GitHub: A tool for social data set development and verification in the cloud

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I can't send you my data b/c I think you might find out I made an error. #overly-honestmethods (tweeted by Charlisle Rainey, 1/10/2013)

A data set is created, analyses are run. An article is published. The data set languishes. The researchers who created the data set move on to other projects. Maybe someone else accesses the data and updates it for their own work. However, these updates are never connected back to the original and the updates are not widely known about. So, multiple people end up wasting time making the same updates. Maybe another researcher finds a mistake in the original data set. They email the authors suggesting corrections. The original authors may or may not make the changes. If changes are made, there are no easily accessible records of this. Some researchers may even be reluctant to make their data available at all partially because of a fear that someone may find a mistake in it and they will have to spend time making tedious corrections.

These are all examples of ways that the development and management of social science data sets do not take advantage of knowledge distributed in the social science community. These problems are partially caused by the data storage tools social scientist use. Despite rapid recent advances in social technologies—Twitter, Facebook, and so on—the tools many social scientists use do not make it easy nor do they provide incentives to collaboratively develop data sets and verify their accuracy, especially for people not involved in creating the original data set.

In this brief article I show that GitHub¹ offers a comprehensive data storage service for social scientists. It has unique tools for *social data set development and verification*. GitHub fits directly into a research workflow, particularly one that also includes R.²

GitHub was originally created and is widely used as a tool for software developers to work together on projects using the Git³ version control system. It has been called the "Facebook of programmers" (Xie, 2011). Though GitHub is already often used by social scientists for statistical package development⁴ it initially seems strange for me to suggest that this service would be useful for social scientists building and maintaining data sets. However, a software program and many social science data sets are fundamentally similar. They are collections of text files. GitHub is a means of storing, version controlling, and collaborating on text files. So if a social scientist's data is (a) in a plain text format, such as comma-separated values (CSV),⁵ and (b) has accompanying variable description files also in a plain text format-e.g. TXT plain text or Markdown (MD),⁶ they can take full advantage of GitHub's features for social data set development and verification.⁷

In this article I begin by discussing the features that we want from a cloud service to store social sci-

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[†]The files used to create this article are available on GitHub at: https://github.com/christophergandrud/PoliSciGitData.

¹https://github.com/

²GitHub can be used to store and develop entire social science research projects, not just data. However, I do not directly address these other topics here.

³http://git-scm.com/

⁴For example the *Zelig* R package (Imai, King and Lau, 2008; Owen et al., 2012) is hosted on GitHub. See https://github.com/IQSS/Zelig.

⁵All major statistical programs as well as Microsoft Excel and Apple's Numbers can save and open files in plain text formats like CSV.

⁶http://daringfireball.net/projects/markdown/

⁷See Bowers (2011, 3) for a discussion of the advantages of storing research files in plain text format.

ence data: stable storage, version control, access, and collaboration. Then I examine how strong these features are in three methods currently used to store social science data. Though each method has their strengths, none of them encourage social data set development and verification.

A few brief notes before beginning: started with Git and GitHub see the GitHub Set Up page: https://help.github.com/articles/ set-up-git. In this article I give examples of how to take advantage of GitHub using their website and from the command line. However, there are also very good graphical user interface (GUI) versions of GitHub for Mac and Windows (see GitHub's Set Up page). I focus on GitHub, but you can also use other services that work with Git such as Bitbucket⁸ for many of the same things. Finally, the tools I discuss in this article are suitable for the small to medium size data sets (i.e. of about 100,000 observations or less) common in social science. Much larger data sets often cannot be efficiently stored in plain text files. GitHub is probably not a suitable storage service for them.

1 What do we need from cloud data storage?

A cloud storage service for social science data needs to enable at least four things: stable storage, version control, access, and collaboration. Obviously a cloud storage service needs to be a stable and reliable platform for keeping data sets on. The service needs to include version control-similar to track changes in a word processing program—so that the development of the data set can be understood and researchers are able to revert to old versions (see Bowers (2011), Healy (2011) and Fredrickson, Testa and Weidmann (2011) for discussions of version control). Data stored on it needs to be accessible to both coauthors and others who wish to reproduce analyses (Fomel and Claerbout, 2009), verify the data sets accuracy, or use the data sets in new projects (Kelly, 2006; King, 1995). Another key part of access is that the storage service makes it easy to both develop and use data sets as part of researchers' workflows. Finally, a cloud data storage service should make collaboration easy. Collaboration should not be limited to coauthors. It should be possible to also collaborate with non-coauthors so that the data set can be improved

by taking advantage of knowledge (and motivation) distributed throughout the social science community. Ideally, this social data set development and verification keeps data sets more up-to-date and more accurate.

2 Data storage: the social science status quo

Social science data is currently often stored in three ways. It may be stored locally on a researcher's computer, on a general purpose cloud storage service such as Dropbox⁹ or Google Cloud Drive, ¹⁰ or on a specialized research hosting service, notably Dataverse. ¹¹ Each of these data storage methods has their strengths in terms of stable storage, access, version control, and collaboration. Nonetheless each are lacking in at least one important way and none of them promote social data set development and verification.

Local storage The least robust form of data storage currently used is storing it on individual researchers' computers. This is not a very stable form of storage as anyone knows if they have lost all of their files when their computer died. Version control with Git or some other program is possible with locally stored data, though if the version control files are lost along with the data, the data is lost. Researchers with access to the computer can easily access the data and use it in their workflow, but access for co-authors is limited. Access by non-coauthors is very cumbersome. Files must be emailed in response to individual requests. These files are not automatically updated with new versions of the data. Because access is limited to users of the computer the data is stored on, collaboration, especially by non-coauthors is extremely limited.

General purpose cloud storage Dropbox and other general purpose cloud storage services offer much more robust storage. These services generally work by syncing files stored on individual computers with those on cloud servers. Dropbox has a basic version control system. If you are using Dropbox for free, old versions are saved for 30 days. ¹² In addition you could use Git with a data set stored on

⁸https://bitbucket.org/

⁹https://www.dropbox.com/

¹⁰https://drive.google.com/

¹¹http://thedata.org/

¹²Old versions are stored for longer with paid accounts.

Dropbox. Access is usually very easy. These services can generally be accessed via desktop programs, mobile apps, and websites. Because files stored on individual computers are automatically synced with cloud serves it is very easy to incorporate them into a workflow. General purpose cloud storage services also make it possible to share files and folders via URL links. Every time you save a file a new version is saved on Dropbox. Collaboration with coauthors is very easy because folders can be shared. This means that official collaborators (those given permission to make changes to the folder, i.e. write access) can easily change files in the folder and these changes are automatically synced for all users. 13 However noncoauthors, without write access to the shared folder cannot easily make changes. They must email one of the researchers with write access to suggest updates. This is not much better than a situation with locally stored files.

Dataverse Many journals-Political Analysis for example-require data used in articles they publish to be uploaded to a specialized research data storage service like Dataverse. Dataverse is a stable form of storage. It does have some version control features. It saves each version that is uploaded to it. Dataverse is easily accessible for anyone with an internet connection. But it is difficult to access data as part of a research workflow, at least using the standard version.¹⁴ Unlike Dropbox, for example, there is no way to automatically update a data set on Dataverse. You have to point and click to upload data files for each version. It is difficult to access data directly from a statistical program such as Stata or R. The data needs to be downloaded by pointing and clicking and then loaded into these programs. Finally, it is difficult to collaborate using Dataverse. Changes to a data set must be uploaded to the site manually, then manually downloaded by collaborators. There are also no direct ways for non-coauthors to suggest changes.

Dataverse is good for what it is designed to do: store a snapshot of a data set for replicating specific published results. However, it is difficult to use it to store and access data as part of an ongoing research workflow. In addition it does not easily enable or incentivize social data set development and verification

3 Data Storage on GitHub

Though admittedly requiring a steeper learning curve, GitHub meets the four criteria for data set storage very well. In this section I will demonstrate how to use GitHub for data set storage, version control, access, and collaboration. Rather than giving a complete introduction, I focus on specific important aspects of each feature in Git and GitHub. In this section I give examples from a data set stored on GitHub. It is a data set of countries' Gallagher Electoral Disproportionality index that I put together from data used in Gallagher (1991, updated through 2011) and Carey and Hix (2011). The data set and more information about it is available at: http://christophergandrud.github.com/Disproportionality_Data/.

3.1 Storing and version control

Version control is an integral part of how GitHub stores files. GitHub remotely stores files in what Git calls "repositories", repos for short. You can think of repos as the folder containing a data set, including the data and description files. Git version controls these files and GitHub hosts them remotely in the cloud.¹⁵

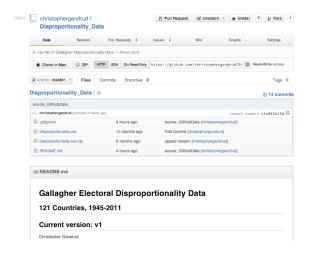
With a GitHub account you can store repositories for free on GitHub if they are in what GitHub calls "public repositories". Anyone can see the contents of public repositories, including all previous versions. ¹⁶ Users can have an unlimited number of public repositories. There is a soft storage size limit of one gigabyte per repository. This should be more than enough for most social science data sets if they are stored in plain text formats. Each repository has a webpage. Here is a portion of the electoral disproportionality repository's main page:

¹³This can create problems if multiple authors are making changes to the same files at the same time. For a discussion of file conflicts see Fredrickson, Testa and Weidmann (2011).

 $^{^{14}\}mathrm{It}$ is possible to build a custom version of Dataverse with some R integration.

¹⁵Specifically, GitHub files are stored on Rackspace (http://www.rackspace.com/).

¹⁶Private repositories are available allowing only official contributors to see their contents. Private repos require a paid account.



Git is a very powerful version control system, especially for text files. For these types of files, Git only saves the actual changes when you (1) add a file to the repository and then (2) commit a version of the file to the repository. This is different from the other cloud storage services we have discussed. They save the whole file for each version. Each "commit" in a repo is given a unique SHA-1 number identifying it. The author of the commit is also saved. Once you commit changes on your computer you can (3) push them to the remote GitHub version of the repo. A repo's GitHub website allows you to view all of the changes that have been made to it. You can browse all versions of the text files, view highlighted changes, and comment on these changes. For example:



To see changes made to a file click on the History button located on the file's GitHub webpage.



This will show the file's entire commit history. Click an individual commit's description to see and comment on the changes.

Contributor analytics An important component of version control on GitHub for us here is that because Git only commits changes and uniquely identifies the changer, it is possible to properly attribute every individuals' contribution to a data set. On the GitHub website you can see every change that was made to each line of a file and who made the change. Git calls this "blame". To view who last changed each line of a file click on the Blame button on its GitHub page next to History.

GitHub has graphical capabilities for organizing and displaying this data. One way to use these is by clicking on the **Graphs** button located near the top of each repository. This will give you access to graphs such as the ones below (I've blurred the contributors' names):



Binary files I mentioned that Git treats non-plain text-binary-files differently. The DTA file format Stata uses to store data and PDFs, are examples of binary file types. Rather than committing specific changes, a new version of the entire binary file is stored with each commit if they are changed. If the files are very very large this could take up a lot of storage space.¹⁷ This is also a problem for all of the other data storage methods with version control discussed here. For very large binary files you will need to use a totally different type of cloud storage system like Rackspace or Amazon's S3. ¹⁸

If you absolutely must have binary files in a repository one solution to the space constraints problem is to have Git ignore it. You do this by including a text file called .gitignore in the repository. In .gitignore simply type the binary file's name. Git will not version control it. This unfortunately also means that the file will not be pushed to GitHub.

Describe the data set with README files Each folder in a GitHub repository can contain a file

¹⁷Though it is important to note that previous commits are compressed, so the amount of storage space a committed binary file take up is less than the size of the original.

¹⁸http://aws.amazon.com/s3/

called README.¹⁹ README files for data sets can contain information about sources, variable descriptions, and so on. GitHub automatically displays the README file on the repository's website in full. If it is written in the Git Flavored Markdown²⁰ markup language and called README.md it will also be automatically formatted. You can see part of the disproportionality data set's README.md above.

Tagging versions Git can tag specific commits. Tags function as bookmarks for major versions of a Git repository. They are particularly useful for demarcating the version of a data set used in a particular publication, for example. Creating tags is simple. Imagine we want to tag the second major version of a data set, the one we used for a publication:

```
git tag -a v2 -m 'Citation Information'
```

The option -a means add, v2 is the version number, and -m adds a message, in this case the publication's citation information. Then simply push the tags to GitHub:

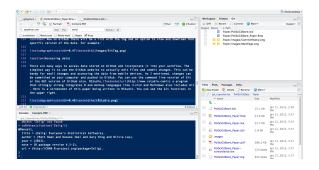
Now on GitHub there will be a list including the tag and an option to view and download this specific version of the data. For example:



3.2 Accessing data

There are many ways to access data stored on GitHub and incorporate it into your workflow. The simplest way is to use the GitHub website to actually edit files and commit changes. This can be handy for small changes, especially from mobile devices. As I mentioned, changes can also be committed on your computer and pushed to GitHub. You can use the command line version of Git or the GUI version of GitHub to push changes. RStudio, ²¹ a program that integrates R and mark-up languages like IATEX and Markdown, also includes Git and the ability to push

a repo to GitHub. So it is possible to make changes to a repository, commit them and push them to GitHub all within the same program. Here is a screenshot of this paper being written in RStudio. The Git functions are in the upper right pane.



Cloning a repo Repositories can be downloaded in full. This is called cloning. You can do this by clicking the Clone in . . . button on the GitHub repositories' website. It can also be done in the command line using the repository's address. For example, the disproportionality data's clone-able address is: https://github.com/christophergandrud/Disproportionality_Data.git.

git clone https://github.com/christophergandrud/
 Disproportionality_Data.git

Note that in real life the address needs to be on the same line. Also, before cloning a repository remember to change the working directory to the folder where you want to save it. In the command line use the cd command to do this.

Once you have cloned the repository you can role back to any previous version with the checkout command. For example if you wanted to roll back to a tag called "v2", type:

git checkout v2

If you have permission to make changes to the repository (see below) you can then push an changes you make back to GitHub.

Access data directly from R Loading data stored on GitHub into R for use in statistical analysis is very easy.²² I have created a function that loads plain text formatted data from GitHub into

 $^{^{19} {\}rm Actually}$ each folder in a repository can contain one as well.

²⁰http://github.github.com/github-flavored-markdown/

²¹http://www.rstudio.com/

 $^{^{22}\}mathrm{You}$ can of course also load data stored on the version of the repo located on your computer.

an R data frame ready for analysis.²³ It's called source_GitHubData and is stored in a GitHub Gist²⁴ at: https://gist.github.com/4466237. It can be loaded into R using the source_gist command from the *devtools* package (Wickham and Chang, 2012).

```
# Load source_GitHubData
# The functions' gist ID is 4466237
devtools::source_gist("4466237")
```

The main argument in <code>source_GitHubData</code> is <code>url.</code> You set this as the URL for the <code>raw</code> version of plain text data file you want to download. The raw version is the version with only the text file. You find this URL by clicking the <code>Raw</code> button on the file's GitHub page. It's next to the <code>Blame</code> button we saw earlier. If you use the URL for the most recent commit of the file, you will always download the most recent version whenever you use <code>source_GitHubData</code>. The raw URLs for each commit and tagged version of the data are also accessible if you want to use a particular version.

The URL for the raw version of the electoral disproportionality data is http://bit.ly/Ss6zDO.²⁵ To download the data using source_GitHubData type:

Note that by default source_GitHubData loads CSV data. You can add the argument sep = "\t" for tab-separated (TSV) data files. You can also specify header = TRUE (default) or header = FALSE.

Citing GitHub data When a researcher uses data they accessed via GitHub how can they cite it? A common practice is to cite the publication the data set was originally used in. However, this is incomplete in at least two ways. First, the version of the data used in the original publication may be different from that used later on. Second, it would be difficult to use this citation to acknowledge contributions made by contributors who did not work on the original data set, but contributed to later versions. One solution is

to use the standard set by Altman and King (2007).²⁶ They propose that data set citations have:

- the author's name,
- the date,
- the data set's title.
- a unique global identifier (UGI),
- a universal numeric fingerprint (UNF),
- a bridge service.

The first three are self explanatory and shared with standard citations for other types of materials. Examples of UGI include Document Object Identifiers (DOI) and the Handel System.²⁷ They uniquely identify the data set. UNF's uniquely number a particular version of the data set. Finally a bridge service allows you to use the DOI and UNF to link to the actual data set. If it is stored on GitHub then it is possible to see every person who has contributed to it.

We can easily use Altman and King's citation standards for GitHub repositories. I would suggest citing specific tags of the data set. If these are not available you can cite the specific commit you used to produce results with.

Showcase a data set with GitHub Pages Each public repository's GitHub website allows anyone full access to the data. However, these pages may be confusing for those without experience using a version control system. To solve this problem GitHub Pages²⁸ allows you to very easily create a simple webpage for the repository. These pages can be used to describe the data and include links to download the entire repository. You can of course also add links to specific files.

To create a page navigate to the repository's normal webpage. Then click Settings \rightarrow Automatic Page Generator. By default it will load the README file as the content of the new page. You can change this, add a Google Analytics tracking ID²⁹ to gather information on who visits the page, and choose an aesthetic style before publishing it. Here is a sample of the disproportionality data's page:

²³The function is based on *devtools*' source_url command.

²⁴Gists host code snippet. See: https://gist.github.com/.

²⁵I used bitly (http://bitly.com/) to shorten the URL so that it would fit on the page more easily. The full URL is: https://raw.github.com/christophergandrud/Disproportionality_Data/master/Disproportionality.csv.

²⁶Dataverse uses a version of this standard.

²⁷ http://www.handle.net/

²⁸http://pages.github.com/

²⁹http://www.google.com/analytics/



3.3 Collaboration

Compared to the status quo social science data storage methods, GitHub is uniquely strong for enabling and incentivizing collaboration both between co-authors and non-co-authors who may be able to help develop and verify the data set.

Co-authors Public repositories can have an unlimited number of what GitHub calls "collaborators". Collaborators have permission to push changes to the repository. There is one important practical issue to note. If multiple collaborators are actively working on a data set, they need to add an extra step to their GitHub commit process. Lach person needs to pull their collaborator's changes and resolve any merge conflicts before pushing committed changes to GitHub. Here is an example:

```
# Add new files to Git
git add .

# Commit the changes
git commit -a -m "A message"

# Pull collaborator's commits
```

git pull

Push changes to GitHub
git push origin master

Collaborating with non-Coauthors: Pull requests People who want to make changes to a repository, but are not collaborators can make "pull requests". All they need are GitHub accounts. Pull requests are specific changes that non-collaborators suggest. Requesters can add comments about why they are suggesting a change. GitHub also has a discussion forum where anyone can discuss these comments.

After a pull request has been made it is up to the repository's collaborators to decide whether or not to accept the requested changes. If a collaborator accepts the changes, they are made instantly and a full record of who made the changes is kept.

There are two ways to make a pull request. If someone notices a small error—e.g. a misspelling, a misclassification—or other improvement that they think should be made to a data set they can make a pull request by navigating to the GitHub page for the specific file they think needs to be changed. Then they click the Edit button located next to the Raw button (see above). Clicking this button "forks" the repository, i.e. gives them a copy they can change. They will also see a window like this:



In this window they can make their proposed changes and add a comment about what the changes are before clicking Propose File Change.

 $^{^{30}\}mathrm{On}$ the repositories' GitHub site click Settings $\rightarrow\!\mathtt{Collaborators}$ and add the co-authors GitHub username.

 $^{^{31}{\}rm This}$ is also true if you make changes to both the GitHub version of the repository and the copy on your computer.

For longer file changes, e.g. a major addition to the data set to bring it up-to-date, it is better to work with the forked repository rather than in the Edit window. To directly fork a data set's repository click the Fork button on the upper right corner of its webpage. You can then make changes to the forked repository as if it were your own. When you are ready to suggest the changes be included, you click the Pull Request button at the top center of the forked repository's GitHub page. For more information on forking and pull requests see the GitHub article on the topic: https://help.github.com/articles/using-pull-requests.

Issues GitHub repositories also have an "Issues" area that allows any other GitHub user to make a comment on the repo. These tend to include general suggests for how to improve the files or questions about the repo that may be of general interest. Anyone can respond in the Issues area creating a discussion thread.

Repo Wiki GitHub provides a very easy tool for creating nicely formatted repository wiki's. For example, a data set repo's wiki could include short articles with details about how the data set was created. Non-official collaborators can add to repo wikis. Like pull requests, these changes are subject to approval by one of the repository's collaborators.

Follow a repository If you are a GitHub member you can follow a public repository, even if you aren't a collaborator, This gives you a Facebook-style newsfeed³² showing any updates made to the repository as well as discussions in the Issues area and pull requests.

Passing on the data set Because of changing time commitments, professional interests, and so on, no one can maintain and update a data set forever. GitHub makes it easy to pass on control of a data set while maintaining it's entire version history. Simply add the new data set maintainer as a collaborator. They will largely have the same privileges to manage the repository as you did. To completely transfer control simply have the new maintainer fork the repository and leave a link to the forked version at the previous location on GitHub.

32See: https://www.facebook.com/help/

Selective incentives Why would people actively contribute to improving publicly available data sets, especially data sets primarily associated with other authors? What's in it for them?

GitHub not only provides technology—particularly pull requests—for social data development and verification, but it also gives Olsonian selective incentives that can motivate people to do so (Olson, 1965). By keeping track of who made what change and providing numerous ways to quantify and visualize each member's contributions it provides strong selective incentives to collaborate. Perhaps one day social scientists could bring GitHub's descriptive statistics of their contributions to hiring and promotion committees.

4 Conclusion

In this article I have tried to show that GitHub is a very good option for storing social science data in the cloud. Admittedly it has more of a learning curve than the incumbent options. However, I hope to have demonstrated that it is worth learning. It is at least as good as the alternatives in terms of stable storage, access, version control, and collaboration. In addition it is far better at enabling social data development and verification. By using a service like GitHub that both makes collaboration easy and provides selective incentives to do so we may improve the social science community's ability to utilize its collective knowledge to create more complete and robust data sets. Better data sets will allow us to better answer our research questions.

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