# Cookiecutter Data Science - VC Edition Documentation

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Victor Calderon <victor.calderon@vanderbilt.edu>

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Part I

## Author

- Victor Calderon (victor.calderon@vanderbilt.edu)
- Driven Data Team

## Part II

# Description

### A logical, reasonably standardized, but flexible project structure for doing and sharing data science work.

#### -Data Science Team

Making your project **reproducible** is one of the key elements when doing research in pretty much any field, including Data Science. This package aims at making it easier to be transparent about what *steps* and *commands* are used when producing tangible results for a research project.

In the following sections, I will provide instructions on how to to use this project tempalte, as well as how to make the most out of this template.

This documentation is part of the repository cookiecutter-data-science-vc, and has been adapated from the Cookiecutter Data Science Project template by Driven Data organization.

## Part III

## Contents

### CHAPTER 1

### **Getting Started**

Author: Victor Calderon (victor.calderon@vanderbilt.edu)

Description: A logical, reasonably standardized, but flexible project structure for doing and sharing data science work.

### **1.1 Structuring your Project**

#### Author: Victor Calderon (victor.calderon@vanderbilt.edu)

Description: A logical, reasonably standardized, but flexible project structure for doing and sharing data science work.

Now that your have a *working* version of **python** on your computer, you can start doing research.

One of the key elements of a project is for it to be **reproducible** by others. Having this in mind when you're structuring your project will allow others to look at your code, understand it well enough to be able to **recreate** your results.

This is a short guide on 2 ways to structure your code, without having to do much of creating documets, etc.

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#### 1.1.1 Cookiecutter and Folder structure

Cookiecutter is a command-line utility that creates projects from cookiecutters (project templates), e.g. Python package projects, LaTeX documents, etc.

Cookiecutter has been widely used for many projects, and each team and organization can create their own *template*. For more information, visit the cookiecutter documentation.

As the famous say goes:

Don't reinvent the wheel!

You can always create your own folder and file structures, and organize your documents the old-fashioned way. The problem with this is that it may **vary** from project to project, and it will be more difficult to be consistent and effective throught your projects.

For this reason, I rely on cookiecutter templates to create the file and folder structure of a project.

There are many different cookiecutter templates out there, but after trying to find the best one that suits my needs in **research** and **programming**, I found one that works great! And after some modifications, I came up with a *version* of this template.

These two templates are shown in *Data Science - Cookiecutter* and *Personal version - Cookiecutter*. But first, let's make sure you have cookiecutter installed correctly.

#### 1.1.2 Requirements to use cookiecutter templates

The minimum rquirements for creating cookiecutter templates are:

- Python 2.7 or 3.5
- Cookiecutter Python package >= 1.4.0: This can be installed with pip or conda depending on how you manage your Python packages.

You can install it by typing this on the terminal

pip install cookiecutter

or via Anaconda:

```
conda config --add channels conda-forge
conda install cookiecutter
```

Now you can use cookiecutter to create new templates for projects and papers!

#### 1.1.3 Data Science - Cookiecutter

Cookiecutter Data Science is best described as

A logical, reasonably standardized, but flexible project structure for doing and sharing data science work.

This folder structure allows everyone looking at your code to understand it right away. It also provides many different functions (as part of a Makefile) that simplify the workflow of your project.

In a nutshell, this cookiecutter includes:

- A **Makefile** file with **\*\***useful functions.
- Documentation to make your project easily accessible and readable
- And more!

In order to use this template, you follow the documentation in Cookiecutter Data Science.

#### 1.1.4 Personal version - Cookiecutter

If you need more than the *normal* Data Science Cookiecutter template, you can use my version. Some of the differences are:

- It includes and easy-to-use environment.yml file that makes it easy to install dependencies.
- Extra functions in the Makefile.
- Choice of what kind of documenation to use. One has the option choose from *traditional* Read The Docs style or the Astropy Sphinx Theme.

You can check how these two styles look like:

- - Read The Docs Version
- - Astropy Version

Next, you can create your own Project based on this cookiecutter version

#### To start a new project and prompts

To start a new project, type the following:

\$ cookiecutter https://github.com/vcalderon2009/cookiecutter-data-science

If you want the **default** project scheme from *DrivenData* (see above), run:

cookiecutter https://github.com/drivendata/cookiecutter-data-science

Depending on what kind of folder structure you want, you might want to choose from the different types.

After running this command, **you will be prompted some questions** regarding the parameters for the project. This will prompt you to answer a few questions like:

Question	Description
project_name	Name of the project. This can be similar to one on
	Github.
	Examples:
	SDSS_analysis
	Lung_cancer_analysis
repo_name	Name of the directory/repository, the project will be
	saved. This field should not contain spaces should not
	contain spaces
	Examples:
	Calderon_Victor_Astro_PhD_Thesis
	Szewciw_Adam_Astro_PhD_Thesis
author_name	Author's first name. It can include spaces
	Examples:
	Adam Sanchez
	Rose Roserberg
	· Rose Roserberg
author_email	Author's email address. Examples:
	• some_email@gmail.com
	_ 0
	• another_email@yahoo.com
short_description	A short description of the project. This can be a brie
	overview of the project.
	Examples:
	<ul> <li>Repository for lung cancer analysis</li> </ul>
	Analysis on galaxies and cosmology
long_description	A longer version of short_description
	Examples:
	<ul> <li>Repository for lung cancer analysis</li> </ul>
	Analysis on galaxies and cosmology
open_source_license	Type of License for the paper. Without this, one cannot
	use any of.
	Options:
	• MIT
	BSD 3-Clause
	• GNU GPL v3+
	Apache Software Licence 2.0
	• BSD 2-Clause*
s3_bucket	Path to AWS storage. This is <b>temporarily</b> disabled!!
aws_profile	AWS profile name. This is <b>temporarily</b> disabled!!
conda_python_env	Name of the project's anaconda environment. In order
	to use the packages of this project, you need to first
	activate this environment.
	Examples:
	<ul> <li>lung_cancer_env</li> </ul>
	SDSS_galaxies_env
github_username	Author's Github username. This will be use tolink the
_	project to the Github repository.
18	Examples: Chapter 1. Getting Starte
	• username
	• username2018

#### Using the Template

Now that one has answered the questions from *To start a new project and prompts*, you just need to populate the project with scripts, notebooks, and **of course**, documentation!!

The structure of the finalized project can be found in the Project Structure section.

#### Editing your environment

Now that you have a working proect from **cookiecutter**, you can start by editing the *environment* of your project.

If you downloaded **my version of cookiecutter**, you should be able to edit the environment.yml file. This file states which packages need to be installed by Anaconda and pip in order to run the scripts of the package.

The environment.yml file looks like the following:

```
name: name_of_environment
channels:
  - defaults
dependencies:
  - python>=3.6
  - ipython
  - anaconda
  - astropy
  - h5py
  - numpy
  - pandas
  - scipy
  - seaborn
  - pip
  - pip:
    - GitPython
    - progressbar2
```

You can edit the environment.yml file to include/exclude packages needed by your project.

After having edited the list of packages needed by your project, you can execute the command

\$ make environment

#### to create the environment.

If you have done this step before, and you want to update the environment, you need to run

\$ make update\_environment

instead.

#### 1.1.5 Adding your Project repository to Github

If you follow the instructions from above, you should have

- Downloaded the repository
- Created your own project with the desired file and folder structure
- · Created your working environment for you project

The next step is to add it to Github and make it accessible.

To do this, your should do the following:

- 1. Create a Github repository with the same name as the repository.
- 2. Type git add remote origin git@github.com:<username>/<project\_name>.git. In here you need to **replace** <username> and project\_name with your details.
- 3. git push origin master This will push your project to Github.

To check that you did this correctly, type

git remote -v

and you should get something that looks like this:

origin https://github.com/<username>/<project\_name>.git (fetch)
origin https://github.com/<username>/<project\_name>.git (push)

where username and project\_name pertain to your repository on Github.

Now all of the files are online on Github, and should be ready to integrate them with Read The Docs.

#### 1.1.6 Documentation for your new project

Now that you have both a working local and online copy of your code, the next step is to create the documentation for the project.

For this, you can easily use Read The Docs (RTD).

You need to do the following:

- Create an account on "Read the Docs"
- Go to your Profile and select My Projects
- From there, you should import the repository *manually* (it's easier). Click on Import a Project and follow the instructions.
- You should add the project with the **same name** as the Github Repo if possible. Otherwise, you might need to **change** the links to the *badges* on the README.md files in the project, among others.
- Make sure that the repository was correctly built by looking at the Builds and see that it compiled correctly. If not, it should tell you if there was an error and what the error was.
- Now you go and change the documentation depending on the project's needs.

#### 1.1.7 Continuous Integration for your Project

**Continuous integration** deals with testing your code for possible errors, and making sure that everything is working as expected. Depending on your project's needs.

This template includes a .travis.yml, which the files used by Travis CI. Travis CI is a *Continuous integration* platform for testing your code, and checking the functionality of your project.

Project based on the modified version of the cookiecutter data science project template.

Project based on the modified version of the cookiecutter data science project template.

### CHAPTER 2

### Commands

The Makefile contains the central entry points for common tasks related to this project.

This section is dedicated towards the functions used through the analysis.

Project based on the modified version of the cookiecutter data science project template.

## chapter $\mathbf{3}$

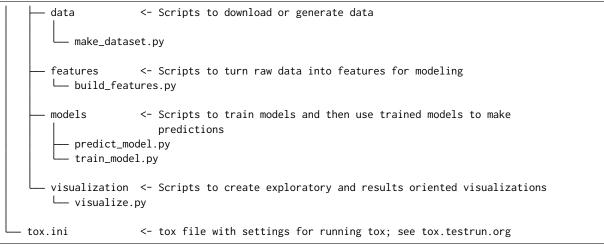
### **Project Structure**

The organization of the project is the following:

LICENSE	
— Makefile	<- Makefile with commands like `make data` or `make train`
README.md	<- The top-level README for developers using this project.
— data	
external	<- Data from third party sources.
interim	<- Intermediate data that has been transformed.
processed	<- The final, canonical data sets for modeling.
raw	<- The original, immutable data dump.
— docs	<- A default Sphinx project; see sphinx-doc.org for details
— models	<- Trained and serialized models, model predictions, or model summaries
	, , ,
— notebooks	<- Jupyter notebooks. Naming convention is a number (for ordering),
	the creator's initials, and a short `-` delimited description, e.g.
	`1.0-jqp-initial-data-exploration`.
- references	<- Data dictionaries, manuals, and all other explanatory materials.
	······································
— reports	<- Generated analysis as HTML, PDF, LaTeX, etc.
└── figures	<- Generated graphics and figures to be used in reporting
5	
- requirements.txt	<- The requirements file for reproducing the analysis environment, e.g.
	generated with `pip freeze > requirements.txt`
environment.yml	<- The Anaconda environment requirements file for reproducing the analysis
→environment.	
	This file is used by Anaconda to create the project environment.
— src	<- Source code for use in this project.
initpv	<- Makes src a Python module

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Project based on the modified version of the cookiecutter data science project template.

Project based on the modified version of the cookiecutter data science project template.