Challenges to collaboration and reproducibility

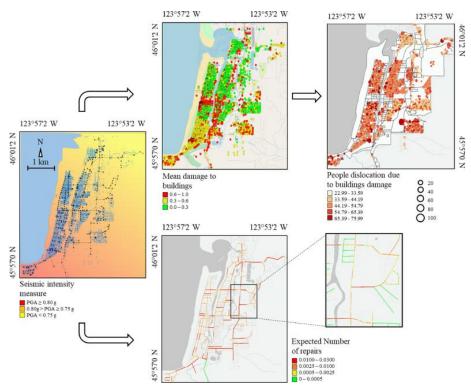
Nathanael Rosenheim, PhD

URSC 689 Feb 1, 2021

Challenges Collaborating Across Universities

Community resilience model that combines work completed by Urban Planning researchers at TAMU and Civil Engineers at the University of Illinois -Urbana-Champaign.

All data shared via email. Models were run independently without shared code or version control.



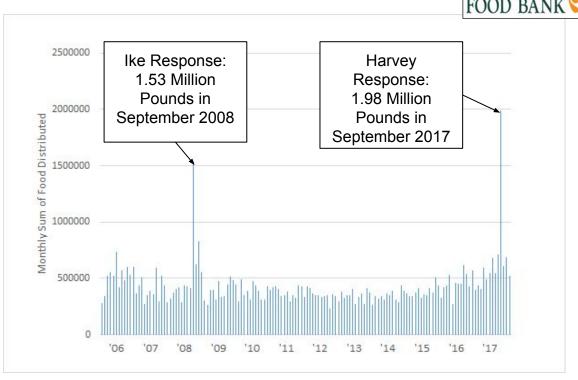
Challenges Collaborating with Communities



Models of post disaster food aid distributions.

Community partner had a massive SQL database, but only way to access information was with a point-and-click user interface.

Took graduate student 40 hours to download data.



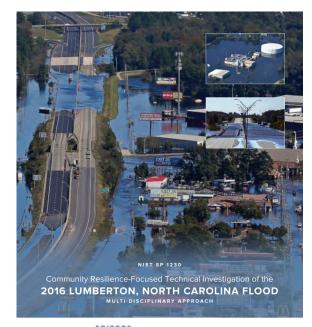
Source Data: Southeast Texas Food Bank Primarius Reports

Challenges Collaborating with Federal Agencies

Post disaster field studies that combine engineering damage assessments with social science household and business level surveys.

Multi-year collaboration between 11 universities through the National Institute for Standards and Technology (NIST).

Diverse range of data collection, required Institutional Review Board (IRB) approval. Requires storage of confidential data with limited means of sharing and citing data.





Validity Mitrani-Reiser

This publication is available free of charge from: https://doi.org/10.6028/NIST.SP.1230



Challenges Collaborating with yourself

The most important collaborator is yourself...

Make it easy for your future self to find files, programs, data, and to continue the work you started.

Provenance: URSC689\Rosenheim_ChallangestoReproducibility_2021_02_01

What is data?

Information obtained by scientific work and used for analysis.¹

- Tabular Data
 - Survey responses
 - Administrative Data
- Metadata -Codebooks
- Relational Databases

	storeid	Q2_1
1	12	2. Manager
2	16	5. Other
3	41	5. Other
4	71	
5	104	5. Other
6	123	2. Manager
7	125	5. Other
8	153	2. Manager
9	154	***
10	165	1. Owner
11	186	5. Other
12	202	5. Other
13	239	
14	279	2. Manager
15	319	2. Manager
16	323	3. Owner and Manager
17	342	2. Manager
18	370	
19	386	2. Manager
20	406	2. Manager
21	448	1. Owner
22	460	
23	474	2. Manager

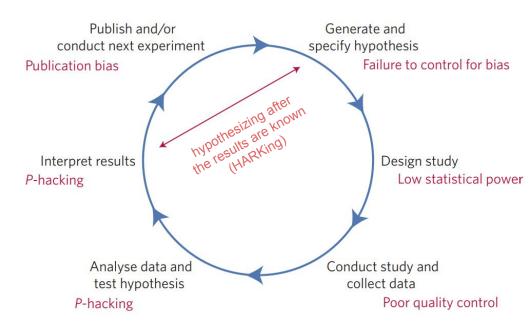
```
type: numeric (int)
                range: [12,3605]
                                                    units: 1
        unique values: 135
                                                 missing .: 0/135
                 mean: 1716.04
             std. dev:
          percentiles:
                                                                    3207
storeid:
 1. [SETX Survey Text] Store ID
 2. Primary Key - unique ID randomly assigned when the sample frame was set.
 3. Use STOREID to merge Coverpage and Response Datasets.
 4. [Citation] Rosenheim, N. et al 2018. Southeast Texas Food Retail Survey.
 5. [Name of Saved Data File]
      RAPID17_1gv1_SNAP_SETX_RetailSurvey_2019-02-01/RAPID17_1gv1_SNAP_SETX_Reta
     > ilSurvey_2019-02-01.dta
 6. [Program to replicate Data File]
      RAPID17 1gv1 SNAP SETX RetailSurvey 2019-02-01.do
 7. [Date data file was created] 1 Feb 2019 16:48:00
02 1
                 type: numeric (byte)
                label: 02 11bl r
                range: [1.5]
                                                    units: 1
        unique values: 5
                                                 missing .: 31/135
           tabulation: Freq.
                                Numeric Label
                                      1 1. Owner
                                      3 3. Owner and Manager
                                      4 4. Assistant Manager
                                      5 5. Other
                                      . Missing
 1. [SETX Survey Text] 1. What is your role with this business? - Selected
      Choice
 2. [Citation] Rosenheim, N. et al 2018. Southeast Texas Food Retail Survey.
 3. [Name of Saved Data File]
      RAPID17 1gv1 SNAP SETX RetailSurvey 2019-02-01/RAPID17 1gv1 SNAP SETX Reta
      > ilSurvey 2019-02-01.dta
 4. [Program to replicate Data File]
      RAPID17_1gv1_SNAP_SETX_RetailSurvey_2019-02-01.do
 5. [Date data file was created] 1 Feb 2019 16:48:00
```

Reference: 1. "data, n.". OED Online. December 2018. Oxford University Press. http://www.oed.com.lib-ezproxy.tamu.edu:2048/view/Entry/296948?rskey=c3az3E&result=1 (accessed February 08, 2019).

What is science?

"Science is an approach to knowledge... that strives to better approximate the state of nature by reducing errors in inferences."

"Conceptualize science is a toolbox of... tools designed to minimize mistakes [or bias]."

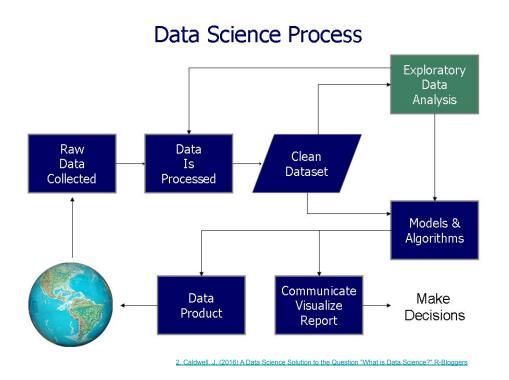


Scientific method with potential threats of bias.²

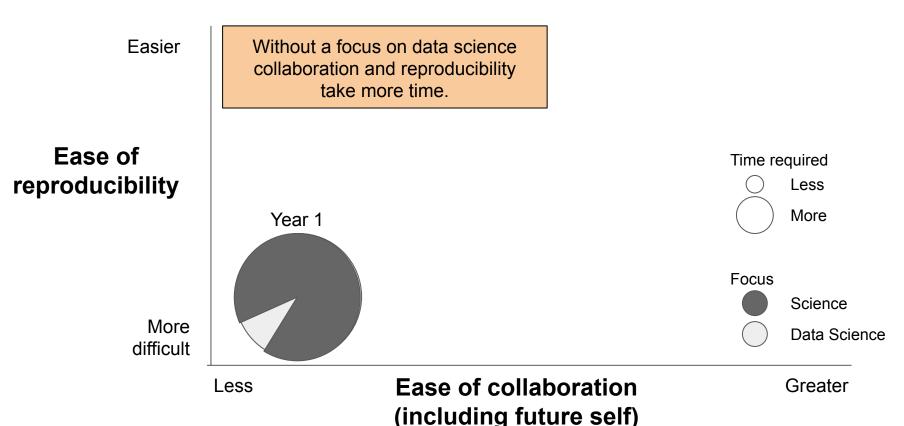
What is Data Science?

Data science is a set of tools designed to minimize bias associated with the analysis of data. "The discipline of turning raw data into understanding."1

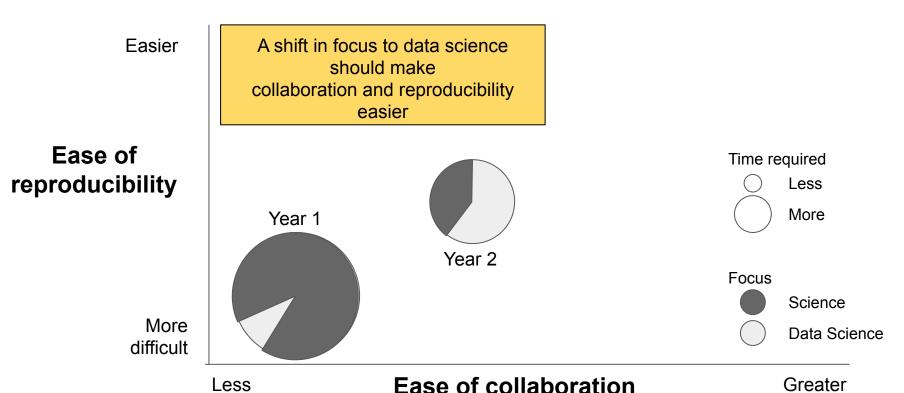
Example Tools/Concepts: Version Control, GitHub, Markdown [RMarkdown or Jupyter Notebook], Workflow, Repositories, Permanent Identifiers e.g. "handle" (hdl) or "digital object identifier" (doi)



Goal: Better science in less time



Goal: Better science in less time



(including future self)

Goal: Better science in less time

Less

Easier Over time the focus on data science shifts to maintenance, as incremental improvements in Year 4 data science tools and techniques become more standard practice. Ease of Time required Year 3 reproducibility Less More Year 1 Year 2 Focus Science More **Data Science** difficult

Ease of collaboration

(including future self)

Greater

Replication Standard - Individual or Social Contract?

Individual Responsibility	Social Contract
If asked a researcher should be able to provide the files to replicate published results.	Files to replicate published results are submitted at time of publication.
Emphasis on trust.	Emphasis on transparency.
Faith in the author.	Focus on openness.
Reinforcement of status.	Distributes power and access.

Course objectives...

Help to overcome challenges to data sharing, documentation, publication, and analytics.

Help to develop a culture that supports a social contract for data replication.

To be a part of a community that bolsters data science and leads to more open, discoverable, reproducible research.

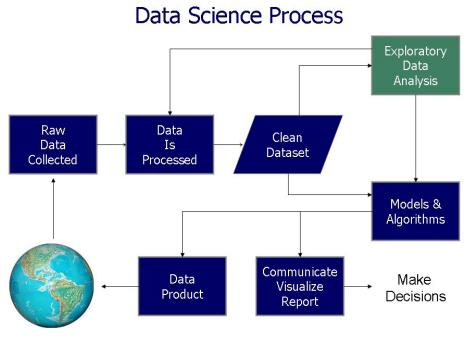
Rosenheim, Nathanael; Day, Wayne; Seong, Kijin (2021) "Automated Neighborhood Characteristics for Community Resilience Planning." DesignSafe-CI. https://doi.org/10.17603/ds2-hj0p-bp40.

Roy, Malini; Rosenheim, Nathanael (2021) "Longitudinal Social Vulnerability Data Exploration for Harris County Census Tracts." DesignSafe-CI. https://doi.org/10.17603/ds2-hn6r-dh03.

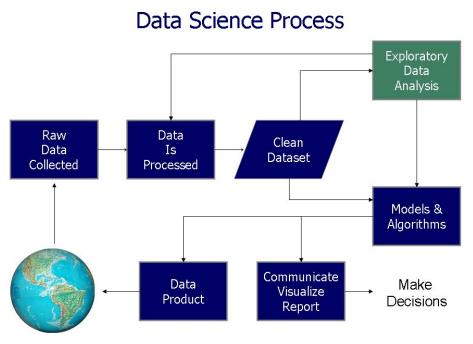
Gu, Donghwan, and Rosenheim, Nathanael. Demographic Analysis Workflow using Census API in Jupyter Notebook: 1990-2000 Population Size and Change. Ann Arbor, MI: Inter-university Consortium for Political and Social Research [distributor], 2020-07-30. https://doi.org/10.3886/E120381V1

Goodman, Cooper, Rosenheim, Nathanael, Day, Wayne, Gu, Donghwan, and Korukonda, Jayasaree. Population Distribution Workflow using Census API in Jupyter Notebook: Dynamic Map of Census Tracts in Boone County, KY, 2000. Ann Arbor, MI: Inter-university Consortium for Political and Social Research [distributor], 2020-07-31. https://doi.org/10.3886/E120382V1

Rosenheim, Nathanael; Day, Wayne; Seong, Kijin (2021) "Automated Neighborhood Characteristics for Community Resilience Planning." DesignSafe-CI. https://doi.org/10.17603/ds2-hj0p-bp 40.



Motivation: Demonstrate the potential of Census API and Jupyter Notebooks for obtaining, cleaning, and exploring Census data.



2. Caldwell. J. (2016) A Data Science Solution to the Question "What is Data Science?" R-Bloggers

Rosenheim, Nathanael; Day, Wayne; Seong, Kijin (2021) "Automated Neighborhood Characteristics for Community Resilience Planning." DesignSafe-CI. https://doi.org/10.17603/ds2-hj0p-bp40.

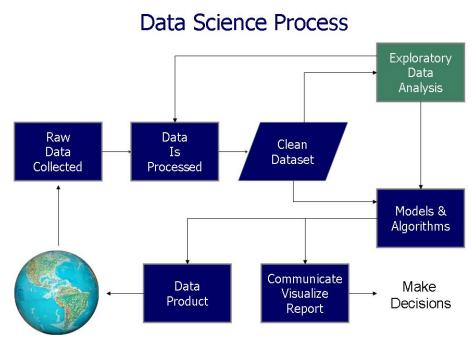
Raw data collection: 2010

Decennial Census data using

Census API and TIGER Shapefiles
for Block Group Boundaries

Program automates data collection:

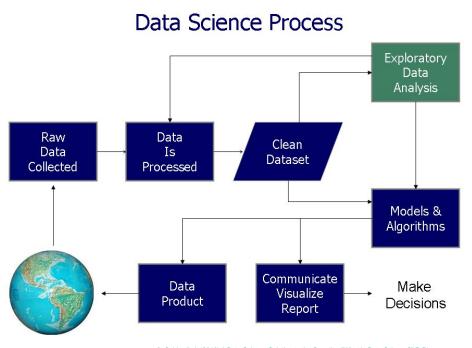
IN-CORE_CENSUSAPI_DWNLD_JOPLIN MSA_2021-01-13.ipynb



Data is Processed:

Jupyter Notebook

IN-CORE_CENSUSAPI_DWNLD_JOPLIN MSA_2021-01-13.ipynb

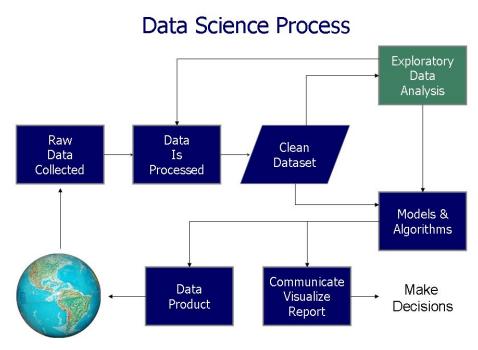


Clean Dataset:

New variables are created and block group data is merged with shapefiles.

Program generates CSV, HTML, and Shp files

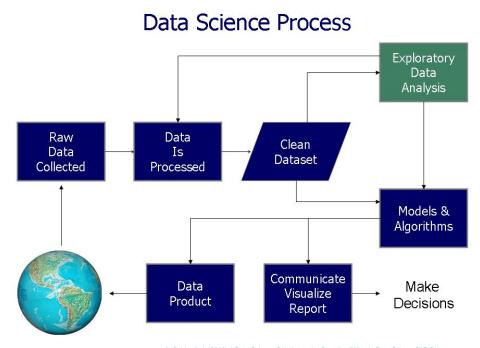
IN-CORE_CENSUSAPI_DWNLD_JOPLIN MSA_2021-01-11.csv



Exploratory Data Analysis:

Jupyter Notebook includes descriptive statistics, tables, and exploratory maps

IN-CORE_CENSUSAPI_DWNLD_JOPLIN MSA_2021-01-11_map.html

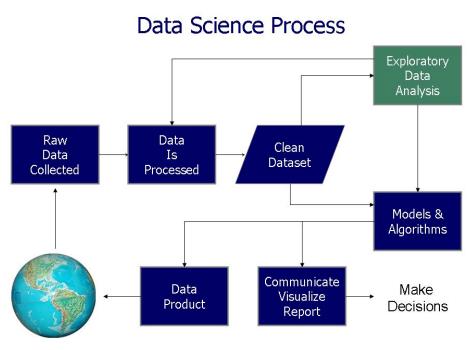


Models and Algorithms:

Workflow does not include models

Algorithm saved as python program file:

IN-CORE_CENSUSAPI_BGMAP_2021-01-19T1403.py

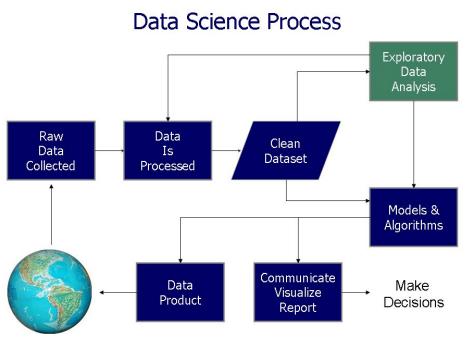


Communicate, Visualize, Report

_ReadMeFirst_CensusAPI_BlockGroup_2 021-01-19.docx

Step by Step Instruction-How to Download Tables as a CSV

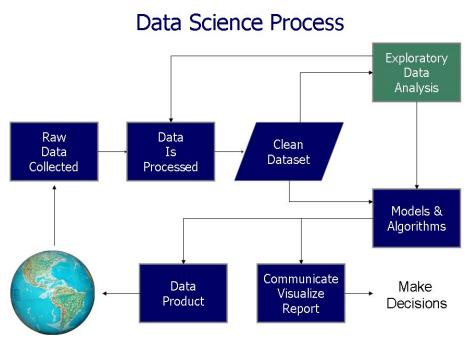
file_2021-01-07.pdf



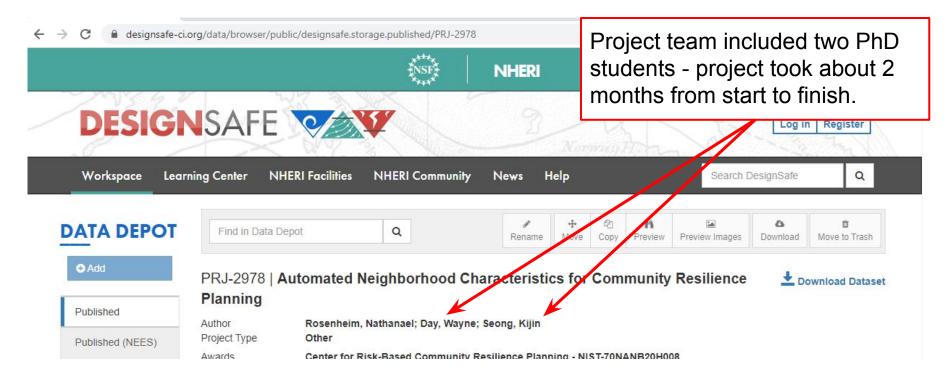
Data Product:

40.

Rosenheim, Nathanael; Day, Wayne; Seong, Kijin (2021) "Automated Neighborhood Characteristics for Community Resilience Planning." DesignSafe-CI. https://doi.org/10.17603/ds2-hj0p-bp



What the project looks like in DesignSafe

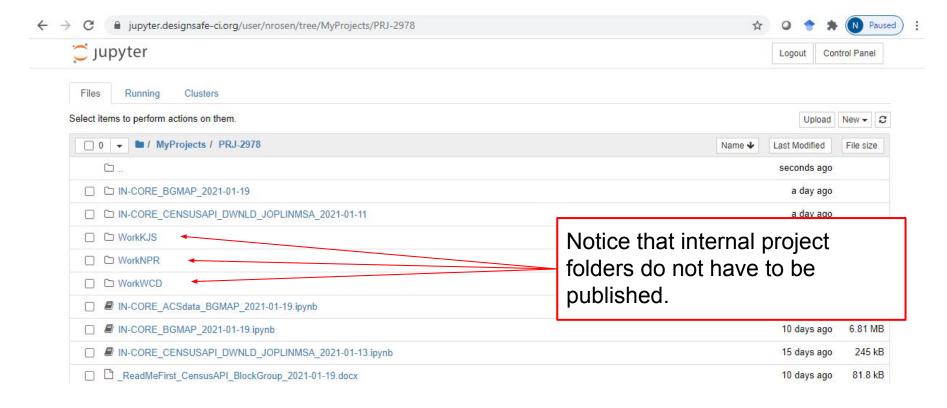


Rosenheim, Nathanael; Day, Wayne; Seong, Kijin (2021) "Automated Neighborhood Characteristics for Community Resilience Planning." DesignSafe-CI. https://doi.org/10.17603/ds2-hj0p-bp40.

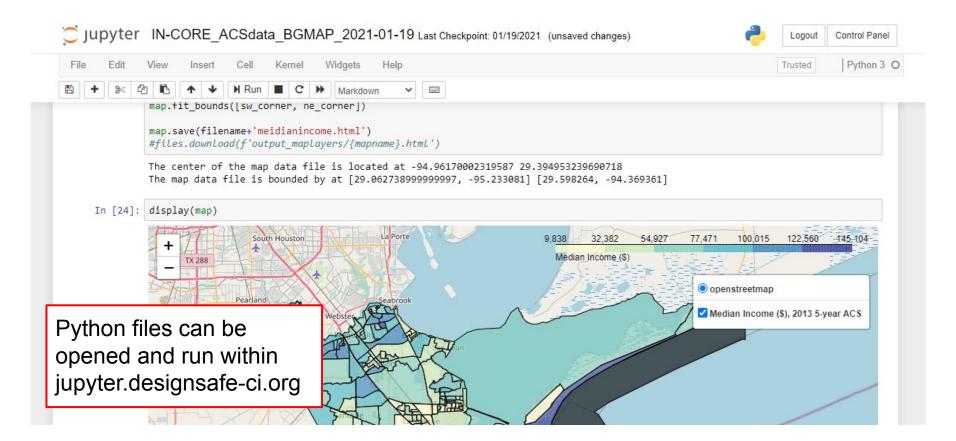
Project files developed and created in DesignSafe



Project Developed Within Jupyter-DesignSafe



Jupyter Notebook uses Python - Free and Open Source



Thank you!

Nathanael Rosenheim

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Thoughts after the workshop

Data science may not be interested in reproducible, open, collaborative research.

What tools help make collaboration and replication easier - "Better science in less time"

Ideal dataset - provides individual level characteristics that identify a person or business. Provides information that helps distinguish and predict outcome. For example, which applicants to Texas A&M will decide to enroll? Need to have enough detail about each applicant to predict the outcome.

Ideal dataset - is linkable to other datasets that provide even more details and information.

Thoughts after the workshop

Some discussion about the drawbacks of synthetic, derived or sanitized data.

Discussion about how "hiding" data leads to bad data. Need to reward making data available and more usefull.

Thoughts about decision makers interpretation of data analytics:

- Each decision maker (example a football coach) will have different ways they want to see the interpretation of data. For example, some like dashboards, others want just a few numbers.
- Emotions in the room make a difference
- The person analyzing the data needs to be very familiar with the data actually needs to see the data collected.
- Need to "read the audience" as the story is being told.

Thoughts after the workshop

Common theme across presentations "Real action comes from individual data". Seemed to be true for the perspectives of public health, transportation, athletics, admissions.

When thinking about data collection start by planning to share the data.

An option for data sharing: Texas Data Repository https://data.tdl.org/