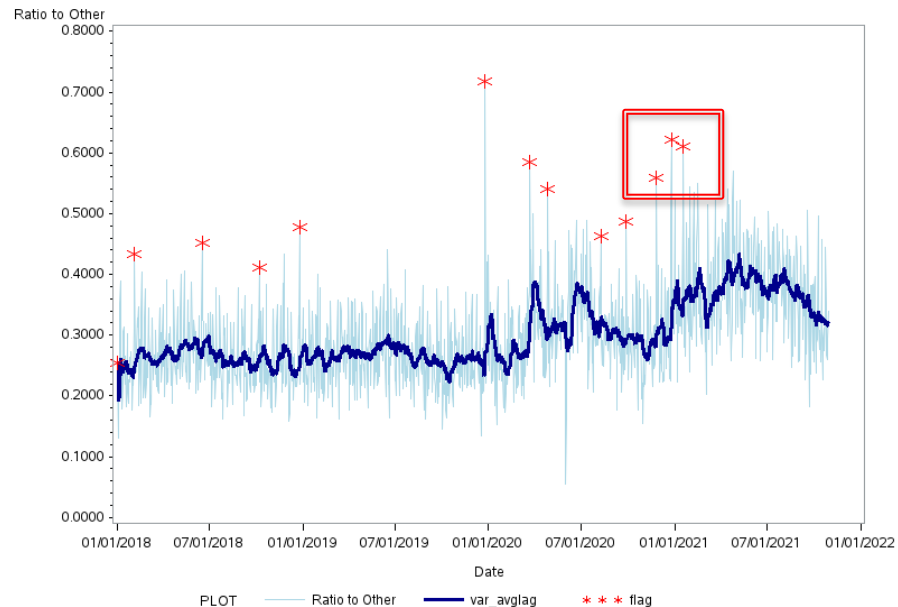
Total Intervals	Flags	Flags Per Hundred
1398	13	0.93

## Top 20 Recent Signals

Date	Ratio to Other	Upper CI	Cases
01/18/2021	0.6106	0.54	69
12/25/2020	0.6218	0.60	74
11/26/2020	0.5583	0.41	67
09/27/2020	0.4868	0.48	92
08/09/2020	0.4633	0.45	82
04/26/2020	0.5403	0.52	67
03/22/2020	0.5856	0.49	65
12/25/2019	0.7183	0.41	102
12/25/2018	0.4779	0.37	65
10/07/2018	0.4109	0.38	83
06/17/2018	0.4518	0.41	103
02/04/2018	0.4326	0.38	61

14-lag Moving Average (ratio\_to\_other) from 01/2018 to 10/2021

Source: chi\_crime\_daily



<https://data.cityofchicago.org/Public-Safety/Crimes-2001-to-Present/ijzp-q8t2>

# Ask Questions

*[nicholas.soulakis@northwestern.edu](mailto:nicholas.soulakis@northwestern.edu)*