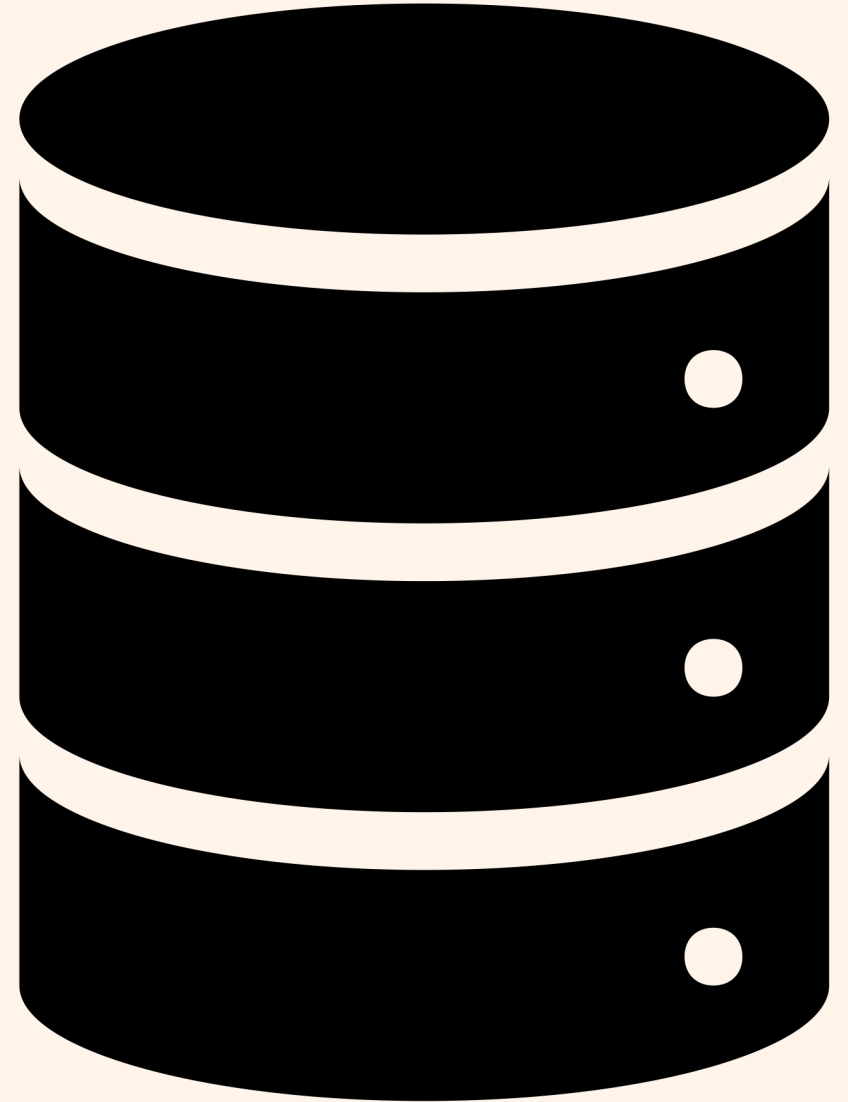


# **BUILDING A DATA SCIENCE TOOLCHAIN**

**(WITH R OR PYTHON)**

RYAN ORSINGER  
TECH DAY 2024

**JUST USE SQL**



# AGENDA

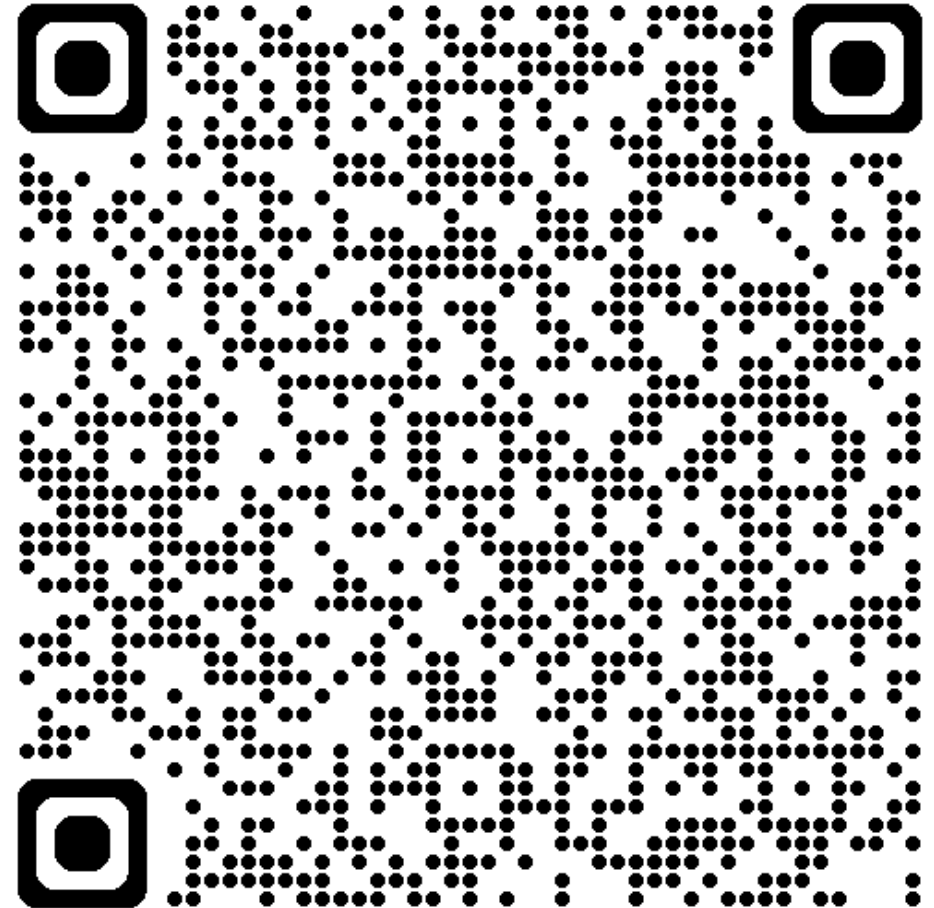
Introduction

Mental Tools

Technical Tools

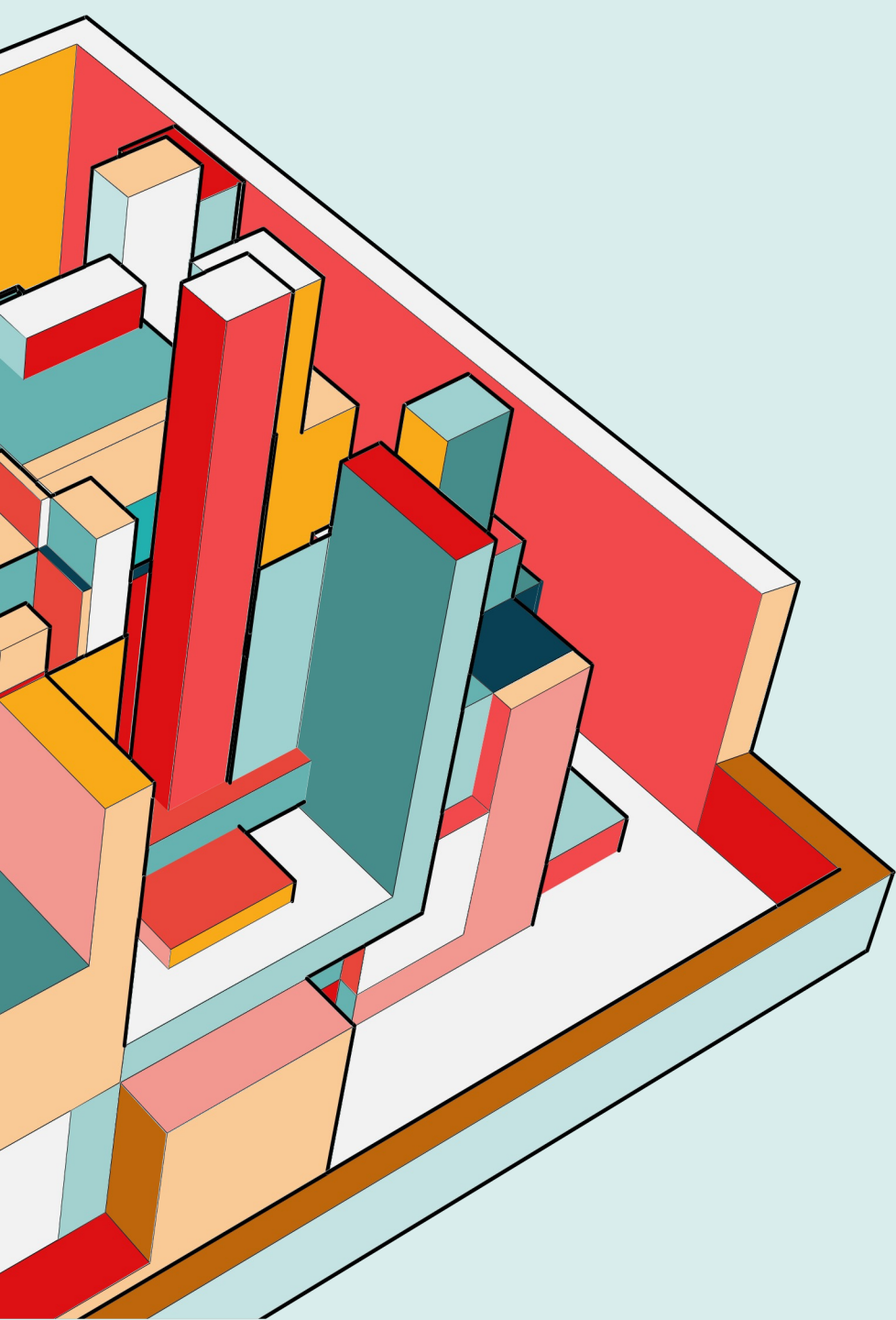
Selected Readings &  
Resources

Next steps (Social Tools)



## ABOUT RYAN

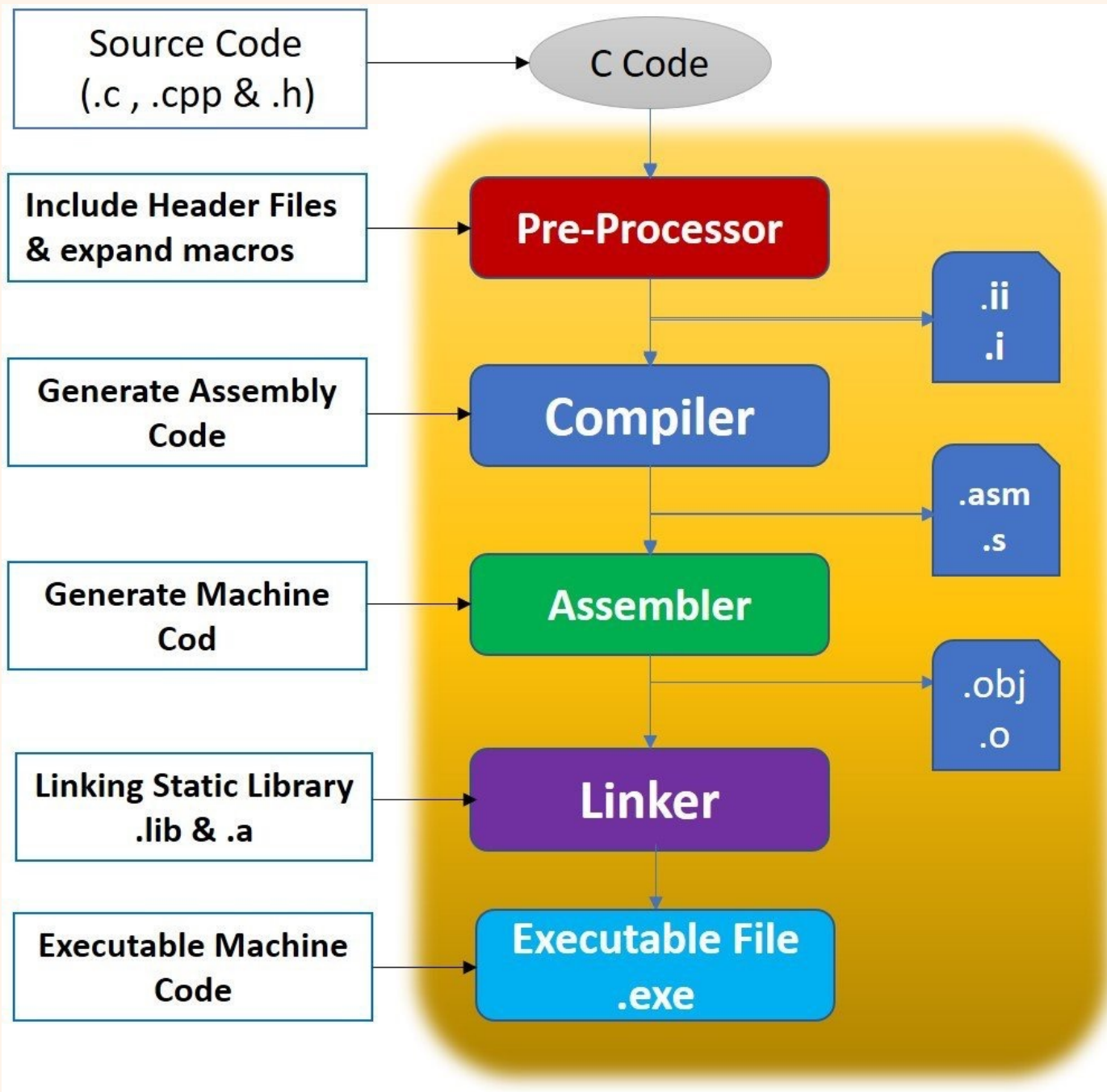




# WHAT IS A TOOLCHAIN?

COLLECTION OF TOOLS USED TO  
PERFORM COMPLEX TASKS

TYPICALLY USED IN SEQUENCE, WITH  
OUTPUT OF ONE TOOL BECOMING  
INPUT FOR THE NEXT TOOL



YOUR DATA SCIENCE  
TOOLCHAIN IS NOT A  
PROGRAMMING  
LANGUAGE  
OR  
A SET OF LIBRARIES.



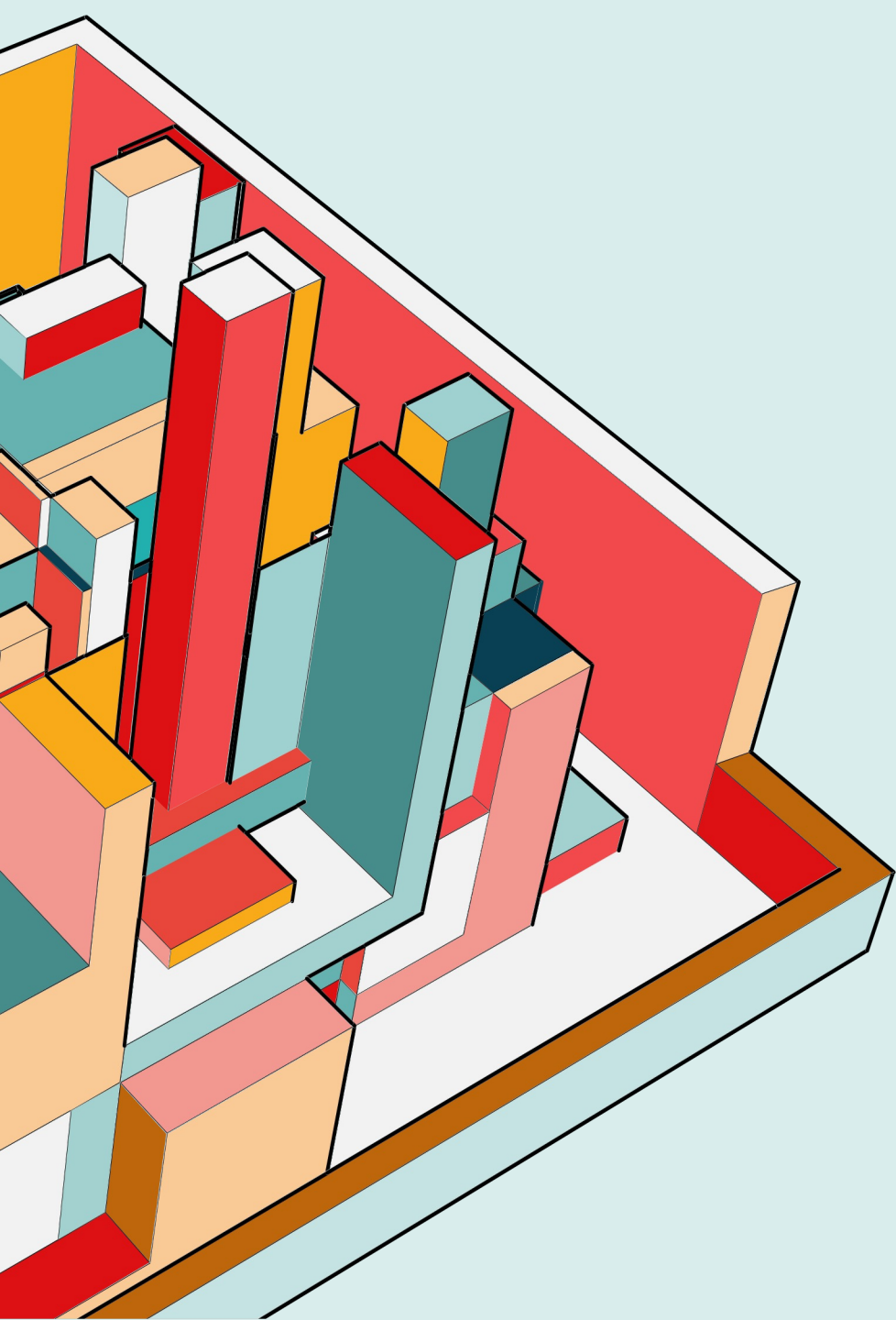




# **YOUR “DATA SCIENCE TOOLCHAIN” IS A WEB, A NETWORK OF:**

- Knowing Your Business Domain
- Mental Models and Habits
- Your Social Ecology
- Feedback & Collaboration
- Technical Tools & Skills
- (Imagine a web with recursion, rather than a linear pipeline)



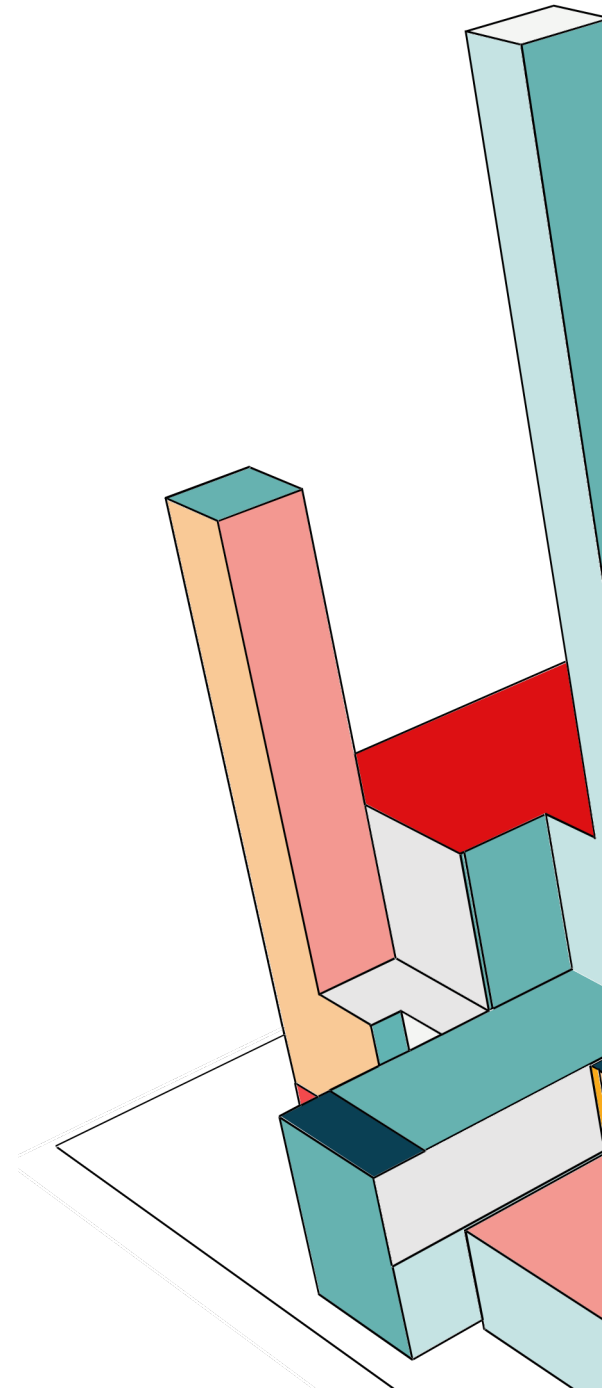


# **1<sup>ST</sup> TOOL IS TO KNOW YOUR BUSINESS.**

YOU MAY KNOW DATA.  
BUT YOU DON'T KNOW YOUR BUSINESS'S  
DATA UNLESS YOU KNOW YOUR BUSINESS.

# KNOW YOUR BUSINESS DOMAIN

- Learn how your business works. There is no substitute.
- Knowing the biz means people will take your recommendations seriously. People can tell if you're an egghead, even if you mean well.
- Interviews > Shadowing > Doing the work of front-line staff
- Work with front-line staff to discover real data quality issues and how your work can help them and the org.
- Connecting with people is great fuel for your relationships, creativity, and problem solving! (#SocialEcology)
- Part of your unique value proposition and problem-solving process.



# CONNECT YOUR ACTIVITIES TO THE LOGIC MODEL OF THE BUSINESS

- Knowing how the business *actually* works (on the ground) and how it *should* work opens the door to:
  - Creativity
  - Better problem solving
  - More tangible impact (Know the people you're helping and how your work helps)
  - Opportunities, Leadership, etc...
- Can your role be done remotely by someone crunching purely technical tickets with no domain knowledge? Say hello to offshoring/LLMs/race to the bottom!

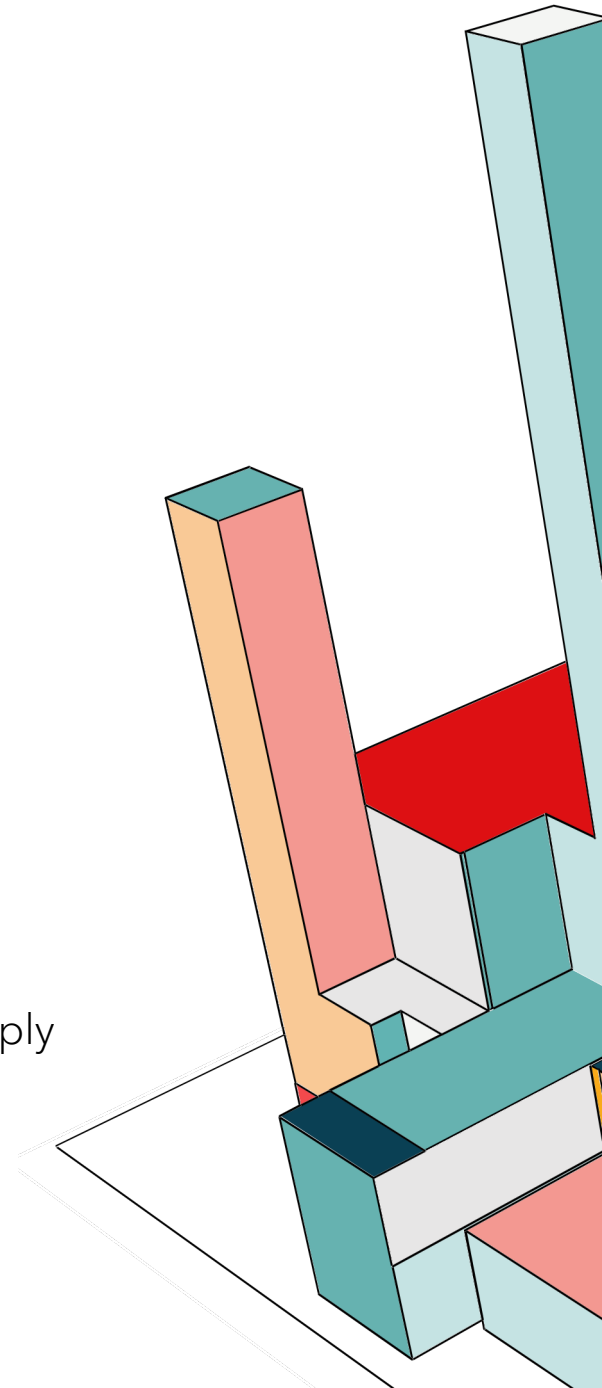
Your work must connect to the [Logic Model](#) of how the organization works.

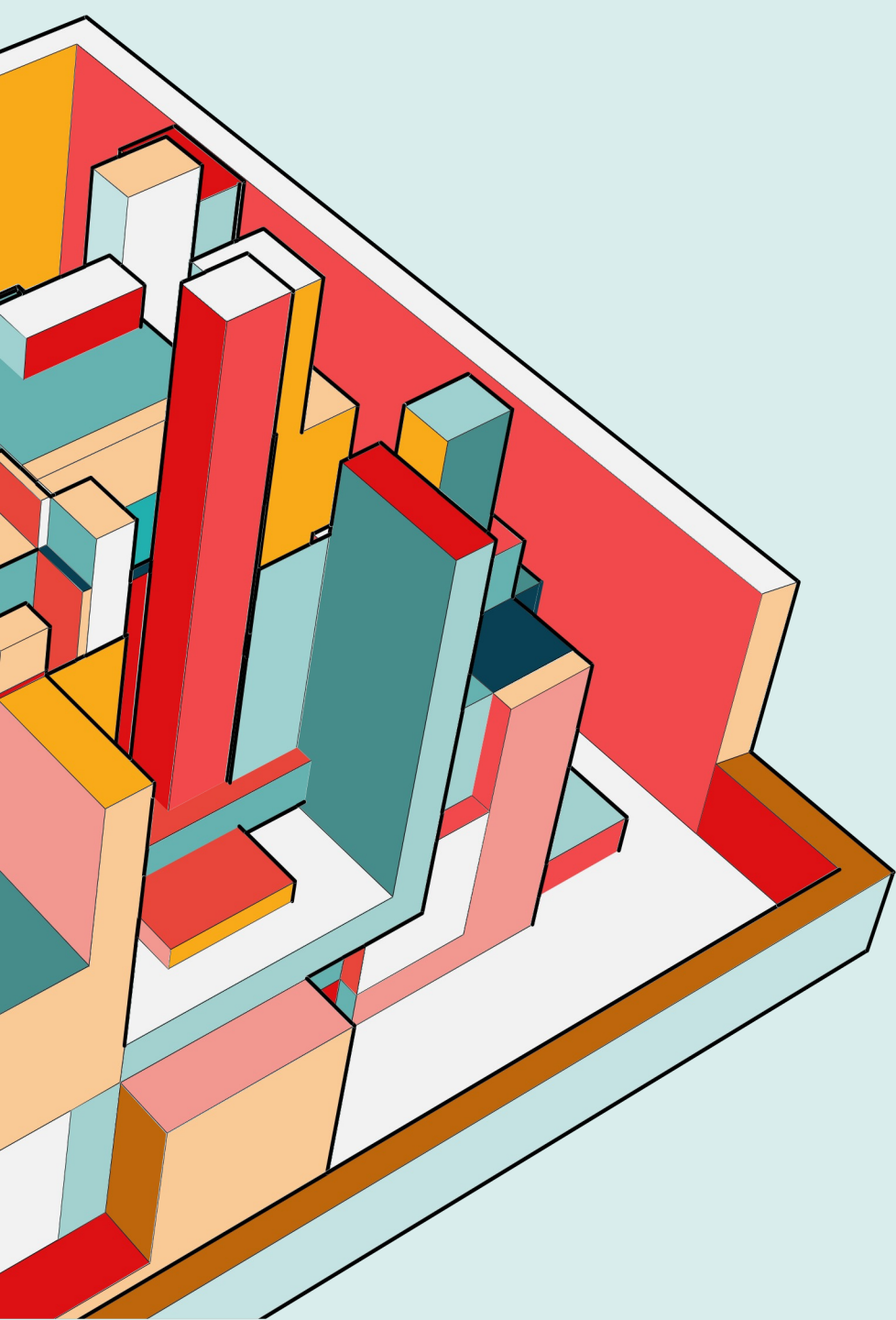
Your analyses must enable the [Business Model](#) of your organization.



# HELPFUL HABITS FOR YOUR TOOL-WEB

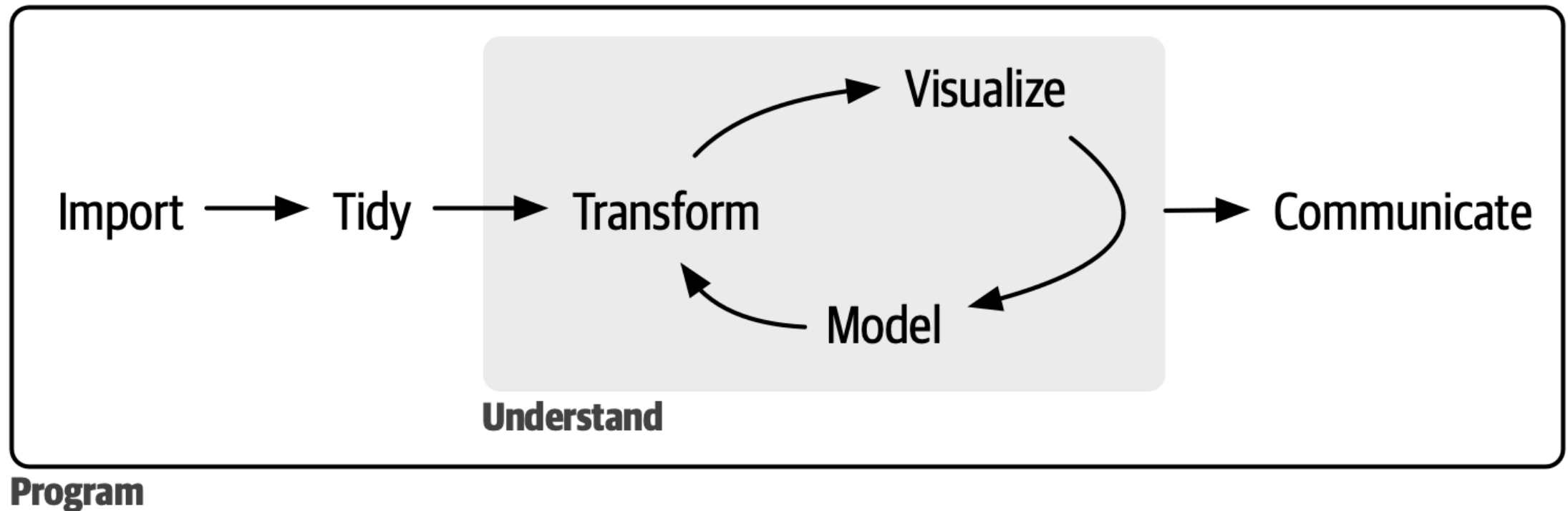
- Rush to a *Minimum Viable Product* ([MVP](#))
  - Rapidly produce low-fidelity prototypes. Done > Perfect. Perfect is the opposite of done. Perfection is asymptotic.
  - Reduces cost of failure. Increases likelihood of success. Easier to pivot.
- Iterate, Seek feedback, Repeat
- Present before you're ready (with a safe environment, when possible)
- When you feel stuck, talk things through!
- Always double check your understanding of things with folks who know the business deeply
- Focus on data definitions, data governance, and the bigger picture.





**TECHNICAL TOOLS!**

# DATA SCIENCE PROCESS IN CODE (WICKHAM)





# PYTHON OR R? YES.

## Python for data science

- Recommend if you already know some Python or a “curly brace” language
- Great for connecting to other data engineering and software development code
- Install [Anaconda](#). This will install Python, too. Batteries included for Mac/Windows/Linux.
- Can run in PowerBI, Tableau, and Excel
- Use [conda environments](#) if you want/need isolated environments/sandboxes

## R for data science

- Recommend if you’re flying solo or starting from scratch
- Install [R](#), [Rstudio](#), and [Tidyverse](#) (batteries included libraries)
- Community can be more beginner friendly
- Can run in PowerBI and Tableau
- Use [renv](#) if you want isolated environments/sandboxes



## A NOTE ON PYTHON “VS.” R

- Use whatever your business already uses. Go with the grain.
- If you're starting from scratch, R code/textbooks get you to the data science faster and with less overhead.
- R is built for statistical computing. Historically, Python has trailed R's development.
- It can be very helpful to learn something you already know in another programming language or paradigm.
- More Python jobs, higher median salaries for Python, but much more competition. R could be a way to “backdoor” some applications.
- What matters most for skill-building is building TONS of projects.
- Real world projects > Toy Projects > Practice > Study



# WHAT ABOUT DEPLOYING YOUR DATA PRODUCTS TO THE CLOUD?

Need to deploy your dashboards, APIs, and models?

If you're using Python, recommend [Anaconda's suite of tools](#).

If you're using R, recommend using [Posit](#)

If you have an IT or DevOps team who will set up environments for you, so you're compliant, secure, etc... that's great, but no need to do this yourself, from scratch.

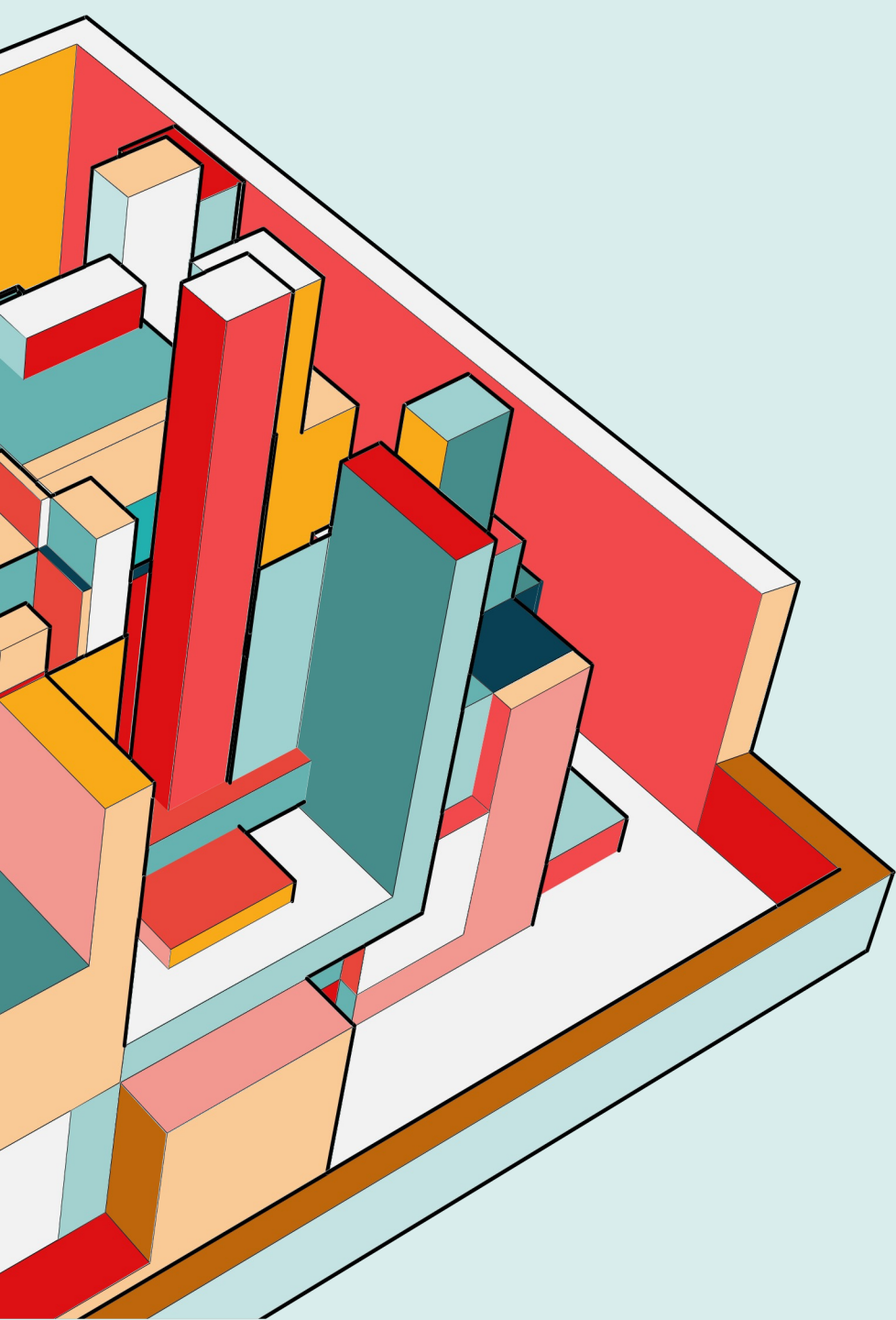
Fussing with infrastructure can slow down the velocity of your data insights.



## COMPLIANCE NOTE ON LLMS

LLM's are very helpful, but they won't *understand* things for you. AGI ain't here.

- Some folks are creating liability/data breaches each time they use an LLM. (HIPPA, for example)
- Be careful about sending material, non-public data to a 3<sup>rd</sup> party...
- 11% of data provided to ChatGPT is confidential, material, non-public
- [Microsoft Copilot for Work](#) advertises security, but your mileage may vary.
- Generate synthetic data with [synthpop](#), [Synthetic Data Vault](#), or LLM itself
- Recommend only feeding synthetic data into an LLM, as examples, to generate code to use in your secure environment.



# **SELECTED READINGS AND RESOURCES**

# SELECTED READINGS

[Think Python](#) by Downey

[Elements of Data Science](#) by Downey

[Python for Data Analysis](#) by McKinney

[Think Stats](#) by Downey

[Think Bayes](#) by Downey

[Causal Inference for the Brave and True](#)

[R for Data Science](#) by Wickham

[Regression and Other Stories](#) by Gelman

[Bayes Rules!](#) by Johnson, Ott, Dogucu

[Statistical Rethinking](#) by McElreath

[ggplot2:: Elegant Graphics for Data Analysis](#) by Wickham

[Tidy Modeling](#) with R by Kuhn and  
Slidge



# ONE PATH TO DATA SCIENCE SKILL BUILDING

- Read lots of books! (books are good)
- Conduct lots of analyses
- Complete lots and lots of projects!
- Real world projects > Toy Projects > Practice > Study
- Your mileage may vary due to motivation and self-accountability

Whether if you learn better on your own or with others, you'll need to do **lots** of projects to build new skills and make new connections!



# OR LEAN INTO SOCIAL ECOLOGY FOR SKILL BUILDING

Community Creates Accountability

Community Creates New Connections

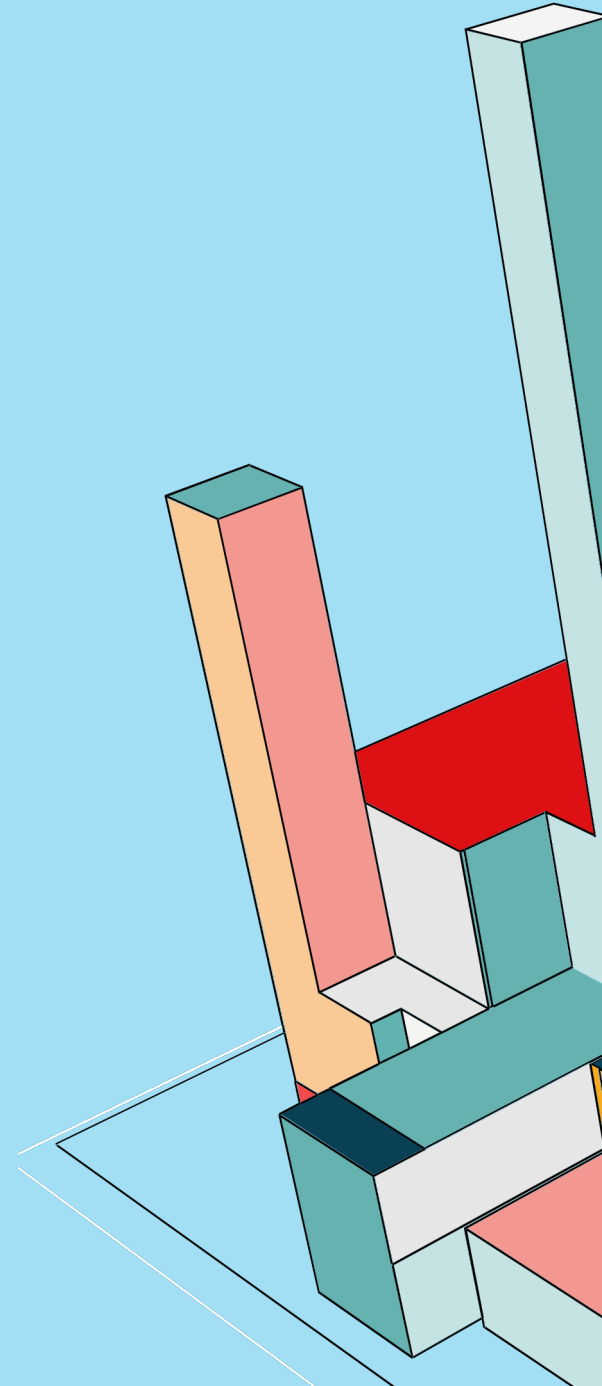
- Data Science Book Club
- Statistics Study Group
- Bayesian Inference Club
- Data Science Project Sharing Group
- Mentoring others, seeking mentors



# FIND A GROUP THAT WORKS FOR YOU!

## TALK AMONGST YOURSELVES!

1. Basic Python Practice Group
2. “Think Stats” Python Book Club
3. “Elements of Data Science” Python Book Club
4. Causal Inference study group (Python)
5. R for Data Science Book Club
6. Bayesian Inference in R book club



# OTHER CONTRIBUTIONS

[101 Exercises](#) for Python and JS

[90 Minutes to Machine Learning](#) video and code

[67 Minutes to NumPy](#) video and code

[Finding the Story in the Data](#)

[Hypothesis Testing in a Nutshell](#)

Contact:

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