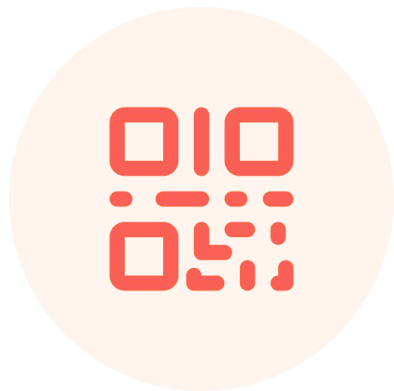




3790542

slido



Join at [slido.com](https://slido.com)  
#3790542

① Click **Present with Slido** or install our [Chrome extension](#) to display joining instructions for participants while presenting.



Human Contexts and Ethics Guest Lecture

# Fairness in Housing Appraisal

Contextualizing the Cook County Assessor's Office Open Data Initiative

*Ari Edmundson*

**Data 100, Fall 2023 @ UC Berkeley**

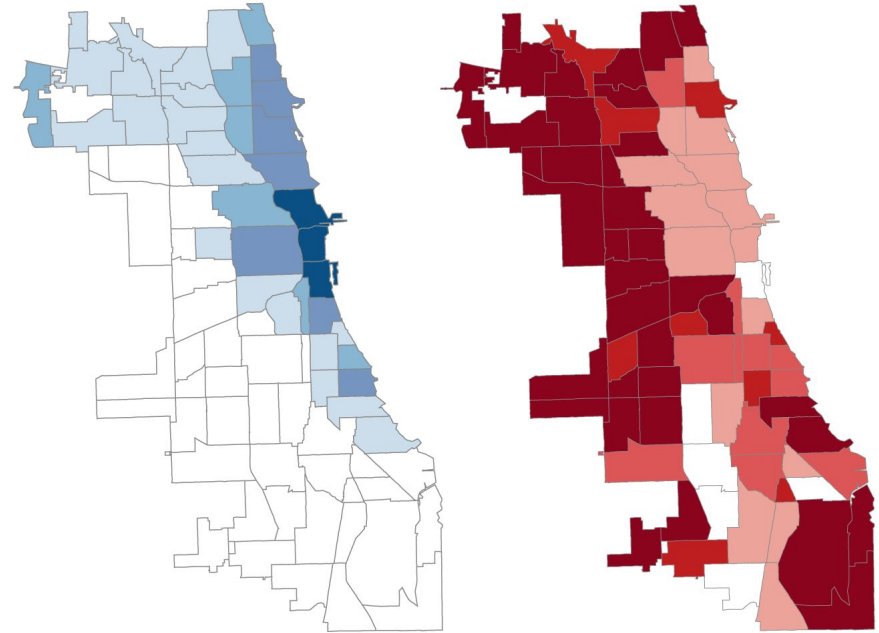


# Case Study: Cook County Assessor's Office and Property Appraisal



Cook County Assessor's Office (CCAO) -  
Chicago, IL and surrounding townships

Charged with assessing property values in order  
to determine property taxes



**COOK COUNTY**  
**GOVERNMENT**



3790542

The Problem  
The Response  
Key Takeaways  
Lessons for Data Science Practice



3790542

# The Problem



Pause

TRIBUNE WATCHDOG: THE TAX DIVIDE

# AN UNFAIR BURDEN

Cook County failed to value homes accurately for years. The result: a property tax system that harmed the poor and helped the rich.

Melrose Park (Terrence Antonio James /  
Chicago Tribune)

By [Jason Grotto](#)

Chicago Tribune

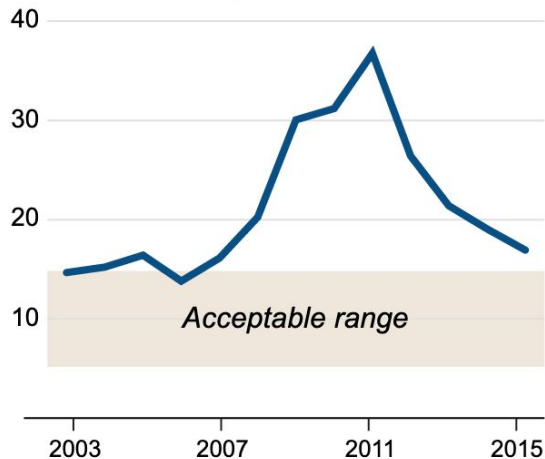
PUBLISHED: JUNE 10, 2017

[The Chicago Tribune](#), June 10, 2017

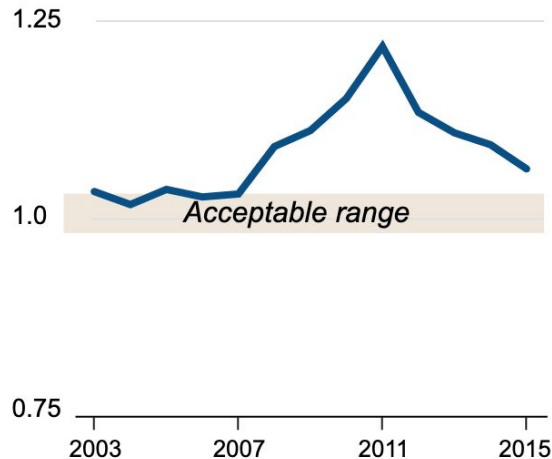


## Standards of accuracy, fairness not met

### Coefficient of dispersion



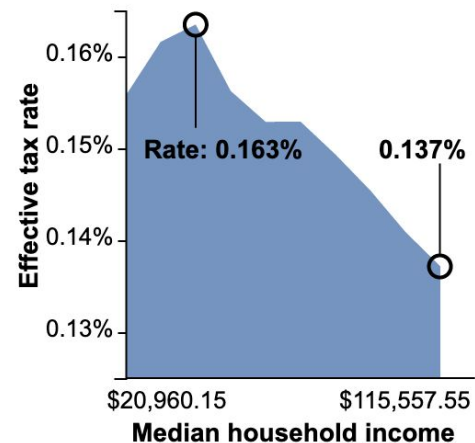
### Price-related differential



Sources: Cook County assessor's office, Illinois Department of Revenue, Tribune analysis

## As income level rises, effective tax rates decline

These rates represent the percentage of homes' value that owners pay toward two taxing districts that cover all of Cook County. They are a small fraction of the overall tax rate but allow for comparisons between communities with widely differing tax bases.



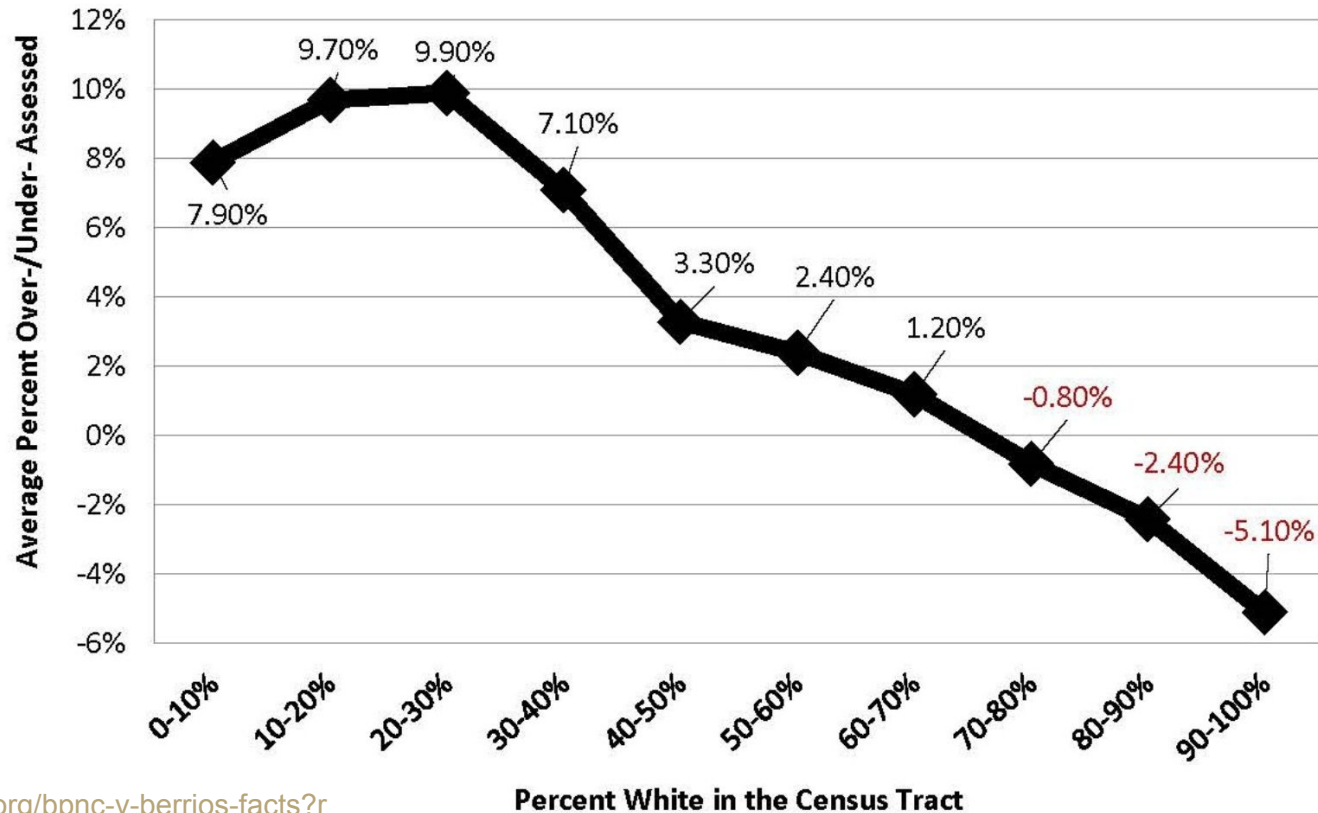
Sources: Cook County assessor's office, Cook County treasurer's office, Tribune analysis

[The Chicago Tribune](#), June 10, 2017





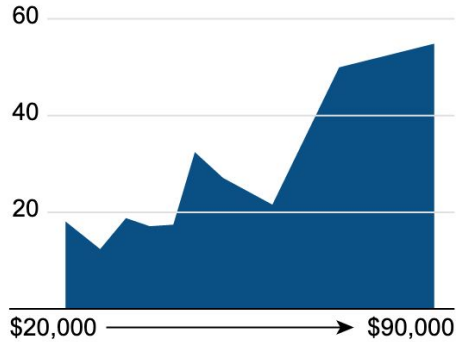
## Chart 2: Average Percent Over/Under Assessment by Percent White, 2011-2015



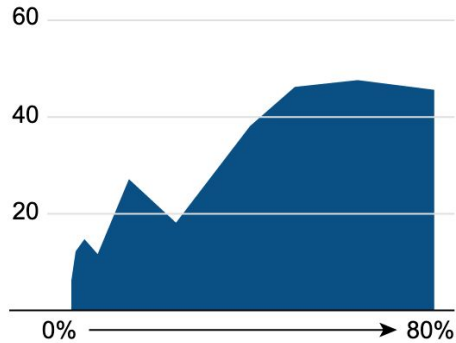


## Neighborhood demographics affect appeals rate

### By median income in 2015

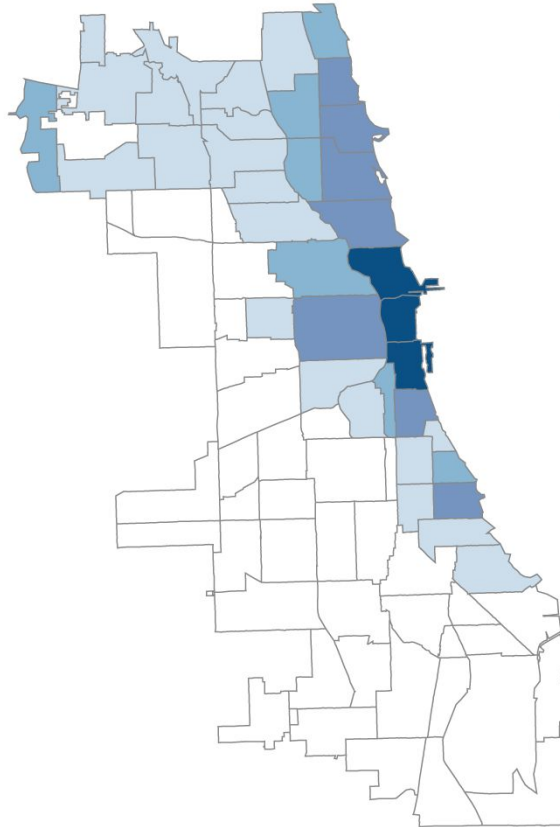
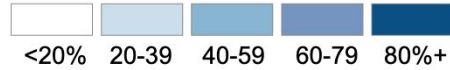


### By percentage white in 2015



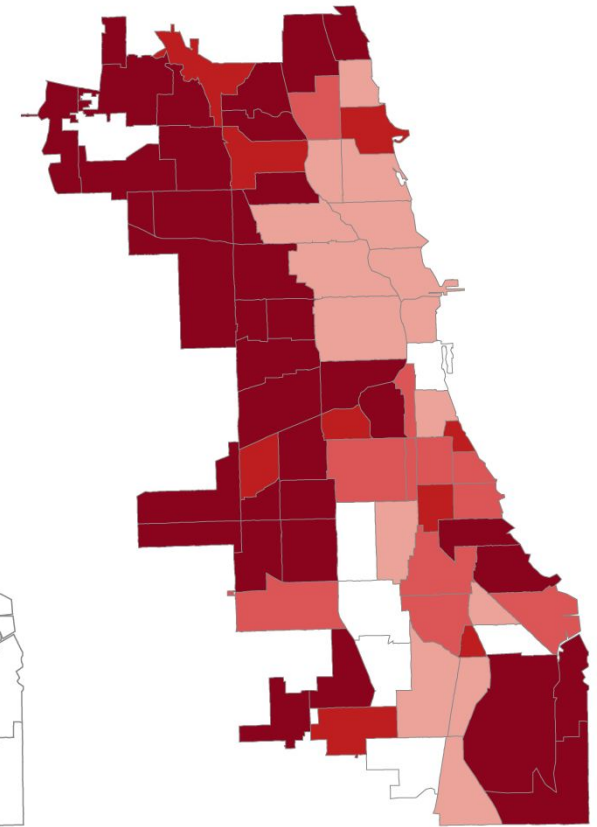
Sources: Cook County assessor's office, U.S. Census Bureau, Tribune analysis

## Percentage of homeowners who appealed, 2011-15



Sources: Cook County assessor's office, Illinois Department of Revenue, Tribune analysis

## Number of years homes were overvalued, 2011-15



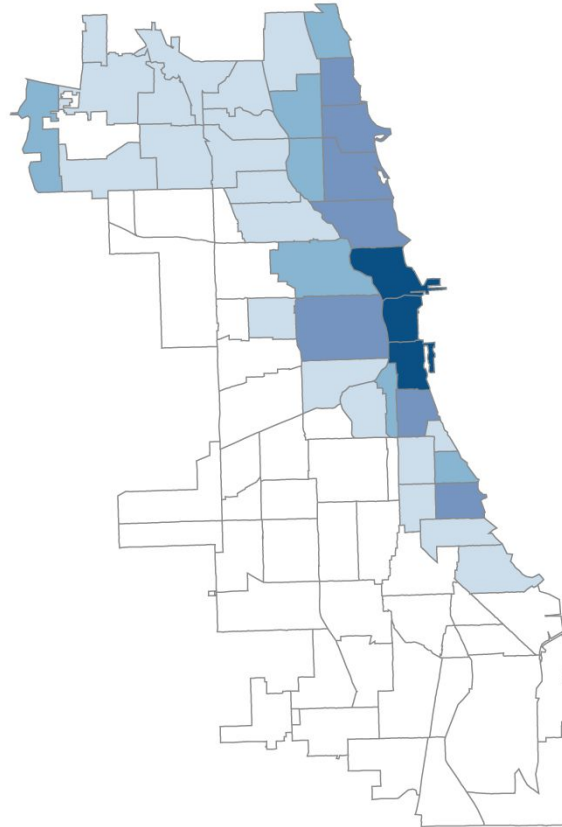
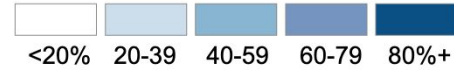
# The role of the appeals process in producing inequity.

“Appeals are a good thing,” Thomas Jaconetty, deputy assessor for valuation and appeals, said in an interview. “The goal here is fairness. We made the numbers. We can change them.”

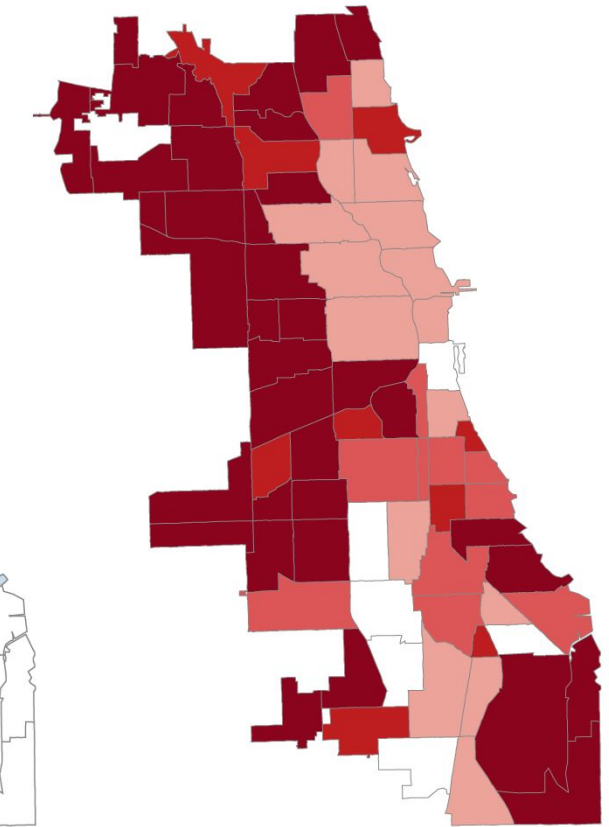
Fairness as equal access: “anyone can appeal” - but that’s not really the case:

Part of a deeper, institutional pattern, potential corruption

**Percentage of homeowners who appealed, 2011-15**



**Number of years homes were overvalued, 2011-15**





# Human impacts



In 2011 Braxton-Williams learned the assessor's office had valued the house at \$147,550. "I love my house, but I know it's not worth that much," she said. "And they know it's not worth that much."  
(Terrence Antonio James/Chicago Tribune)

[The Chicago Tribune](#), June 10, 2017

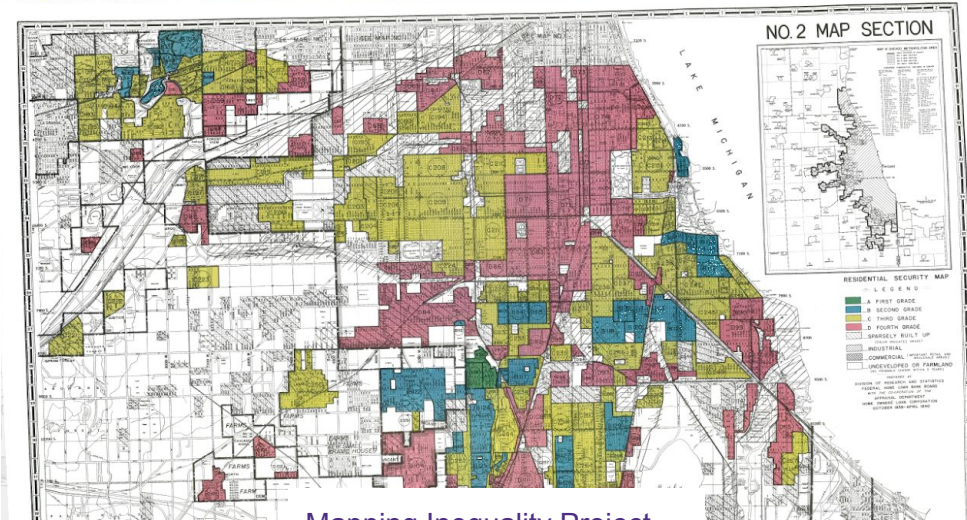
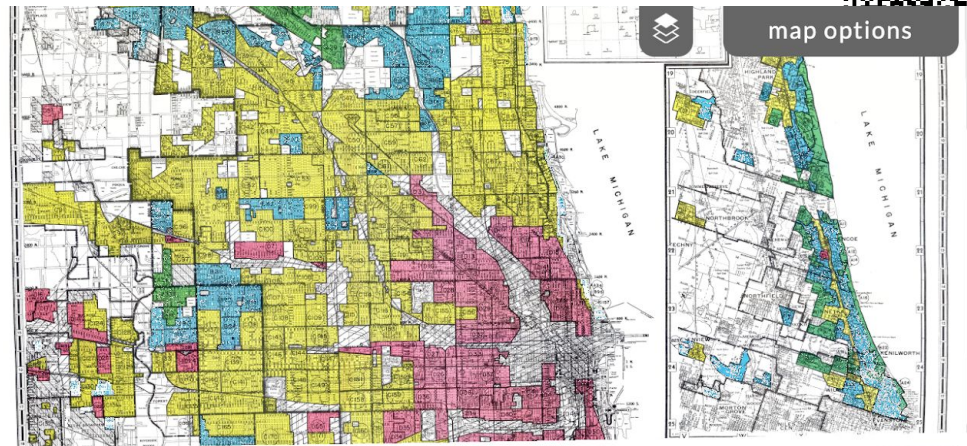
# Real estate and racial inequality in the United States

Housing has been a key motor of racial inequality in modern US History

Segregation and credit-market racism

**Redlining:** making it difficult or impossible to get a federally-backed mortgage to buy a house in specific neighborhoods coded as “risky” (red).

What made them “risky” according to the makers of these maps? Their racial composition...





# Real estate and racial inequality in the United States

Segregation was not only a result of federal policy, but developed by real estate professionals

Real estate industry “professionalized” in the 1920’s and 1930’s by aspiring to become a science guided by strict methods and principles.

These methods centered on creating objective rating systems (information technologies) for the appraisal of property values...

which encoded race as a factor of valuation and which, in turn, influenced federal policy and practice

ADDITIONS		<i>Per Cent</i>
Site.....		15
Type of neighborhood and social factors.....		20
View and climate.....		15
Public utilities and schools.....		5
Streets and alleys; distance to work in city.....		10
Contour and soil.....		5
Physical environment.....		20
Restrictions and planning.....		10

A table of common deductions reducing the above percentages of value follows:

DEDUCTIONS		<i>Per Cent</i>
Noise and dirt, up to.....		25
<u>Racial and foreign neighbors, up to.....</u>		<u>60</u>
Adjacent vacancy, up to.....		20
Poor architecture, up to.....		20
Obsolescence, up to.....		70
Distances from city, work, schools, etc., up to.....		100
Nuisances (funerals, freight, trucks, etc.), up to.....		100
Dead-end streets, up to.....		15

Table of common deductions from a 1937 Appraising Manual (image from Colin Koopman, *How We Became Our Data* (2019) p. 137)





3790542

# The Response

CCAO's mandate under new Assessor, Fritz Kaegi

- Distributional equity in property taxation = properties of same value treated alike during assessments
- Creates new Office of Data Science,

# Why the Cook County Assessor's Office made its residential assessment code and data public — voluntarily

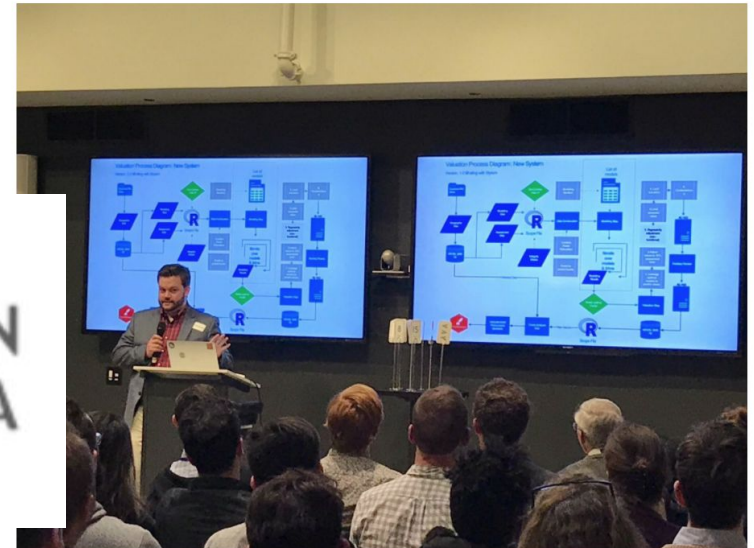
 Cook County Assessor Apr 17, 2019 · 4 min read



By Robert Ross  
 Chief Data Officer, Cook County Assessor's Office



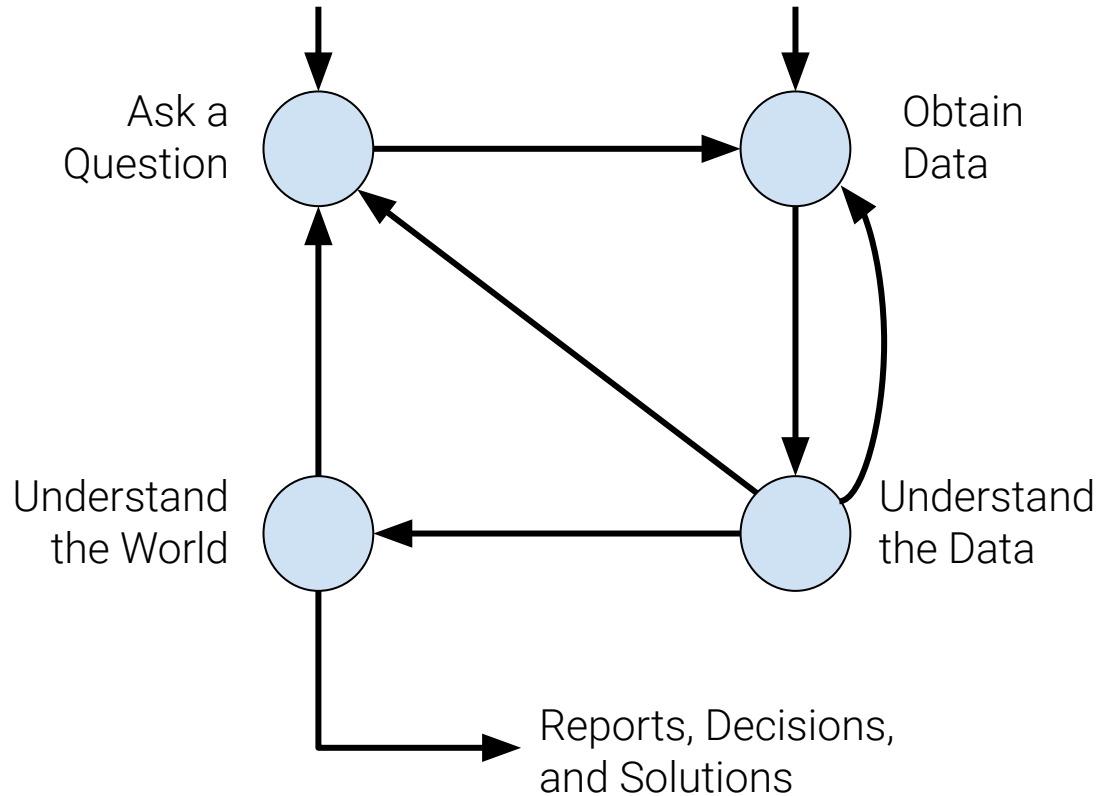
**COOK COUNTY GOVERNMENT** | **OPEN DATA**



Rob Ross, Chief Data Officer with the Cook County Assessor's Office discusses the residential assessment modeling used by the CCAO during a presentation at Chi Hack Night on April 16th, 2019.



# Data science lifecycle



The data science lifecycle is a **high-level description** of the data science workflow.

Note the two distinct entry points!





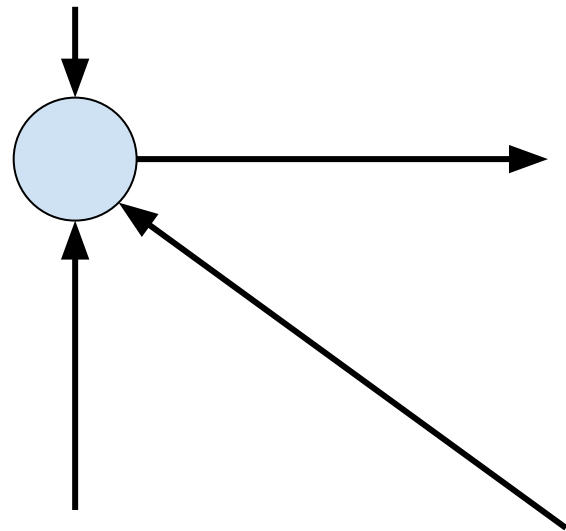
# 1. Question/Problem Formulation

What problems are we trying to solve?

1. **Accurately, uniformly, and impartially** assess the value of a home
  - a. → accurately predict the sale price of a home within the next year
2. Create a “robust pipeline” that assesses at **scale**, and and is **fair** to all people, across perceived racial and income differences

What are our metrics for success?

Ask a Question

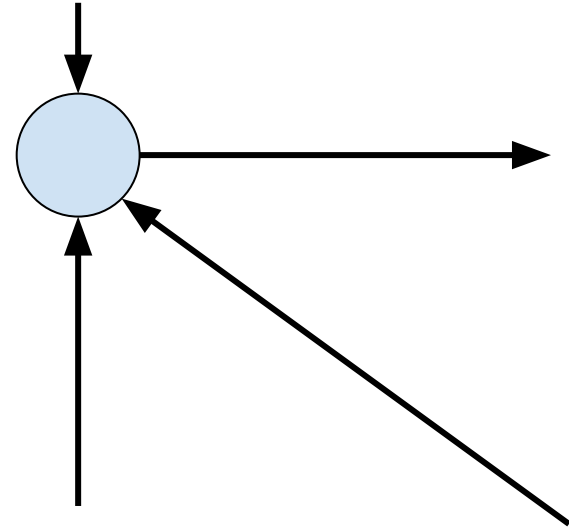




# 1. Question/Problem Formulation

- a. What is an assessment of a home's value?
- b. What makes one assessment more accurate than another?
- c. What makes one batch of assessments more accurate than another batch?

Ask a Question





# 1. Question/Problem Formulation

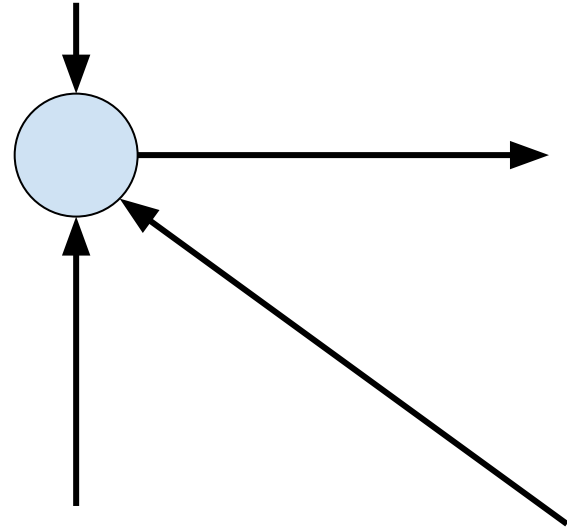
## a. What is an assessment?

An estimate of the **value** of a home

What is the value of a home? What determines it? How do we know?

One can imagine many different responses...

Ask a Question





# 1. Question/Problem Formulation

## a. What is an assessment?

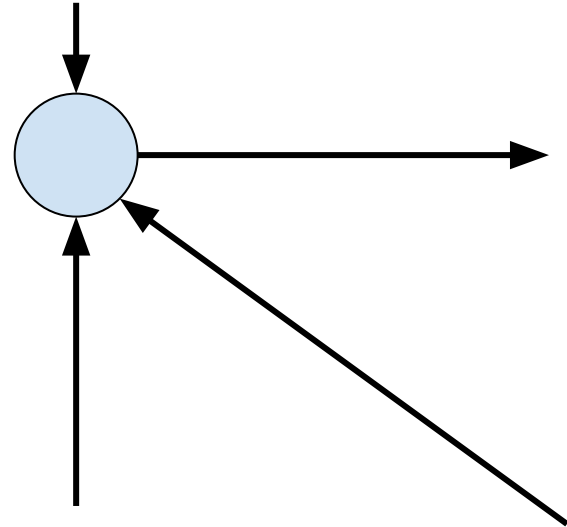
An estimate of the **value** of a home

What is the value of a home? What determines it? How do we know?

Here: its **market value**

The “*true value*” of a home is only determined only at the moment of its sale (at least as far as property tax assessors are concerned)

Ask a Question





# 1. Question/Problem Formulation

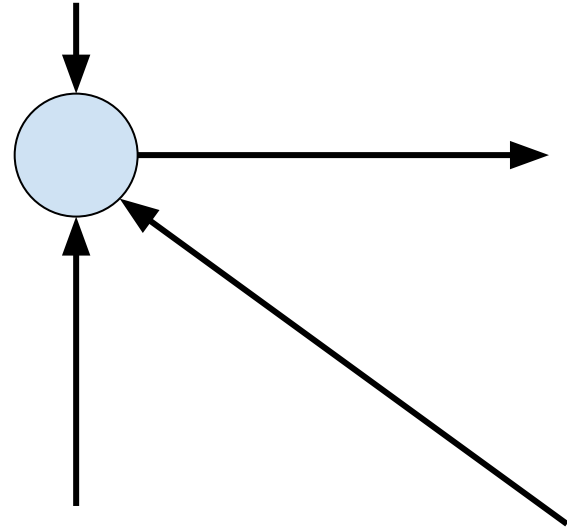
## a. What is an assessment?

Obviously the assessor can't make everyone sell their house every year.

So how do they generate a reliable estimate?

Make predictions

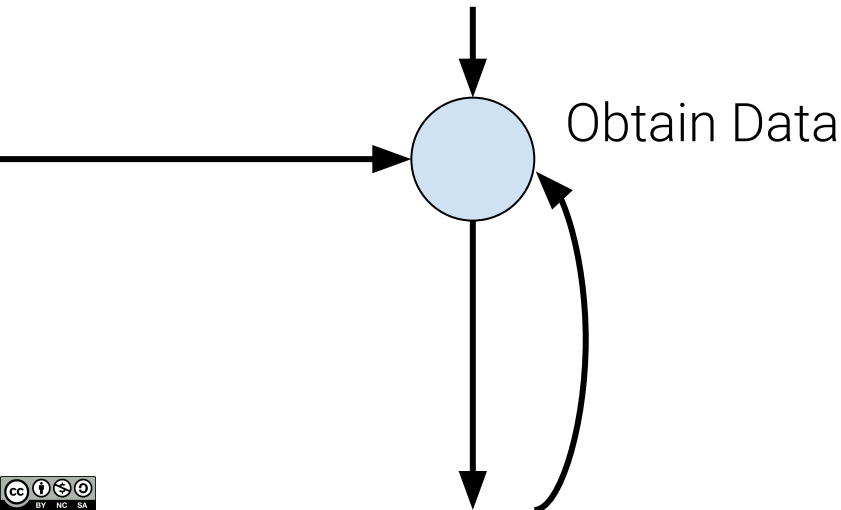
Ask a Question





## 2. Data Acquisition and Cleaning

- What data do we have and what data do we need?
- How will we sample more data?
- Is our data representative of the population we want to study?



# What's in the data?

1. Sales data
  - a. All recorded sales data 2013-2019
2. Property characteristics
  - a. Property Identification Number
  - b. Physical characteristics (Age, Bedroom, Baths, Square feet, Neighborhood, Site Desirability, etc.)

How was this data collected? When? By whom? For what purposes? How and why were particular categories created?

Feature Name	Category	Type
Age	Characteristic	numeric
Central Air Conditioning	Characteristic	categorical
Apartments	Characteristic	categorical
Attic Finish	Characteristic	categorical
Attic Type	Characteristic	categorical
Bedrooms	Characteristic	numeric
Building Square Feet	Characteristic	numeric
Basement	Characteristic	categorical
Basement Finish	Characteristic	categorical
Wall Material	Characteristic	categorical
Full Baths	Characteristic	numeric
Fireplaces	Characteristic	numeric
Garage 1 Area	Characteristic	categorical
Garage 1 Attached	Characteristic	categorical
Garage 1 Material	Characteristic	categorical

# What's in the data?

Are these attributes differentially reported?

How are “improvements” (i.e. renovations) tracked and updated?

How might these attributes be differentially reported?

Which data is missing, and for which neighborhoods or populations is it missing? And how do you know?

What other data sources might be valuable?

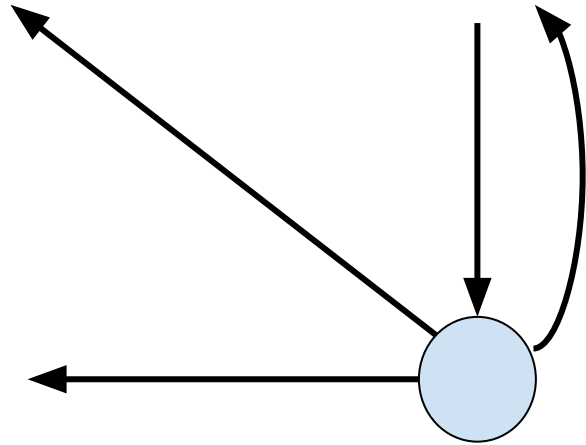
Creating new attributes (flood plains, airport flight path)

Feature Name	Category	Type
Age	Characteristic	numeric
Central Air Conditioning	Characteristic	categorical
Apartments	Characteristic	categorical
Attic Finish	Characteristic	categorical
Attic Type	Characteristic	categorical
Bedrooms	Characteristic	numeric
Building Square Feet	Characteristic	numeric
Basement	Characteristic	categorical
Basement Finish	Characteristic	categorical
Wall Material	Characteristic	categorical
Full Baths	Characteristic	numeric
Fireplaces	Characteristic	numeric
Garage 1 Area	Characteristic	categorical
Garage 1 Attached	Characteristic	categorical
Garage 1 Material	Characteristic	categorical





# What's in the Data?



Understand the Data

Which attributes are most predictive of sales price?

Is the data uniformly distributed? Do all neighborhoods have up to date data? Same granularity? Or do some neighborhoods have missing or outdated data?

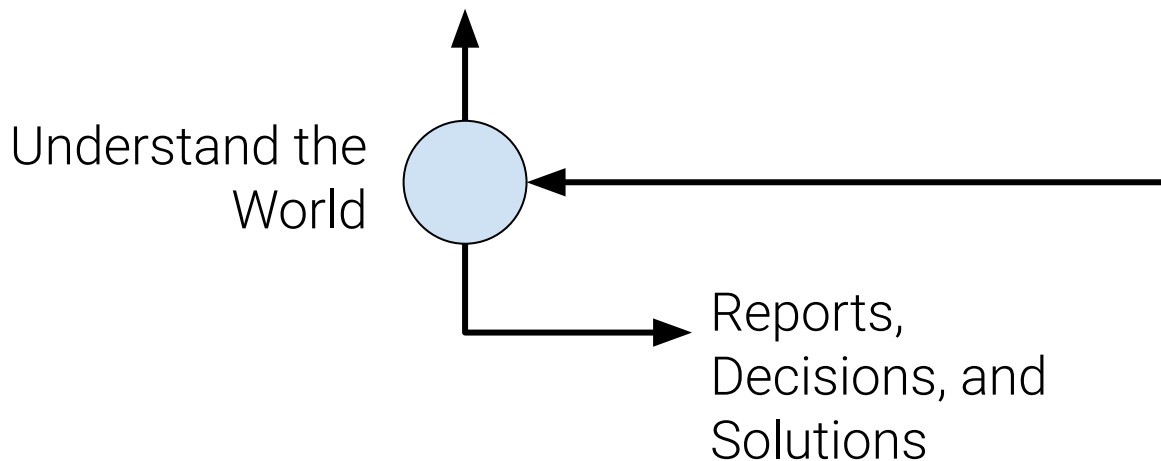
CCAO noticed that low income neighborhoods had disproportionately spottier data

Need to develop new data collection practices--including finding new sources of data



## 4. Prediction and Inference

- What does the data say about the world?
- Does it answer our questions or accurately solve the problem?
- How robust are our conclusions and can we trust the predictions?





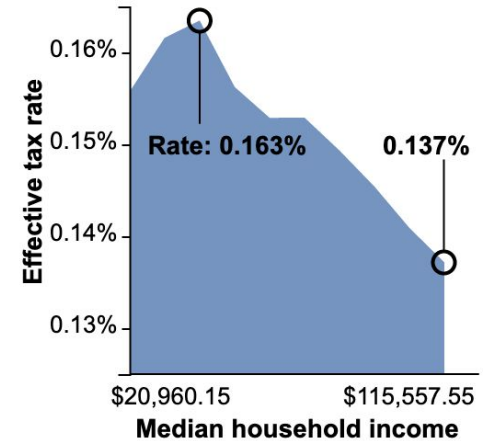
# Predicting housing prices at scale (“mass appraisal”)

Predict sale price (“fair market value”) of unsold properties by discovering patterns in data sets containing known sale prices and characteristics of similar and nearby properties.

How do I know if my assessment is accurate? What counts as getting close? Who decides?

## As income level rises, effective tax rates decline

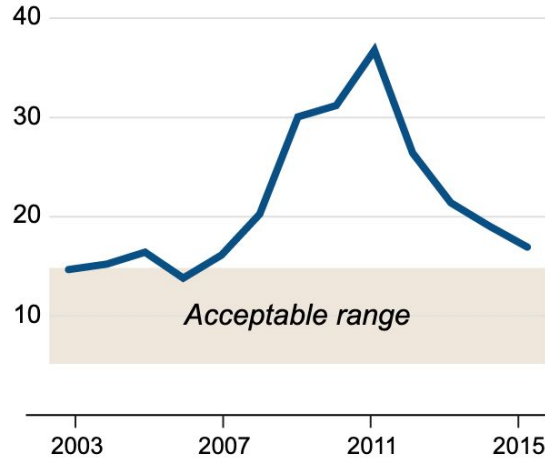
These rates represent the percentage of homes' value that owners pay toward two taxing districts that cover all of Cook County. They are a small fraction of the overall tax rate but allow for comparisons between communities with widely differing tax bases.



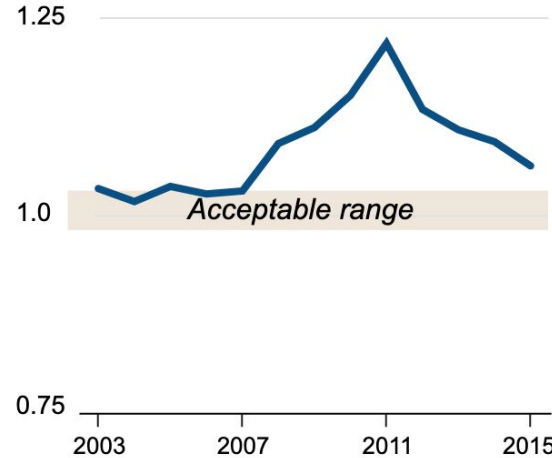
Sources: Cook County assessor's office, Cook County treasurer's office, Tribune analysis

## Standards of accuracy, fairness not met

### Coefficient of dispersion



### Price-related differential



Sources: Cook County assessor's office, Illinois Department of Revenue, Tribune analysis

[The Chicago Tribune](#), June 10, 2017

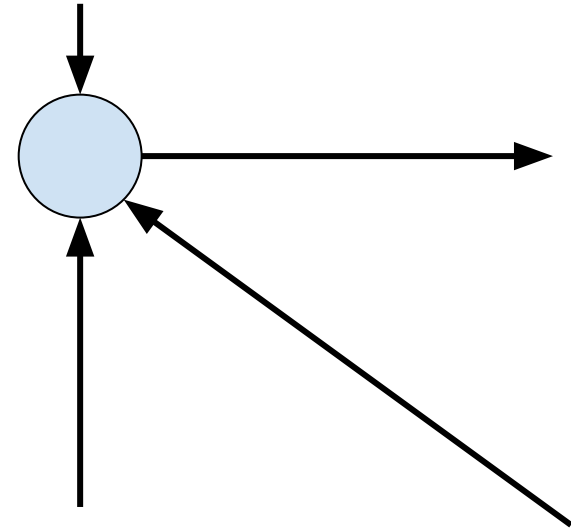


# 1. Question/Problem Formulation

- a. What is an assessment?
- b. what makes one assessment more accurate than another?
- c. what makes one batch of assessments more accurate than another batch?

What is the value of a home? Its relational. Determined by a **sociotechnical** performance, depends upon **trust**.

Ask a Question





# Social trust and translating “accuracy” into fairness

Why should any individual believe that the model is accurate for *their* property?

The CCAO counts on its performance of “**transparency**” (putting data, models, pipeline onto GitLab) to foster public **trust**, which would help it equate the production of “accurate assessments” with “fairness”

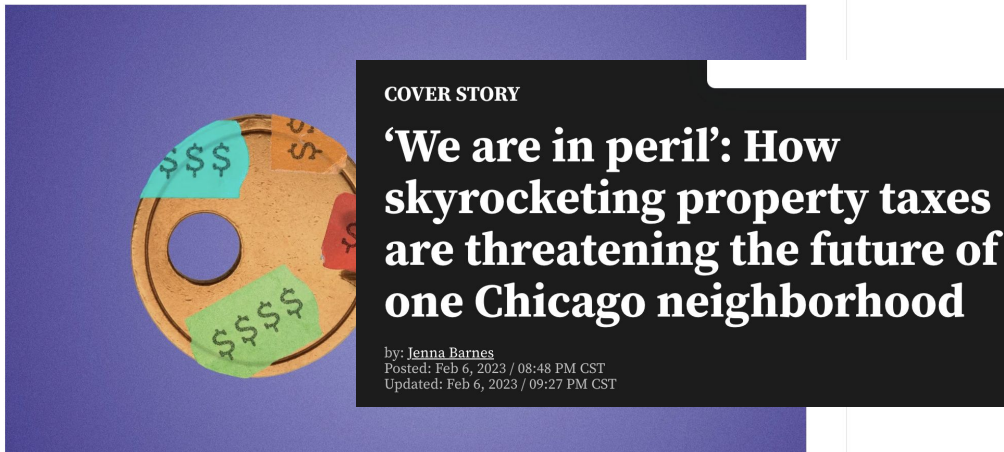
## Transparency

**Defined as:** the ability of the data science department to share and explain pipeline results and decisions to both internal and external stakeholders

# The headache of appealing Cook County property tax assessments



Monica Eng

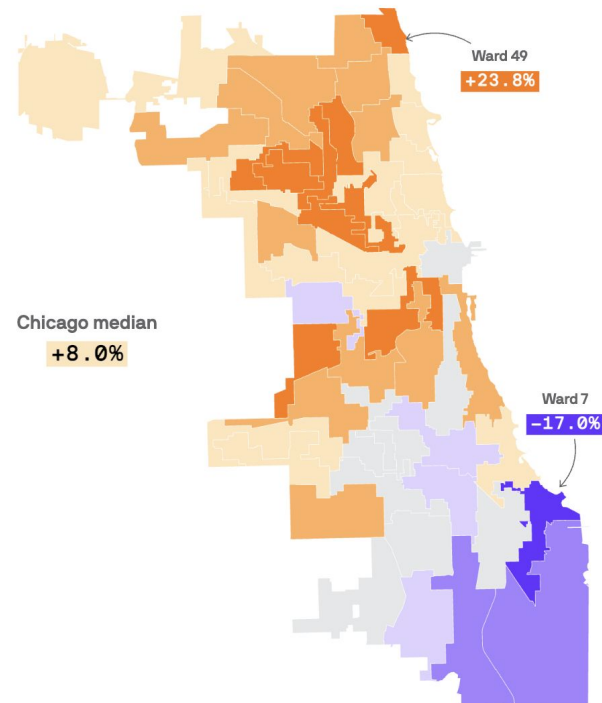
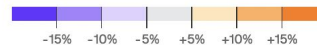


<https://www.axios.com/local/chicago/2023/05/22/cook-county-property-tax-appeal-process>

<https://wgntv.com/news/cover-story/tax-pilsen/>

## Change in total property taxes, by ward

2020-2021



Source: Cook County Treasurer; Map: Erin Davis/Axios Visuals

<https://www.axios.com/local/chicago/2022/12/01/why-chicagos-property-tax-bills-so-high>





3790542

# Key Takeaways





1. Accuracy is a necessary, but not sufficient, condition of a fair system.



2. Fairness and transparency are context-dependent and sociotechnical concepts



3. Learn to work with contexts, and consider how your data analysis will reshape them



## 4. Keep in mind the power, and limits, of data analysis



*What is a home  
"worth"?*

In 2011 Braxton-Williams learned the assessor's office had valued the house at \$147,550. "I love my house, but I know it's not worth that much," she said. "And they know it's not worth that much."  
(Terrence Antonio James/Chicago Tribune)

[The Chicago Tribune](#), June 10, 2017



3790542

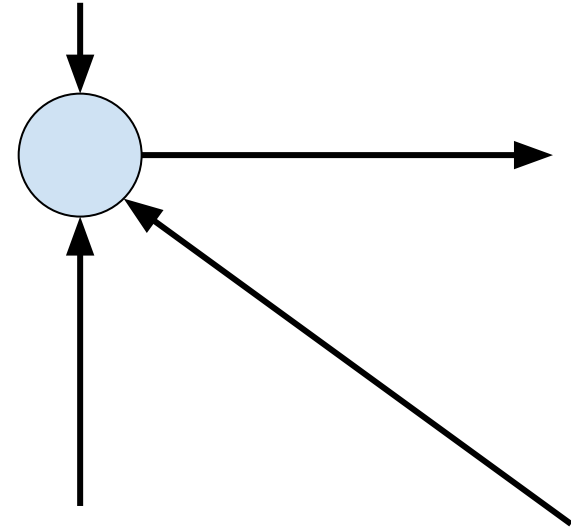
# Lessons for Data Science Practice



# 1. Question/Problem Formulation

- What do we want to know?
- What problems are we trying to solve?
- What are the hypotheses we want to test?
- What are our metrics for success?

Ask a Question

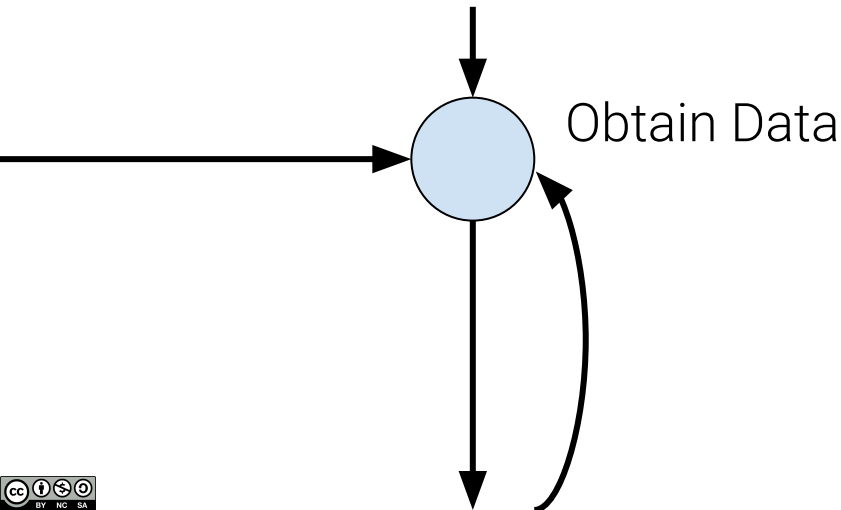


- Who is responsible for framing the problem?
- Who are the stakeholders? How are they involved in the problem framing?
- What do you bring to the table? How does your positionality affect your understanding of the problem?
- What are the narratives that you're tapping into?



## 2. Data Acquisition and Cleaning

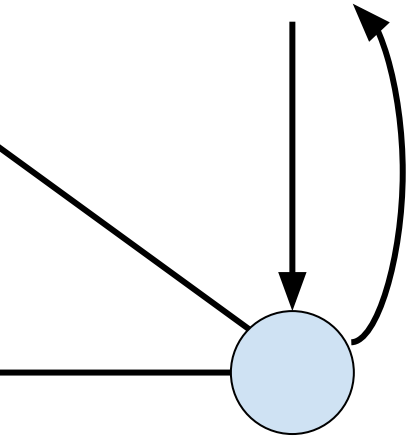
- What data do we have and what data do we need?
- How will we sample more data?
- Is our data representative of the population we want to study?



- Where does the data come from?
  - Who collected it? For what purpose?
- What kinds of collecting and recording systems and techniques were used?
- How has this data been used in the past?
- What restrictions are there on access to the data? What enables you to have access?



### 3. Exploratory Data Analysis & Visualization



- How is our data organized and what does it contain?
- Do we already have relevant data?
- What are the biases, anomalies, or other issues with the data?
- How do we transform the data to enable effective analysis?

#### Understand the Data

- What kind of personal or group identities have become salient in this data?
- Which variables became salient, and what kinds of relationship obtain between them?
- Do any of the relationships made visible lend themselves to arguments that might be potentially harmful to a particular community?

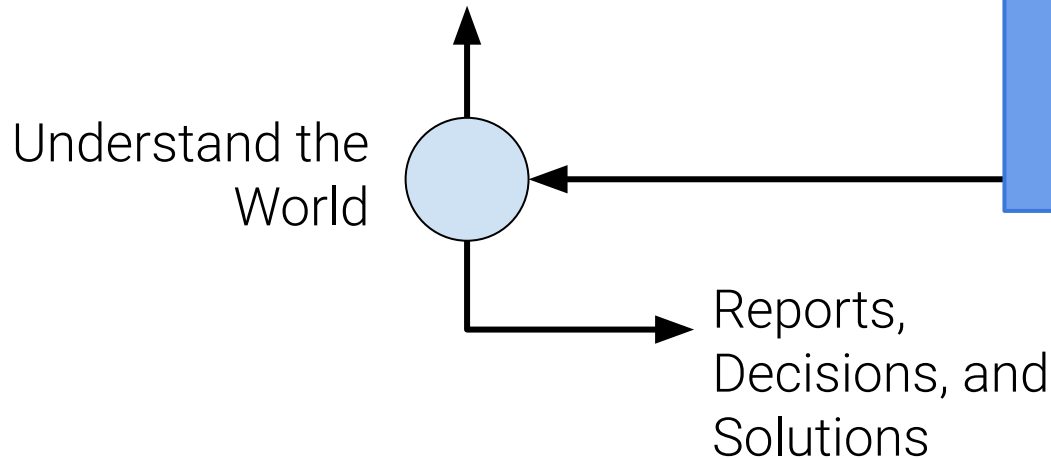




## 4. Prediction and Inference

- What does the data say about the world?
- Does it answer our questions or accurately solve the problem?
- How robust are our conclusions and can we trust the predictions?

- What does the prediction or inference *do* in the world?
- Are the results useful for the intended purposes?
- Are there benchmarks to compare the results?
- How are your predictions and inferences dependent upon the larger system in which your model works?





slido



# Write down one take-away from today's lecture

- ① Click **Present with Slido** or install our [Chrome extension](#) to activate this poll while presenting.