Extracting Key Phrases and Relations from Scientific Publications

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Main Goals

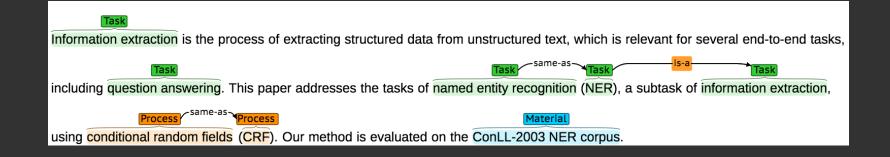
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Provide a solution to SciencelE (from SemEval 2017) - Natural Language Processing 2

Produce a 'product like' proof of concept presenting a use of data produced

What is SciencelE?

- Key Phrase (KP) identification (A)
- KP classification into (B)
 - Task
 - Process
 - Material
- Relation extraction (C)
 - Hyponym-of
 - Synonym-of



Example

Formatted Example

Original Text

In this paper a comparison between two popular feature extraction methods is presented. Scale-invariant feature transform (or SIFT) is the first method. The Speeded up robust features (or SURF) is presented as second. These two methods are tested on set of depth maps. Ten defined gestures of left hand are in these depth maps. The Microsoft Kinect camera is used for capturing the images [1]. The Support vector machine (or SVM) is used as classification method. The results are accuracy of SVM prediction on selected images.

BRAT Annotated Format

11	Process 88 121	Scale-invariant feature				
transf	orm					
T2	Process 126 130	SIFT				
T3	Process 157 183	Speeded up robust features				
T4	Process 188 192	SURF				
*	Synonym-of T1 T2					
*	Sýnoným-of T3 T4					
T5	Material 257 267	depth maps				
T6	Process 398 420	Support vector machine				
T7	Process 425 428	SVM				
*	Synonym-of T6 T7					
T8		arison between two popular				
feature extraction methods						
T9	Process 332 355	Microsoft Kinect camera				
T10	Process 492 495					
T22	Task 47 65 featu	re extraction				
T23	Process 441 455					
T27		classification method				
R2	Hyponym-of Arg1:	T7 Arg2:T27				

SciencelE Data

- ▶ 50 development documents
- > 350 training documents
- ▶ 100 test documents
- BRAT annotation format
- Short documents
 - ► Not full publications!

Background

- At SciencelE
 - Support Vector Machine (SVM)
 - Neural Networks
 - Recurrent Neural Networks
 - ▶ Long Short Term Memory networks
 - Conditional Random Fields
 - Gazetters
- Other
 - Clustering
 - Rule based systems for relation extraction

Background - Word2Vec

- A vector space containing word vectors, with similar words being near each other
- E.g.: "knee is to leg" as "elbow is to [forearm, arm, ulna_bone]" (produced using Word2Vec)
- Several large pretrained models:
 - Google News (vocabulary size of 3 million different tokens)
 - Freebase (vocabulary size of 1.4 million)
 - Extension could be automatically build a model based on just scientific papers
 - ▶ All data from SciencelE was too small to build a meaningful vector space
- I use the 'Deep Learning for Java' (DL4J) Word2Vec library

Project Architecture

- Java 8
- Maven 3
 - Supports importing libraries
 - Supports exporting this as a library (for part 2 of the project)



Preprocessing

- All data passed through Stanford's CoreNLP. This produces:
 - Sentences
 - Tokenisation
 - Parse trees of sentences
- I utilise Java serialisation to help cut down experiment run times



Key Phrase Extraction

- Support Vector Machine (libsvm implementation)
 - ► Radial Basis Function, C = 50, $\gamma = 0.5$
 - 12 support vectors
 - Training data tokens labelled as key phrases
 - Mainly based around position in text (and relation to other words)
 - Word2Vec distances and other token attributes (e.g. TF-IDF) also considered
 - Cross Validation to help with parameter optimisation
- Post processing
 - Remove badly formed key phrases
 - Remove low TF-IDF words

Key Phrase Extraction (Other)

- Also attempted hierarchal clustering
 - Bag-of-words approach
 - Word2Vec to define spacing
 - Didn't work well...

Key Phrase Classification

- Word2Vec distances
 - Average and closest distance between bag-of-words tested
 - Option to ignore unimportant words
 - Default classification (catches words not in dictionary)
 - 221 couldn't be classified: 3 "task", 81 "process" and 137 "material"
 - Attempted to see if these were in WordNet but unfortunately not
- Also prove the position of words doesn't really give any indication of class (similar SVM to before)

Relation extraction

- Based on Word2Vec relative distances
- SVM
 - One for hyponyms (218 pairs), one for synonyms (207 sets)
 - ► Around 50,000 possible combinations
 - Relative distances between phrases
 - Angel and length of distance also tested between phrases
- Simple rules

Testing

- How to test key phrases?
- SciencelE evaluation scripts use BRAT libraries
- My evaluation
 - Key phrase extraction comes in flavours:
 - ► Harsh (exact phrase matches)
 - ▶ Reasonable (matching gold and predicted phrases)
 - Generous (gold phrases are equal to or subphrases of predicted)
 - Classification compared gold to predicted (on gold key phrase extraction)
 - Relation correctness between gold key phrases

Results - ScienceIE Scripts (F1 Scores)

C L'	SciencelE		My Evaluation			
Section	Individual Best / Average	End-To-End Best / Average	Individual Best	End-To-End Best		
KP Extraction	0.56 / 0.38	0.56 / 0.38	0.20	0.20		
KP Classification	0.67 / 0.57	0.44 / 0.26	0.55	0.11		
Relation extraction	0.64 / 0.43	0.28 / 0.07 0.1		0.28 / 0.07 0.1		0.02
Overall	N/A	0.43	N/A	0.11		

Results - My Evaluation

- Key phrase extraction F1 scores:
 - Very strict: 0.2 (I agree with the SciencelE score)
 - When matching on just the phrase: 0.36
 - When being generous: 0.74

The Product: The Website: ExtractorIE

- Java based web project, with the NLP discussed as a library through Maven
- MySQL database
- Uses Spring Boot 1.5
 - Produces convenient self-contained jar
 - Integrated Apache Tomcat
 - Useful libraries for MySQL
 - Hibernate
- Interesting Visualisations:
 - Donut chart (using d3.js)
 - Word Cloud (using jqcloud.js)
- Custom search prioritising search results with key phrases connected to query

See ExtractorIE

http://tomclarke.xyz

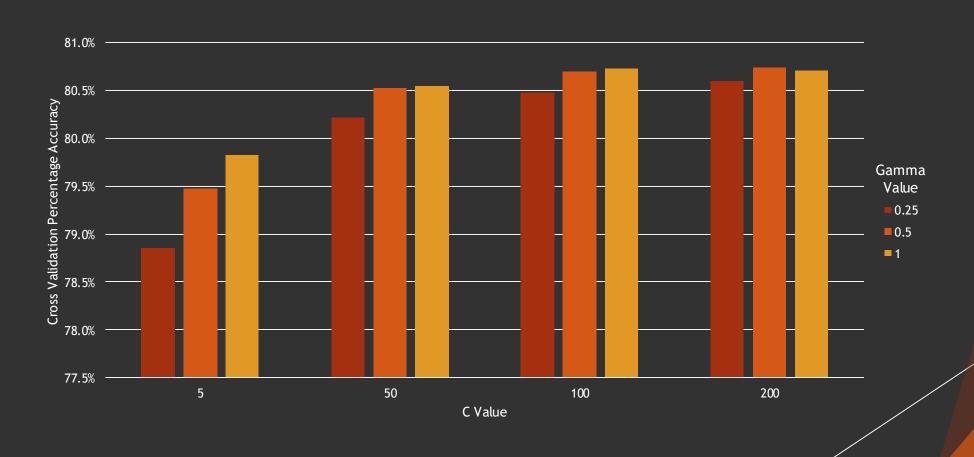
Thank you for listening!

Any questions?

Key Phrase Extraction SVM: Support Vectors

- Word length / max word length
- Part-Of-Speech type
- 3. TF-IDF
- Whether the word is a stop word
- Depth in document (position / total tokens)
- 6. Depth in sentence
- 7. Whether the token is in the first sentence
- 8. Whether the token is in the last sentence
- Parse tree depth
- Whether the previous token was a key phrase
- 11. Similarity to "task"
- 12. Similarity to "process"
- 13. Similarity to "material"

Key Phrase Extraction SVM: Cross Validation





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ExtractorIE

Extracting Key Phrases and Relations from Scientific Publications

For more information on what the project is based upon, see the ScienceIE shared task.

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Hyponyms

14

Synonyms

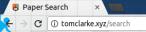


Search for papers

Enter terms to search for and focus on task, process or material related items:

Search terms or phrases

Focus on: ■ Task ■ Process ■ Material



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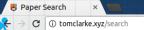
Enter terms to search for and focus on task, process or material related items:

support vector machine Submit

Focus on: ■ Task ■ Process ■ Material

Search for "support vector machine" completed in 0.257 seconds, finding 7 papers.

ID	Paper Title		KPs / Rels	
	S221267161400105X Kinect camera is used for capturing the images [1]. The support vector machine (or SVM) is used as classif	•	15/0	•
84	S1877750315000460 and build HemeLB on any remote resource, to reuse machine-specific configurations, and to organize a	④	22 / 6	•
	S0021999112003579 We order the discrete unknowns so that the vector of unknowns, xPS=[X,L], contains the nx unk	•	14/0	•
	\$0098300414000259 event that future versions of Hadoop are optimized to support paradigms other than MapReduce, Pig script	④	19/0	•
	S107158191630074X the simulation (e.g., the input buffer of a certain machine at the time of refilling it). A relatively	•	22 / 0	•
	S002002551630384X hulls, such as Vizster [22]. However, they do not support visualizing set overlaps	•	12/0	•
	S1877750313001269 is itself very important. Grid and cloud computing support different interaction models. In grid comp	•		•



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Search for papers

Enter terms to search for and focus on task, process or material related items:

test

Focus on: Task Process Material

Search for "test" completed in 0.613 seconds, finding 19 papers.

ID Paper Title

S0263822312000657 bamboo shows that they are all graded with their greatest strength on the outside, in areas where the g...

S0301679X14003272

Paper Title

S0301679X14003272

Paper Title

S0301679X14003272

Paper Title

S0301679X14003272

	\$0263822312000657 bamboo shows that they are all graded with their greatest strength on the outside, in areas where the g	•	18 / 1	•
	\$0301679X14003272 Q, is measured and recorded throughout the entire test by a piezoelectric load cell which is connect	•	19/0	•
	S0021999113006955 The test cases confirm that the high-order discretisat	•	34 / 0	•
	S0370269303017222 making the predicted Roper mass heavier than the lightest negative parity baryon mass. Pairwise spin-de	•	19/0	•
	\$0011227514002136 thermometry (which was not available at the time of test ing but which will be used for the mKCC), as t	•	13/0	•
	S0021999113004555 Three Runge–Kutta IMEX schemes were tested by Ullrich and Jablonowski [23] for the HEV	•	19/0	•
	S0257897213004131 Fig. 7 shows the relationship between the test ing time and friction coefficients of various	•	19/0	•
89	\$0011227515000648 #1/#2 and the others is the most influential on the test results. The redesign and upgrade to 110-nm p	•	16/0	•
	S0301679X14000449 RH ceramics was smaller than Sc,critical under all test ed conditions during the initial stage of fric	•	28 / 0	•
	S221267161400105X SURF) is presented as second. These two methods are test ed on set of depth maps. Ten defined gestures	•	15/0	•

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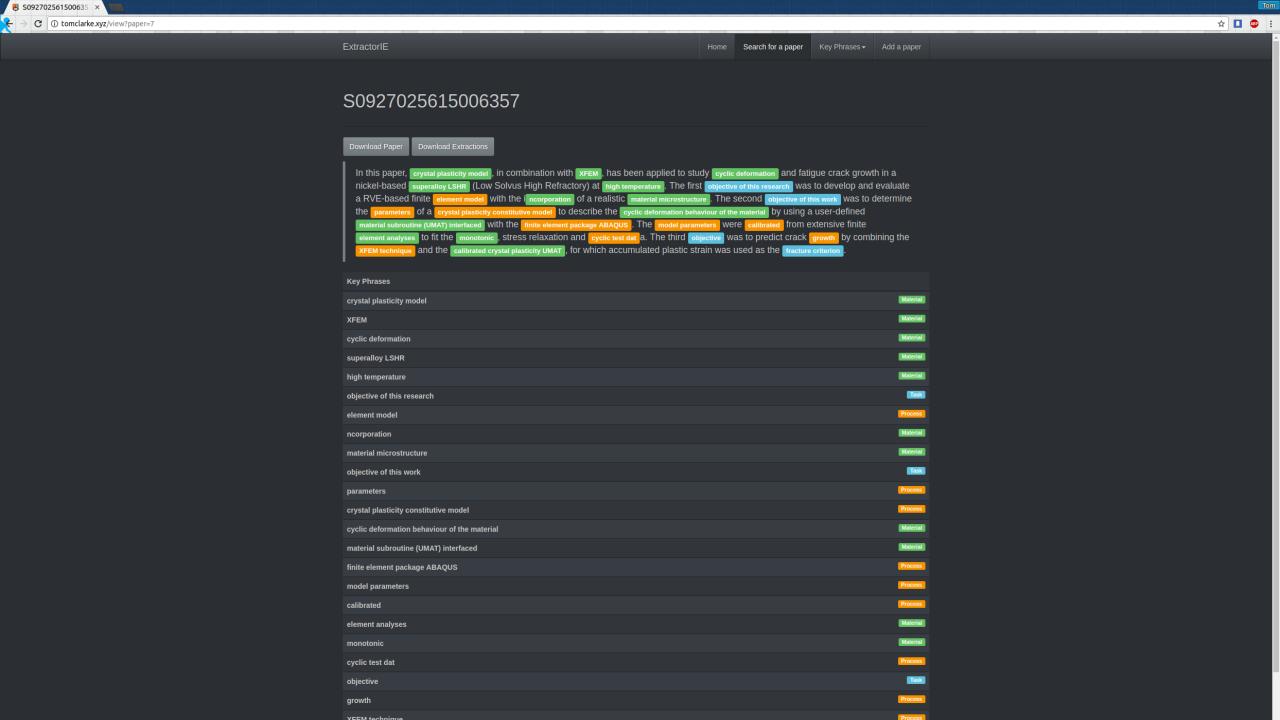
Search for papers

Enter terms to search for and focus on task, process or material related items:

test

Focus on: ☑ Task ☐ Process ☐ Material

Search	Search for "test" completed in 0.597 seconds, finding 19 papers.					
ID	Paper Title		KPs / Rels			
	S0263822312000657 bamboo shows that they are all graded with their greatest strength on the outside, in areas where the g	⊕	18 / 1	•		
	S0301679X14003272 Q, is measured and recorded throughout the entire test by a piezoelectric load cell which is connect	•	19/0	•		
	S0021999113004555 Three Runge–Kutta IMEX schemes were tested by Ullrich and Jablonowski [23] for the HEV	•	19/0	•		
98	S0301679X14000449 RH ceramics was smaller than Sc,critical under all test ed conditions during the initial stage of fric	④	28 / 0	•		
	S221267161400105X SURF) is presented as second. These two methods are test ed on set of depth maps. Ten defined gestures	•	15/0	•		
68	S0021999113006955 The test cases confirm that the high-order discretisat	•	34 / 0	•		
44	S0370269304012638 SU(N?1)×U(1). There have appeared two independent F6 tests of this conjecture [19,20], with conflicting	④	15/0	•		
	S0370269303017222 making the predicted Roper mass heavier than the lightest negative parity baryon mass. Pairwise spin-de	•	19/0	•		
	S0011227514002136 thermometry (which was not available at the time of test ing but which will be used for the mKCC), as t	•	13/0	•		
	S221450951400014X According to Fig. 2 and the results of the Marshall tests , the optimum bitumen measures decrease signi	•	28 / 0	•		
	Show 9 more					



🛮 🕲 E All found key phrases, grouped by classification. To generate the below graphs, key phrases are broken up into tokens, and the size of a token is relative to it's TF-IDF value. dimensional rgument gale rchaeological self bargmann_{brane} squares first mages spring dualitybsorbance nucleus slider camerloop nitial vs ssue dashpotnformalntspectr ads ndustry principles. Il su least urban denotes stigmatism kernin unitarity processing multi nterpretation lisflood na http www.kiber assembly.nput http wwwkiber assermby doption mplementations ssessment spe doption cm_although mol co cryo su rearving de bility brillouin nalysis com tems spin aaal energy ccuracy az cumg automatimproved redshift mplicit nfrastructurekirkwood ____nitial __ntensity ssociatedhigh ccomplished half mproved fferent pseudo repproach distinct benzyl transconductance received benzyl ggregate pre nformal sigmadistance catalysts quasiends xial silic jileading short non-self non-self dirtefactsgrown coriented cullet soot stigens wave nitial nelasticity mvs^{gain} ilipplication elaz assembles complementary posterior coord hydrozincte imidazolium starchy aa soil agglomerates medi hydroxyethylpyrrolidinonium