



Australian Research Data Commons

Automatically classifying data records using ANZSRC-FoR subject headings

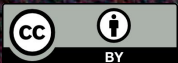
Mingfang Wu

Australian Research Data Commons, Australia

2021-10-13

DCMI 2021 Virtual

#ARDC_AU



The Australian Research
Data Commons is enabled
by NCRIS.



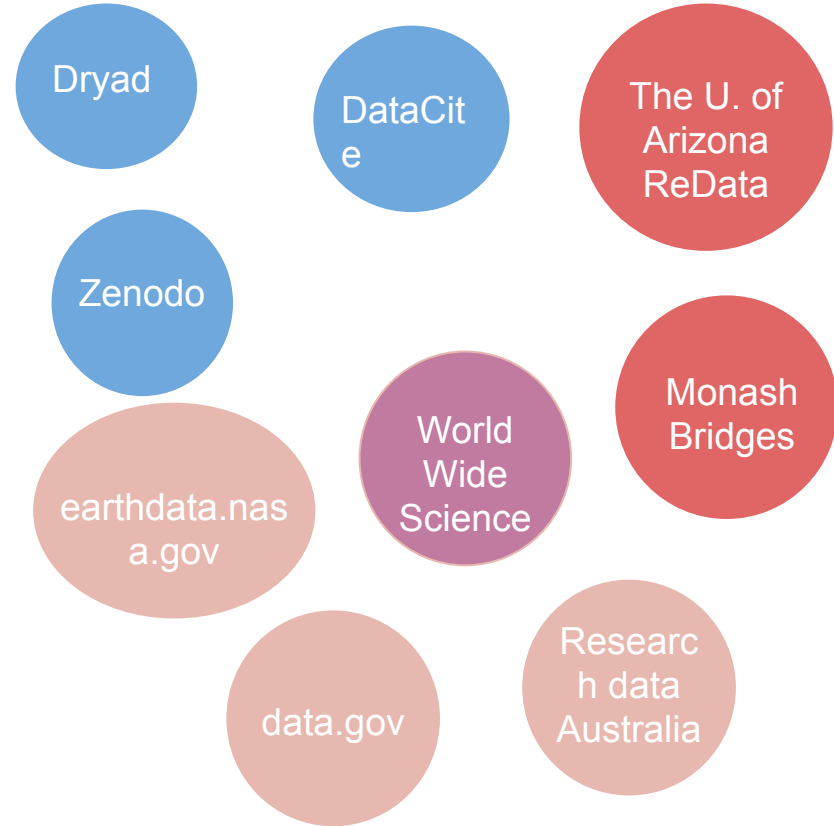
Outlines

- ANZSRC-FoR subject headings: A background about the use of subject metadata in the data catalogue: Research Data Australia
- Experiment: Automatic labelling/classification of data records with the schema - Approach & Result & Implication

Research data repositories/catalogues

There has been a growing number of data repositories & catalogues for publishing and sharing data

re3data.org, the Registry of Research Data Repositories, had 23 repositories when it went online in 2012; the number quickly increased to over 1,200 data repositories from across the globe in three years and, by February 2020, the registry had more than 2450 repositories.



Subject metadata

- Dublin Core Definition:
 - A topic of the content of the resource.
 - Typically, subject will be expressed as keywords, key phrases, or classification codes that describe a topic of the resource. Recommended best practice is to select a value from a controlled vocabulary or formal classification scheme (e.g. Faceted Application of Subject Terminology).
- Benefit: Subject metadata is a powerful way of knowledge organisation and linkage of (distributed) resources for interoperability and discovery
 - Subject metadata is included in almost all metadata schemas. For schemas describing dataset, most of them include subject metadata as an optional not mandatory field.
- Cost: Manually labelling resources with subject metadata is not efficient and may introduce inconsistency and omission.

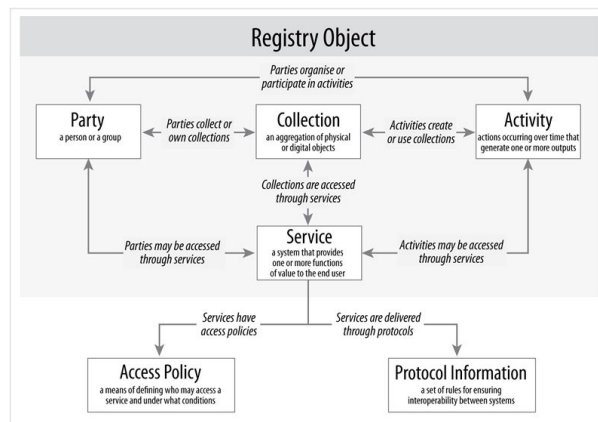
Research Data Australia - A National Data Catalogue

The screenshot shows the Research Data Australia homepage. At the top, it says "Find data for research" and "Find, access, and re-use data for research - from over one hundred Australian research organisations, government agencies, and cultural institutions." Below this is a search bar. The main section is "Browse By Subjects" with a grid of 10 categories: Humanities and Social Sciences, Business, Economics and Law, Medical and Health Sciences, Engineering, Computing and Technology, Built Environment and Design, Biological Sciences, Agricultural and Veterinary Sciences, Environmental Sciences, Earth Sciences, Physical, Chemical and Earth Sciences. Below this is a section "Who Contributes to Research Data Australia" with logos for Griffith University, The University of Queensland, The University of Western Australia, Deakin University, and RMIT University. At the bottom, it says "Research Data Australia is the data discovery service of the Australian Research Data Commons (ARDC). The ARDC is supported by the Australian Government through the National Collaborative Research Infrastructure Strategy Program."

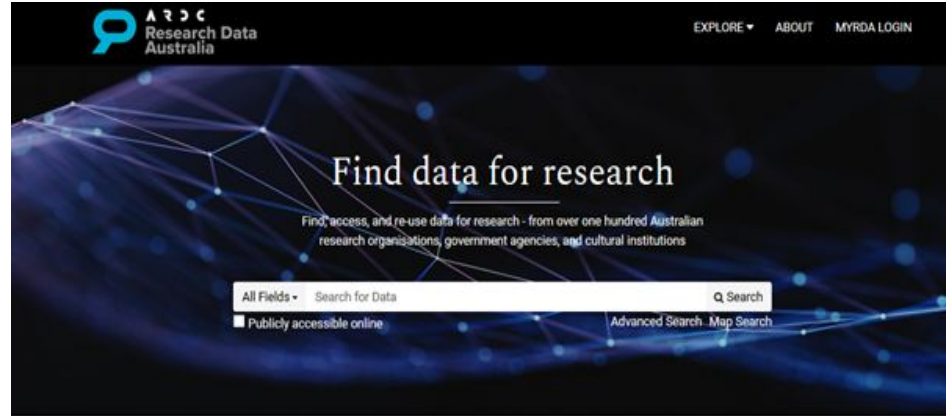
186K+ metadata records of datasets
60K+ research grants
100 Contributors



Schema: The Registry Interchange Format - Collections and Services (RIF-CS, ISO 2146:2010)



Browse by subjects



Browse By Subjects



Subject headings

766 results (48 milliseconds) Records selected: 0

All Fields - gene x Search

Publicly accessible online Advanced Search Map Search

Save Records Export

Current Search

Data

All Fields

gene x

Save Search Clear Search

Refine search results

Add more keywords Go

Type

☐ Data 762

☐ Software 4

Facet filter

Subject

- ☐ Biological Sciences 330
- ☐ Medical And Health Sciences 240
- ☐ Agricultural And Veterinary... 26
- ☐ Environmental Sciences 21
- ☐ Information And Computing S... 13

View More

Data Provider

- ☐ Monash University 235
- ☐ Australian Ocean Data Network 86

☐ Select All

☐ **Gene Sherman Collection**
Museum Metadata Exchange
A collection of Japanese fashion owned and worn by G...
The **Gene** Sherman collection is made up of approximat...
http://www.powerhousemuseum.com/collection/datab...
Gene Sherman (in Subject)

☐ **Disease gene prediction database**
Deakin University
This database includes **gene** predictions for disease ph...
... primers for phenotype-specific resequencing of patient...
Development of a bioinformatic tool for the rapid identifi...
Inherited Diseases (incl. **Gene** Therapy) (in Subject)

☐ **Play to Cure: Genes in Space**
Atlas of Living Australia
We know that faults in our **genes** can lead to cancer cel...
... to the amount of **genes** in our cells - sometimes we h...
Play to Cure: **Genes** in Space (in Related Organisations)

☐ **Lactation related gene expression d...**
Deakin University
RNA sequencing and **gene** expression data related to la...
The data was automatically generated from sequencing...
gene expression (in Subject)

☐ **Antibiotic resistance gene cassettes**
University of New South Wales
Gene cassettes and cassette arrays... (in Description)

Advanced Search

Filters

Search Terms ✓

Type

Subject

Data Provider

Access

Access Method

Licence

Time Period

Location

Review ✓

Help


Vocabulary ANZSRC FOR -

- ☐ Agricultural And Veterinary Sciences (26)
- ☐ Biological Sciences (330)
- ☐ Built Environment And Design (1)
- ☐ Chemical Sciences (2)
- ☐ Commerce, Management, Tourism And Services (1)
- ☐ Earth Sciences (1)
- ☐ Economics (1)
- ☐ Education (0)
- ☐ Engineering (2)
- ☐ Environmental Sciences (21)
- ☐ History And Archaeology (1)
- ☐ Information And Computing Sciences (13)
- ☐ Language, Communication And Culture (1)
- ☐ Law And Legal Studies (0)
- ☐ Mathematical Sciences (1)

Search for Data -

7




Record view



Disease gene prediction database

Deakin University

Dr Merridee Wouters (Aggregated by) Mr Martin Oti (Aggregated by)

   Viewed: 946 Accessed: 15

Access the data

Cite Save to MyRDA

Licence & Rights:
Other [view details](#)

Access:
Other [view details](#)

Contact Information
Postal Address:
School of Life and Environmental Sciences,
Deakin University, 75 Pigdons Road, Waurn
Ponds, Victoria 3216 Australia

Full description

This database includes gene predictions for disease phenotypes based on published Genome-Wide Association Data. May be used to choose primers for phenotype-specific resequencing of patient DNA.

For each prediction for following data is listed: phenotype, predicted gene, significant SNP, datasource, datasource reference.

Notes

The data was generated by a computer from clinical data, and some data from HuGE (<http://hugenavigator.net/HuGENavigator/home.do>) was used. The data is organised within a searchable

Subjects

Facet search
(vocabulary + keyword)

Biological Sciences | Clinical Health (Organs, Diseases and Abnormal Conditions) | Genetics | Genetics Not Elsewhere Classified | Health | Inherited Diseases (Incl. Gene Therapy) | database | genetic databases | genome-wide association study | humans | polymorphism | protein disease/genetics | single nucleotide | software |

8

Types of subject classification schemas

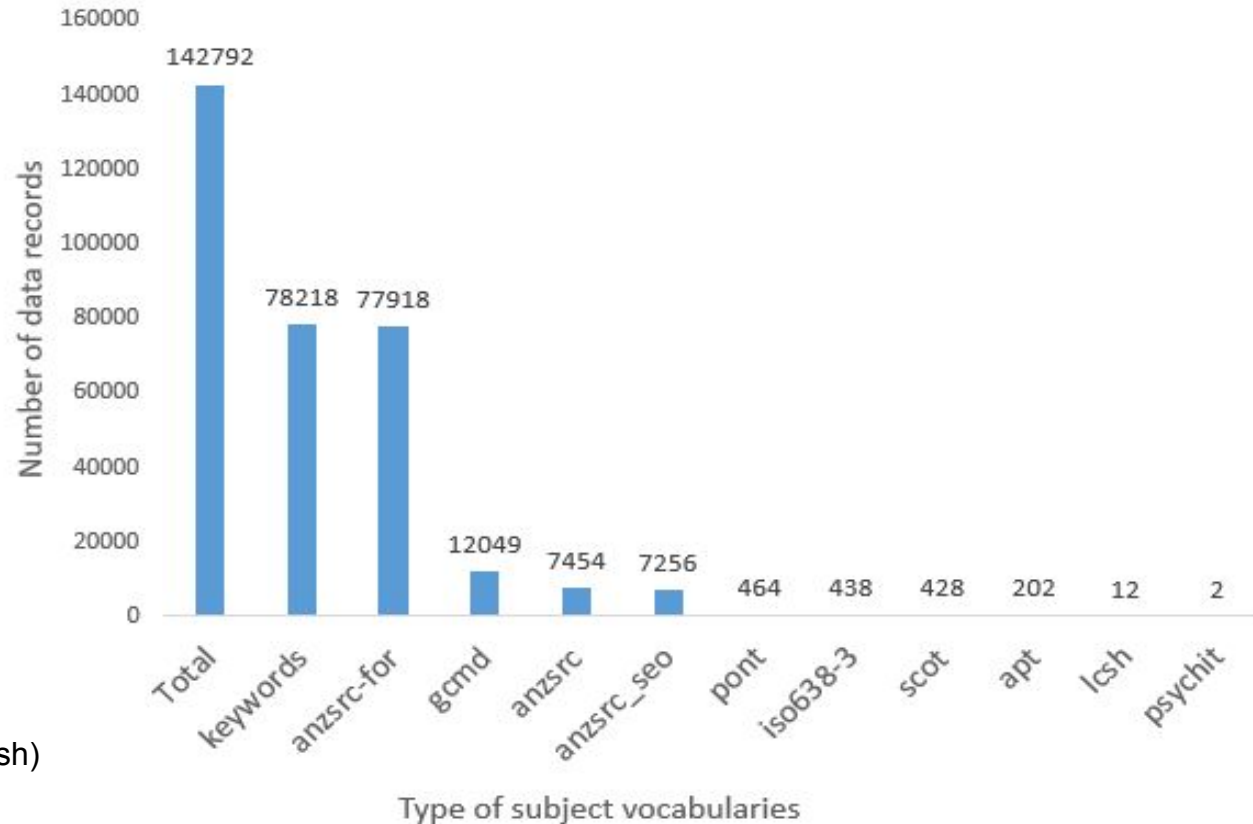
ANZSRC-FoR: The Australian and New Zealand Standard Research Classification (ANZSRC, fields of research)

Global change master directory (GCMD) keywords

Australian Pictorial Thesaurus (apt)

Thesaurus of Psychological Index Terms (psychit)

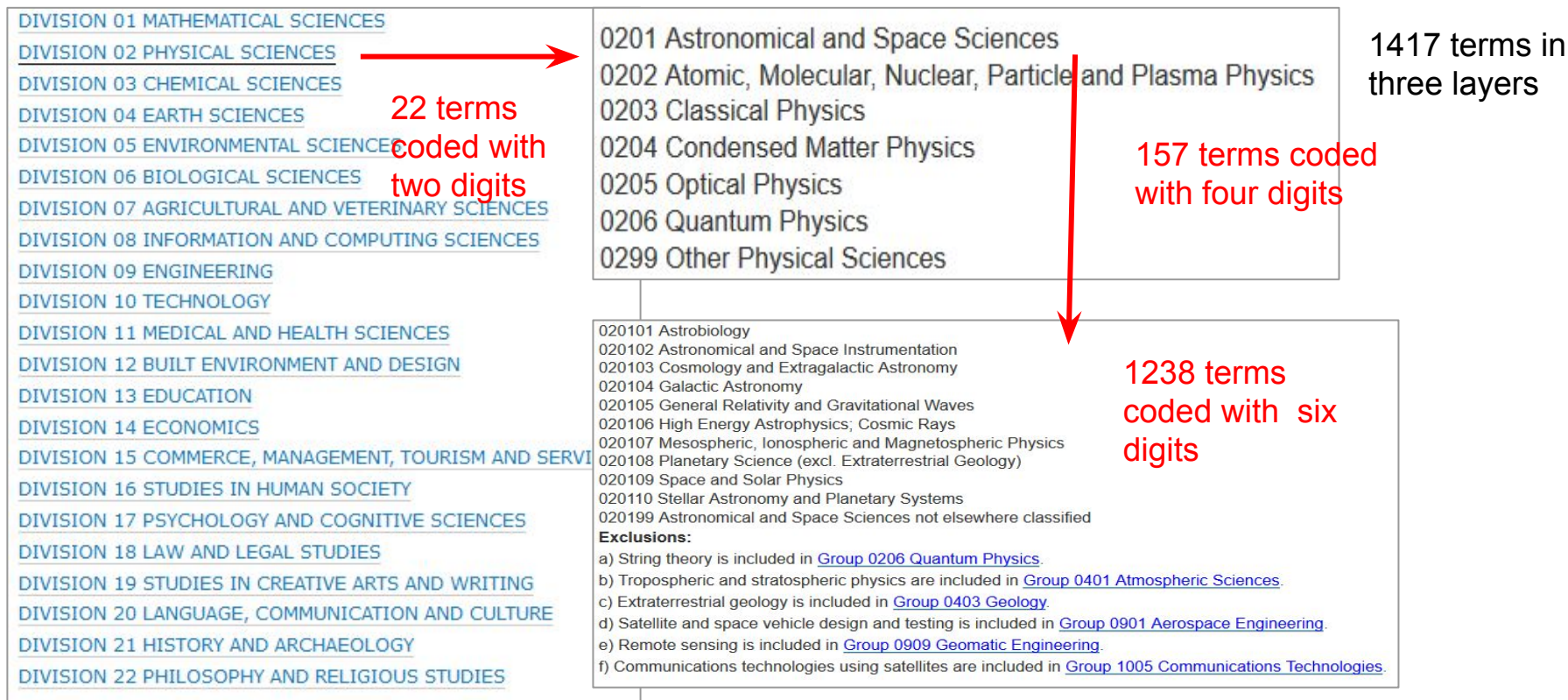
Library of Congress Subject Headings (lcsch)



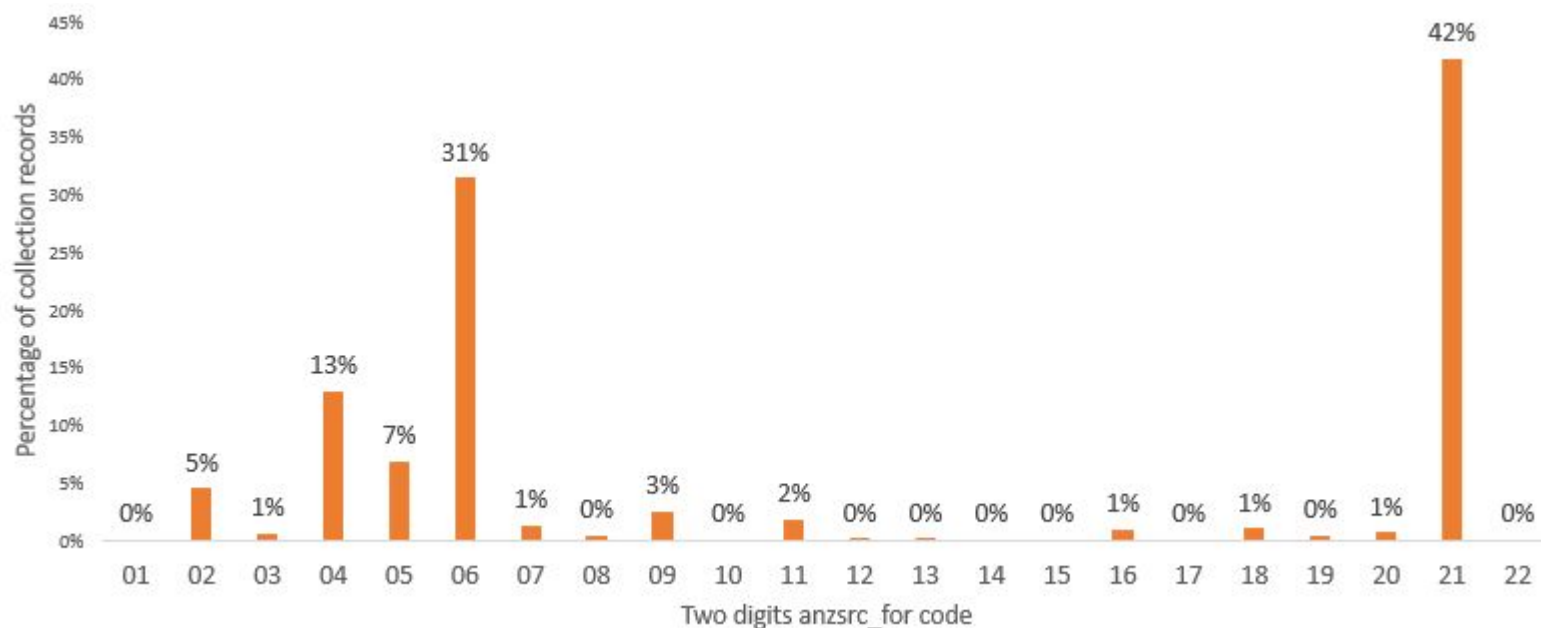
ANZSRC-FoR: The Australian and New Zealand Standard Research Classification - Fields of Research

- ANZSRC ensures that R&D statistics collected are useful to governments, educational institutions, international organisations, scientific, professional or business organisations, business enterprises, community groups and private individuals in Australia and New Zealand.
- ANZSRC-FoR include major fields and related sub-fields of research and emerging areas of study investigated by businesses, universities, tertiary institutions, national research institutions and other organisations.

ANZSRC-FoR: The Australian and New Zealand Standard Research Classification - Fields of Research) (2008 version)



Number of records per anzsrc-for two digits



04: Earth Sciences

06: Biological Sciences

21: History and Archaeology

Machine learning for classifying/labelling subject metadata

- Assign ANZSRC-FoR code to unlabelled records automatically
 - Aim to improve search experience for both human and machine
 - Understand domain coverage of the collection
- Train models, three components are essential for the training:
 - Classifier - four supervised machine learning methods:
 - multinomial logistic regression (MLR), multinomial naive bayes (MNB), K Nearest Neighbors (KNN), Support Vector Machine (SVM)
 - Labels - top layer 22 labels from the ANZSRC-FoR code
 - Data - (~78k) records with anzsrc-for code
 - Extract title & description from each metadata record
 - Split into two sets: training set, test set
- Apply model(s)/best prediction to test set

Result

Four models: multinomial logistic regression (MLR), multinomial naive bayes (MNB), K Nearest Neighbors (KNN), Support Vector Machine (SVM)

| Model | Training Set Accuracy | Test Set Accuracy |
|-------|-----------------------|-------------------|
| MLR | 0.76 | 0.70 |
| SVM | 0.70 | 0.67 |
| KNN | 0.92 | 0.66 |
| MNB | 0.70 | 0.66 |

Performance per label/category

| 2 digits code | MLR | SVM | KNN | MNB | down size | all data |
|---------------|------|------|------|------|-----------|----------|
| 01 | 0.29 | 0.00 | 0.41 | 0.33 | *111 | 111 |
| 02 | 0.97 | 1.00 | 1.00 | 0.92 | 300 | 3537 |
| 03 | 0.73 | 0.61 | 0.60 | 0.59 | 499 | 499 |
| 04 | 0.96 | 0.98 | 0.92 | 0.90 | 600 | 10147 |
| 05 | 0.61 | 0.63 | 0.68 | 0.49 | 400 | 5417 |
| 06 | 1.00 | 1.00 | 0.64 | 0.96 | 600 | 24520 |
| 07 | 0.63 | 0.52 | 0.77 | 0.42 | 200 | 1032 |
| 08 | 0.45 | 0.22 | 0.53 | 0.26 | *386 | 386 |
| 09 | 1.00 | 1.00 | 0.94 | 1.00 | 200 | 2031 |
| 10 | 0.29 | 0.00 | 0.20 | 0.00 | *128 | 128 |
| 11 | 0.68 | 0.69 | 0.63 | 0.64 | 400 | 1409 |
| 12 | 0.61 | 0.95 | 0.67 | 0.66 | *174 | 174 |
| 13 | 0.58 | 0.91 | 0.69 | 0.67 | *148 | 148 |
| 14 | 0.41 | 0.00 | 0.58 | 0.57 | *122 | 122 |
| 15 | 0.21 | 0.00 | 0.18 | 0.00 | *76 | 76 |
| 16 | 0.56 | 0.50 | 0.55 | 0.54 | 300 | 723 |
| 17 | 0.40 | 0.00 | 0.32 | 0.67 | *112 | 112 |
| 18 | 1.00 | 1.00 | 0.99 | 0.98 | 400 | 849 |
| 19 | 0.82 | 0.69 | 0.76 | 0.54 | *343 | 343 |
| 20 | 0.89 | 0.85 | 0.26 | 0.81 | 300 | 553 |
| 21 | 0.97 | 0.96 | 0.99 | 0.88 | 600 | 32592 |
| 22 | 0.34 | 0.00 | 0.65 | 0.44 | *79 | 79 |
| micro ave | 0.70 | 0.67 | 0.66 | 0.66 | 4799 | 84988 |
| macro ave | 0.65 | 0.57 | 0.63 | 0.60 | | |
| weighted ave | 0.76 | 0.71 | 0.70 | 0.68 | | |

Most correlated unigrams:

| Code | Top 5 | Bottom 5 |
|------|---|---|
| 04 | earth airborne geophysical mount ign | al unit two australia region |
| 15 | study financial survey university dataset | given number received document expert |

04: Earth Science

15: Commerce, Management, Tourism and Services

What we have learnt

- Large proportion of records from the catalogue don't have the subject metadata **a known issue**
- Those with subject metadata are biased toward a few categories
 - **Encourage underrepresented subject areas to publish and share data**
- Automatic classification works for some categories
 - **Explore correlation and improvement, and other machine learning methods**

Discussion

- There does not exist a classifier that fits for all text classification tasks or all collections. This leads to a generalisation issue, it is hard for organisations who want to improve subject metadata but have no resources to train classification models for their collection.
- Traditional ML methods require a good amount of training dataset with quality annotation. Some new technologies (e.g. deep learning and transfer learning) may provide a way, for example, word embedding technique by learning word association from other similar big collections.
- Practically there is a question about how to use ML outputs: MLs suggest labels based on probability, human expertise may be necessary to look at a range of suggestions and make a decision, this may lead to improve ML intelligence and classification accuracy - we help the ML community to help ourselves.

Acknowledgement

Ying-Hsang Liu, Oslo Metropolitan University, Norway

Rowan Brownlee, Australian Research Data Commons

Jenny Xiuzhen Zhang, RMIT University, Australia




Australian Research Data Commons

Thank you

Contact:

mingfang.wu@ardc.edu.au

www.ardc.edu.au

 [australian-research-data-commons](#)

 [ARDC_AU](#)

The Australian Research
Data Commons is enabled
by NCRIS.

