

Joshua Vendrow, Jamie Haddock, Elizaveta Rebrova, Deanna Needell

Department of Mathematics, UCLA

Background

Nonnegative Matrix Factorization (NMF) is an approach typically applied in unsupervised tasks such as dimensionality-reduction, latent topic modeling, and clustering. ► Given nonnegative data matrix $X \in_{>0}^{m \times n}$ and a user-defined target dimension $k \in \mathbb{N}$, NMF seeks nonnegative factor matrices $A \in_{>0}^{m \times k}$, and $S \in_{>0}^{k \times n}$ such that $X \approx AS$, formulated as $\arg \min ||X - AS||_F^2$. A≥o,S≥o **Semi-supermised Nonnegative Matrix Factorization (SSNMF)** jointly factorizes a data matrix $X \in_{>0}^{m \times n}$ and a supervision information matrix $Y \in_{>0}^{c \times n}$, formulated as arg min $||X - AS||_F^2 + \lambda ||Y - BS||_F^2$. Reconstruction Error Method \triangleright We denote a seed topic as a vector $v = (v_1, \ldots, v_m)$ (where m denotes the vocabulary size), where $v_i = 0$ if the *i*th word in the vocabulary is not in the seed topic and some positive weight otherwise. ▶ Let the data matrix $X \in M^{\times n}$ have examples along the columns and features along the rows and suppose we have seed topics $v^{(1)}$, $v^{(2)}$, $\ldots, v^{(c)} \in R^m.$ Let the seed matrix be $Y = [v^{(1)}, v^{(2)}, \dots]$ Guided NMF is formulated as $\min_{A \ge 0, S \ge 0, B \ge 0} \|X - AS\|_F^2$ Figure 1 shows a visualization of Guided NMF with seed word space. Bag of words representation n ~ X m m 'Space" 🔶 Ĩ \boldsymbol{m} m

Figure 1: Visualization of Guided NMF

Conclusions

We propose an NMF-based model, Guided NMF, which incorporates seed guide learned topics towards meaningful and coherent sets of features. illustrate the promise of this model in text-based topic modeling application

On a Guided Nonnegative Matrix Factorization

(1)

Classification Error

$$\left[\begin{array}{c} \mathbf{v}^{(c)} \right] \in_{\geq 0}^{m \times c} \tag{3}$$

$$\left[\begin{array}{c} \mathbf{e} \\ \mathbf{e} \end{array} + \lambda \| \mathbf{Y} - \mathbf{AB} \|_{F}^{2} \end{array} \right] \tag{4}$$



topic supervision to
Our initial experiments
ations.

Contact Information

jvendrow@math.ucla.edu Ś www.joshvendrow.com

O github.com/jvendrow

Results

(2)

The 20 Newsgroups dataset is a collection of approximately 20,000 text documents containing the text of messages from 20 different newsgroups on the distributed discussion system Usenet





> We use Guided NMF to encourage topics to form around specific newsgroups (e.g. space, medicine, religion).

Rank 4 Guided NMF on the 20 Newsgroups dataset with Rank 4 Guided NMF on the 20 Newsgroups dataset with seed word *pitch*, *medical*, and *space*. seed words *motorcycle*, *sale*, and *religion*.

pic 3 Topic 4
ligion people
ristian know
daism think
edom time
istians use
slam new
pulsion space
avi see
life say
npoint us

The Twitter political data set is a data set of tweets sent by political candidates during the 2016 election season. > We use Guided NMF to encourage topics to form around *issues* rather than *candidates*.

		er politicat ua	la sel.	Kalik o Gui						
Topic 1	Topic 2	Topic 3	Topic 4	words eco	words economy and obamacare.					
thank	govpencein	gopdebate	tedcruz	Topic 1	Topic 2	Topic 3	Topic 4			
trump2016	indiana	imwithhuck	cruz	economy	obamacare	govpencein	gopdebate			
maga ¹	indiana_edc	jeb	cruzcrew	jobs	fullrepeal	indiana	kasich			
great	state	tonight	ted	tax	repeal	indiana_edc	randpaul			
america	jobs	president	choosecruz	plan	replace	state	john			
Topic 5	Topic 6	Topic 7	Topic 8	create	fight	jobs	tonight			
kasich	hillary	randpaul	fitn	Topic 5	Topic 6	Topic 7	Topic 8			
john	trump	iowa	new	tedcruz	hillary	johnkasich	people			
johnkasich	people	iacaucus	hampshire	thank	trump	new	need			
ohio	donald	caucus	johnkasich	cruz	donald	fitn	must			
gov	president	tonight	nh	cruzcrew	clinton	kasich	berniesanders			
¹ Here "mag	a" abbreviate	s "makeamer	icagreatagain."	ted	president	hampshire	country			

Rank 8 NMF on the Twitter political data set

Ablation and Comparison

> We explore the impact of adding additional seedwords and varying the rank of the factorization, and compare to SeededLDA Guided NMF consistently has an AUC above 0.8 for all rank and number of seedword choices. For few seed words and/or only a small rank, Guided NMF significantly out-performs Seeded LDA.

AUC s docun	cores for 20 nents from sp	New ace c	sgrou lass.	ps da	ataset	AUC : docui	scores for 20 ments from ba	New Aseba	sgrou Il clas	ips d ss.	atase
Rank	Method	# Seed words			Dank	Mathad	# Seed words				
		1	2	4	8	Kallk	Method	1	2	4	8
4	Guided NMF	0.83	0.88	0.88	0.87	,	Guided NMF	0.89	0.9	0.9	0.9
	Seeded LDA	0.31	0.42	0.74	0.86	4	Seeded LDA	0.31	0.42	0.74	0.86
6	Guided NMF	0.86	0.87	0.88	0.87	6	Guided NMF	0.9	0.9	0.9	0.9
	Seeded LDA	0.37	0.5	0.91	0.89	0	Seeded LDA	0.37	0.5	0.91	0.89
10	Guided NMF	0.88	0.89	0.89	0.89	10	Guided NMF	0.87	0.9	0.9	0.9
	Seeded LDA	0.45	0.95	0.95	0.95	10	Seeded LDA	0.45	0.95	0.95	0.95

Rank 8 Guided NMF on Twitter political data set with seed

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