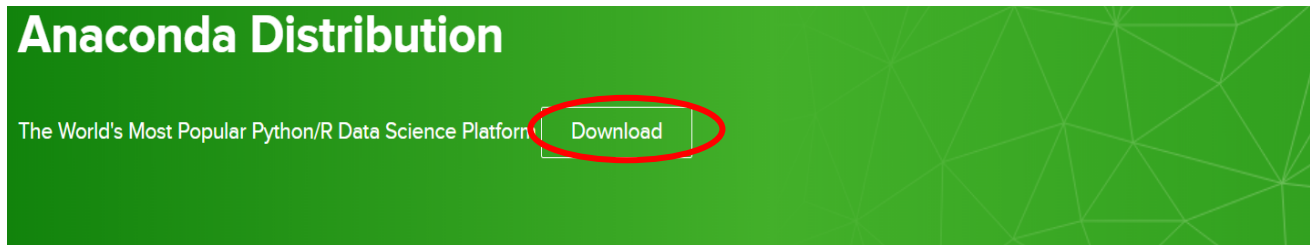


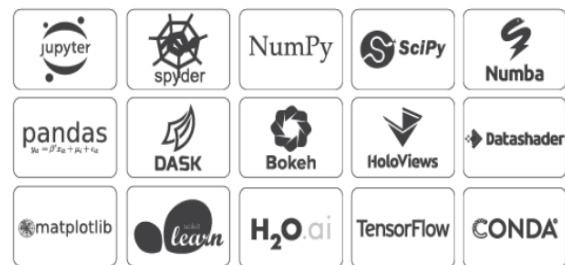
Getting started with Anaconda and Jupyter Notebooks

1. Go to Anaconda website: <https://www.anaconda.com/distribution/>
Download Anaconda Python

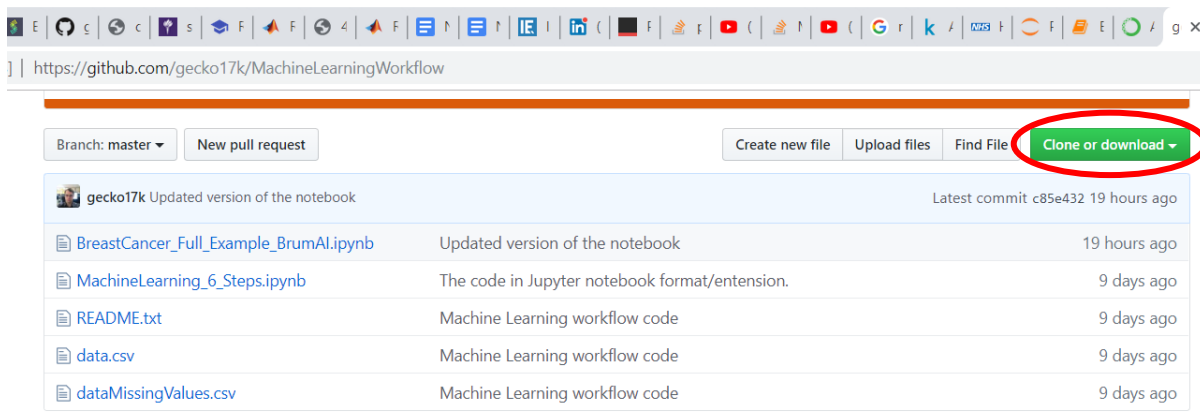


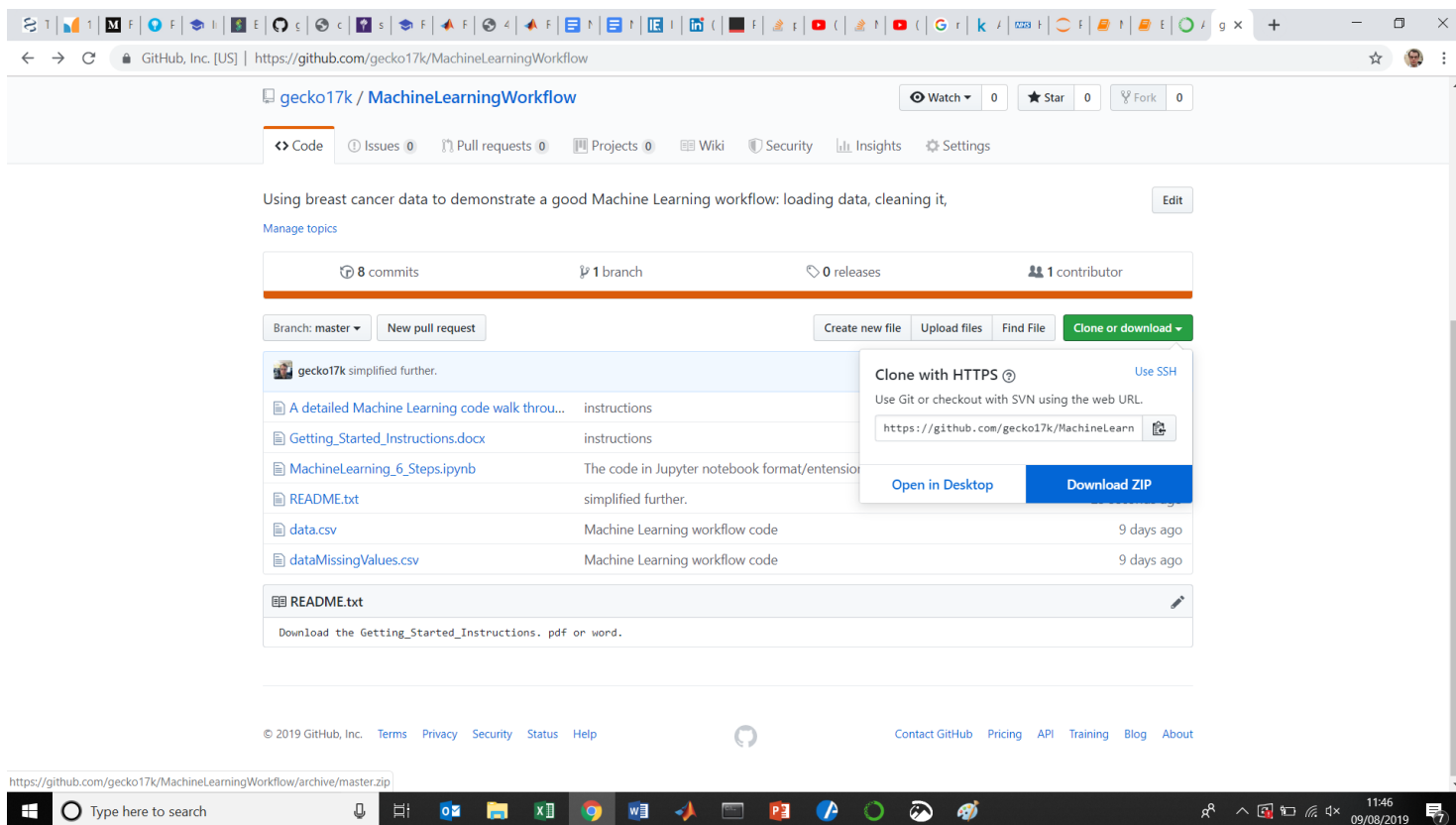
The open-source [Anaconda Distribution](#) is the easiest way to perform Python/R data science and machine learning on Linux, Windows, and Mac OS X. With over 15 million users worldwide, it is the industry standard for developing, testing, and training on a single machine, enabling *individual data scientists* to:

- Quickly download 1,500+ Python/R data science packages
- Manage libraries, dependencies, and environments with [Conda](#)
- Develop and train machine learning and deep learning models with [scikit-learn](#), [TensorFlow](#), and [Theano](#)
- Analyze data with scalability and performance with [Dask](#), [NumPy](#), [pandas](#), and [Numba](#)
- Visualize results with [Matplotlib](#), [Bokeh](#), [Datashader](#), and [HoloViews](#)

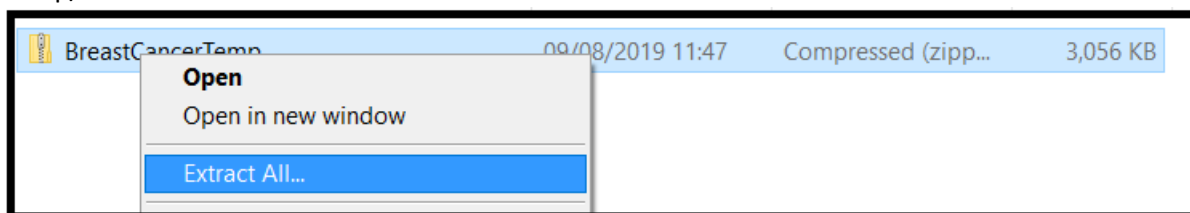


1. Go to GitHub, get the repo: <https://github.com/gecko17k/MachineLearningWorkflow>
Clone or download the files shown below, easiest to get the ZIP file.



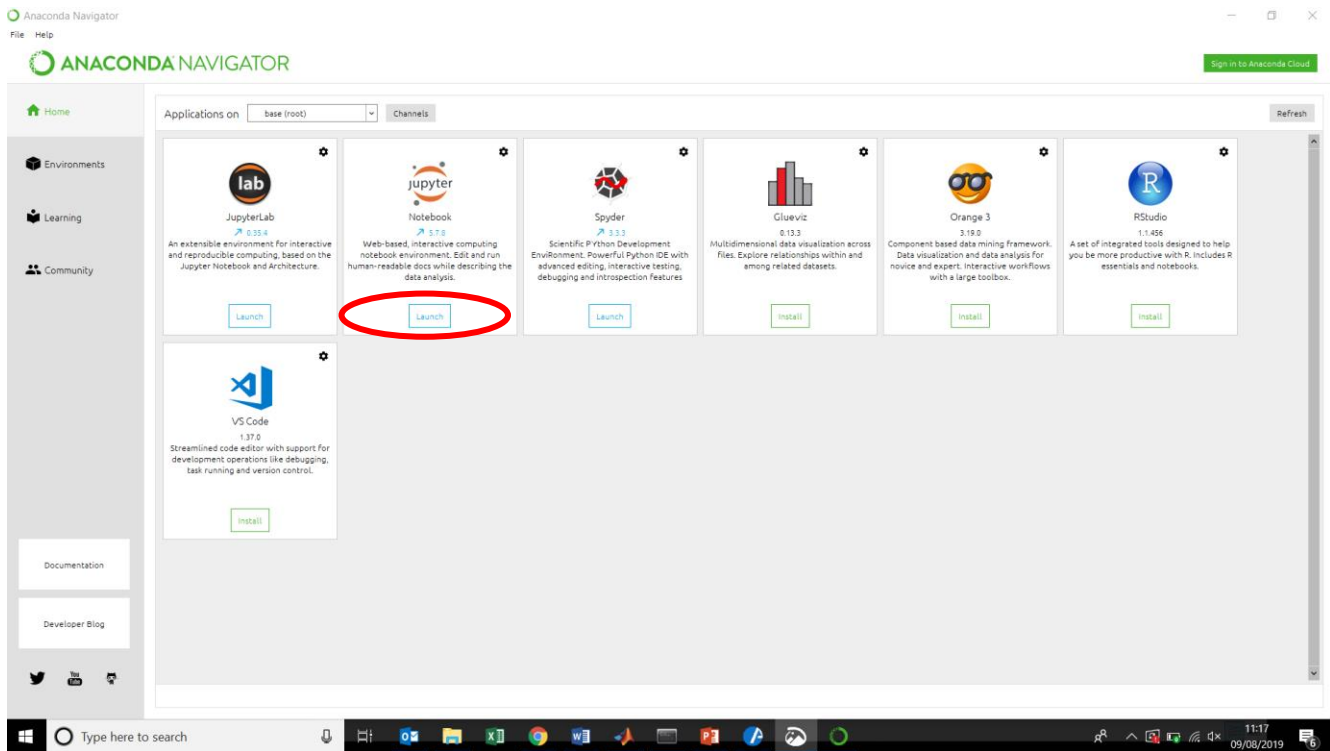


2. Unzip/Extract

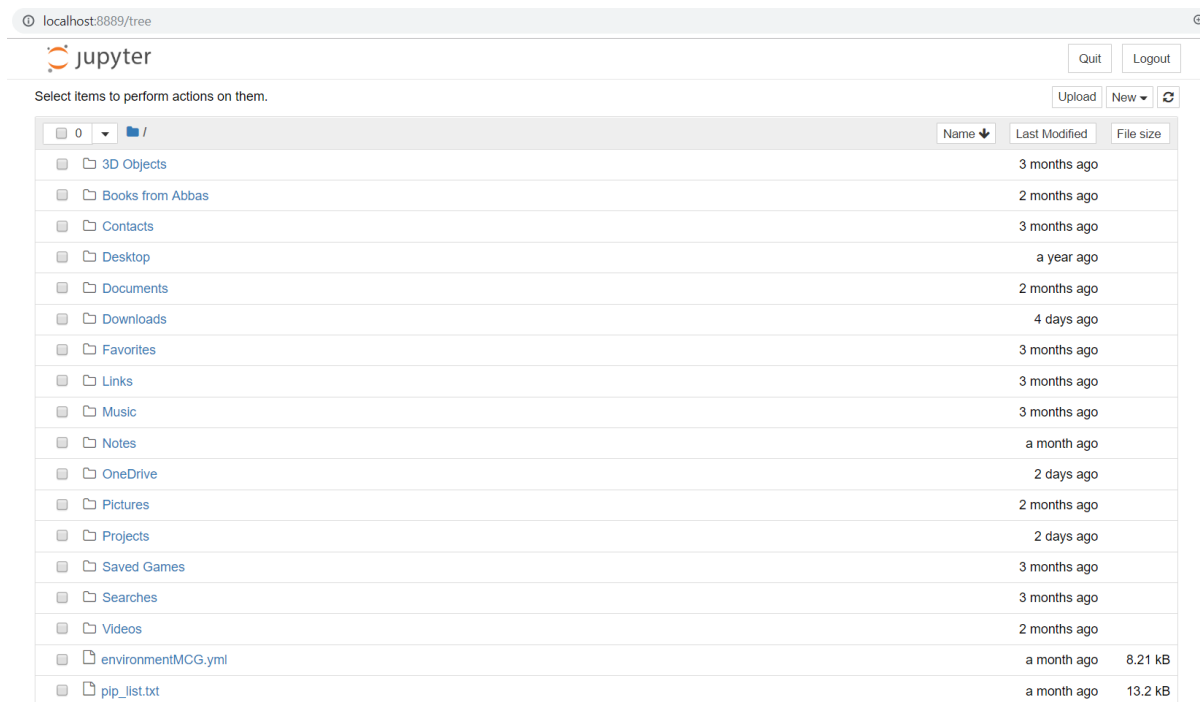


Once you unzip, the relevant folder will be called "MachineLearningWorkflow-master".

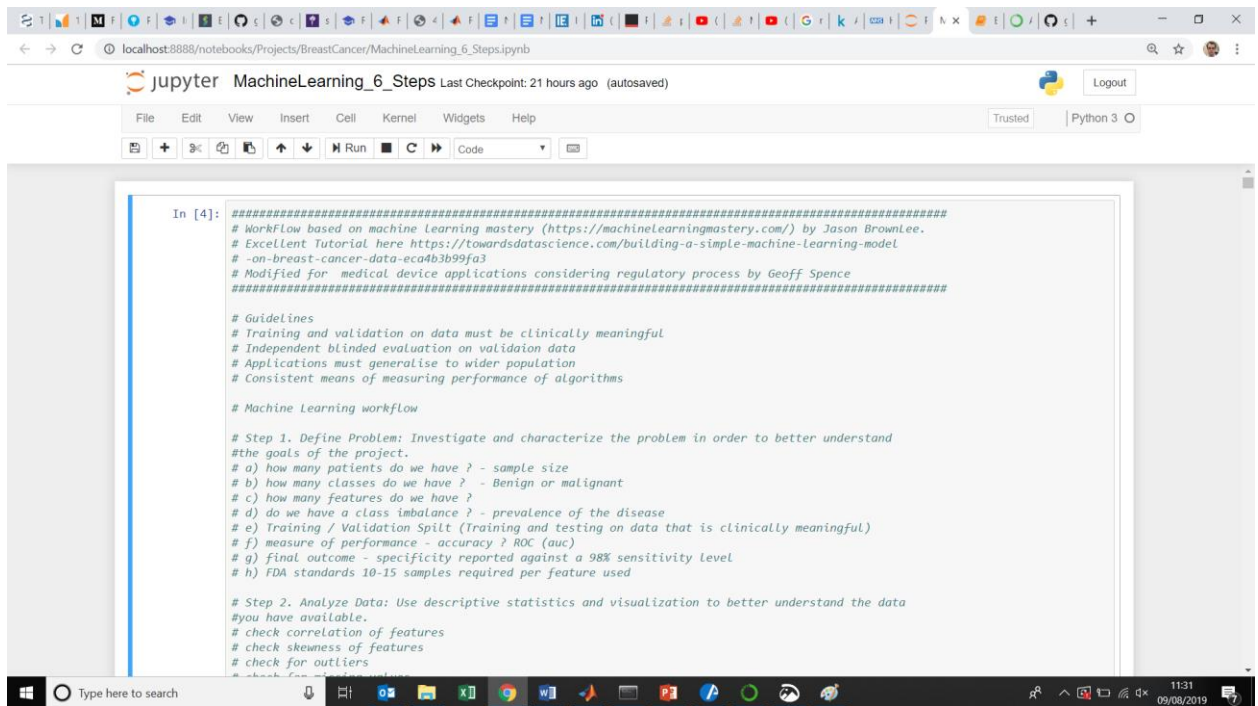
3. Place in a single folder on your laptop, e.g. C:\Users\Name\Projects\BreastCancer or a folder called "BreastCancer".
4. Open Anaconda Navigator
Launch Jupyter Notebook



5. Jupyter opens in your browser.



6. Navigate to where you put the files from GitHub and open up “-MachineLearning_6_Steps.ipynb”.



The screenshot shows a Jupyter Notebook titled "MachineLearning_6_Steps" running on a local host. The code cell contains the following text:

```
In [4]: #####  
# Workflow based on machine learning mastery (https://machinelearningmastery.com/) by Jason Brownlee.  
# ExcelLent Tutorial here https://towardsdatascience.com/building-a-simple-machine-learning-model  
# -on-breast-cancer-data-eca4b3b99fa3  
# Modified for medical device applications considering regulatory process by Geoff Spence  
#####  
  
# Guidelines  
# Training and validation on data must be clinically meaningful  
# Independent blinded evaluation on validation data  
# Applications must generalise to wider population  
# Consistent means of measuring performance of algorithms  
  
# Machine Learning workflow  
  
# Step 1. Define Problem: Investigate and characterize the problem in order to better understand  
# the goals of the project.  
# a) how many patients do we have ? - sample size  
# b) how many classes do we have ? - Benign or malignant  
# c) how many features do we have ?  
# d) do we have a class imbalance ? - prevalence of the disease  
# e) Training / Validation Split (Training and testing on data that is clinically meaningful)  
# f) measure of performance - accuracy ? ROC (auc)  
# g) final outcome - specificity reported against a 98% sensitivity level  
# h) FDA standards 10-15 samples required per feature used  
  
# Step 2. Analyze Data: Use descriptive statistics and visualization to better understand the data  
# you have available.  
# check correlation of features  
# check skewness of features  
# check for outliers
```

We are now ready to start.