

Bibliometrics: Predicting Publication Success

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Clemson University

Introduction

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Knowledge is power

- Sir Francis Bacon

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- 2.5 million scientific papers published each year¹
- Research trends
- Research funding

¹The STM Report; Fourth Edition, March 2015

Background: Bibliometrics

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Background: Bibliometrics

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Defining the Problem

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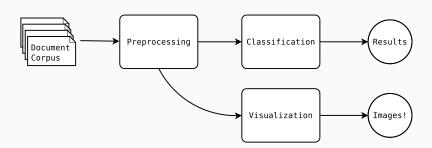
- Given: a corpus of academic publications
- Goal: predict which papers will be "successful"
- Success Metric: citation count
- Success Definition: citation count > median citation count for that cluster
- Evaluation Metric: accuracy

Overview

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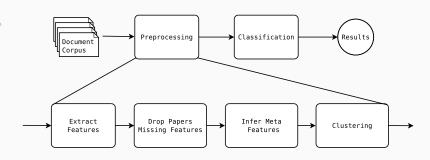
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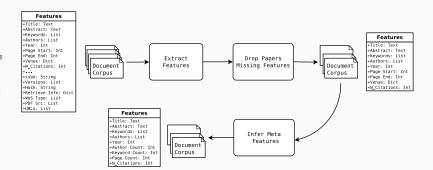


Extraction, Thinning, and Meta Features

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- Initial Corpus: 22,588
- Removed:
 - ☐ Papers after 2013
 - □ Outliers
 - ☐ Papers missing features
- Final Corpus: 4,914

Clustering

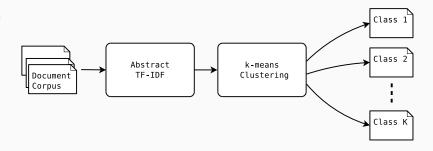
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Clustering: Choosing K

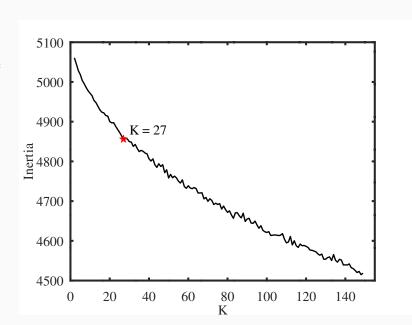
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Clustering: Good Clusters

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Cluster	Count	Keywords
Clustel	Count	reywords
0	15	alice, bob, girls, programming, communication
4	106	video, videos, 3D, quality, streams, users
15	170	query, XML, search, data, databases
17	869	design, people, user, information, research
20	93	internet, TCP, network, protocol, congestion
22	60	privacy, private, data, information, awareness

Visualization

Yearly Publications

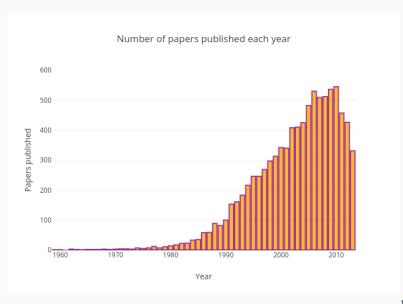
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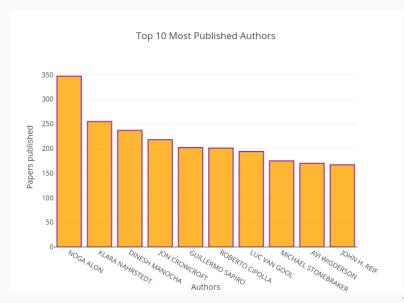


Author Publications

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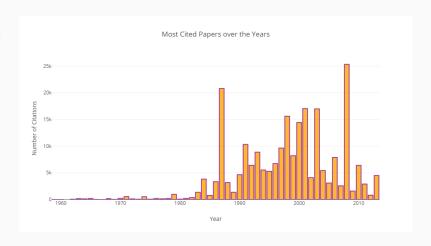
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Popular Publications

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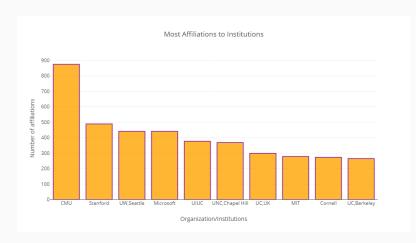
Institution Publications

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Classification

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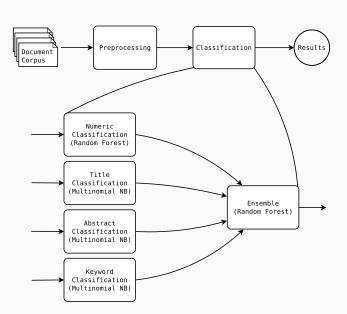
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Random Forest

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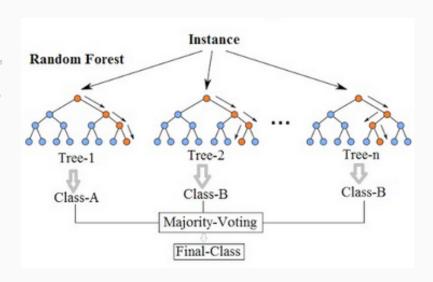
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Multinomial Naive Bayes

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$$\begin{split} \hat{c} &= \underset{c \in C}{\operatorname{argmax}} \ P(d|c)P(c) \\ P(d|c) &= P(f_1, f_2, ..., f_n|c) = P(f_1|c)P(f_2|c)...P(f_n|c) \\ P(c) &= \frac{N_c}{N_d} \end{split}$$

- c: Class
- d: Document
- f_i : Feature
- N_c : Number of words in class c
- $-N_d$: Number of words in document d

Classifier Accuracy

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$$\label{eq:accuracy} \text{Accuracy} = \frac{\text{TP} + \text{TN}}{\text{TP} + \text{TN} + \text{FP} + \text{FN}}$$

Training Method

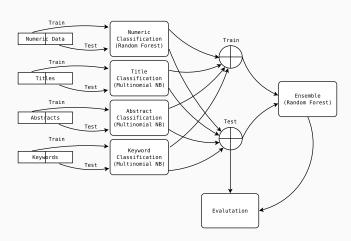
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- test_train_split: 50%
- GridSearchCV
- CV = 5
- RF Params: n_estimators and max_features

Results

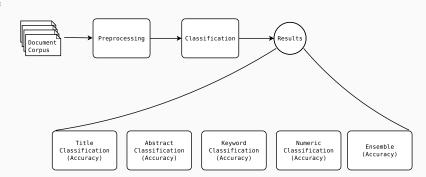
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Classification Parameters

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	Numeric Classifier Parameters			
Cluster	n_estimators	max_features		
4	10	2		
15	10	6		
17	8	2		
20	12	2		
22	6	4		

Classification Parameters

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	Ensemble Classifier Parameters			
Cluster	n_estimators	max_features		
4	1	1		
15	4	1		
17	1	1		
20	1	1		
22	1	1		

Feature Importance

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Numeric Feature Weight Cluster Title Abstract Keyword Year Author Page (length) (length) (length) Count Count 4 0.089 0.117 0.143 0.240 0.188 0.223 0.160 0.175 15 0.157 0.121 0.321 0.065 17 0.136 0.253 0.183 0.089 0.165 0.174 20 0.087 0.214 0.242 0.109 0.228 0.121 22 0.140 0.038 0.066 0.556 0.155 0.045

Feature Importance

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	Ensemble Feature Weight					
Cluster	Numeric	Title	Abstract	Keyword		
	Classifier	Classifier	Classifier	Classifier		
4	1	0	0	0		
15	0.420	0.271	0.059	0.250		
17	0.917	0	0.083	0		
20	1	0	0	0		
22	1	0	0	0		

Classification Accuracy

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	Accuracy (%)				
Cluster	Numeric	Title	Abstract	Keywords	Ensemble
4	60.4	45.8	56.3	58.3	60.4
15	63.0	64.2	65.4	55.6	61.7
17	64.2	59.1	59.6	62.5	68.5
20	64.3	42.9	35.7	38.1	64.3
22	64.3	0.5	0.5	0.5	64.3

- Overall Accuracy (All Clusters): 57.7%

- Overall Accuracy (4,15,17,20,22): 66.5%

Conclusions

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Classificatio

- Cleaning data is important
- Some clusters are better than others
- 66.5% prediction accuracy in optimal clusters
- Future work: targeted dataset
- Controlling for year

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WIN ARE HATS SO EXPOSIVE IN UNITY DO TUNIS HAVE DIFFERENT FINEERPRINTS SUITY IS HITTPS (ROSSED OUT IN REL BURY DO YOUR BOOGS HURT? WHY ARE AMERICANS AFRAID OF DRAGONS SUITY IS THERE A LINE. THROUGH HITMS OF DRAGONS SUITY BOOGS HURT? WHY ARE AMERICANS AFRAID OF DRAGONS SUITY IS THERE AD UNE RISKOVED HURTS ON PREBOO MHY IS HTTPS IMPORTANT WHY AREN'T MY ARMS GROWING Y POI FEEL WHY ARE THERE DOORS ON THE FREEWAY # WHY ARE THERE KICKING IN MY STOMACH WHY ARE THERE KICKING IN MY STOMACH WHY ARE THERE TWO SUASHES AFTER HTTP WHY ARE TEXT MESSAGES BLUE LIHY ARE THERE **GHOSTS** SOUIRRELS UHY DO Q TIPS FEEL GOOD WHY AREN'T 100 THERE GUNS IN WHY IS SEX HARRY POTTER 50 IMPORTANT PROSPITING SO HARD O WHY AREN'T THERE E GRADES THESE A O CHI RESSIDE O WHY IS ISOLATION BAD OF LIM OF REPRESENTED SOURCE OF THE TOTAL CONTINUE AND THE TOTAL CONTIN

https://xkcd.com/1256

WHY AREN'T THERE ANY FOREIGN MILITARY BASES IN AMERICA

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WHY ARENT BULLETS SHARP TO WHY IS LYING GOOD F