Interactive

September 10, 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 0:00:33.673711
Repos loaded in 0:00:05.437692
Owners loaded in 0:00:00.777930
Notebook imports loaded in 0:00:43.048986
Errors loaded in 0:00:06.050018
Cell stats loaded in 0:00:04.898696
Cell order loaded in 0:00:35.426024
Outputs loaded in 0:00:06.025272
Statuses loaded in 0:00:03.127332
Cell stats loaded in 0:00:01.842215
Collaboration statuses loaded in 0:00:00.071805
Special functions loaded in 0:00:17.884992
Framework uses loaded in 0:00:11.105059
Educational status loaded in 0:00:00.185618
```

DONE IN 0:09:20.538907

```
[]: # 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.

```
[11]: 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. Only looking at notebooks made using Google Colaboratory. 132,162 (3.4%) notebooks fit your criteria.

1.2.1 Summary Statistics

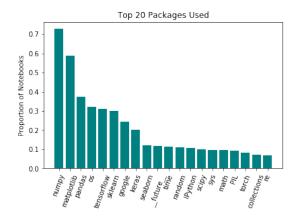
	num_cells	forks_count	open_issues_count	stargazers_count	\
mean	31.95	6.6	0.24	11.68	
median	23.00	0.0	0.00	0.00	
min	0.00	0.0	0.00	0.00	

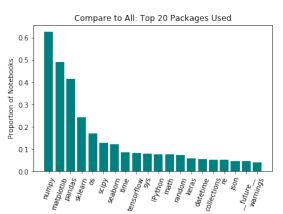
max	1641.00 58	374.0	427.00	18627.00
	subscribers_count	watchers_count	lines_of_code	num_words
mean	1.51	11.68	198.81	379.86
median	0.00	0.00	133.00	70.00
min	0.00	0.00	1.00	0.00
max	1375.00	18627.00	37455.00	25923.00

Compare to all:

	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



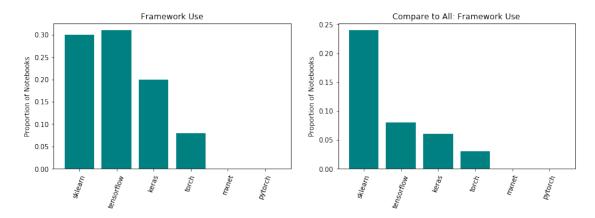


1.2.3 Framework Use

67.95% of these notebooks use at least one framework.

Compare to all:

35.48% of all notebooks use at least one framework.

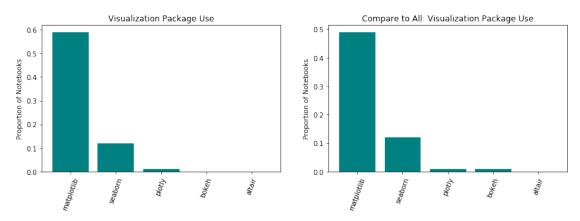


1.2.4 Visualization Package Use

59.7% 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 Number of Errors per Notebook

mean	0.19
median	0.00
min	0.00
max	52.00

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

1.2.6 Ratio of Markdown to Code

mean 4.74 median 0.66 min 0.00 max 8400.00

Name: ratio_wl, dtype: float64

Compare to all:

mean 6.45
median 0.69
min 0.00
max 18029.00

Name: ratio_wl, dtype: float64

1.2.7 Execution Order

57.86% of these notebooks have cells run in order.

84.85% of these notebooks have at least one output, 50.48% of which are run in order.

80.81% of these notebooks were able to be parsed with Python AST.

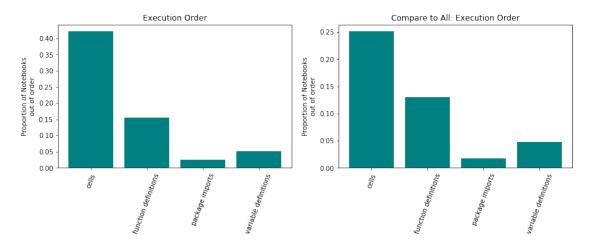
Of these, 15.55% had a function used before it was defined, 2.59% had a package used before it was imported, and 5.1% used a variable before it was defined.

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.

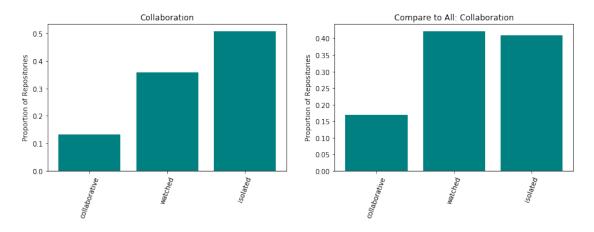


1.2.8 Collaboration

13.08% of these repositories are collaborative, containing 17.56% of these notebooks.

Compare to all:

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



1.2.9 Educational Status

22.15% of these repos are educational, holding 26.26% 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