

Modeling Community Question-Answering Archives Zainab Zolaktaf, Fatemeh Riahi, Mahdi Shafiei and Evangelos Milios Email : <u>z.zolaktaf@gmail.com</u>

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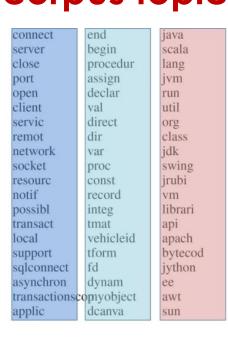
Simple intuition

- Corpus contains two types of topics, Question topics (Qtopics) and Answer topics (A-topics)
- Each question is a mixture of Q-topics
- Each answer is a mixture of A-topics

Inference and Parameter Estimation

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- Inference Output

Per-corpus topic distributions



Community Question-Answering Archives

Abstract

Community Question Answering (CQA) services contain large archives of previously asked questions and their answers. We present a statistical topic model for modeling Question-Answering archives. We refer to our model as Question-Answering Topic Model or QATM. The model explicitly captures relationships between questions and their answers by modeling topical dependencies. To analyze the model, we use it for Question Answering and Automatic Tagging. Our model achieves improved performance in retrieving the correct answer for a query question compared to the LDA model. In addition, for large numbers of topics, our model achieves better clustering performance in terms of F-measure.

Example Question-Answer Pair

Question:

How do I replace all occurrences of a word in a document with another word? Any solution is welcome!

Answer:

Regular expressions(regex) allow you to search and manipulate text based on patterns. In some languages, standard string manipulation functions are available e.g. replaceAll() method from Java's string class. In other languages, such as Perl, regex's are integrated into the language itself. Utilities such as grep, vi, etc can also perform this type of pattern matching and replacement.

Properties of Question-Answers

- Answer topics influenced by question topics
- Answer topics more technical and specific
- Answers may contain additional topics that are correlated with question topics

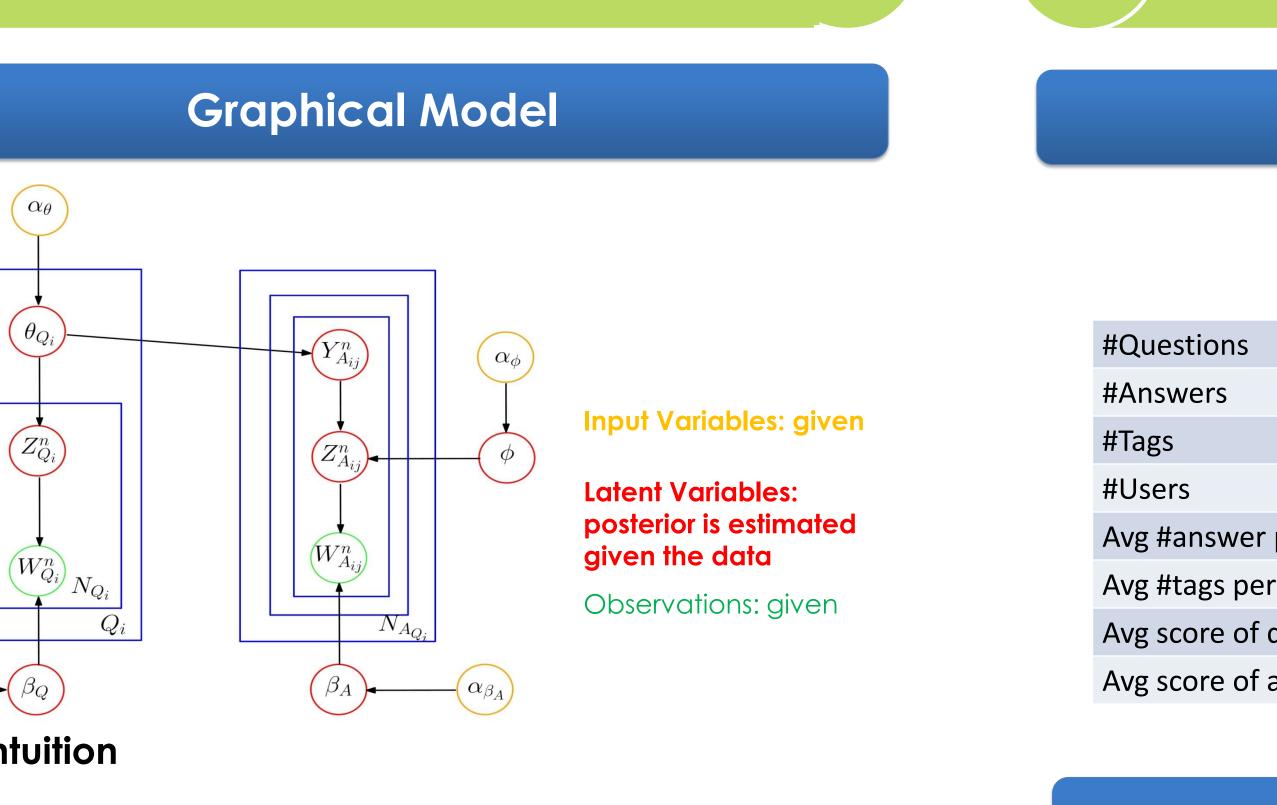
Goal

Discovering the topical structure of CQA archives

- Automatic Tagging: Annotating the content with topics discovered to help browse and understand the archive.
- Question Answering: Given a newly submitted question, use estimated topics to retrieve relevant question-answers from the archive.

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QATM



Documents exhibit multiple topics

Topics in answers are influenced by topics in question

Generative Process

Documents are generated by realizations of random variables that are sampled from probability distributions

Model captures the dependency between question topics and answer topics by conditioning each A-topic in an answer on Q-topics drawn from the topic distribution of the corresponding question.

Essentially, an optimization problem: Model specifies a joint probability distribution with some parameters. What are the optimal values of the model parameters?

To fit the model to data, the generative process is inverted and parameter values are generated from given observations(the words)

Exact inference is intractable. Gibbs sampling, an approximate inference algorithm is used.

Per-word topic assignments

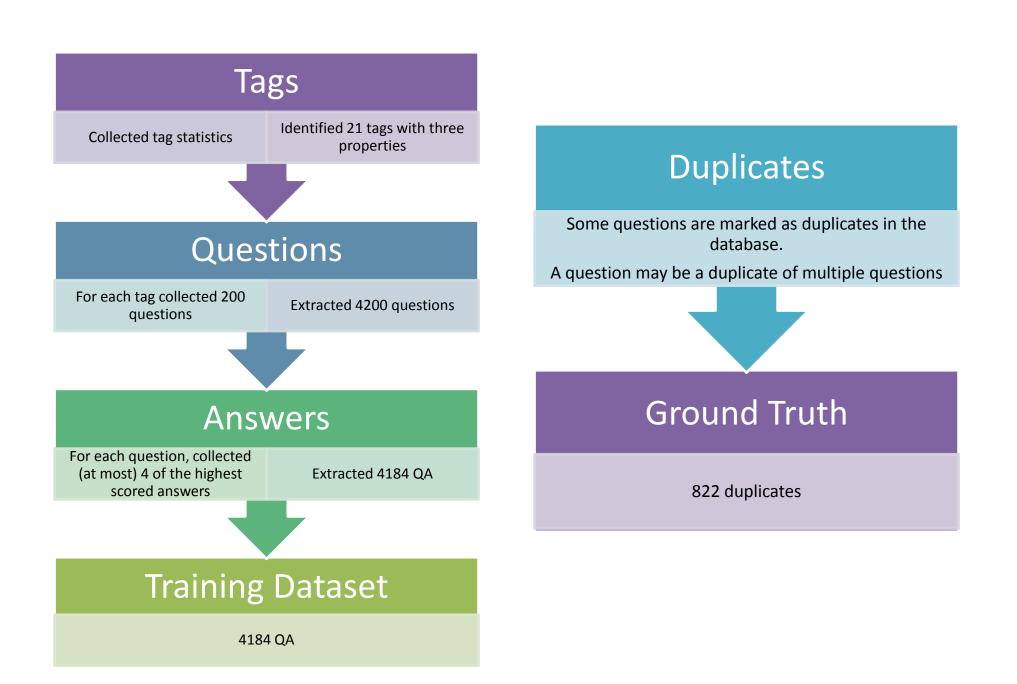
onnect end java jvm server open il close direct data o nload mail declare clien network fd local jython sun dyna 0.45 0.4 0.35 -Per-document topic 0.3 0.25 proportions 0.2 0.15

0.05 -

Dataset

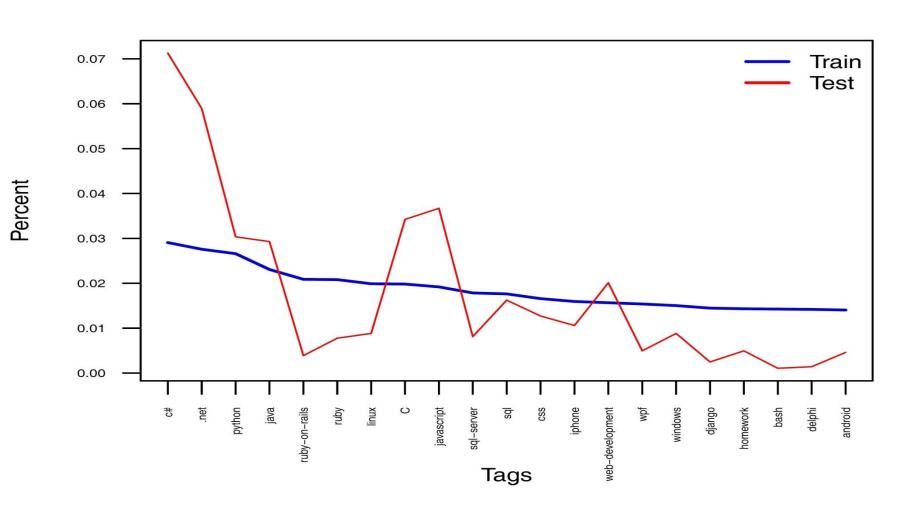
Stackoverflow Statistics

#Questions	1,188,585				
#Answers	2,939,767				
#Tags	27,659				
#Users	440,672				
Avg #answer per question	2.4818				
Avg #tags per question	2.9254				
Avg score of questions	1.4967				
Avg score of answers	1.8478				



0.11	_			
C#	.net	Sql		
Sql-server	Java-script	 css Ruby-on-rails Web-development Delphi C 		
Java	Ruby			
Wpf	Iphone			
Android	Windows			
Django	Python			
Bash	Linux	Homework		

Tag Distribution in Training Set and Test Set



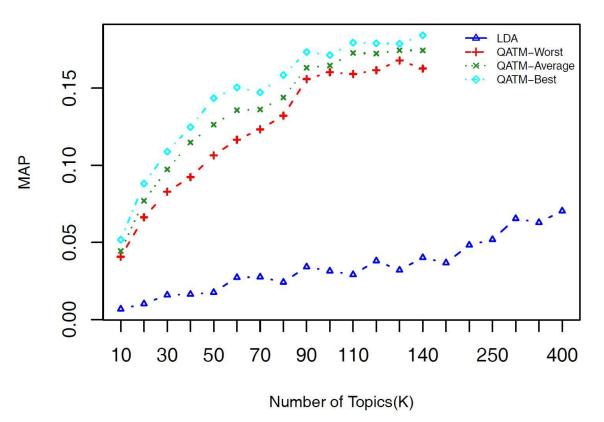




Experiments – Question Answering

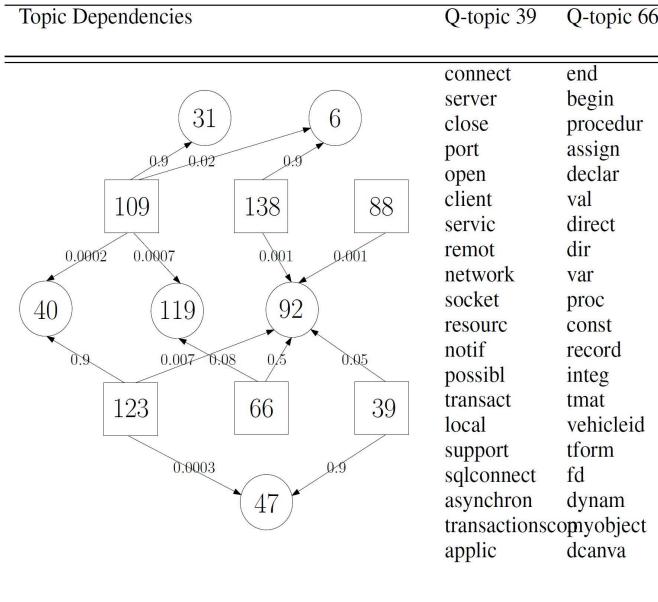
Stackoverflow

QATM retrieval performance compared to LDA model in terms of Mean Average Precision and TopN measures



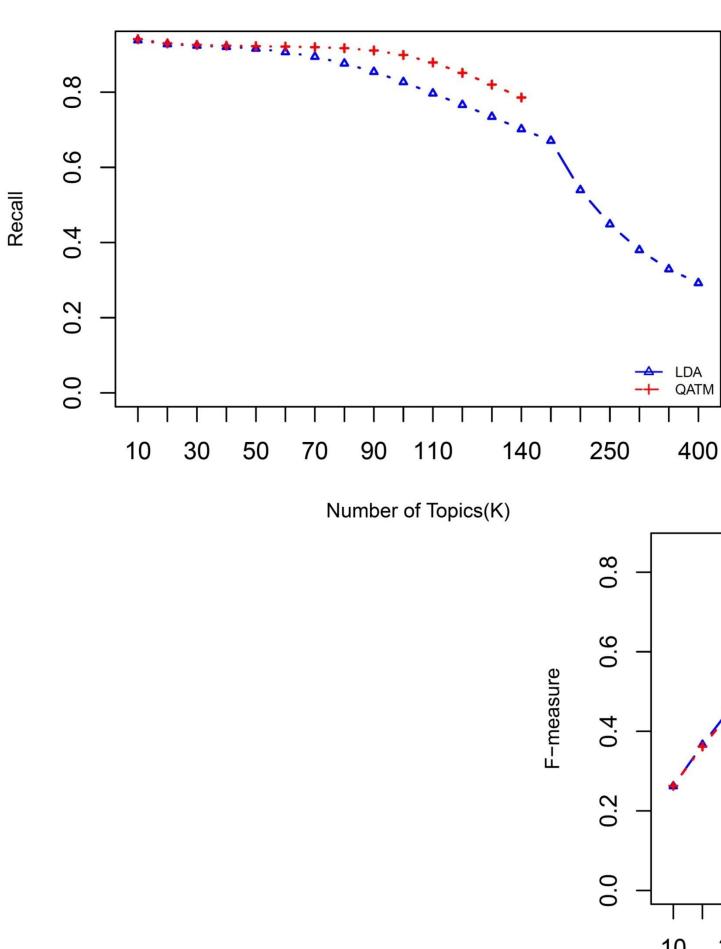
first 20 most probable words

Topic Dependencies





Clustering results in terms of Precision, Recall and F-measure.



Dataset Creation

Selected Tags



Results on the Stackoverflow dataset

	Top1	Top2	Тор3	Top4	Top5
LDA	0.023	0.026	0.029	0.03	0.032
QATM-Worst	0.108	0.127	0.138	0.144	0.148
QATM-Average	0.122	0.142	0.151	0.156	0.16
QATM-Best	0.131	0.152	0.161	0.164	0.168

Topical dependencies captured by QATM with examples of Q-topics and A-topics represented by their

6	Q-topic 88	Q-topic 109	Q-topic 123	Q-topic 138	A-topic 6	A-topic 31	A-topic 40	A-topic 47	A-topic 92	A-topic 119
	java	email	stream	http	http	part	stream	connect	thi	answer
	scala	piec	buffer	googl	WWW	email	buffer	socket	wai	question
	lang	send	mp	WWW	googl	send	data	port	ani	point
	jvm	contact	audio	blog	org	home	length	network	onli	case
	run	split	call	site	html	target	byte	send	doe	order
	util	mail	plai	org	blog	work	close	server	make	actual
	org	input	data	stackoverflow	articl	contact	mp	data	someth	follow
	class	final	download	en	en	messag	flush	client	check	give
	jdk	subject	decod	html	link	mail	reader	receiv	exampl	ye
	swing	sm	sound	sampl	aspx	sourc	upload	remot	doesn	expect
	jrubi	separ	util	wiki	wiki	sm	audio	summari	code	sinc
	vm	pop	flush	place	post	relpath	output	proxi	befor	assum
	librari	millisecond	upload	link	stackoverflov	v split	memorystrea	melose	wa	consid
	api	msxml	close	twitter	doc	address	payment	conn	note	happen
	apach	receiv	clap	msdn	archiv	smtp	dest	timeout	find	otherwis
	bytecod	attach	audiomanag	home	msdn	gmail	bankhead	param	possibl	howev
	jython	regist	programmat	forum	librari	receiv	input	packet	abov	side
	ee	programmat	listen	articl	github	subject	result	open	don	correct
2 2	awt	current	video	aspx	found	spam	assert	implement	problem	inform
	sun	previousposit	inputstream	archiv	amp	bodi	filepath	host	put	sai

Results on the Stackoverflow dataset

📥 LDA 10 30 50 250 400 140 90 110 70 Number of Topics(K) 1.+·+·+·+ 📥 LDA -+ QATM 10 30 50 140 250 110 70 90 400 Number of Topics(K)