

Data acquisition and pilot study on BioRxiv and MedRxiv full text data to facilitate comprehensive data mining on biomedical literature

A project funded by
Elizabeth Blackwell Institute
Rapid Research Call

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Background

- The proposed project idea originated from our earlier work ASQ (Liu & Gaunt, 2022)
- We scraped and analysed MedRxiv abstracts from 2020-01-01 to 2021-12-31
- Limits:
 - Web scraping 24 months of *just* abstracts was time consuming
 - MedRxiv API only keeps one version of metadata
- We have known BioRxiv / MedRxiv kept full text data for text mining

1 Triangulating evidence in health sciences with 2 Annotated Semantic Queries

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7 ABSTRACT

Integrating information from data sources representing different study designs has the potential to strengthen evidence in population health research. However, this concept of evidence “triangulation” presents a number of challenges for systematically identifying and integrating relevant information. We present ASQ (Annotated Semantic Queries), a natural language query interface to the integrated biomedical entities and epidemiological evidence in EpiGraphDB, which enables users to extract “claims” from a piece of unstructured text, and then investigate the evidence that could either support, contradict the claims, or offer additional information to the query. This approach has the potential to support the rapid review of pre-prints, grant applications, conference abstracts and articles submitted for peer review. ASQ implements strategies to harmonize biomedical entities in different taxonomies and evidence from different sources, to facilitate evidence triangulation and interpretation. ASQ is openly available at <https://asq.epigraphdb.org>.

8 1 Introduction

10 Researchers in health sciences are encouraged to seek multiple strands of complementary ev-
11 idence to minimise the risk of bias creating false positives. This has been referred to as the
12 *triangulation*¹ of evidence, which may combine results from different study designs with differ-
13 ent sources of bias, including from established findings in the literature. Platforms which offer
14 a portal to integrated heterogeneous data such as Open Targets² and EpiGraphDB³ are highly
15 valuable sources which have the potential to support evidence triangulation by integrating evi-
16 dence with relevant information from a range of dedicated data providers, including biomedical
17 ontologies^{4,5}, genetic associations⁶ and literature-derived evidence⁷. One of the main objec-
18 tives for the web interface of such integrated data platforms is to present users with focused in-
19 formation from various integrated sources in order to facilitate the fast navigation and discovery
20 of evidence. However, there is a need to improve accessibility of such complex data resources
21 for less experienced users and to improve the interpretability of data, transforming source data
22 into comprehensible evidence and knowledge regarding a research question. There are several
23 challenges in order for these issues to be addressed, such as: how can a research question be
24 represented so that evidence can be retrieved for triangulation, how should we integrate biomed-



About the project

- This is a seedcorn funding project funded by the Elizabeth Blackwell Institute 2023 Rapid Research Call to us (**Yi Liu**, Tom Gaunt)
- We proposed to acquire the full text data of BioRxiv and MedRxiv preprints and conduct pilot studies on the acquired data
- Data and results from this project will lead to our next stage projects involving cross-Faculty collaborations (in progress)
- Code on processing and analysis <https://github.com/mrcieu/biorxiv-medrxiv-tdm>



Timeline (2023-06-05 -- 2023-07-31)

- 05-31 Award confirmed; 06-08 Budget code generated;
- 06-09 -- 06-23 Tried to sort out payment mechanisms
 - Finance; IT; Finance; Procurement; AWS; Procurement
- 06-26 -- 06-29 AWS access setup;
Data transfer from S3 to epi-franklin
- 07-01 -- 07-31 Exploratory analysis
- 07-01 -- 07-17 GDPR compliance check;
- 07-19 -- 07-27 Contact ACRC on purchasing RDSF;
- 07-02 -- 07-27 Setting up MyERP things (requisition, purchase order, etc.) to get the invoice paid
- 08-02 -- ... Chasing Finance to get the invoice paid

Why didn't I ...

- Set up AWS things sooner?
 - Need to appropriately set up University procurement / payment process
 - Need experiments on costs with small batches
- Contact ACRC sooner?
 - GDPR compliance check on individual identifiable information

Lessons

- Should have asked around for prior experiences & lessons more



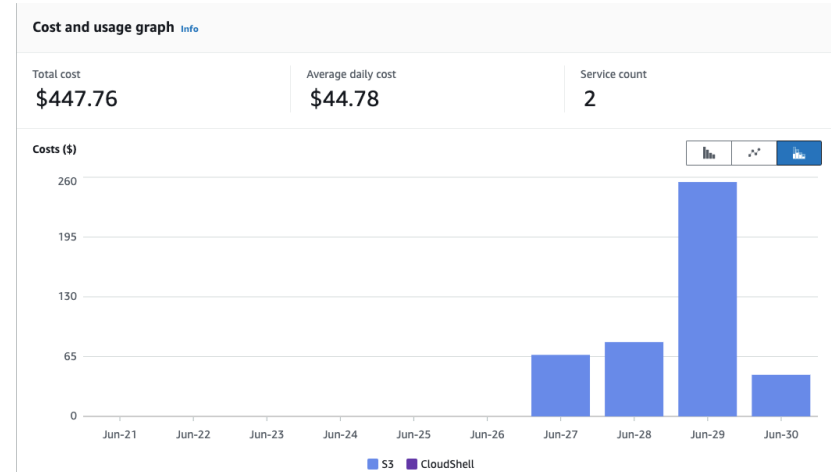
Text/data mining BioRxiv / MedRxiv

- Via API: 1) List of submitted preprints based on a query time interval 2) metadata on individual preprint
- Via web scraping, based on the doi known from metadata
- Via full text data archives
 - Hosted on as AWS S3 Requester Pays buckets -- Requester pays for the costs associated with the data transfer
 - BioRxiv s3://biorxiv-src-monthly; MedRxiv s3://medrxiv-src-monthly



Amazon AWS

- Root user: University invoice account
 - MFA by a Google Authenticator 😊
- IAM user: sub-user specific for S3 access
- Could reuse for future projects
- Originally budgeted for
 - Transferring and storing on our buckets
 - Transferring out of AWS
 - 15T per month, two months



- Traffics on 4.5T + experiment batches of data access



Acquired Dataset

```
> du -sh */
4.3T   biorxiv/
216M   examples/
305G   medrxiv/
```

```
> tree -L 1 biorxiv medrxiv
biorxiv
├── Back_Content
└── Current_Content
medrxiv
├── Back_Content
└── Current_Content
```

```
> tree -L 1 medrxiv/Current_Content
medrxiv/Current_Content
├── April_2021
├── April_2022
├── April_2023
├── August_2021
├── August_2022
├── December_2020
├── December_2021
├── December_2022
├── February_2021
├── February_2022
├── February_2023
├── January_2021
├── January_2022
├── January_2023
├── July_2021
├── July_2022
├── June_2021
├── June_2022
├── March_2021
├── March_2022
├── March_2023
├── May_2021
├── May_2022
└── May_2023
```

```
> tree -L 1 medrxiv/Current_Content/May_2023 | head -20
medrxiv/Current_Content/May_2023
├── 00064708-6f3a-1014-90ed-ad6eec8c97dc.meca
├── 001336c2-6c35-1014-8c47-83f0078374ce.meca
├── 008bc184-6ded-1014-a6e5-9a5a78f32b1a.meca
├── 00bcaab1-6e27-1014-8c3e-ad234583f629.meca
├── 00db2d3b-6c35-1014-96a9-f763932db72c.meca
├── 00deb104-6d80-1014-98d9-824dc2017398.meca
├── 016d8cf2-6ded-1014-a32e-8855e5b212f4.meca
├── 018524aa-6e27-1014-b804-88e79350a88e.meca
├── 018acf86-6ce0-1014-aaa6-ac750c63a96a.meca
├── 01d60259-6c35-1014-9c29-f3b71f96beae.meca
├── 02507972-6f3a-1014-afa4-a0ce8743f22c.meca
├── 02521bb4-6ce0-1014-b1a7-91b22fabe95b.meca
├── 02591458-6ded-1014-95d4-f366c0ec52aa.meca
├── 029491e8-6c35-1014-b9a6-800a1519d027.meca
├── 029aed2f-6e27-1014-b90f-a89b6dc6880c.meca
├── 03054bf5-6ce0-1014-bffc-a41fde5862ab.meca
├── 03328666-6ded-1014-91d3-d08c0dcfc278.meca
├── 033862af-6d80-1014-802b-aa418a535807.meca
└── 03591fa6-6e27-1014-b1c6-c354fe55fdd2.meca
```



A data archive

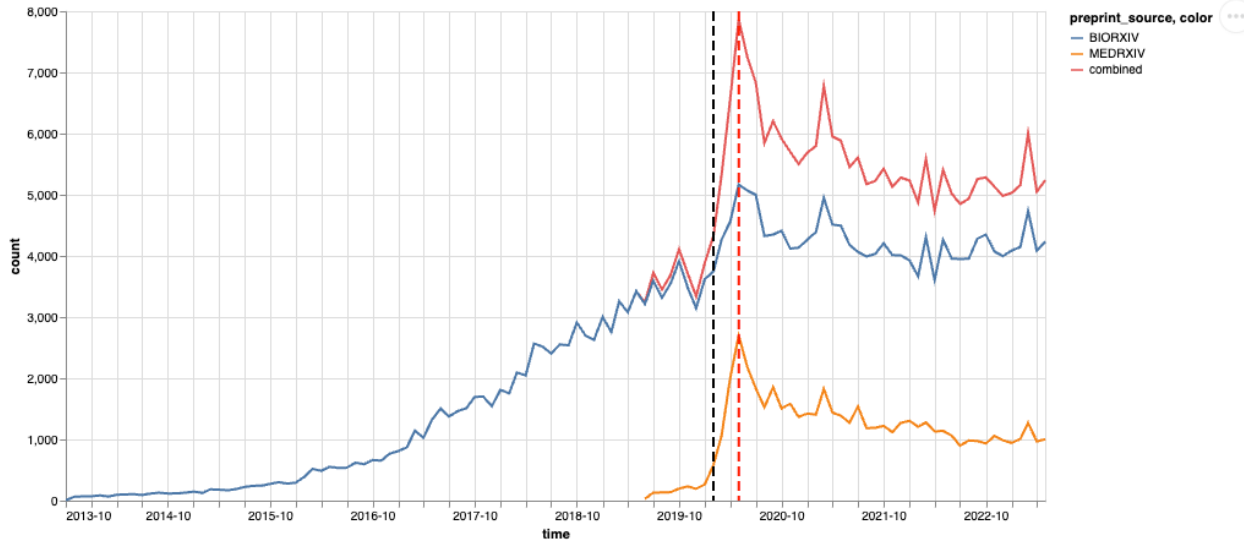
```
ik18445@ssc-franklin2.cse.bris.ac.uk:~/d/i/p/b/data/local-source-data/biorxiv/Current_Content/April_2020
Back_Content April_2019 @a0be6aa-6ce9-1014-a73c-cba5fe5880abc.meca
Current_Content April_2020 @a3bd842-6fae-1014-935a-cf7d7d807662.meca
April_2021 @a285ccc-6c51-1014-8ed1-934969bea46e.meca
April_2022 @a365fbb-6c30-1014-9ba9-a3a6a18c377d.meca
August_2019 @a32503f-6eb0-1014-bd7e-9aa79c949194.meca
August_2020 @a958632-6cc5-1014-9b63-9aedefe64693.meca
August_2021 @8c3f700-6c71-1014-a083-df666cea9c6c.meca
August_2022 @a3e2212-6f2d-1014-aace-c74e8a708088f.meca
December_2018 @a6bf0be-6fa6-1014-8c6c-82282fdaabe.meca
December_2019 @a13cfb5-6f34-1014-bacb-9f41581acc9b.meca
December_2020 @a74e56d-6c8d-1014-a4d5-9c0892437914.meca
December_2021 @a63d9ca-6d26-1014-be46-dd0ddac7e462.meca
December_2022 @a2394cd-6da5-1014-bb25-fae9579eefb0.meca
February_2019 @ac0f608-6c98-1014-a6e8-b5765352cc6.meca
@a8c7b63-6d37-1014-b185-e3e3e744f963.meca
```

```
ik18445@ssc-franklin2.cse.bris.ac.uk:~/d/i/p/b/d/l/a/0a9a5225-6c3e-1014-b429-f3eaae354361/content
@a2ef310-6c- content 225656.pdf
@a3e547c-6c- directives.xml 225656.xml
@a9a5225-6c- manifest.xml
@a797d56-6c- mimetype 225656_fig1.tif
@8c1a93e7-6c- transfer.xml 225656_fig2.tif
1c135323-6c-
1d7e60e6-6c-
24887b80-6c-
@a2ef310-6c-
@a3e547c-6c-
@a9a5225-6c-
@a797d56-6c-
@8c1a93e7-6c-
1c135323-6c-
1d7e60e6-6c-
24887b80-6c-
```

```
<?xml version="1.0" encoding="UTF-8"?>
<!DOCTYPE article PUBLIC "-//NLM//DTD JATS (Z39.96) Journal Archiving and Interchange DTD v1.2d1 2
0170631//EN" "JATS-archivearticle1.dtd">
<article xmlns:mml="http://www.w3.org/1998/Math/MathML" xmlns:xlink="http://www.w3.org/1999/xlink"
xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance" article-type="article" dtd-version="1.2d1"
specific-use="production" xml:lang="en">
<front>
<journal-meta>
<journal-id journal-id-type="publisher-id">BIORXIV</journal-id>
<journal-title-group>
<journal-title>bioRxiv</journal-title>
<abbrev-journal-title abbrev-type="publisher">bioRxiv</abbrev-journal-title>
</journal-title-group>
<publisher>
<publisher-name>Cold Spring Harbor Laboratory</publisher-name>
</publisher>
</journal-meta>
<article-meta>
<article-id pub-id-type="doi">10.1101/225656</article-id>
<article-version>1.1</article-version>
<article-categories>
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<subject>Regular Article</subject>
</subj-group>
<subj-group subj-group-type="heading">
<subject>New Results</subject>
</subj-group>
<subj-group subj-group-type="hwp-journal-coll">
<subject>Microbiology</subject>
</subj-group>
</article-categories>
<title-group>
<article-title>Drug repurposing for yellow fever using high content screening</article-title>
</title-group>
<contrib-group>
<contrib contrib-type="author">
<name>
```



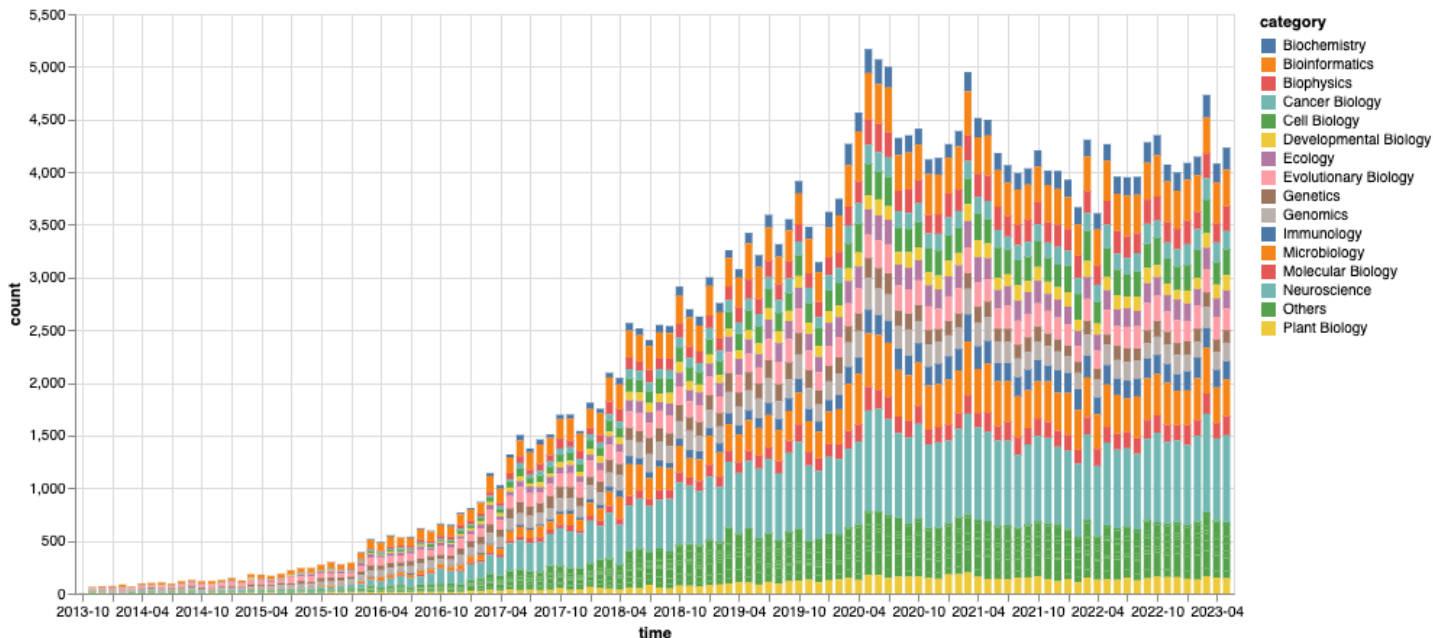

EDA: Volume



- The separation occurred in mid 2019
- Black line: 2020-02-01
- Red line: 2020-05-01



EDA: Categories, BioRxiv



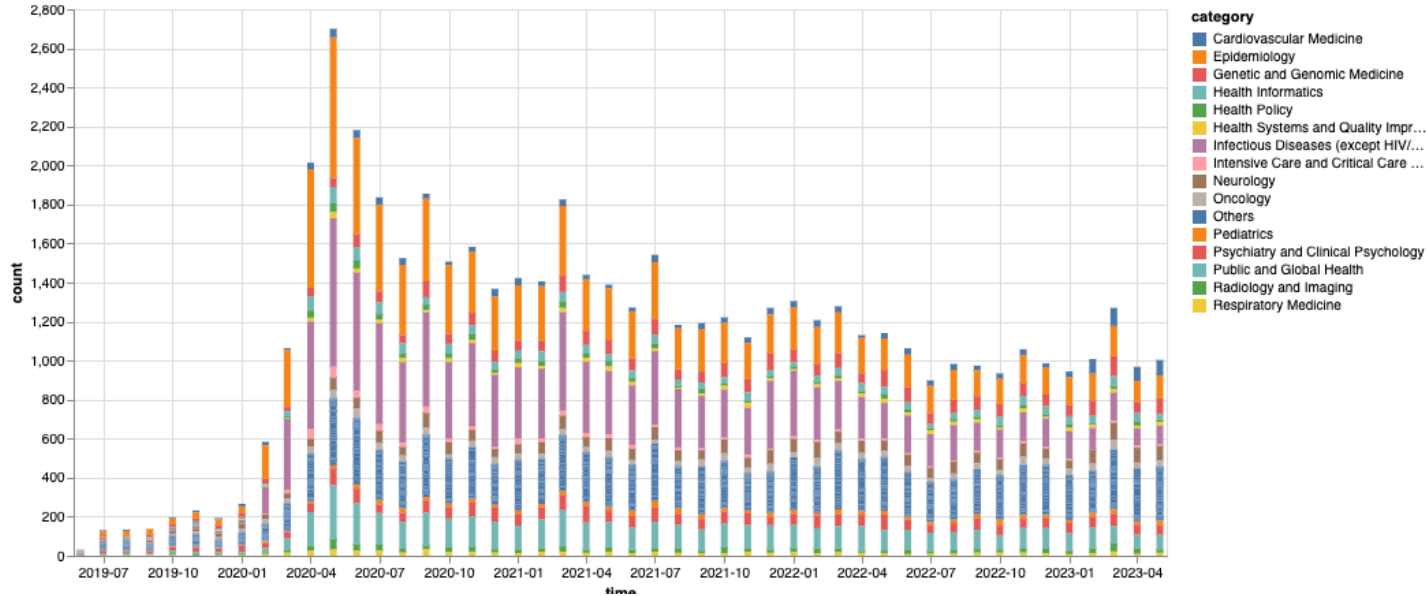
Top15 categories
+ Others

Most populous
categories:

- Bioinformatics
- Microbiology
- Neuroscience



EDA: Categories, MedRxiv



Top 15 categories
+ Others

Most populous
categories:

- Epidemiology
- Infectious Diseases (except HIV/AIDS)
- Public and Global Health



EDA: revisions

	num_versions	count
0	1	177751
1	2	43092
2	3	11806
3	4	3328
4	5	999
5	6	374
6	7	146
7	8	58
8	9	35
9	10	13

10	11	13
11	12	5
12	13	3
13	14	1
14	15	2
15	16	1
16	19	2
17	25	1
18	26	1

	doi	num_versions
0	10.1101/2020.07.09.20143164	26
1	10.1101/290825	25
2	10.1101/066423	19
3	10.1101/016840	19
4	10.1101/2020.05.26.104687	16
...



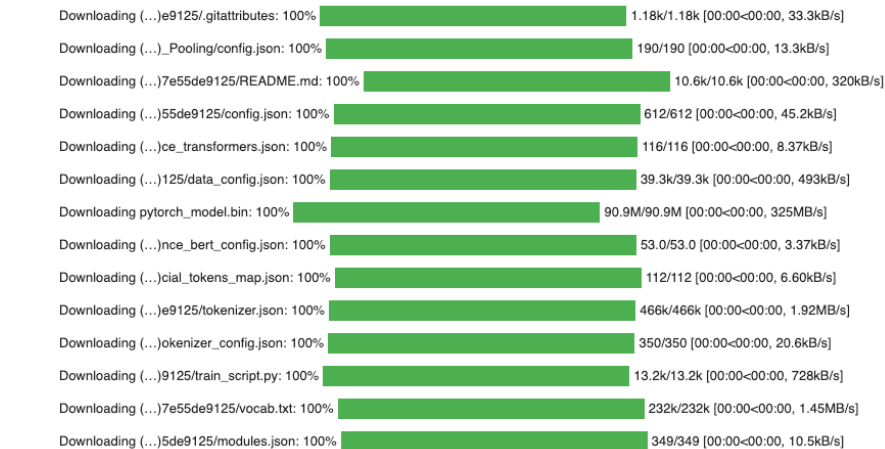
EDA: Topic analysis

BERTopic

- Done on Google Colab in ~1 hour (20 mins init, 20 mins model fit, 20 mins analysis)
- Topics from titles
- No preprocessing

```
[12] from bertopic import BERTopic

[13] topic_model = BERTopic(min_topic_size=35, verbose=True)
      topics, _ = topic_model.fit_transform(df["title"].to_list())
```



File	Progress	Size	Speed
(...)e9125/.gitattributes	100%	1.18k/1.18k	33.3kB/s
(...)_Pooling/config.json	100%	190/190	13.3kB/s
(...)7e55de9125/README.md	100%	10.6k/10.6k	320kB/s
(...)55de9125/config.json	100%	612/612	45.2kB/s
(...)ce_transformers.json	100%	116/116	8.37kB/s
(...)125/data_config.json	100%	39.3k/39.3k	493kB/s
pytorch_model.bin	100%	90.9M/90.9M	325MB/s
(...)nce_bert_config.json	100%	53.0/53.0	3.37kB/s
(...)cial_tokens_map.json	100%	112/112	6.60kB/s
(...)e9125/tokenizer.json	100%	466k/466k	1.92MB/s
(...)okenizer_config.json	100%	350/350	20.6kB/s
(...)9125/train_script.py	100%	13.2k/13.2k	728kB/s
(...)7e55de9125/vocab.txt	100%	232k/232k	1.45MB/s
(...)5de9125/modules.json	100%	349/349	10.5kB/s

```
freq = topic_model.get_topic_info()
freq
```

Topic	Count	Name	Representation	Representative_Docs
0	-1	141048	-1_covid19_for_and_the	[covid19, for, and, the, of, to, in, on, with,... [An Unsupervised Learning Method for Disease C...
1	0	3605	0_drosophila_melanogaster_larval_mushroom	[drosophila, melanogaster, larval, mushroom, w... [Identification of Microbiota-Induced Gene Exp...
2	1	2549	1_sarscov2_transmission_seroprevalence_2020	[sarscov2, transmission, seroprevalence, 2020,... [An integrated analysis of contact tracing and...
3	2	2305	2_biodiversity_ecological_forest_species	[biodiversity, ecological, forest, species, cl... [Predicting coexistence in experimental ecolog...
4	3	2160	3_gut_microbiota_microbiome_fecal	[gut, microbiota, microbiome, fecal, intestina... [Genetics of human gut microbiome composition,...
...
1006	1005	35	1005_myocarditis_pericarditis_myopericarditis_...	[myocarditis, pericarditis, myopericarditis, r... [Systematic review of spontaneous reports of m...
1007	1006	35	1006_lipidomics_lipidomic_batl_lipidome	[lipidomics, lipidomic, batl, lipidome, lipid... [BATL: Bayesian annotations for targeted lipid...
1008	1007	35	1007_cdc42_cytokinesis_gefs_polarity	[cdc42, cytokinesis, gefs, polarity, pak1depen... [A novel interplay between GEFs orchestrates C...
1009	1008	35	1008_japan_declaration_tokyo_testingisolation	[japan, declaration, tokyo, testingisolation, ... [Interim estimation for the effect of the thir...
1010	1009	35	1009_egfr_egf_epidermal_crossconservation	[egfr, egf, epidermal, crossconservation, down... [Single EGF mutants unravel the mechanism for ...

```
topic_model.get_topic(-1)
```

```
[('covid19', 0.0010036547552193563),
 ('for', 0.0009881032695176456),
 ('and', 0.000983067659642955),
 ('the', 0.0009795698563111266),
 ('of', 0.000975412825242444),
 ('to', 0.0009715365303141605),
 ('in', 0.000956200428659259),
 ('on', 0.0009529458489471976),
 ('with', 0.0009451296541405764),
 ('from', 0.0009329334366575781)]
```

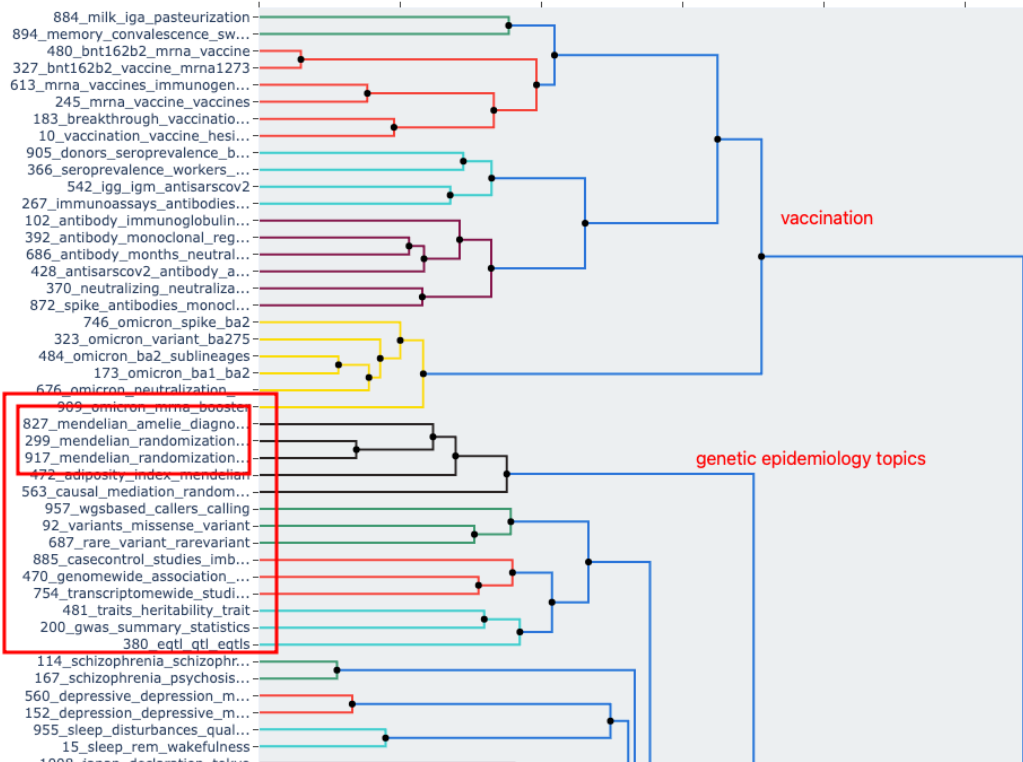
```
topic_model.get_topic(0)
```

```
[('drosophila', 0.04944922119078785),
 ('melanogaster', 0.02600333219223589),
 ('larval', 0.005590239439452415),
 ('mushroom', 0.0049347742511025965),
 ('wing', 0.004475129523336172),
 ('olfactory', 0.003962517347463073),
 ('flies', 0.0038724102058763934),
 ('suzukii', 0.0037570194657399834),
 ('adult', 0.0036421584365652804),
 ('fly', 0.003383645800590048)]
```

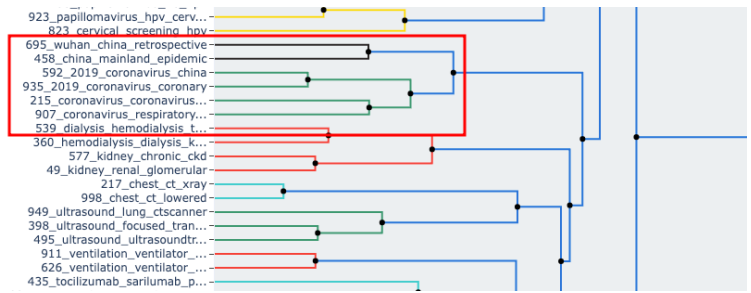
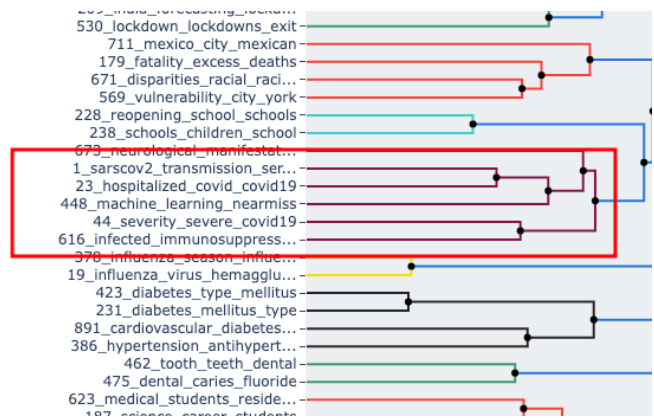
```
topic_model.get_topic(1)
```

```
[('sarscov2', 0.01813446682184406),
 ('transmission', 0.00829385784658454),
 ('seroprevalence', 0.0062970073461358494),
 ('2020', 0.006149735262666744),
 ('concern', 0.00533943029702636),
 ('surveillance', 0.005330693738371139),
 ('spread', 0.005168875294245271),
 ('b117', 0.005042158495342891),
 ('infection', 0.004969304993161452),
 ('2021', 0.00493603442912085)]
```

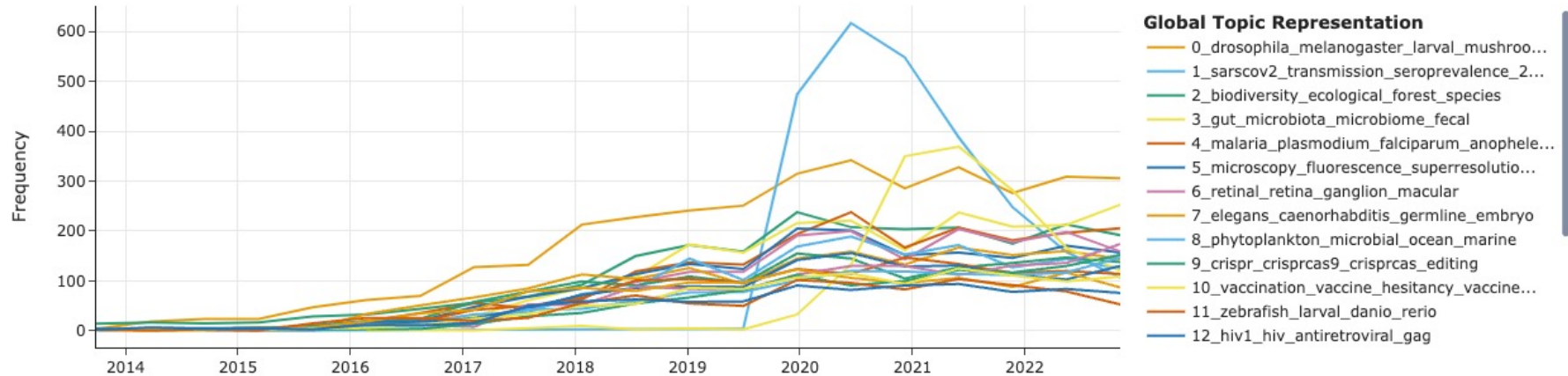
Hierarchical Clustering



Covid severity Coronavirus



Topics over Time





Next steps

- Write a blog about the project
- Put the text data into use, with collaboration projects
- Topics
 - Use the scientific text to train language models
 - Appropriate use of clustering methods to analyse research topics
 - What factors lead to successful publication of a biomedical preprint?
 - Assessment of risk-of-bias on preprints
- Questions, comments, suggestions welcome!

Acknowledgement

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Elizabeth Blackwell Institute
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