University of Massachusetts Medical School

eScholarship@UMMS

National Network of Libraries of Medicine New England Region (NNLM NER) Repository England Region

2019-03-29

Data Science Training for the Future: Building a Carpentries Consortium

Julie Goldman Harvard Medical School

Et al.

Let us know how access to this document benefits you.

Follow this and additional works at: https://escholarship.umassmed.edu/ner

🔮 Part of the Library and Information Science Commons, and the Public Health Commons

Repository Citation

Goldman J, New England Software Carpentry Library Consortium. (2019). Data Science Training for the Future: Building a Carpentries Consortium. National Network of Libraries of Medicine New England Region (NNLM NER) Repository. https://doi.org/10.13028/q2bx-q484. Retrieved from https://escholarship.umassmed.edu/ner/68

Creative Commons License

This work is licensed under a Creative Commons Attribution 4.0 License.

This material is brought to you by eScholarship@UMMS. It has been accepted for inclusion in National Network of Libraries of Medicine New England Region (NNLM NER) Repository by an authorized administrator of eScholarship@UMMS. For more information, please contact Lisa.Palmer@umassmed.edu.

Data Science Training for the Future

Building a Carpentries Consortium

Julie Goldman on behalf of the New England Software Carpentry Library Consortium NNLM NER eScience Forum | March 29, 2019

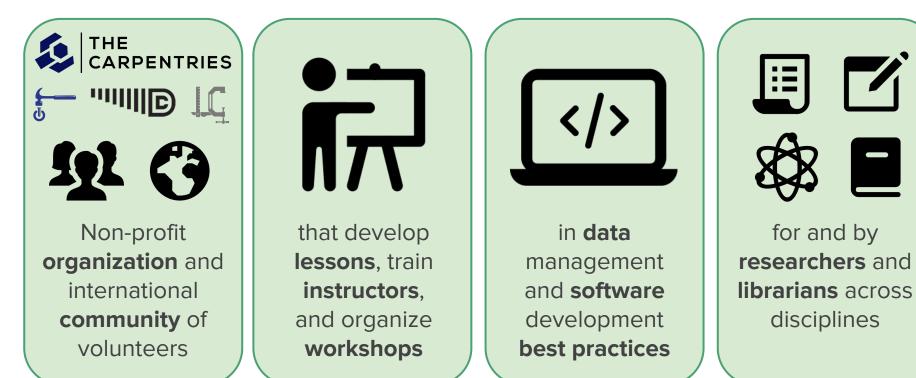
Community of Practice

- Desire for more advanced/technical training for librarians
 - For teaching students
 - For librarians to gain skills too
- Wanted training to teach coding and data skills
 - Carpentries offered this opportunity
 - Long waiting list for training
 - Shorter with a membership
- Expensive to go alone
 - Less risk as a group
- Conversation at RDAP 2017
 - Could we bring together institutional contributions?
 - What would be the benefits to our institutions?



Created by Vectors Market from Noun Project

The Carpentries



https://carpentries.org

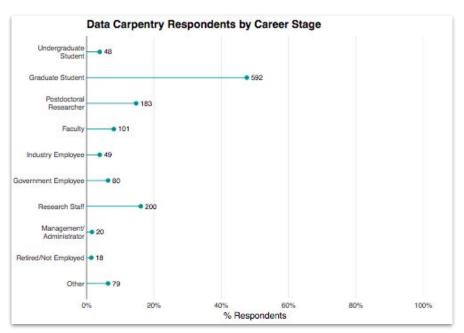
Workshops

- 2-days, active/hands-on learning
- Trained/certified instructors
- Feedback to learners throughout the workshop (Surveys, post-its, +/-)
- Friendly learning environment (Code of Conduct)



Who takes workshops?

66% of Data Carpentry workshop attendees are early career



Analysis of Software and Data Carpentry's Pre- and Post-Workshop Surveys https://doi.org/10.5281/zenodo.1325463

What's the impact?

85% report they gained more confidence after a workshop

	Figure: Perc	eption of	Workshop	o Impact			
Reproducible	3.73 (1.00)	4.2%	8.1%	18.2%	49.7%	19.9%	
Recognition	3.72 (0.96)	3.9%	4.4%	26.6%	45.8%	19.2%	
Productivity	3.45 (1.00)	4.6%	9.7%	35.0%	37.0%	13.7%	Percen
Motivation	2.86 (1.03)	9.0%	26.9%	39.4%	17.9%	6.7%	- 75 50
Confidence	4.12 (0.88)	2.8%	2.5%	9.5%	50.7%	34.5%	25 0
Coding	3.67 (1.00)	4.4%	6.9%	25.4%	44.3%	18.9%	
Career	3.57 (1.02)	4.2%	7.9%	33.7%	35.3%	18.9%	
21	Mean (SD) Str	ongly disagre	e Disagree	Neutral	Agree	Strongly agree	

Analysis of the Carpentries Long-Term Feedback Survey: Perception of Workshop Impact https://carpentries.github.io/assessment/learner-assessment/archives/2018/code/2018_Ja nuary_long_term_report.html

Forming a Consortium

New England Software Carpentry Library Consortium (NESCLiC)

- Gathered a group of like minded individuals
- Library and IT staff from academic institutions

Administrative Steps

- Written proposals for each school
- Signing/Billing Institution & Letters of Intent
- Gold Tier Membership
 - \$15,000 per year
 - \$1000 per instructor trained





NESCLiC Members



Andrew Creamer Bruce Boucek Patrick Rashleigh

Kelsey Sawyer Ashley Champagne



Kristin Lee Ari Gofman Kyle Monahan

Susan Remondi

UCONN UNIVERSITY OF CONNECTICUT

Jennifer Chaput



DARTMOUTH

Lora Leligdon Christian Darabos James Adams

Pamela Bagley Stephen Gaughan Yale

Joshua Dull Catherine DeRose Kate Nyhan

David Cirella Barbara Esty Sawyer Newman UMASS AMHERST

Thea Atwood

Ann Kardos Erin Jerome

HARVARD LIBRARY Julie Goldman Ceilyn Boyd



Sarah Oelker James Burke



Sally Gore Tess Grynoch

Meghan Kerr

Instructor Training

- Virtual instruction in two locations
 - Brown & UMass Amherst
- Introduce you to evidence-based best-practices of teaching
 - A different way to look at pedagogy!
- Create a positive environment for learners at your workshops
- Provide opportunities for you to practice and build your teaching skills
- Help you become integrated into the Carpentries community



Example Lesson: *OpenRefine*

N	
	Library Carpentry OpenRefine
	Introduction to OpenRefine
Overview	
feaching: 15 min Exercises: 0 min	Questions What is OpenRefine? What can it do? Objectives Explain what the OpenRefine software does Explain how the OpenRefine software can help work with data files
What is OpenR	efine?
penRefine is described as "a po	over tool for working with messy data" David Huynh - but what does this mean? It is probably easiest to describe the kinds of data OpenRefine is ts of problems it can help you solve.
	you have data in a simple tabular format such as a spreadsheet, a comma separated values file (csv) or a tab delimited file (tsv) but with internal mats, or where data appears, or in terminology used. OpenRefine can be used to standardize and clean data across your file. It can help you:
	it data set, for example standardizing date formatting nore granular parts, for example splitting up cells with multiple authors into separate cells
 Match local data up to other Enhance a data set with data 	data sets, for example in matching local subjects against the Library of Congress Subject Headings a from other sources
	a from other sources
 Enhance a data set with data one common scenarios might t Where you want to know how Where you want to know how 	a from other sources
 Enhance a data set with data one common scenarios might t Where you want to know how Where you want to know how 	a from other sources be: w many times a particular value (name, publisher, subject) appears in a column in your data w values are distributed across your whole data set
Enhance a data set with data orme common scenarios might t Where you want to know how Where you want to know how Where you have a list of date	a from other sources be: w many times a particular value (name, publisher, subject) appears in a column in your data w values are distributed across your whole data set ss which are formatted in different ways, and want to change all the dates in the list to a single common date format. For example:
Enhance a data set with data ome common scenarios might t Where you want to know how Where you want to know how Where you have a list of date Data you have	a from other sources be: w many times a particular value (name, publisher, subject) appears in a column in your data w values are distributed across your whole data set s which are formatted in different ways, and want to change all the dates in the list to a single common date format. For example: Desired data
Enhance a data set with data ome common scenarios might t Where you want to know how Where you want to know how Where you have a list of date Data you have list January 2014	a from other sources be: w many times a particular value (name, publisher, subject) appears in a column in your data w values are distributed across your whole data set s which are formatted in different ways, and want to change all the dates in the list to a single common date format. For example: Desired data 2014-01-01
Enhance a data set with data ome common scenarios might t Where you want to know how Where you want to know how Where you have a list of date Data you have Ist January 2014 D1/01/2014	a from other sources be: w many times a particular value (name, publisher, subject) appears in a column in your data w values are distributed across your whole data set so which are formatted in different ways, and want to change all the dates in the list to a single common date format. For example: Desired data 2014-01-01 2014-01-01
Enhance a data set with data ome common scenarios might t Where you want to know how Where you want to know how Where you have a list of data Data you have Ist January 2014 Jan 1 2014 Jan 1 2014 2014-01-01	a from other sources be: wr many times a particular value (name, publisher, subject) appears in a column in your data wr values are distributed across your whole data set es which are formatted in different ways, and want to change all the dates in the list to a single common date format. For example: Desired data Desired data 2014-01-01 2014-01-01 2014-01-01
Enhance a data set with data ome common scenarios might t Where you want to know how Where you want to know how Where you have a list of data Data you have Ist January 2014 Jan 1 2014 Jan 1 2014 2014-01-01	a from other sources be: w many times a particular value (name, publisher, subject) appears in a column in your data w values are distributed across your whole data set as which are formatted in different ways, and want to change all the dates in the list to a single common date format. For example: Desired data 2014-01-01 2014-01-01 2014-01-01 2014-01-01

London

•	All		Chemical eleme
		1.	hydrogen
		2.	Choose new match helium
			Choose new match
		3.	Rhium Choose new match
		4.	beryllum Choose new match
		5.	boron Choose new match
		6.	carbon
		7.	Choose new match ntrogen
			Choose new match
		8	oxygen Choose new match
		9	fluorine Choose new match
		10.	neon Choose new match
		11	sodium
		12.	Choose new match magnesium
		13	Choose new match aluminum
			Choose new match
		14.	silicon Choose new match
		15.	francium Choose new match
		16.	phosphorus Choose new match
		17.	iron
		18	Choose new match sulfur
			Choose new match
		19.	chlorine Choose new match
		20.	argon Choose new match
		21.	potassium Choose new match
		22.	calcium
		23.	Choose new match lead
121	-10	-	Choose new match

https://librarycarpentry.org/lc-open-refine

London]

Unexpected Benefits

- Not just instructor training
 - Keep data/coding skills sharp
 - A group of people who help each other
- Expanded Local Network
 - Communicate and collaborate on other projects
- Collaboration with other regional organizations
 - NEASIST
 - NNLM New England Region



Moving Forward



- The Carpentries
 - Evolving as an organization
- Changing roles within libraries
 - Evolution of Library Carpentry
 - More 'non-traditional' library training
- Train-the-Trainer
 - Train others in Carpentries pedagogy
- Further participation in lesson development & curriculum areas
 - Becoming lesson maintainers
 - Digital Humanities at Yale
 - Biodiversity at Harvard

nesclic.github.io/home

NESCLIC About Members Publications Workshops

The New England Software Carpentry Library Consortium or NESCLIC is an association of academic libraries, joined as a community of practice focused on building data science skills in research computing, and extracting, wrangling, storing, analyzing, and visualizing data. NESCLIC allows member institutions to share in a Gold-level membership with the Carpentries.

Our Model

NESCIIC member institutions share a Gold-level Carpentries membership, which affords NESCLIC 6 free coordinated workshops, a 50% discount on the cost for additional coordinated workshops, no charge for self-organized workshops, and fifteen instructors trained, per year.

NESCLIC provides the opportunity for members to:

- Develop instructional materials
- Build a network of data fluent researchers and instructors
- · Refine skills by participating as instructors at other member institutions

Financial commitment: Cost of shared Carpentry membership (\$1000/person) plus training expenses (travel or hosting)

Time commitment: (2) - two-day training workshops plus committee work

NESCLIC has established bylaws and members contribute to the consortium by participating in one of several roles:

- Carpentries liaison
- Membership/Recruiting
- Assessment
- Outreach
- Workshop coordination

Become a Member

Interested in joining NESCLIC?

Email: nesclic@googlegroups.com

Workshops

Check back for future workshops offered by NESCLiC members.

This list includes official Carpentries' workshops and workshops adapted from the Carpentries' materials where NESCLIC members served as hosts, instructors, or helpers.

Venue	Dates	Lessons		
Tufts University	April 12- 13, 2019	Unix, Git, Programming with Python		
Tufts University	February 15-16, 2019	Data Organization in Spreadsheets, OpenRefine for Data Cleaning, R for Social Scientists		
Brown University	January 16-17, 2019	Data Organization in Spreadsheets, OpenRefine for Data Cleaning, Introduction to R, Data Analysis and Visualization in R, Data Management with SQL		
Harvard University	January 15-16, 2019	Data Intro for Librarians, Shell Lessons for Libraries, Tidy Data for Librarians, OpenRefine for Librarians		
UMass Amerhest	January 8- 9, 2019	Unix, Python, Git		
Dartmouth College	November 29-30, 2018	Unix, Git, Python		
Tufts University	November 9-10, 2018	Unix, Git, Programming with Python		
Brown University Library (Hosts & Helpers), Yale and Tufts (Instructors)	October 22-23, 2018	Data Intro for Librarians, Shell Lesson for Librarians, Git Intro for Librarians, OpenRefine for Librarians		
Yale University Library	August 15-16, 2018	Tidy Data for Librarians, Shell Lesson for Librarians, OpenRefine for Librarians, Python Intro for Libraries		