## Interactive

September 4, 2019

Data Science Workflow < | > Hall of Fame

# 1 Exploring Subsets Interactively

In this notebook you can select criteria for a subset to inspect and compare to the entire dataset.

#### **Instructions:**

- 1. Run Cell 1, wait for "DONE IN <time>" message before continuing.
- 2. Run Cell 2, fill out criteria before continuing.
- 3. Run Cell 3, view report.
- 4. Repeat steps 2 and 3 with new criteria, if desired.

#### **Prerequisites:**

- four final CSV file local in ./data\_final
- all aggregations created by aggregate.py local in ./analysis\_data
- ipywidgets
  - pip install ipywidgets
- nodejs
  - conda install nodejs
- npm
  - pip install npm
- labextension
  - for jupyter lab: jupyter labextension install @jupyter-widgets/jupyterlab-manager)
  - for jupyter notebook: notebook extension (jupyter nbextension enable --py
    widgetsnbextension

#### 1.1 Run Cell 1

This takes about 10 minutes. Wait for "DONE IN <time>" message before continuing!

```
[1]: import interactive
  import load_data
  import datetime
  from IPython.core.display import HTML
```

```
# 10 minutes to load data
     start = datetime.datetime.now()
     data_frames = interactive.data()
     end = datetime.datetime.now()
     print('\n'+'-'*80+'\n'+'DONE\ IN\ \{0\}'.format(end - start))
    Notebooks loaded in 0:00:28.177134
    Repos loaded in 0:00:04.412145
    Owners loaded in 0:00:00.694790
    Notebook imports loaded in 0:00:41.461839
    Errors loaded in 0:00:03.000942
    Cell stats loaded in 0:00:01.782402
    Cell order loaded in 0:00:27.241687
    Outputs loaded in 0:00:01.941591
    Statuses loaded in 0:00:00.993714
    Cell stats loaded in 0:00:01.961678
    Collaboration statuses loaded in 0:00:00.030213
    Special functions loaded in 0:00:12.218296
    Framework uses loaded in 0:00:08.789084
    Educational status loaded in 0:00:00.219593
    DONE IN 0:09:19.337819
[4]: | # query = interactive.interactive(data_frames)
```

#### 1.2 Run Cell 3

This takes about 1 minute. View report! Repeat Cells 2 and 3 with different criteria.

```
[3]: data_frames_sub = interactive.subset(data_frames, query)
    print('\n'+'-'*73+'\n')
    interactive.report_comparisons(data_frames_sub, data_frames)
```

Subsetting to Python notebooks pushed between 2011-10-24 and 2019-07-14. Limiting to those in repositories with at least 0 stargazers, 0 watchers, 1 forks, and 0 issues.

Only looking at notebooks created by educational or not educational organizations.

166,903 (4.3%) notebooks fit your criteria.

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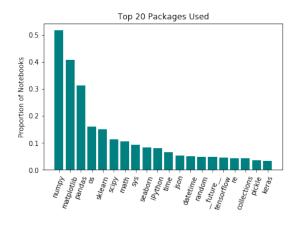
#### 1.2.1 Summary Statistics

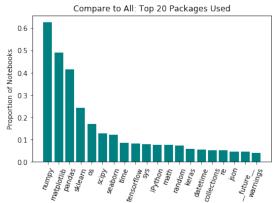
num\_cells forks\_count open\_issues\_count stargazers\_count \

mean	26.83	4	1.41		7.15	58.96
median	18.00		5.00		0.00	4.00
min	0.00		1.00		0.00	0.00
max	573.00	569	8.00		2003.00	12680.00
	subscribers	count	watchers	_count	lines_of_code	num_words
mean		17.64		58.96	129.25	457.38
median		6.00		4.00	78.00	132.00
min		0.00		0.00	1.00	0.00
max	2	2446.00	12	2680.00	12933.00	55725.00

	num_cells	forks_c	ount	open_issue	s_count	starga	zers_count	\
mean	28.76		5.56		0.49		9.88	
median	19.00		0.00		0.00		0.00	
min	0.00		0.00		0.00		0.00	
max	1641.00	1763	4.00		2003.00		22831.00	
	subscriber	s_count	watc	hers_count	lines_o	f_code	num_words	
mean		2.13		9.88		147.26	405.18	
median		1.00		0.00		88.00	60.00	
min		0.00		0.00		1.00	0.00	
max		2446.00		22831.00	462	118.00	200404.00	

## 1.2.2 Package Use

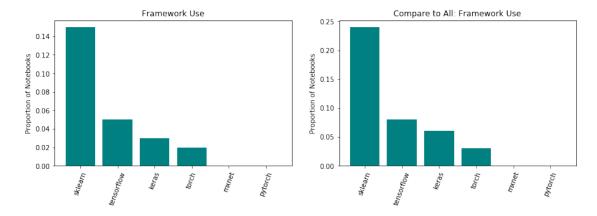




#### 1.2.3 Framework Use

21.69% of these notebooks use at least one framework.

35.48% of all notebooks use at least one framework.

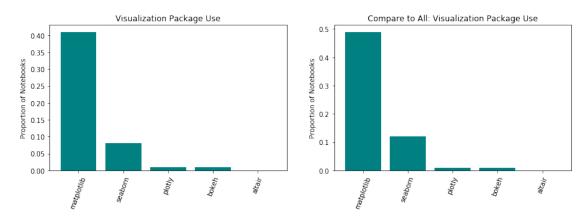


### 1.2.4 Visualization Package Use

42.86% of these notebooks use at least one visualization package.

#### Compare to all:

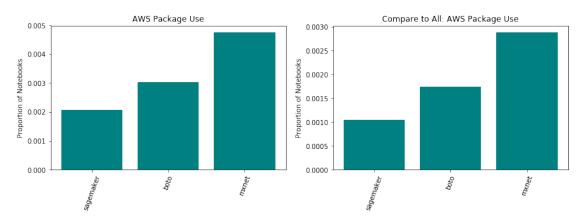
50.71% of all notebooks use at least one visualization package.



### 1.2.5 AWS Package Use

0.806% of these notebooks import at least one of sagemaker, boto, or mxnet.

0.493% of all notebooks import at least one of sagemaker, boto, or mxnet.



### 1.2.6 Number of Errors per Notebook

mean 0.14 median 0.00 min 0.00 max 50.00

Name: num\_errors, dtype: float64

## Compare to all:

 mean
 0.22

 median
 0.00

 min
 0.00

 max
 9104.00

Name: num\_errors, dtype: float64

#### 1.2.7 Ratio of Markdown to Code

mean 13.46
median 1.68
min 0.00
max 18029.00

Name: ratio\_wl, dtype: float64

mean	6.45
median	0.69
min	0.00
max	18029.00

Name: ratio\_wl, dtype: float64

#### 1.2.8 Execution Order

84.1% of these notebooks have cells run in order.

70.85% of these notebooks have at least one output, 77.91% of which are run in order.

85.04% of these notebooks were able to be parsed with Python AST.

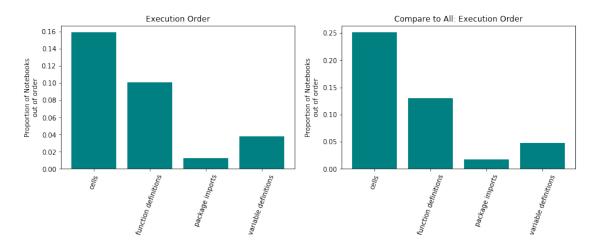
Of these, 10.11% had a function used before it was defined, 1.25% had a package used before it was imported, and 3.83% used a variable before it was defined.

#### Compare to all:

74.92% of all notebooks have cells run in order.

84.4% of all notebooks have at least one output, 70.58% of which are run in order.

86.41% of all notebooks were able to be parsed with Python AST. Of these, 13.0% had a function used before it was defined, 1.8% had a package used before it was imported, and 4.76% used a variable before it was defined.

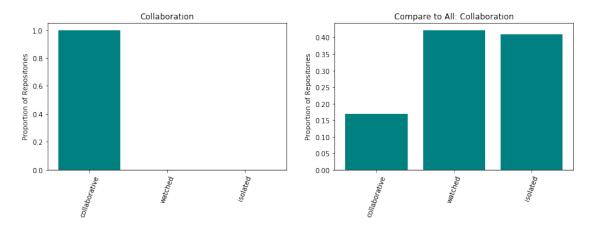


#### 1.2.9 Collaboration

100.0% of these repositories are collaborative, containing 100.0% of these notebooks.

#### Compare to all:

16.85% of all repositories are collaborative, containing 22.76% of all notebooks.



#### 1.2.10 Educational Status

26.9% of these repos are educational, holding 28.35% of these notebooks

### Compare to all:

23.67% of all repos are educational, holding 29.21% of all notebooks Data Science Workflow < | > Hall of Fame