

# The Data Science Corps Wrangle/Analyze/Visualize (DSC-WAV) project: Building data science curricula and connections between two- and four- year institutions

Nicholas J. Horton, Amherst College

April 9, 2021, [nhorton@amherst.edu](mailto:nhorton@amherst.edu)



Image source: heylagostechie

```
31 def __init__(self, settings):
32     self.file = None
33     self.fingerprints = set()
34     self.logdups = True
35     self.debug = debug
36     self.logger = logging.getLogger(__name__)
37     if path:
38         self.file = open(os.path.join(path, "requests.log"),
39                          "a")
40         self.file.seek(0)
41         self.fingerprints.update(settings.get("request_fingerprints", []))
42
43 @classmethod
44 def from_settings(cls, settings):
45     debug = settings.getbool("DEBUG", False)
46     return cls(job_dir(settings), debug)
47
48 def request_seen(self, request):
49     fp = self.request_fingerprint(request)
50     if fp in self.fingerprints:
51         return True
52     self.fingerprints.add(fp)
53     if self.file:
54         self.file.write(fp + os.linesep)
55
56 def request_fingerprint(self, request):
57     return request_fingerprint(request)
```

Image source: Wikicommons



Image source: Concord Consortium

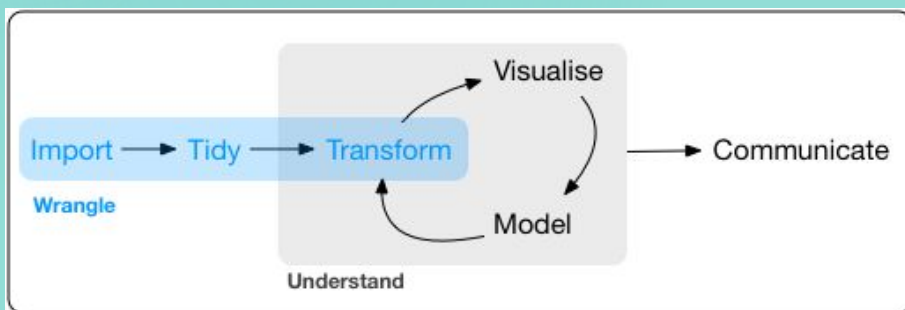


Image source: Hadley Wickham and Garrett Golemund

Slides and links at <https://dsc-wav.github.io/ma-ds-pathways>

# Acknowledgements

- ▶ NSF grant I923388 (DSC-WAV), <https://dsc-wav.github.io/www>
- ▶ co-PIs Ileana Vasu (Holyoke Community College), Brian Candido (Springfield Technical Community College), Eben Afarikumah (Greenfield Community College), Ben Baumer, Matt Rattigan, Valerie Barr, and Jaime Davila
- ▶ also thanks to Jo Hardin, Maria Tackett, Matt Beckman, Andy Zieffler, Jie Chao, Bill Finzer, Colin Rundel, Mine Çetinkaya-Rundel, and others...

# Plan

- ▶ What is Data Science and why should we care?
- ▶ Insights from the NASEM (2018) report
- ▶ Options for Introductory Data Science courses
- ▶ Data science/analytics programs at two-year colleges
- ▶ Co-curricular structures and the DSC-WAV project
- ▶ Next steps

# Zoom poll #1

- ▶ How would you define Data Science?

Please provide a succinct description in the chat window but wait to share it until I say

# DATA SCIENCE FOR UNDERGRADUATES

## Opportunities and Options

consensus report published in 2018  
free download from  
<https://nas.edu/envisioningds>

**Study funded by the  
National Science Foundation**



# Key Insights NASEM (2018): Undergraduate Data Science

- ▶ There must be **multiple pathways** for undergraduates to study data science
- ▶ The undergraduate experience should cater to and **promote diversity** – demographic and intellectual – in the students it serves
- ▶ There are some core competencies that all data science students (and, ideally, all undergraduates) should have
  - ▶ They should develop **data acumen**
  - ▶ Ethical problem-solving is a key component of data acumen

# A Central Finding

**Finding 2.3** A critical task in the education of future data scientists is to instill **data acumen**. This requires exposure to key concepts in data science, real-world data and problems that can reinforce the limitations of tools, and ethical considerations that permeate many applications. Key concepts involved in developing data acumen include the following:

- ▶ Mathematical foundations
- ▶ Computational foundations
- ▶ Statistical foundations
- ▶ Data management and curation
- ▶ Data description and visualization
- ▶ Data modeling and assessment
- ▶ Workflow and reproducibility
- ▶ Communication and teamwork
- ▶ Domain-specific considerations
- ▶ Ethical problem solving.

# Mathematical concepts

Key **mathematical** concepts/skills that would be important for all students in their data science programs and critical for their success in the workforce are the following:

- ▶ Set theory and basic logic,
- ▶ Multivariate thinking via functions and graphical displays,
- ▶ Basic probability theory and randomness,
- ▶ Matrices and basic linear algebra,
- ▶ Networks and graph theory, and
- ▶ Optimization.



# Computational concepts

While it would be ideal for all data scientists to have extensive coursework in computer science, new pathways may be needed to establish appropriate depth in **algorithmic thinking and abstraction** in a streamlined manner. This might include the following:

- ▶ Basic abstractions,
- ▶ Algorithmic thinking,
- ▶ Programming concepts,
- ▶ Data structures, and
- ▶ Simulations.

# Statistical concepts

Important **statistical foundations** might include the following:

- ▶ Variability, uncertainty, sampling error, and inference;
- ▶ Multivariate thinking;
- ▶ Nonsampling error, design, experiments (e.g., A/B testing), biases, confounding, and causal inference;
- ▶ Exploratory data analysis;
- ▶ Statistical modeling and model assessment; and
- ▶ Simulations and experiments

# Data management concepts

Key **data management and curation** concepts/skills that would be important for all students in their data science programs and critical for their success in the workforce are the following:

- ▶ Data provenance;
- ▶ Data preparation, especially data cleansing and data transformation;
- ▶ Data management (of a variety of data types);
- ▶ Record retention policies;
- ▶ Data subject privacy;
- ▶ Missing and conflicting data; and
- ▶ Modern databases.

# Data visualization concepts

Key **data description and visualization** concepts/skills that would be important for all students in their data science programs and critical for their success in the workforce are the following:

- ▶ Data consistency checking,
- ▶ Exploratory data analysis,
- ▶ Grammar of graphics,
- ▶ Attractive and sound static visualizations,
- ▶ Dynamic visualizations and dashboards.

# Data modeling concepts

Key **data modeling and assessment** concepts/skills that would be important for all students in their data science programs and critical for their success in the workforce are the following:

- ▶ Machine learning,
- ▶ Multivariate modeling and supervised learning,
- ▶ Dimension reduction techniques and unsupervised learning,
- ▶ Deep learning,
- ▶ Model assessment and sensitivity analysis, and
- ▶ Model interpretation (particularly for black box models).

# Workflow and reproducibility concepts

Key **workflow and reproducibility** concepts/skills that would be important for all students in their data science programs and critical for their success in the workforce are the following:

- ▶ Workflows and workflow systems,
- ▶ Reproducible analysis,
- ▶ Documentation and code standards,
- ▶ Source code (version) control systems, and
- ▶ Collaboration.

# Communication and teamwork concepts

Key **communication and teamwork** concepts/skills that would be important for all students in their data science programs and critical for their success in the workforce are the following:

- ▶ Ability to understand client needs,
- ▶ Clear and comprehensive reporting,
- ▶ Conflict resolution skills,
- ▶ Well-structured technical writing without jargon, and
- ▶ Effective presentation skills.

# Ethical concepts

Key aspects of **ethics** needed for all data scientists (and for that matter, all educated citizens) include the following:

- ▶ Ethical precepts for data science and codes of conduct,
- ▶ Privacy and confidentiality,
- ▶ Responsible conduct of research,
- ▶ Ability to identify “junk” science, and
- ▶ Ability to detect algorithmic bias.



# Ethical concepts

Key aspects of **ethics** needed for all data scientists (and for that matter, all educated citizens) include the following:

- ▶ Ethical precepts for data science and codes of conduct,
- ▶ Privacy and confidentiality,
- ▶ Responsible conduct of research,
- ▶ Ability to identify “junk” science, and
- ▶ Ability to detect algorithmic bias.

# Zoom poll #2

- ▶ What Data Science or Data Analytics initiatives are underway at your institution?

Please provide a succinct description in the chat window but wait to share it until I say

# Where to start?

- ▶ Introductory data science courses
- ▶ Certificate programs
- ▶ Associates to workforce
- ▶ Associates to transfer
- ▶ Flexible pathways for students
- ▶ My goal: ensuring that the mathematical sciences are engaged in these exciting developments

# Revised Guidelines for Assessment and Instruction in Statistics Education (GAISE) College report (2016)

- ▶ Teach statistical thinking.
- ▶ Teach statistics as an investigative process of problem-solving and decision-making.
- ▶ Give students experience with multivariable thinking.
- ▶ Focus on conceptual understanding.
- ▶ Integrate real data with a context and purpose.
- ▶ Foster active learning.
- ▶ Use technology to explore concepts and analyze data.
- ▶ Use assessments to improve and evaluate student learning.

# Acknowledging some realities

- ▶ There's a pandemic
- ▶ There's never been sufficient resources allocated to our public two-year colleges
- ▶ The enrollment challenges of 2020-2021 make things even more challenging
- ▶ Use of appropriate technology not widespread in two-year college introductory statistics courses
- ▶ How to find a path forward? (Possible answer: not require everyone to develop everything on their own)

# Zoom poll #3

- ▶ Is there an introductory data science course planned or offered at your institution? If so, which department(s) are offering it?

Please provide a succinct description in the chat window but wait to share it until I say

# Introductory data science (all OER!)

- ▶ UC/Berkeley's Data 8: <http://data8.org>
- ▶ Data Science in a Box: <https://datasciencebox.org>
- ▶ Modern Data Science with R (2e, shameless plug!): <https://mdsr-book.github.io/mdsr2e>
- ▶ Hardin et al (Data Science in Statistics Curricula: Preparing Students to “Think with Data”, TAS, 2015, <https://www.tandfonline.com/doi/full/10.1080/00031305.2015.1077729>)

# Next steps for programs

- ▶ Certificate programs
- ▶ Associates to workforce
- ▶ Associates to transfer
- ▶ Work to develop flexible pathways for students

2018 “Two Year College Data Science Summit” report  
(<https://www.amstat.org/ASA/Education/Two-Year-College-Data-Science-Summit.aspx>)



# Developing data acumen is hard!

“Integrating Computing in the Statistics and Data Science Curriculum: Creative Structures, Novel Skills and Habits, and Ways to Teach Computational Thinking” (Horton and Hardin, *Journal of Statistics and Data Science Education*, 2021):

<https://www.tandfonline.com/doi/full/10.1080/10691898.2020.1870416>

We need to think creatively about how to give students repeated practice with the entire data science analysis cycle

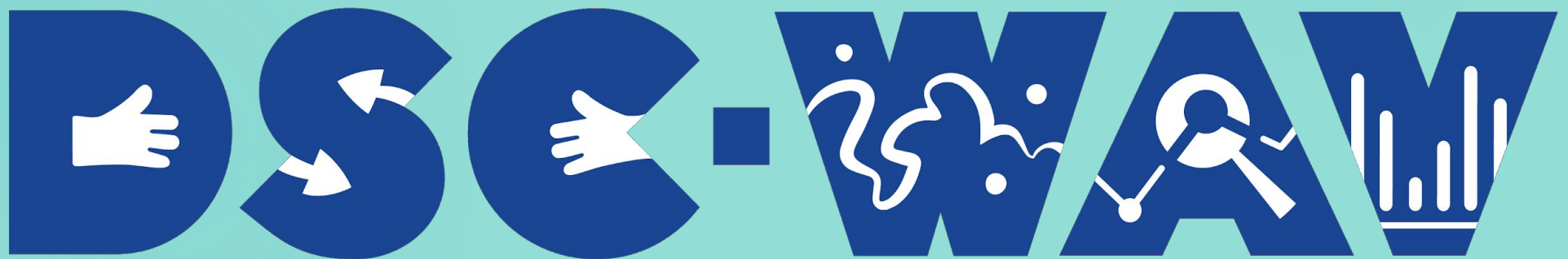
Requires some new courses

Requires reformulation of other courses to develop data acumen

# DSC-WAV (Wrangle-Analyze-Visualize)

- ▶ NSF funded effort from the Harnessing the Data Revolution (HDR) Data Science Corps (DSC) initiative:  
<https://dsc-wav.github.io/www>

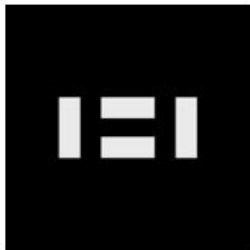
DATA SCIENCE CORPS



WRANGLE•ANALYZE•VISUALIZE

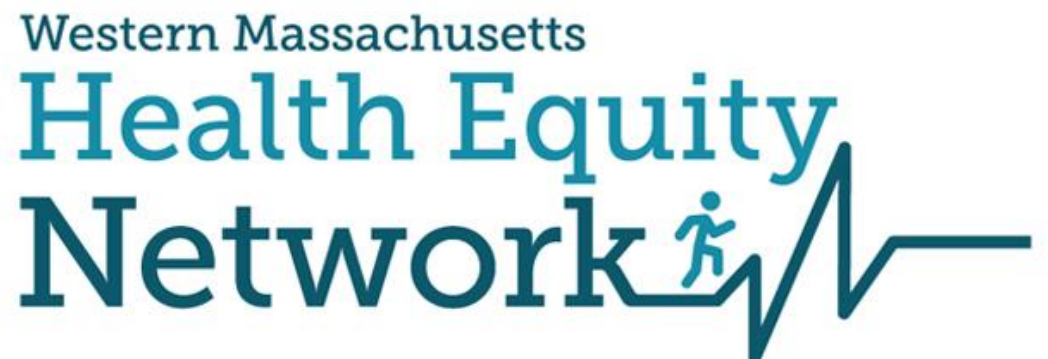
# DSC-WAV (Wrangle-Analyze-Visualize)

- ▶ <https://dsc-wav.github.io/www>
- ▶ Collaborative project with Five Colleges (Amherst, Smith, Hampshire, Mount Holyoke, and UMass/Amherst), Greenfield Community College, Holyoke Community College, Springfield Technical Community College, and the University of Minnesota



# DSC-WAV (Wrangle-Analyze-Visualize)

- ▶ Goal I: create opportunities for undergraduate students to work on Data Science for Social Good projects for community organizations



**girls  
inc.**

of the Valley



The Nature  
Conservancy  
Protecting nature. Preserving life.™

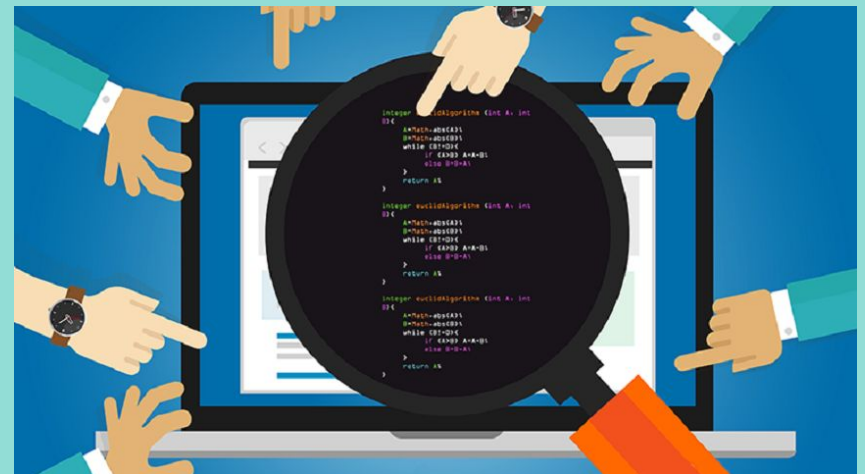


# Agile and scrum for undergraduates

- ▶ Horton et al (2021, HDSR)

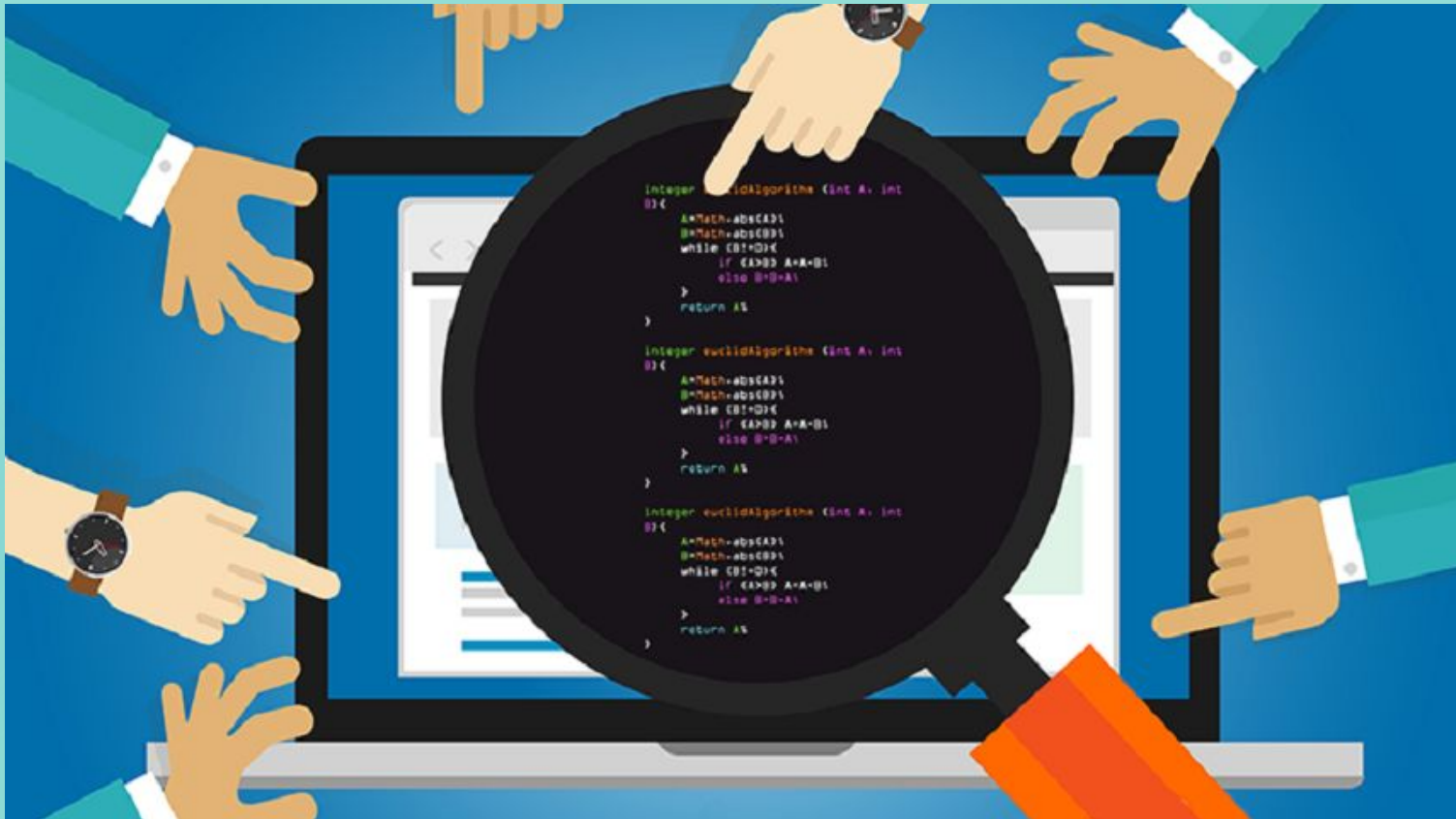
<https://hdsr.mitpress.mit.edu/pub/nvflcexe/release/1>

“While many of these courses and programs teach students relevant data science skills, we can expect coursework to develop students’ data acumen only so far. It is unclear whether coursework alone is enough to provide students with the experiences with data and computing they need to be successful in tomorrow’s workplace.”



Source: techgig.com

# Agile and scrum for undergraduates



Source: techgig.com

# Agile and scrum for undergraduates

The work is organized into a series of short sprints to break up large tasks.

- ▶ Subtasks are organized into a backlog to identify priorities for that stage of the analysis.
- ▶ The team and stakeholders (faculty and community organization liaison) meet regularly to share results and make adjustments in advance of the next sprint.
- ▶ Kanban project boards, implemented using Trello or GitHub Projects, are used to review the backlog and team progress.
- ▶ Code review, implemented using GitHub pull requests, is included as a regular part of the process.

# Agile and scrum for undergraduates

- ▶ Code review, implemented using GitHub pull requests, is included as a regular part of the process.



Source: smartbear.com



Source: Esti Alvarez, see also <https://teachdatascience.com/pairprogramming>



# Agile and scrum for undergraduates

## Goal of code review (simpleprogrammer.com)

1. An evaluation method used to identify code errors
2. Should reveal and remove bugs/errors
3. Should help improve code and documentation quality
4. Should (ideally) build developer/analyst skills and self-confidence

Two heads are better than one!

But still hard for undergraduates

# 9 Code Review Best Practices



Source:kinsta.com

1. Know what to look for in a code review
2. Build and test (before review)
3. Don't review for longer than 15 minutes
4. Check no more than 400 lines at a time (smaller PR are better?)
5. Give feedback that helps (not hurts)
6. Communicate goals and expectations
7. Include everyone in the code review process
8. Foster a positive culture
9. Automate to save time

<https://www.perforce.com/blog/qac/9-best-practices-for-code-review>

# DSC-WAV (Wrangle-Analyze-Visualize)

- ▶ Goal 2: foster curricular innovations and connections between two and four year colleges to teach data science
- ▶ work to develop intro data science courses and programs at the two year colleges
- ▶ provide faculty development opportunities
- ▶ facilitate transfer and articulation for students at 2Y colleges
- ▶ team teaching
- ▶ “Junior Fellow” program

DATA SCIENCE CORPS



WRANGLE•ANALYZE•VISUALIZE

# DSC-WAV Lessons Learned

- ▶ Many challenges to helping students develop the ability to “think with data”
- ▶ Our courses and programs need to adapt to give them necessary workforce skills as analysts
- ▶ DSC-WAV projects have provided a starting point but more reinforcement is needed
- ▶ Lots of work needed to scale out programs at two- and four-year schools

**DATA SCIENCE CORPS**



**WRANGLE • ANALYZE • VISUALIZE**

# Big picture

- ▶ **What are we hoping that students will learn?**

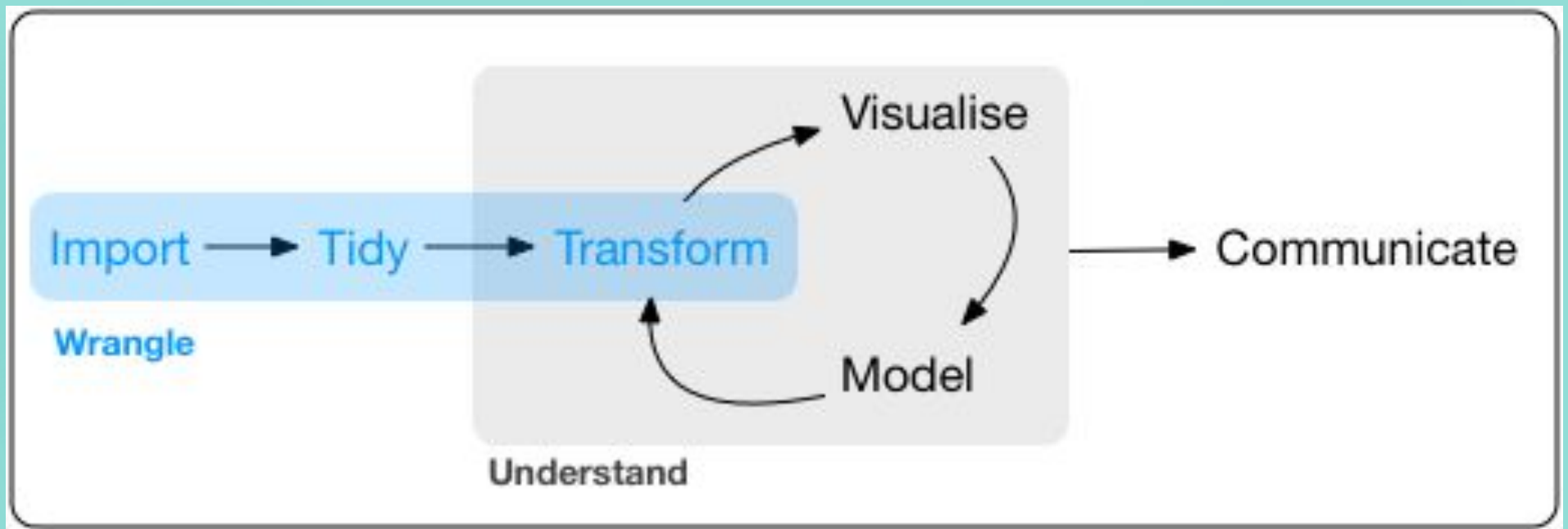


Image source: Hadley Wickham and Garrett Golemund

# Next steps for data science education

- ▶ Focus on computational thinking early and often (key role of multivariate thinking and data acumen)
- ▶ Embrace simplified computational interfaces and approaches to minimize cognitive load and scaffold reproducibility
- ▶ Embrace cloud computing to minimize barriers to technology
- ▶ Integrate and adopt high impact practices and active learning techniques (e.g., pair programming, group- and project- based learning)
- ▶ Creatively scale up faculty development and training

# Zoom poll #4

- ▶ What's the biggest barrier for data science education at your school? What would help address that barrier and support your efforts?

Please provide a succinct description in the chat window but wait to share it until I say

# Back to NASEM (2018)

**Recommendation 2.1:** Academic institutions should embrace data science as a vital new field that requires specifically tailored instruction delivered through majors and minors in data science as well as the development of a cadre of faculty equipped to teach in this new field.

**Recommendation 2.2:** Academic institutions should provide and evolve a range of educational pathways to prepare students for an array of data science roles in the workplace.



# Back to NASEM (2018)

**Recommendation 2.3:** To prepare their graduates for this new data-driven era, academic institutions should encourage the development of a basic understanding of data science in all undergraduates.

**Recommendation 3.1:** Four-year and two-year institutions should establish a forum for dialogue across institutions on all aspects of data science education, training, and workforce development.

# Massachusetts Data Science Pathways

- ▶ Newly founded organization
- ▶ Goal: to foster connections between Massachusetts educators and other stakeholders focused on data science pathways from high school through two- and four- colleges and universities
- ▶ Plans: resource sharing and occasional convenings
- ▶ Sign up for our low-volume mailing list: send mail to : [ma-ds-pathways-join@cs.umass.edu](mailto:ma-ds-pathways-join@cs.umass.edu)
- ▶ Minimal website: <https://dsc-wav.github.io/ma-ds-pathways>

# Zoom poll #5

- ▶ Are there next steps that we as a community can help to support?

Please chime in via the chat window as you have ideas to share.

# The Data Science Corps Wrangle/Analyze/Visualize (DSC-WAV) project: Building data science curricula and connections between two- and four- year institutions

Nicholas J. Horton, Amherst College

April 9, 2021, [nhorton@amherst.edu](mailto:nhorton@amherst.edu)



Image source: heylagostechie

```
31 def __init__(self, settings):
32     self.file = None
33     self.fingerprints = set()
34     self.logdupes = True
35     self.debug = debug
36     self.logger = logging.getLogger(__name__)
37     if path:
38         self.file = open(os.path.join(path, "requests.txt"),
39                         "a")
40         self.file.seek(0)
41         self.fingerprints.update(settings.get("request_fingerprints", []))
42
43 @classmethod
44 def from_settings(cls, settings):
45     debug = settings.getbool("DEBUG_REQUEST_FINGERPRINTS")
46     return cls(job_dir(settings), debug)
47
48 def request_seen(self, request):
49     fp = self.request_fingerprint(request)
50     if fp in self.fingerprints:
51         return True
52     self.fingerprints.add(fp)
53     if self.file:
54         self.file.write(fp + os.linesep)
55
56 def request_fingerprint(self, request):
57     return request_fingerprint(request)
```

Image source: Wikicommons



Image source: Concord Consortium

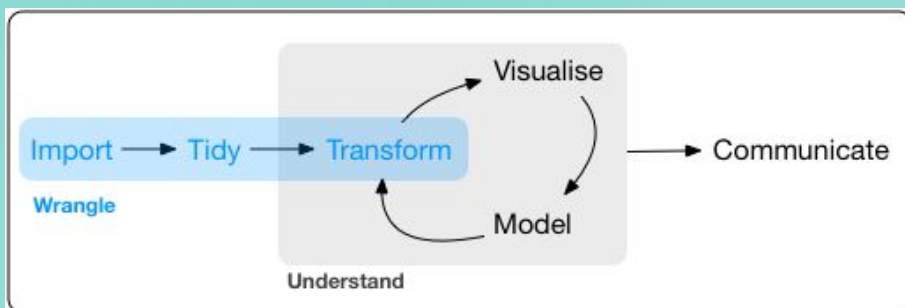


Image source: Hadley Wickham and Garrett Golemund

Slides and links at <https://dsc-wav.github.io/ma-ds-pathways>