Category-Level Transfer Learning from Knowledge Base to Microblog Stream for Accurate Event Detection

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@DASFAA 2017. Suzhou. China

Motivation

The conditional probability p(w|d) is the combination of **word-topic** distribution $p(w|t_i)$ and **topic-document distribution** $p(t_i|d)$.

$$p(w|d) = \sum_{i=1}^{K} p(w|t_i)p(t_i|d)$$
 (1)

Equation (1) equals to the following form:

$$p(w|d) = \phi(w)\theta^{T}(d) \tag{2}$$

, where $\phi(w)$ is the row vector **word-topic distribution** $[p(w|t_1), \cdots, p(w|t_K)]$, and $\theta(d)$ is the row vector **topic-document distribution** $[p(t_1|d), \cdots, p(t_K|d)]$.

Challenges (1/2)

What we have already known:

What we maybe know, but didn't pay much attention: In equation (2), the distributions $\phi(w)$ and $\theta(d)$ have the following explanation:

- **1** $\phi(w)$ functions as the look-up layer for words with the **sigmoid** activation functions.
- ② $\theta(d)$ functions as the look-up layer for documents with the **softmax** activation function.

Challenges (2/2)

Statistics of some labeled events on Edinburgh twitter corpus.

Table: My caption

Date	Event Size
05/08/2011	656
21/07/2011	595
25/07/2011	485
07/09/2011	286
23/07/2011	283
23/07/2011	260
24/08/2011	246
09/08/2011	174
12/09/2011	135
15/08/2011	127
04/08/2011	124
22/07/2011	114
06/09/2011	25
13/07/2011	21
12/07/2011	18
10/08/2011	16
21/08/2011	13
	05/08/2011 21/07/2011 25/07/2011 07/09/2011 23/07/2011 23/07/2011 24/08/2011 09/08/2011 12/09/2011 15/08/2011 04/08/2011 06/09/2011 13/07/2011 12/07/2011 10/08/2011

How about existing methods on dealing with challenges?

An Example

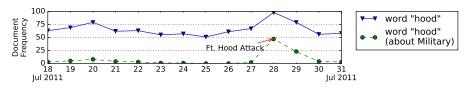


Figure: The comparison of the time series between the raw word *hood* and the *Military* related word *hood*, computed on the *Edinburgh twitter corpus*. The rise of document frequency on July 28th, 2011 is corresponding to the event mentioned in

https://en.wikipedia.org/wiki/Fort_Hood#2011_attack_plot.

The insights on the example

Overview of our method

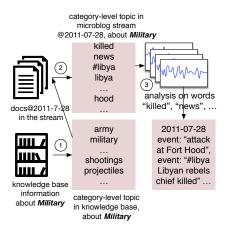


Figure: TransDetector's processing flow, taking *Military* related events in microblogs as an example.

TRANSDETECTOR: Phrase 1

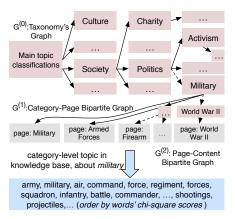


Figure: Extracting Category-Level Topics in Knowledge Base via its three fold hierarchical structure, taking *Military* as an example.

TRANSDETECTOR: Phrase 2

TRANSDETECTOR: Phrase 3

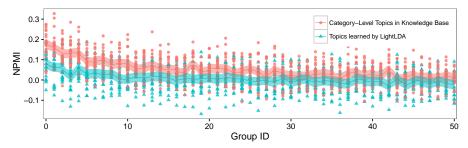


Figure: More topics are compared at the NPMI metrics between our method and LightLDA

some

Table: The comparison on the topic coherence(NPMI) between our method and LightLDA, taking *Aviation* as an example. (NPMI is computed on a group of ten words. \sim stands for the top five words.)

Category-Level Topics extracted from Wikipedia by TRANSDETECTOR					Topics Learned from Wikipedia by LightLDA						
GID	#words*	words	NPMI	GID #words* words			NPMI				
-	1-5	aircraft air airport flight airline	-	-	1-5	engine aircraft car air power	-				
0	1-5, 6-10	~, airlines aviation flying pilot squadron	0.113	0	1-5, 6-10	~, design flight model production speed	0.112				
1	1-5, 11-15	~, flights pilots raf airways fighter	0.155	1	1-5, 11-15	~, system vehicle cars engines mm	0.062				
2	1-5, 16-20	~, boeing runway force crashed flew	0.092	2	1-5, 16-20	~, fuel vehicles designed models type	0.072				
3	1-5, 21-25	~, airfield landing passengers plane aerial	0.179	3	1-5, 21-25	~, version front produced rear electric	0.035				
4	1-5, 26-30	~, bomber radar wing bombers crash	0.137	4	1-5, 26-30	~, space control motor standard development	0.085				
5	1-5, 31-35	~, airbus airports operations jet helicopter	0.189	5	1-5, 31-35	~, film range light using available					
6	1-5, 36-40	~, squadrons base flown havilland crew	0.088	6	1-5, 36-40	~, wing powered wheel weight launch	0.087				
7	1-5, 41-45	~, combat luftwaffe aerodrome carrier fokker	0.159	7	1-5, 41-45	~, developed low test ford cylinder	0.007				
8	1-5, 46-50	~, planes fly engine takeoff fleet	0.186	8	1-5, 46-50	~, equipment side pilot hp aviation	0.091				
9	1-5, 51-55	~, fuselage helicopters aviator naval aero	0.157	9	1-5, 51-55	~, systems us sold body drive	-0.051				
10	1-5, 56-60	~, glider command training balloon faa	0.166	10	1-5, 56-60	~, gear introduced class safety seat	0.069				
		***				***					
18	1-5, 96-100	~, scheduled carriers military curtiss biplane	0.131	18	1-5, 96-100	~, transmission special replaced limited different	0.059				
19	1-5, 101-105	~, accident engines iaf albatross rcaf	0.068	19	1-5, 101-105	~, features machine nuclear even unit	0.011				

Experiment Settings

Table: Category-Level Topics extracted from knowledge base and the corresponding topics on microblog stream learned from CTrans-LDA. The words in **bold** font are newly learned on the microblog stream by the transfer learning.

Aviation		Health		Middle East		Military		Mobile Phones	
Knowledge	Microblog	Knowledge	Microblog	Knowledge	Microblog	Knowledge	Microblog	Knowledge	Microblog
Base	Stream	Base	Stream	Base	Stream	Base	Stream	Base	Stream
aircraft	air	health	weight	al	#syria	army	killed	android	iphone
air	plane	patients	loss	israel	#bahrain	military	news	mobile	apple
airport	flight	medical	diet	iran	people	air	#libya	nokia	android
flight	time	disease	health	arab	israel	command	libya	ios	app
airline	airlines	treatment	cancer	israeli	police	force	rebels	phone	ipad
airlines	news	hospital	lose	egypt	#libya	regiment	people	samsung	samsung
aviation	boat	patient	fat	egyptian	#egypt	forces	police	game	mobile
flying	airport	clinical	tips	ibn	news	squadron	war	арр	blackberry
pilot	force	symptoms	treatment	jerusalem	#israel	infantry	libyan	iphone	tablet
squadron	fly	cancer	body	syria	world	battle	attack	htc	apps

Table: Overall Performance on Event Detection

Table. Overall I choffinance on Event Detection									
Method	Number of Events to be Evaluated	Recall@ Benchmark1	Precision@ Benchmark2	Recall@ Benchmark2	F@ Benchmark2	DERate ^a (Duplicate Event Rate)@ Benchmark2			
LSH	500	0.704	0.788	0.651	0.713	0.348			
TimeUserLDA	100	0.370	0.790	0.177	0.289	0.114			
Twevent	375	0.741	0.808	0.658	0.725	0.142			
EDCoW	349	0.556	0.748	0.511	0.607	0.226			
BurstyBTM	200	0.667	0.825	0.384	0.497	0.079			
TransDetector	457	0.889	0.912	0.876	0.894	0.170			

^a DERate = (the number of duplicate events) / (the total number of detected realistic events)

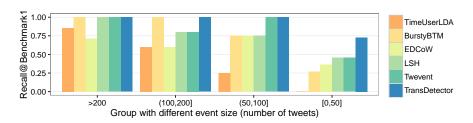


Figure: The relation between the recall and the event size

Table: Events about *military* detected by systems between 2011-07-22 and 2011-07-28

Date	Event key words	Representative event tweet	Number of	Methods ^a					
		Representative event tweet	event tweet	L	TU	TW	E	В	TD
7/22/11	Norway, Oslo, attacks, bombing	Terror Attacks Devastate Norway: A bomb ripped through government offices in Oslo and a gunman http://dlvr.it/cLbk8	557	✓	√	√	√	✓	√
7/23/11	Gunman, rink	Gunman Kills Self, 5 Others at Texas Roller Rink http://dlvr.it/cLcTH	43	-	-	✓	✓	-	✓
7/26/11	Kandahar, mayor, suicide, attack	TELEGRAPH]: Kandahar mayor killed by Afghan suicide bomber: The mayor of Kandahar, the biggest city in south _	47	√	-	√	✓	-	√
7/28/11	Ft., Hood, attack	Possible Ft. Hood Attack Thwarted http://t.co/BSJ33hk	52	-	-	-	-	-	✓
7/28/11	Libyan, rebel, gunned	Libyan rebel chief gunned down in Benghazi http://sns.mx/prfvy1	44	-	-	-	-	-	√

^a L=LSH, TU=TimeUserLDA, TW=Twevent, E=EDCoW, B=BurstyBTM, TD=TRANSDETECTOR.

Thanks!

Q&A

⁰This slide and more data are available at http://q-r.to/bajx8I ODASFAA 2017, Suzhou, China 19/19/19 Weijing Huang, Tengjiao Wang, Wei Chen, YCategory-Level Transfer Learning from Knowl