

ICML + EAHIL 2017 Dublin, 12-16 June

James Thomas, Claire Stansfield, Alison O'Mara-Eves, Ian Shemilt Evidence for Policy and Practice Information and Co-ordinating Centre (EPPI-Centre) Social Science Research Unit

UCL Institute of Education

University College London







Acknowledgements & declaration of interest

- Many people, including: Sergio Graziosi and Jeff Brunton (EPPI-Centre); National Centre for Text Mining (NaCTeM, University of Manchester); Chris Mavergames and Cochrane IKMD team; Julian Elliott and others on Cochrane Transform project; Iain Marshall (Kings College); Byron Wallace (Northeastern University)
- James Thomas receives funding from Cochrane and the funders below for this and related work; co-lead of Project Transform; lead EPPI-Reviewer software development.
- Parts of this work funded by: Cochrane, JISC, Medical Research Council (UK), National Health & Medical Research Council (Australia), Wellcome Trust, Bill & Melinda Gates Foundation. All views expressed are our own, and not necessarily those of these funders.



Objectives

- Provide an overview of text-mining technologies for searching and study selection
- Try out machine learning technologies
- Discuss methodological issues, their implementation and software options



Automation in systematic reviews – what can be done?

- Study identification:
 - Assisting search development
 - RCT classifier
 - Citation screening
 - Updating reviews
- Mapping research activity
- Data extraction
 - Risk of Bias assessment
 - Other study characteristics
 - · Extraction of statistical data
- Synthesis and conclusions





Technology

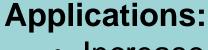
- Icomiology									
	Term recognition / text analytics	Automatic clustering	Automatic classification						
Developing search strategies									
Obtaining sub- sets of citations									
'Mapping' research literature quickly									
Reducing workload during citation screening		$\overline{\checkmark}$							

Use in a review



Assisting search development

Purpose: to explore linkages or words in text or controlled vocabulary



- Increase precision
- Increase sensitivity
- Aid translation across databases
- "Objective" search strategies
- Integrated search and screen systems





Sample of citations

Citation elements (title, abstract, controlled vocabulary, body of text, etc)

Text analysis

Word frequency counts, phrases or nearby terms in text

Generic tools

Database specific (PubMed) tools

Term extraction and automatic clustering

Statistical analysis

Statistical and linguistic analysis

TF-IDF

TerMine

Automatic Clustering

Word or phrase lists Visualisation

Revise search elements



Humans assess relevance and impact to search





Mapping research activity

- It is possible to apply 'keywords' to text automatically, without needing to 'teach' the machine beforehand
- This relies on 'clustering' technology – which groups studies which use similar combinations of words
- Very few evaluations
 - Can be promising, especially when time is short
 - But users have no control on the terms actually used

Original Article Received 23 November 2012. Revised 21 November 2013. Accepted 21 April 2013 Published arkine in Wiley Ordine Library (wileyonlinelibrary.com) DOE 10.1002/jrsmt.1082 'Clustering' documents automatically to support scoping reviews of research: a case study Claire Stansfield,** James Thomas and Josephine Kavanagh

Background: Scoping reviews of research help determine the feasibility and the resource requirements of conducting a systematic review, and the potential to generate a description of the literature quickly is attractive.

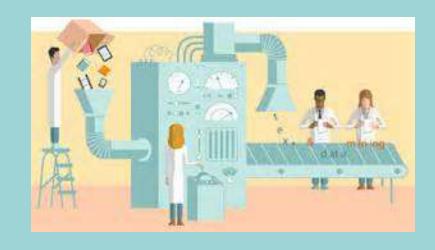
Aims: To test the utility and applicability of an automated clustering tool to describe and group research

dethods: A retrocoertive study of two completed scoping reviews was conducted. This compared th

studies to improve the efficiency of scoping reviews.



Demo – Topic modelling pyLDAvis



http://eppi.ioe.ac.uk/ldavis/index.ht ml#topic=6&lambda=0.63&term=



Technologies for identifying subsets of citations

- Different families of techniques
 - Fairly simple approaches which examine term frequencies to group similar citations
 - More complex approaches, such as Latent Dirichlet Allocation (LDA)
- The difficult part is finding good labels to describe the clusters
 - But are labels always needed?
- Visualisations are often incorporated into tools

Citation screening

C'Mera-Eves et al. Systematic Reviews 2013, 4:5 http://www.systematicreviews/pounal.com/content/4/1/5



RESEARCH

Open Access

Using text mining for study identification in systematic reviews: a systematic review of current approaches

Alison O'Mara-Eves*, James Thomas*, John McNaught*, Makoto Miwa* and Sophia Ananiadou*

Abstract

Background: The large and growing number of published studies, and their increasing rate of publication, makes the task of identifying relevant studies in an unbiased way for inclusion in systematic reviews both complex and time consuming. Text mining has been offered as a potential solution; through automating some of the screening process, reviewer time can be saved. The evidence base amount the use of text mining for screening has not yet been pulled together systematically this systematic review fills that research gap Trousing mainly on non-recheical issues, the review aims to increase awareness of the potential of these technologies and promote further collaborative research between the computer science and systematic review communities.

Methods: Five research questions led our review, what is the state of the evidence base how has workload reduction been reviousled; what are this purpose of serie-automation and how effective are they, how have key concessual problems of anotherin test mission to the sottenatio review that them published, and what children in

Citation screening

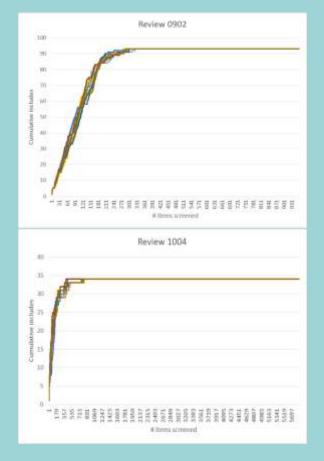
- Has received most R&D attention
- Diverse evidence base; difficult to compare evaluations
- 'semi-automated' approaches are the most common
- Possible reductions in workload in excess of 30% (and up to 97%)

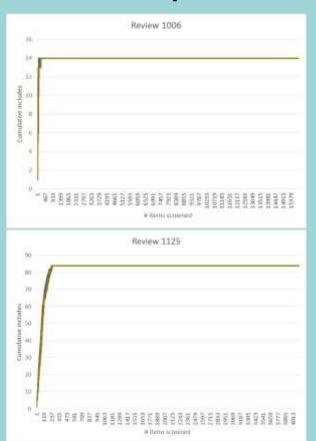
Summary of conclusions

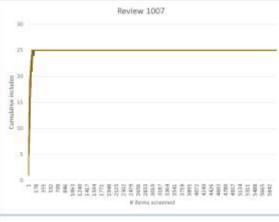
- Screening prioritisation
 - · 'safe to use'
- Machine as a 'second screener'
 - Use with care
- Automatic study exclusion
 - Highly promising in many areas, but performance varies significantly depending on the domain of literature being screened



Does it work? e.g. reviews from Cochrane Heart Group











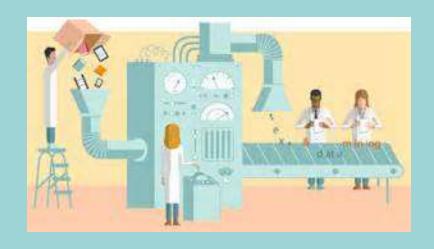


What does a classifier do?

- It takes as its input the title and abstract describing a publication
- It outputs a 'probability' score between 0 and 1 which indicates how likely the publication is to being the 'positive class' (e.g. is an RCT)
- Classification is an integral part of the 'evidence pipeline'



Demo – Topic modelling and classification in RobotAnalyst



http://www.nactem.ac.uk/robotanalyst/



RobotAnalyst

- Systematic review software designed by National Centre for Text Mining at the University of Manchester:
 - Topic modelling, term extraction, search in text and metadata,
 - Automatic classification based on user's decisions
- Currently being evaluated (users welcome! contact NaCeTM); to be released soon

http://www.nactem.ac.uk/robotanalyst/



Pre-built or build your own

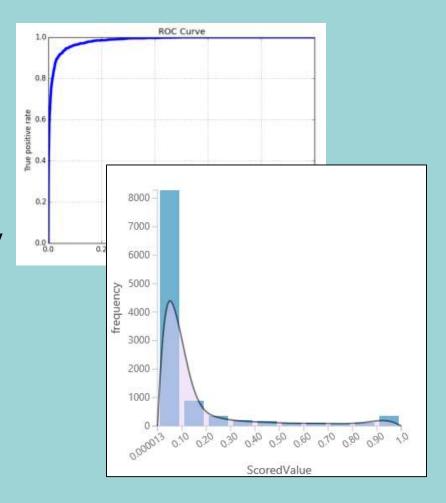
- Pre-built in EPPI-Reviewer
 - Developed from established datasets
 - RCT model
 - Systematic review model
 - Economic evaluation
- Build your own





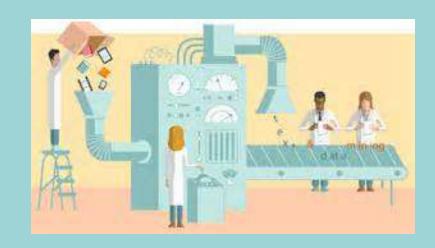
Pre-built classifier

- An RCT classifier was built using more than 280,000 records from Cochrane Crowd
- 60% of the studies have scores < 0.1
- If we trust the machine, and automatically exclude these citations, we're left with 99.897% of the RCTs (i.e. we lose 0.1%)
- Is that good enough?
- Systematic review community needs to discuss appropriate uses of automation





Demo - pre-built RCT classifier EPPI-Reviewer 4



http://eppi.ioe.ac.uk/eppireviewer4/

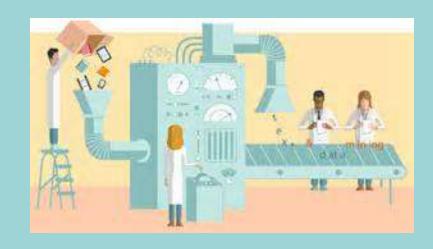


Testing models for TRoPHI register of health promotion controlled trials

N=9,431 records	Pre-built RCT classifier		Build your own classifier Best Second best			
Items scored 11-99:	RCTs	NonRCTs	RCTs	NonRCTs	RCTs	NonRCTs
Precision	12%	3%	17%	5%	12%	4%
Recall	99%	86%	99%	99%	99%	100%
Screening reduction	43%		58%		41%	



Demo – Triaging studies using Cochrane CRS



https://crsdemo.metaxis.com/



A few tools to try...



Resources Machine learning in systematic reviews

Resources for machine-assisted searching and study selection in systematic reviews (Dublin, June 2017)

https://eppi.ioe.ac.uk/cms/Default.aspx?tabid=3677

Slides for this workshop

Identifying sub-sets of citations (clustering / topic identification)

- Termine: http://www.nactem.ac.uk/software/termine/
- Topic modelling: http://eppi.ioe.ac.uk/ldavis/index.html#topic=6&lambda=0.63&term=
- Carrot2 search: http://search.carrot2.org/stable/search

Classification: RCT Classifier; and 'custom' classifier

- EPPI-Reviewer: http://eppi.ioe.ac.uk/eppireviewer4
- (Cochrane CRS-Web: https://crsdemo.metaxis.com/)

Workload reduction during citation screening (using 'active learing')

- Rayyan systematic reviews tool: http://rayyan.qcri.org/reviews/5
- Microsoft Azure Machine Learning (James can demo for those interested)

Automated risk of bias and PICO classification:

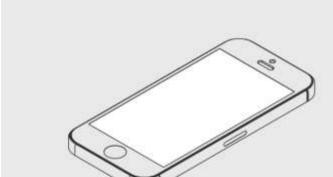
- RobotReviewer: https://robot-reviewer.vortext.systems/
- Cochrane PICO Finder: https://uat-data.cochrane.org/pico-finder/

https://www.mentimeter.com/app

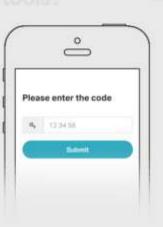


Discussion

Go to www.menti.com and use the code 80 60 84



www.menti.com



Grab your phone

2 Go to www.menti.com

Enter the code 80 60 84 and vote! What methods and processes will need to be developed to use these tools?

What are your concerns?

What do you think are the potential benefits?

Research registers

Review

Efficiency

types Skills Topic modelling

Reduce recall

Information

Literacy

Software

Risk

mapping Transparency

and

Processes

Availability

Acceptability

Opportunities

Selected bibliography



- SR Toolbox http://systematicreviewtools.com/
- Paynter R, et al. (2016). EPC Methods: An Exploration of the Use of Text-Mining Software in Systematic Reviews. <u>AHRQ Research White Paper</u>.
- O'Mara-Eves A, et al. (2015). Using text mining for study identification in systematic reviews: a systematic review of current approaches. Syst Rev 4: 5.
- Thomas J, et al. (2011). Applications of text mining within systematic reviews. Res Synth Meth 2(1): 1-14.
- Shemilt I, et al. (2016) Use of cost-effectiveness analysis to compare the efficiency of study identification methods in systematic reviews. Syst Rev 5: 140.
- Stansfield C, et al. (in press) Text mining for search term development in systematic reviewing: a discussion of some methods and challenges. Res Synth Meth.
- Stansfield C, et al. (2015) Reducing systematic review workload using text mining: opportunities and pitfalls. J. EAHIL 11(3): 8-10.



Thank you

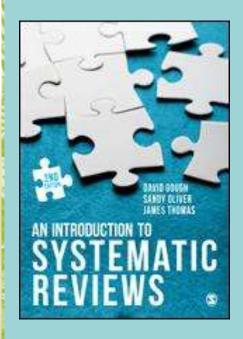
EPPI-Centre website: http://eppi.ioe.ac.uk

Email

j.thomas@ucl.ac.uk c.stansfield@ucl.ac.uk a.omara-eves@ucl.ac.uk i.shemilt@ucl.ac.uk



The EPPI-Centre is part of the Social Science Research Unit at the UCL Institute of Education, University College London



EPPI-Centre

Social Science Research Unit Institute of Education University of London 18 Woburn Square London WC1H 0NR

Tel +44 (0)20 7612 6397 Fax +44 (0)20 7612 6400 Email eppi@ioe.ac.uk Web eppi.ioe.ac.uk/