



# Unveiling Experts in Data Science for Team Composition: Insights from Mining Software Repositories

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## Characterize Experts in Data Science

- □ Explore the activity of software repositories
  - Identify experts in data science and programming, that we call data science programmers.
- □ Identify their roles in team composition

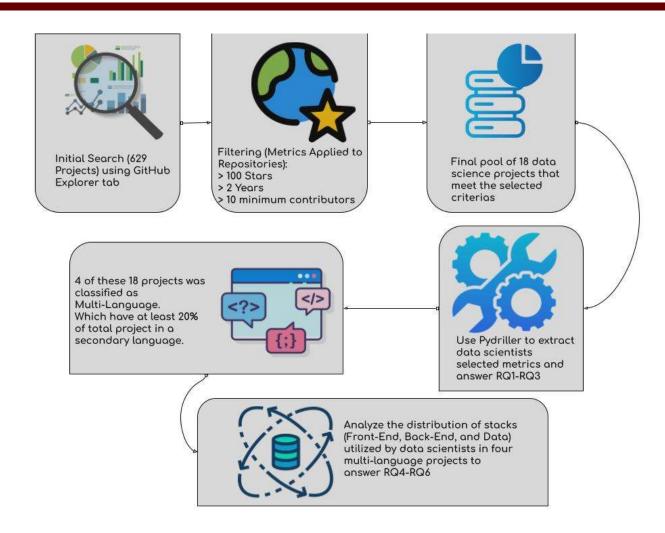
## Research Questions

- □ RQ1: What information about data science programmers do commit-based metrics provide?
- □ RQ2: What are the characteristics of experts in data science software projects?
- □ RQ3: Which programming languages are mostly used by data science programmers?

## Research Questions

- □ RQ4: What is the general composition of project teams?
- □ RQ5: What are the roles and responsibilities of experts within the project teams?
- □ RQ6: What are the defining characteristics of the project teams?

## Steps for Criteria Evaluation



## Used GitHub Repositories

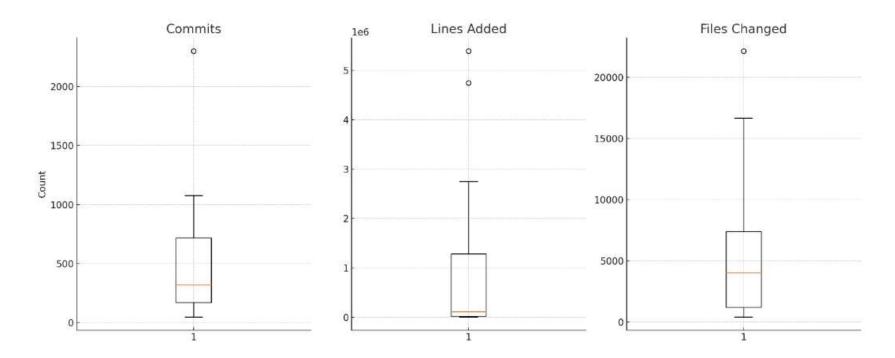
Repository	Stars	Contributors	Forks	Watching	
kedro-org/	9.3k	211	936	105	
OpenMined/PySyft	9.2k	423	2k	196	
goplus/gop	8.8k	39	549	177	
Netflix/metaflow	7.5k	88	824	293	
google/deepvariant	3.1k	24	741	151	
quadratichq/quadratic	2.7k	22	195	30	
colour-science/colour	1.9k	45	266	85	
NannyML/nannyml	1.7k	29	153	23	
apache/systemds	1k	180	482	85	
visualpython/visualpython	799	6	115	19	
LineaLabs/lineapy	653	21	57	20	
googleapis/python-aiplatform	520	93	360	76	
IBM/lale	320	25	81	21	
nebari-dev/nebari	254	63	98	15	
vertica/VerticaPy	214	16	47	15	
EpistasisLab/Aliro	219	20	63	23	
mithril-security/bastionlab	165	12	11	4	
microsoft/MLOS	123	18	71	12	

## Metrics

□ Collect data with Pydriller, including developer emails and commit messages.

- Number of Commits
- □ Lines of Code
- Number of Changed Files

□ RQ1: What information about data science programmers do commit-based metrics provide?



- □ RQ2: What are the characteristics of individuals identified as experts in data science projects?
- □ Analysis of 69 Data Scientists across 18 projects
- □ Progressive Data Scientist (53% of all DS)
  - Spend 60% in additions and no more than 40% in deletions
- □ Ordinary Data Scientist (25% of all DS)
  - Not exceed 60% in additions or deletions of their total activity
- □ Conservative Data Scientist (22% of all DS)
  - Spend 60% in deletions and no more than 40% in additions

- □ RQ2: What are the characteristics of individuals identified as experts in data science projects?
- □ Mono-Language Data Scientists (68% of all DS)
  - No more than 40% of their work performed on a secondary language
- □ Multi-Language Data Scientists (32% of all DS)
  - At least 40% of their work performed on a secondary language

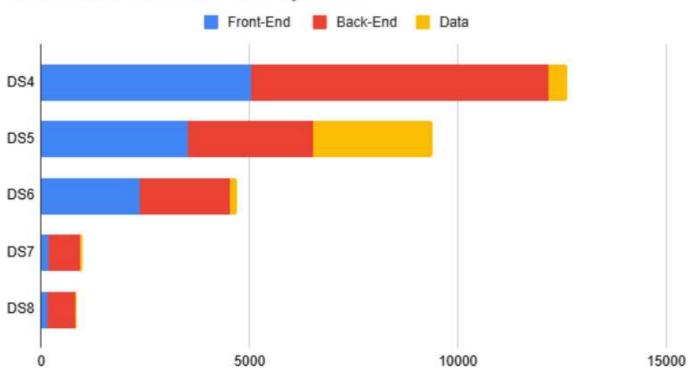
□ RQ4: What is the general composition of project teams?

Projects	Data Scientists	.py	.js	.ts	ı.r	.json	.java	.rs	.cc / .h
DV	DS1	361	110	145	11	110	0	681	0
DV	DS2	80	0	20	0	0	0	159	0
DV	DS3	81	31	67	1	31	0	70	0
QD	DS4	259	239	4618	201	185	0	0	7135
QD	DS5	2703	621	2688	158	208	8	0	2996
QD	DS6	14	145	2092	152	116	0	0	2166
QD	DS7	32	31	120	14	24	0	0	753
QD	DS8	1	32	108	21	11	0	0	681
NB	DS9	731	28	352	27	24	0	0	0
NB	DS10	316	22	130	3	17	0	0	0
BL	DS11	322	0	43	0	0	0	0	213
BL	DS12	298	0	0	9	0	0	0	327
BL	DS13	181	0	0	2	0	0	0	9
BL	DS14	65	4	17	0	0	4	0	4
BL	DS15	44	0	0	0	0	0	0	143

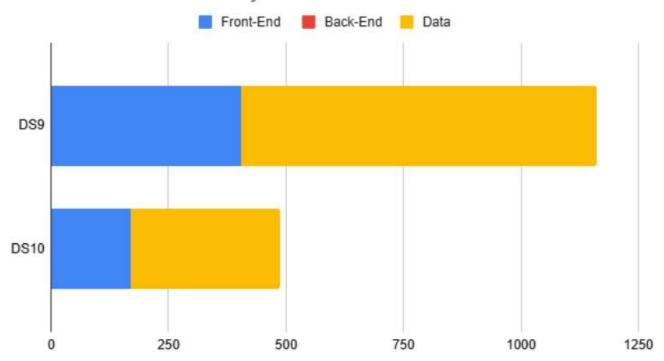
□ RQ5: What are the DS's roles and responsibilities within the project teams?

- □ (Front-End, Back-End, and Data)
  - Multi-language projects: Quadratic, Nebari, BastionLab and DeepVariant

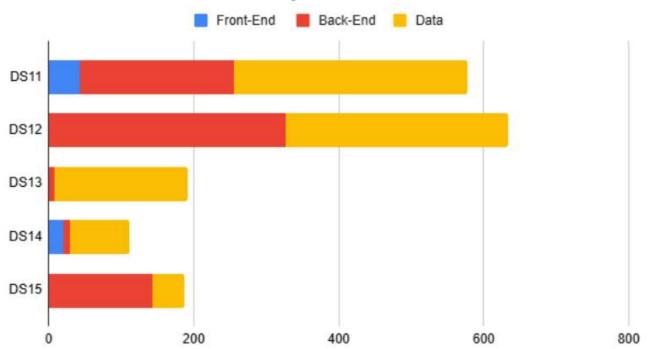
#### Quadratic Data Scientists by Stack

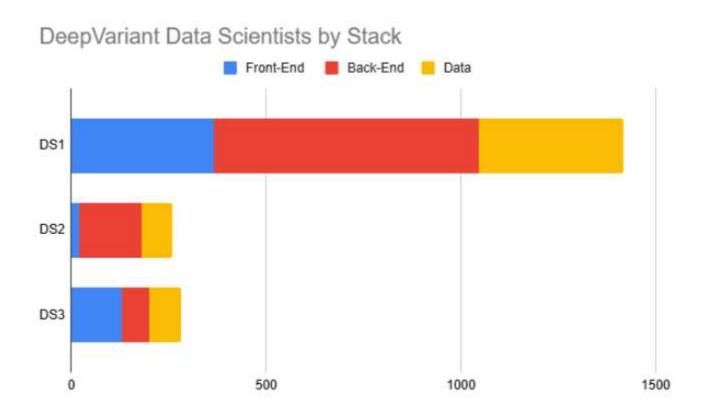


#### Nebari Data Scientists by Stack



#### BastionLab Data Scientists by Stack





□ Closer look at Quadratic

■ We selected Quadratic because it has a clear workflow of tasks assigned to developers.

- □ Enhancements New Feature or Request
- □ Bug something is broken
- Papercut annoying but not the end of the world

Assignes	Enhancements	Bug	Papercut
DS4	5	1	1
DS5	1	0	0
DS6	2	4	4
DS7	3	0	0
DS8	2	4	0

Labels				
Priority	Types	Assignes		
high priority	bug	jimniels		
high priority	enhancement	ddimaria		
high priority	enhancement	jimniels		
high priority	enhancement	HactarCE		
high priority	enhancement	davidfig		
high priority	enhancement	AyushAgrawal-A2		
high priority	enhancement	davidfig, ddimaria		
high priority	enhancement	davidfig, ddimaria		
high priority	enhancement	davidfig		
high priority	enhancement	davidkircos, jimniels, luke-quadratio		
high priority	enhancement	AyushAgrawal-A2		
QA	bug	AyushAgrawal-A2		
prioritized	bug	AyushAgrawal-A2		
NA	bug	HactarCE		
NA	bug	HactarCE		
NA	bug	HactarCE		
NA	bug	AyushAgrawal-A2		
NA	bug	davidfig		
NA	bug, papercut	jimniels		
NA	bug, papercut	jimniels		
NA	bug, papercut	jimniels		
NA	enhancement	HactarCE		
NA	enhancement	davidfig		
NA	papercut	davidfig		
NA	papercut	jimniels		
NA	bug, papercut	AyushAgrawal-A2, HactarCE		

## Threats to Validity

- □ Small sample of Projects
- Developers with fewer commits but impactful contributions maybe undervalued
- Number of Stars and Contributors

## Conclusion

- □ Future work could address these limitations by expanding the dataset.
- □ Incorporating more diverse metrics.

□ How LLM would classify data scientists





## Obrigado!