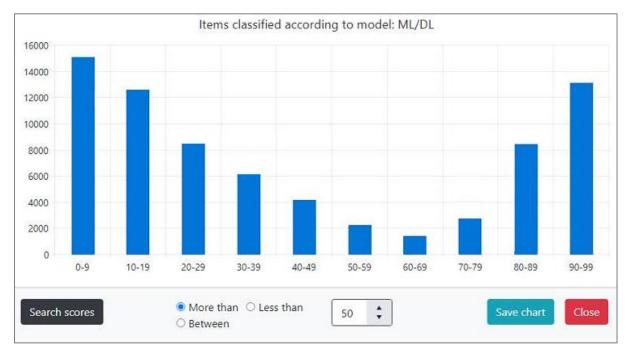
Classifiers in EPPI Reviewer

Background

Classifiers organise references, sorting them by the probability that they are relevant or not, according to existing data (i.e. an existing set of items already screened and thus coded with include or exclude codes).

Each item will be given a "probability of being relevant" score and can be presented in order - sorted on this score or banded into decile bands of probability (0-9% likely to be relevant, 1-10% likely to be relevant, through to 90-99% likely to be relevant).



If you have a dataset (references and coding, or references in batches according to your inclusion / relevant and exclusion / irrelevant criteria), you can create classifiers based on this information.

(This can be screening or coding done on items already in an existing review, whereby data from EPPI Reviewer's screening system can thus be taken to build a classifier. You can amalgamate multiple exclusion or inclusion criteria to create two codes summarising whether an item is relevant or not.

Alternatively it can be done by giving codes to subsets of your dataset, according to whether the items are relevant or irrelevant, thus you can also work with data imported from outside EPPI Reviewer i.e. without screening or coding the items within ER.)

You can then apply the classifier to new items, automatically classifying them according to knowledge the system has learnt from your existing screening / coding.

There are also several pre-built classifiers available in EPPI Reviewer, designed to find, for example, Cochrane RCTs, original RCTs, or economic evaluations.

So, the models can classify items according to the probability they are an original Random Controlled Trial, a Systematic Review, etc.

The models are built from thousands of records and should be very accurate when used on items in the same subject area i.e. biomedical records of human studies, such as RCT records found on PubMed.

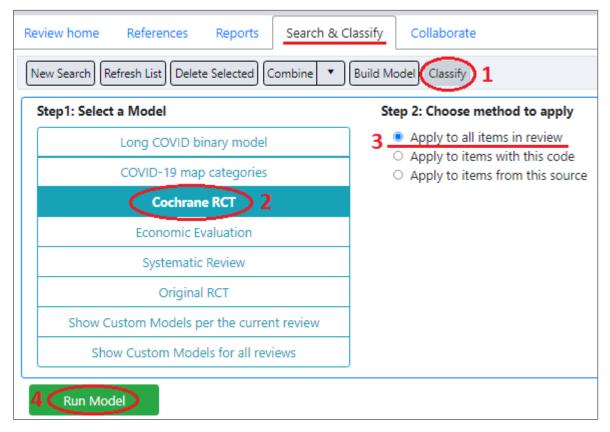
(Note that we are continually expanding and improving these in-built classifiers; for example, we have recently included specific models relating to Covid and Long Covid. We have incorporated current methods such as BERT models (Bidirectional Encoder Representations from Transformers), utilising a machine-learning technique developed for natural language processing by Google.)

Review home References Reports Search &	Classify Collaborate
New Search Refresh List Delete Selected Combine	Build Model Classify
Step1: Select a Model	Step 2: Choose method to apply
Long COVID binary model	Apply to all items in review
COVID-19 map categories	 Apply to items with this code Apply to items from this source
Cochrane RCT	
Economic Evaluation	
Systematic Review	
Original RCT	
Show Custom Models per the current review	
Show Custom Models for all reviews	
Run Model	

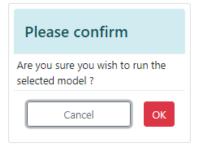
Using an in-built classifier

Under the *Search & Classify* tab, click the *Classify* button (1).

Select the in-built model you wish to use (2), then which items you wish to apply the classifier to (3). Finally click the *Run Model* button (4).



You will be asked to confirm the action, as it may take several minutes to process.



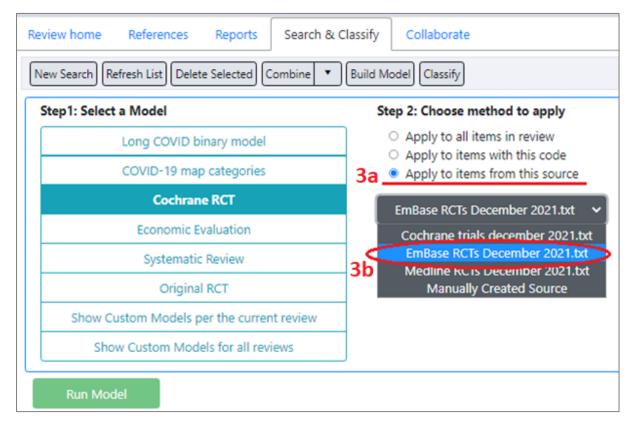
You will then see a status message - "Refresh List to see results".

(!) Refresh List to see results ×

Note that you can also apply the model to all items with a particular code, by selecting the relevant option, followed by the relevant codeset (in the code tree on the right).

ew Search Refresh List Delete Selected Combine 🔻	Build Model Classify	SIGN Information Scientist Sift
Step1: Select a Model	Step 2: Choose method to apply	 SIGN Methodology Checklist 1:
Long COVID binary model	 Apply to all items in review Apply to items with this code 	Systematic Reviews and Meta-analy
COVID-19 map categories	 Apply to items from this source 	Allocations
Cochrane RCT	Current code: All items for HSRs	Group 1 Group 2
Economic Evaluation		All items for HSRs 3b
Systematic Review		SIGN HSR Screen on Title & Abst
Original RCT		
Show Custom Models per the current review		
Show Custom Models for all reviews		

Alternatively, you can apply the model to all items from a particular source (just select the relevant source from the drop-down list).



Once the system has processed your reference (applied the model to your items and classified them accordingly), the results will be shown under your **Search & Classify** tab, along with your searches.

As mentioned, this can take several minutes. You may need to click the *Refresh List* button to see the classifier appear in your list.

Classifiers are indicated by the *True* link in the right-most "*Classifier*" column. To show a visualisation of the results, click the *true* link to the right of the name (in the *Classifier* column).

Revie	Review home References Reports Search & Classify Collaborate							
New	New Search Refresh List Delete Selected Combine 🔻 Build Model Classify							
	No	Name	Created By	Date	Hits	Classifier		
	14	Items classified according to model: RCT	Zak Ghouze	4 May 2022	41552	true		
	13	With at least one document uploaded.	Zak Ghouze	31 Aug 2021	1	false		
	12	"role" (in Abstract)	Zak Ghouze	12 Apr 2021	4275	false		

The results are shown in probability bands (10 bands according to how likely an item is likely to be relevant or irrelevant, from 0-9% to 90-99%).

Note that you can search for particular score ranges or save the visualisation as a graphic (for reuse elsewhere).



You can display the probability score on the *References* tab. To show a list of the items in the classifier, click on the *Hits* number adjacent to the classifier Name.

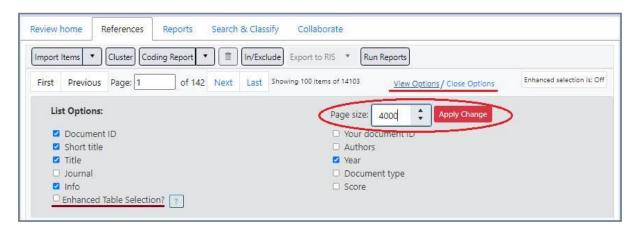
ew	Search	efresh List) Delete Selected) Combine 💌 Build Model) Cla	essify				
	No	Name	Created By	Date	Hits	Classifier	
0	381	Search #377 scores between 40 and 49	Mark Engelbert		32377	false	í
	380	Search #377 scores between 50 and 59	Mark Engelbert		18616	false	
	379	Search #377 scores between 60 and 69	Mark Engelbert		4405	false	1
	378	Search #377 scores between 70 and 79	Mark Engelbert		170	false	
	377	Items classified according to model: 2017 T-A data (pub year 2017 only)	Mark Engelbert		66309	true	
	376	Not coded with: All screened (previous projects) 2021- 03-31	Mark Engelbert		66317	false	
	375	Not coded with: All screened (previous projects) 2021- 03-31	Zafeer Ravat		66356	false	
0	374	Not coded with: All screened (previous projects) 2021- 03-31	Zafeer Ravat		66361	false	
	372	Not coded with: All screened (previous projects) 2021- 03-31 (TO RE-RUN)	Zafeer Ravat		66398	false	

Alter the *View Options* to show *Score*, then you can sort by score (from lowest to highest or vice versa with each click).

mport Items		ter Coding Report		
	evious Pag			inhanced selection is: On
List Option	nent ID title al	lection?	Page size: 100 Your document ID Authors Year Document type Score	
	ns classified acc	ording to model: 2017	7 T-A data (pub year 2017 only)	ED
howing Item				-
ihowing Iten	ID	Short title1	Title	Year Score
ihowing Item		Short title1 Amstutz (2020)	Title Home-based oral self-testing for absent and declining individuals during a door-to-door HIV testing campaign in rural Lesotho (HOSENG): a cluster-randomised trial	Year Score

First	Pre	evious Pag	e: 1 of 17	Next Last Showing 4000 items of 66309 View Options En	hanced selection is: Off
Showir	ng Item	is classified acc	ording to model: 2017 T-	A data (pub year 2017 only)	ED
		ID	Short title	Title	Year Score1
GO	• I	56148203	Jain (2018)	Three Essays on Health and Aging	2018 60
60	1	56193340	Jahan (2020)	Awareness Development and Usage of Mobile Health Technology Among Individuals With Hypertension in a Rural Community of Bangladesh: Randomized Controlled Trial	2020 60

(You can alter the number of items shown per page via the *View Options* link. A list of items can be selected in one operation via the checkbox at the top of the list.)



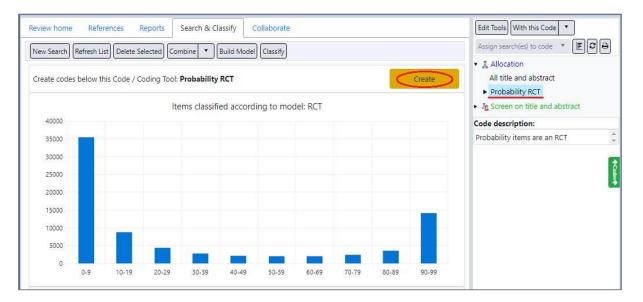
mport Item	ns 🔹 Clus	ster) Coding Report 🔹 🏦 In/I	Exclude Export to RJS Run Reports		
First Pr	revious Pag	ge: 1 of 20 Next Last	Showing 4000 items of 77102 View Options	Enhanced sele	ction is: Of
Showing Ite	ems classified a	ccording to model: RCT		I E D	
\odot	ID	Short title†	Title	Year Info	Score
50 🗹 I	61387832	"Prospective, Randomized, (Sungurtekin)	"Prospective, Randomized, Controlled Trial of Ultra-modified Internal Sphincterotomy vs Closed Lateral Internal Sphincterotomy for Chronic Fissure-in-Ano"	5	99
30 🗹 1	6 <mark>14</mark> 27941	[A randomized study	[A randomized study of prophylactic intravesical instillation of		99
		(Hashimura)	pirarubicin (THP) prior to transurethral resection of superficial		

(You can also "filter" on score in ER 4, if you enable the column in your view options, so can get a specific subset of items by score from your classifier.)

				Grosstabs Reports Meta-analysis Collaborate My info 09 in this list in total).				Codes		
howing	nowing: Items classified according to model: 2017 T-A data (pub year 2017 only)							i Ly 🛈) C 🗕 🌱	* X
I	E	Ð	999	🍟 👸 🔚 🏵 👞 🚔 🍕 📰 🔳 🤼 🐥 Filter:					Title/abstract s	-
			Authors	Title	Year 🟹	Score	7		Allocations	
Go		I	Homsy Jaco ; Kir	Primary HIV prevention in pregnant and lactating Ugandan women: A randomized trial	2019	79	Sel	ect All		
Go		I	Samburu Betty N	Effectiveness of the baby-friendly community initiative in promoting exclusive breastfeeding amor	2020	76	66			status
Go		I	Nguyen B T; Sas	The effect of universal health insurance for children in Vietnam	2019	76	67			status partially code
Go		I	Sharma V ; Leigl	Effectiveness of a culturally appropriate intervention to prevent intimate partner violence and HIV	2020	75	68			status Immunisation
Go		I	Nwaozuru Ucheo	By her own hands: Combination income-generating HIV prevention intervention to promote HIV r	2021	75	69			status protocols FSN
Go		I	Heath R ; Hidrob	Cash transfers, polygamy, and intimate partner violence: experimental evidence from Mali	2020	75	70			ers (DO NOT TOUCH
Go		I	Heath Rachel ; H	Cash transfers, polygamy, and intimate partner violence: Experimental evidence from Mali	2018	75	71			
Go		I	Effendy Devi Sav	Nutrition education in Southeast Sulawesi Province, Indonesia: A cluster randomized controlled st	2020	75	72			
Go		I	Myer Landon ; Pl	Integration of postpartum healthcare services for HIV-infected women and their infants in South	2018	74	74			
Go		I	Maman Suzanne	Results from a cluster-randomized trial to evaluate a microfinance and peer health leadership inte	2020	74	75			
Go		I	Maldonado L Y; E	Improving maternal, newborn and child health outcomes through a community-based women's h	2020	74	76		-	
Go		I	de Hoop Jacobus	Women's Economic Capacity and Children's Human Capital Accumulation	2017	74	Show r	ows with v	alue that	
Go		I	Blakstad M M; M	Home gardening improves dietary diversity, a cluster-randomized controlled trial among Tanzania	2020	74	Is equ	al to	•	
Go		I	Settergren Susar	Cluster randomized trial of comprehensive gender-based violence programming delivered through	2018	73				
Go		I	Saleem Naeem H	Do conditional cash transfers help to improve the access of maternal health care services in provi	2019	73	And		•	
Go		I	Ogum Alangea ;	Evaluation of the rural response system intervention to prevent violence against women: findings	2020	73	Is equ	al to	•	
4	1	2	3 4 5 6 7		Page	1				tatistics
	-	4		4/03/2021 User: Zak Ghouze Review: REV ID = 26065 (SiteAdmin manual access)			F	ilter	Clear Filter	tensuts

You can use the decile bands EPPI Reviewer produces to automatically create 10 codes within a codeset which are assigned to each item according to their band of probable relevance, useful for creating allocation codes or focusing on specific bands. (For example, you may wish to exclude all those items in the lower few bands en masse i.e. all items unlikely to be relevant / likely to be irrelevant, or you may wish to manually screen the items around the middle bands, where the system isn't sure if they will be includes or excludes, or you may wish to work with only those items the system judges likely to be relevant i.e. those in the uppermost bands.)

If you select a code in the code tree on the right, you can then get the system to automatically make 10 child codes for the 10 bands by clicking the *Create* button. (You can easily create a code specifically for this purpose.)



The child codes will be applied to your items according to which band they are in.



So, for example, you can list and operate on the items in the 90-99% band by simply listing the items with the relevant code.

Import II First		ster Coding Report 🔹 🔳	Last Showing 100 items of 14103 View Options	Enhanced selection is: On	Allocation All title and abstract
Showing	90-99% range.	Short title:	Title	I E D Year Info	 Probability RCT 0-9% range 10-19% range
60 🗆	I 61387832	"Prospective, Randomized, (Sungurtekin)	"Prospective, Randomized, Controlled Trial of Ultra- modified Internal Sphincterotomy vs Closed Lateral Internal Sphincterotomy for Chronic Fissure-in-Ano		20-29% range 30-39% range 40-49% range
<u>;</u>	1 61427941	(A randomized study (Hashimura)	[A randomized study of prophylactic intravesical instillation of pirarubicin (THP) prior to transurethra resection of superficial bladder cancer]	(50-59% range 60-69% range 70-79% range 80-89% range
30	1 61350240	[Application of Alpha1- adrenergic (Wang)	[Application of Alpha1-adrenergic antagonist with extracorporeal shock wave lithotripsy for lower uret stone]	eral	90-99% range
50 🗆	1 61383198	[Approach to percutaneous (Sedano- Portillo)	[Approach to percutaneous nephrolithotomy. Comparison of the procedure in a one-shot versus i sequential with metal dilata]	the	Code description: FROM: Items classified according to model: RCT

(You can see the items' coding via a *Frequency Chart*. Clicking on the numbers brings up a list of those items.)

Review hor	me References Reports Search & Classify	Collaborate	Edit Tools With this Code 🔹 🗷 🖨
Frequencie	is and crosstabs Configurable reports V Run Reports		Allocation Get Frequence All title and abstract
Rows: Columns:	Probability RCT Set Not set (only used for Crosstabs) Set	Get Frequencies	Probability RCT
Filter:		Filter Clear Filter	Age Screen on title and abstract Code description:
Get Frequ			Probability items are an RCT
Show re	sults as: 💽 Table O Pie chart O Bar chart 🛛 🖉 Show	'None of the codes above' Export D	A
0-9% ra	nge	35319	
10-19%	range	8769	
20-29%	range	4268	
30-39%	range	2704	
40-49%	range	2098	
50-59%	range	1966	

Creating Custom Classifiers

Further to the built-in classifiers described above, you can also create your own classifiers based on existing data.

You may have a batch of coded or screened items in the review that you want to use as the basis for a model. You may have sets of references elsewhere that you wish to import, coding each batch with an appropriate code for the model to work with. (You could also, for example, take references from another review, or combine your existing priority screening coding into two codes (*relevant* for items with an include code and *irrelevant* for items with an exclude code) and use that as the basis for a model.)

Your model will learn to classify items according to whether they are relevant or irrelevant, based on the dataset (refs and coding) you base the model on.

Review home References	Reports Search & Classify	Collaborate				
New Search Refresh List Delete Selected Combine 🔹 Build Model Classify						
Run Search	With this code 👻					
	INCLUDE on title & abstract					
	Included O Ex	cluded				

To create a new model, click the *Build Model* button (still under the *Search & Classify* tab).

Next, select the 2 codes which define the items which are relevant and those that are irrelevant. Give the model a name and click the next *Build Model* button.

	Build	Model
Learn to apply this code:	Relevant 🝷	
Distinguish from this code:	irrelevant 👻	
Name for your model:	Example	
Delete Selected Model(s)	Build Model	Refresh Models

As the process takes some time, you will see a confirmation that your request was submitted.

request was submitted ×

The system will work through the data and build its model. Whilst it is processing, you can use other functions within the software.

(If you wish to check if the model has been built, you can click the *Refresh Models* button. Whilst it is still building, the entry in the list of models will show "in progress" and the associated figures will all show as zero.)

Learn	Learn to apply this code:		Rele	vant 👻				Close
Distin	guish from this c	ode:	Irrelevant 🔫					
Name	e for your model:	E	xample					
Del	Delete Selected Model(s)		Build Model	C	efresh Models			
	ReviewId	Title	Att On	Att Not On	Accuracy	AUC	Precision	Recall
	25830	Example (in progress)	Relevant	Irrelevant	0	0	0	0

When the processing has completed, you will see the model listed with associated figures. (These figures are detailed later in this document.)

	ReviewId	Title	Att On	Att Not On	Accuracy	AUC	Precision	Recall	
	25830	Example	Relevant	Irrelevant	0.907	0.94	0.306	0.807	*

Applying Custom Classifiers

To apply your newly built classifier, click the *Classify* button in the normal way.

Review home	References	Reports	Search & Classify	Collaborate
New Search Re	efresh List Delet	e Selected	Combine 🔻 Build Ma	odel Classify

Click the option to *Show Custom Models per the current review* (1) option, select the relevant model from the list below (2), then select which items are to be classified (3).

Revie	Review home References Reports Search & Classify Collaborate									
New	New Search Refresh List Delete Selected Combine 🔻 Build Model Classify									
Ste	ep1: Select a	Model		Step 2: Choose method to apply						
	L	ong COVID b.	inary model		3	Apply to a O Apply to i	all items in rev tems with this			
	C	COVID-19 maj	o categories				tems from thi			
		Cochran	e RCT							
		Economic E	valuation							
		Systematic	Review							
		Origina	I RCT							
1	Show Cust	om Models p	er the currer	nt review						
	Show	Custom Mod	els for all revi	ews						
C	ReviewID	ModelID	Title	Applies	Compar	Precision	Recall	Rebuild		
୭	25830	2268	Example	Relevant	Irrelevant	0.306	0.807	Rebuild		
	25830	1202	Most used Exclude	INCLUDE on title & abstract	Most used excludes	0.256	0.636	Rebuild		

You can also create models in other reviews and apply them in your current review. (You may have other data or existing models you wish to apply to items in a different project.)

Step1: Select a Model

Long COVID binary model
COVID-19 map categories
Cochrane RCT
Economic Evaluation
Systematic Review
Original RCT
Show Custom Models per the current review
Show Custom Models for all reviews

(Equally, you can apply the model (or classify) items with a particular code, or items from a particular source – as with the built-in models.)

Step 2: Choose method to apply

- Apply to all items in review
- Apply to items with this code
- Apply to items from this source

Please select a code from the right.

When you have the parameters as you wish, click the *Run Model* button.

Step 1: Select a Model Step 2: Choose method to apply								
Long COVID binary model					l items in review			
F	CO	/ID-19 map categor	ies		ems with this code ems from this sour			
		Cochrane RCT		Current code	: Unscreened 17-5			
	E	conomic Evaluation						
		Systematic Review						
		Original RCT						
	Show Custom	Models per the cu	Irrent review					
Show Custom Models for all reviews								
		Stoff Would for all						
3	ReviewID	ModelID	Title	Applies	Compared wi	Precision	Recall	Rebuild
	ReviewID 25830			Applies Relevant	Compared wi Irrelevant	Precision 0.306	Recall 0.807	Rebuild Rebuild
		ModelID	Title					

You will be asked to confirm the action, as it can take some time to complete.

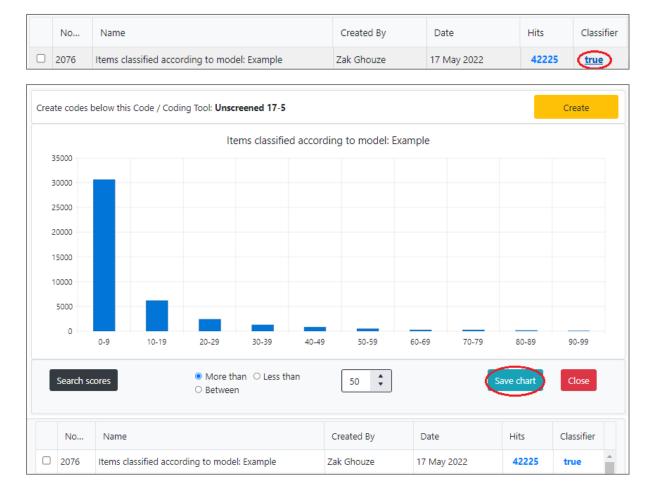
Please confirm						
Are you sure you wish to run the selected model ?						
Cancel						

You will then see a message, suggesting you refresh the Search list to check for the results coming in. (Again, you can continue to use other functions in EPPI Reviewer whilst the system runs the model.)



When the results are ready, you will see them listed, with the line showing *true* in the "Classifier" column. Click the *true* link to visualise their distribution according to probability of being relevant.

The chart can be saved as a graphic via the *Save chart* button.



You can search for certain ranges of probabilities; the results will be listed in your searches. Enter the parameters and click the *Search scores* button.

	Search sc	ores	○ More than ● Less than ○ Between	60		Save chart	Close	
	No	Name		Created By	Date	Hits	Classifier	
	2079	Search #771394 score	s less than 60	Zak Ghouze	17 May 2022	41675	false	

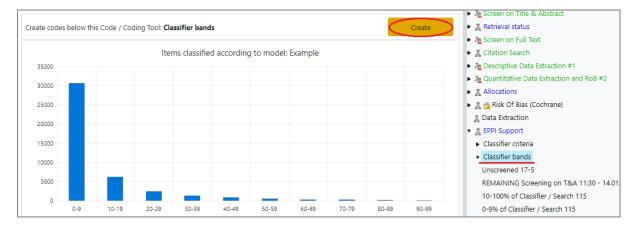
You can also list the items classified in order of probable relevance by clicking on the number of *Hits*, as described for built-in classifiers earlier.

	No	Name			Created By	Date	Hits	Classifie	r	
	2076	Items cla	assified according to mode	el: Example	Zak Ghouze	17 May 2022	42225	true	*	
Rev	Review home References Reports Search & Classify Collaborate									
Im	port Items	Clus	ster Coding Report 🔹 🔳	In/Exclude Export to	RIS * Run Reports					
Fi	irst Pre	vious Pag	ge: 1 of 11 Next	Last Showing 4000 ite	ms of 42225 View Optio	ns	Enhanced	selection is: C)ff	
s	howing Iten	ns classified a	eccording to model: Example				I E I	þ		
		ID	Short title	Title			Year	Score 1		
G	0 0 1	54977122	Ban, (2020)	Self-help groups, savin in Cambodia	gs and social capital: Ev	idence from a field expe	eriment 2020	98		
G	0 0 1	54834975	Adoho (2014)	The impact of an adole Liberia	escent girls employment	program: The EPAG pro	oject in 2014	1 97		
G	0 0 1	54978725	Naved, (2018)		controlled trial to assess st women and girls in sl			3 96		
G	0 0 1	54977125	McIntosh (2012)	The CLP's impact on w	omen's empowerment		2012	2 95	1	
G	• • •	54977119	Muhammad, (2012)	Women empowerment	t and microfinance: A ca	se study of Pakistan.	2012	2 95		
G	•	54834794	Buehren, (2017)	Adolescent Girls' Empo Evidence from South S	owerment in Conflict-Aff udan	fected Settings: Experim	ental 2017	7 95		

Finally, the system can automatically create 10 codes based on the decile bands of probability, with each item being given one of the 10 codes accordingly.



Select the parent code where you want the 10 new codes to appear, then click the *Create* button.



The codes will be created and applied as requested.

Rows:	Classifier bands Set Get Frequenci	es		Retrieval status
Columns:	Not set (only used for Crosstabs) Set		۰.	Litation Search
Filter:	Not set (optional) Set Filter Clear Filt	er	• 2	le Descriptive Data Extraction #1
Get Freq	uencies Get CrossTab	Current code: Classifier bands		le Quantitative Data Extraction and RoB
Get Fleq	dendes der crossrab	Current code. Classifier ballds	•	& Allocations
Show re	sults as: 💿 Table 🔿 Pie chart 🔿 Bar chart 🖉 Show 'None of the codes above' 🗈	kport 🕅	•	🤱 🙆 Risk Of Bias (Cochrane)
		A	1	👢 Data Extraction
Code		Count	۰.	🙎 EPPI Support
0-9% ra	nge	30576	1.1	 Classifier criteria
			· ·	 Classifier bands
10-19%	range	6109		0-9% range
				10-19% range
20-29%	range	2399		20-29% range
				30-39% range
30-39%	range	1304		40-49% range
40-49%		794		50-59% range
40-4970	Tange	/ 34		60-69% range
50-59%	range	493		70-79% range
	-			80-89% range
en enn/	52520	247		90-99% range

The numbers

The Recall and Precision figures are built in the following way -:

- Upon receiving the training data, a proportion (10%) of randomly selected records is set aside for the purpose of calculating the "performance" figures including precision and recall. (The algorithm will "fail" if it doesn't receive enough records to train and then evaluate the result.)
- The remaining portion (the 90% majority of records) will then be used for training. Once the classification model is built, the initial "set aside" portion is used to evaluate performance. For this, precision and recall are calculated by setting the cut-off threshold at 0.5 (or "50%").
- (NOTE: Rebuilding models, especially when there isn't a great deal of training data, can produce slightly different precision and recall numbers, depending on what gets randomly selected for "evaluation".)
- When you apply a model, you will hopefully find the majority of references that scored more than 50% are "true positives" i.e. have been classified correctly. You may also find a tail of true positives below the 50% score, generally accounting for a small minority of all true positives.
- Accuracy is usually calculated using precision / recall, and is produced from the same 10% records used for the other performance stats. (The specific formula used for the Accuracy score is documented here: <u>https://scikit-</u> <u>learn.org/stable/modules/model_evaluation.html#accuracy-score</u>

$$extbf{accuracy}(y, \hat{y}) = rac{1}{n_{ ext{samples}}} \sum_{i=0}^{n_{ ext{samples}}-1} 1(\hat{y}_i = y_i) \, .$$

You'll find some background information here <u>https://paulvanderlaken.com/2019/08/16/roc-auc-precision-and-recall-visually-explained/</u> and formal definitions here <u>https://en.wikipedia.org/wiki/Precision_and_recall</u>. (Excuse the source, but it's not a bad summary.)

We have a document on the Machine Learning available within ER Web available at https://eppi.ioe.ac.uk/CMS/Portals/35/machine_learning_in_eppi-reviewer_v_7_web_version.pdf.

(Other information on automation in ER can be found at <u>https://eppi.ioe.ac.uk/cms/Default.aspx?tabid=3772</u>. Other information about the software can be found at <u>https://eppi.ioe.ac.uk/cms/Default.aspx?tabid=3822</u>.