Text mining assignment - Topic modeling

Guillem Bagaria

Objective To analyze the results output and performance from two different Topic Modeling algorithms, one algebraic (NMF) and one probabilistic (LDA).

Non-negative Matrix Factorization is a linear algebraic model for bag-of-words topic modeling, which require large and sparse matrices and are a hard problem to tackle both computationally and statistically that may cause unstable feature selection.

The goal is to find two more tractable, non-negative matrices, obtained through numeric approximation, such that their product results in the original non-negative matrix.

$$W_{m \times k} \times H_{k \times n} \approx A_{m \times n}$$

Both W and H are initialized with random, non-negative values and are iteratively updated up until convergence within a fixed threshold.

$$H_{ij} \leftarrow Hij \frac{(W^T A)_{ij}}{(W^T W H)_{ij} + \epsilon}$$

$$W_{ij} \leftarrow Wij \frac{(AH^T)_{ij}}{(WHH^T)_{ij} + \epsilon}$$

Latent Dirichlet Allocation is a probabilistic model that finds the most likely membership of a document from k number of topics. The learning process uses variational Bayes in order to update the distribution of the likelihood for each word and thus, document.

One of the expected results is that LDA would perform better for big data. Time complexity is polynomial in NMF while In LDA, the time complexity is lower, proportional to $\mathcal{O}(n*iterations)$ where n is the number of documents.

The data This dataset is standard in the SciKitLearn python package and contains around 1800 newsgroup posts from 20 threads, which correspond to the 20 different topic labels. For this application, only the first level group will be used, in order to simplify the analysis.

'alt.atheism', 'comp.graphics', 'comp.os.ms-windows.misc',

'comp.sys.ibm.pc.hardware', 'comp.sys.mac.hardware', 'comp.windows.x',

Default topics 'misc.forsale', 'rec.autos', 'rec.motorcycles', 'rec.sport.baseball', 'rec.sport.hockey',

'sci.crypt', 'sci.electronics', 'sci.med', 'sci.space', 'soc.religion.christian',

'talk.politics.guns', 'talk.politics.mideast','talk.politics.misc', 'talk.religion.misc'

Atheism, Christianity, Computer Hardware, Windows OS, For Sale, Motor Sports,

Target topics Team Sports, Cryptography, Electronics, Medicine, Space, Gun Politics, Middle East

Politics, Other Politics.

The raw text from the posts is then processed and appropriately tokenized. As this is an unsupervised learning approach, all the data will be used for model training.

Results Findings of the comparison between different algorithms and settings. The selected number of topics was reduced from the original specific 20 to 14 more general categories for each of the topics in order to produce slightly more manageable results while remaining sensible.

LDA tf NMF tf-idf

players win league Topic 8: law people government gun state right rights control guns crime Topic 9: windows use program window software file dos using output image Topic 10: god people jesus believe does say think christian bible don Topic 11: ax max end air information drivers video conference new use Topic 12:	play hockey win league Topic 4: drive scsi drives hard disk ide floppy controller cd tape Topic 5: tiam cars engine miles price new speed condition good bike Topic 6: windows file files dos use window program using problem running Topic 7: key chip encryption clipper keys escrow government use public algorithm Topic 8: people government israel armenian israeli jews armenians rights state law Topic 9: edu soon cs com email university ftp send internet david Topic 10: card video monitor cards drivers bus vga sale color driver Topic 11: just wondering thought don wanted mean tell fine oh maybe
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	-
video conference new use	just wondering thought don wanted mean
Topic 12:	tell fine oh maybe
thanks know does use problem help work	Topic 12:
like need ve	does know anybody don let mean doesn
Topic 13:	help program appreciated
mr president university research health	1 1 0 11
april national center program stephanopoliks sounds looks look sound lot things	
I to the second second second	new doing sell

1 7	
Topic 0: ax max end air follow firearms fit floppy folks young Topic 1: edu com available graphics ftp pub image mail data send Topic 2: db cs al bits higher gas lower bit west east Topic 3: people said know don didn just like went say think Topic 4: file gun control firearms states united mr house law crime Topic 5: mr stephanopoulos president know going don think said did package Topic 6: jpeg image gif file color images format quality version files Topic 7: output file entry program stream build rules section info line Topic 8: hockey league new nhl team season edu games vs division Topic 9: internet privacy anonymous information email use mail computer pub electronic Topic 10: use widget window subject application available xt motif set used Topic 11: disk drive drives hard bios rom control: card floppy supports Topic 12: space launch satellite new data nasa	Topic 5: car cars engine miles price new speed condition good bike Topic 6: windows file files dos use window program using problem running Topic 7: key chip encryption clipper keys escrow government use public algorithm Topic 8: people government israel armenian israeli jews armenians rights state law Topic 9: edu soon cs com email university ftp send internet david Topic 10: card video monitor cards drivers bus vga sale color driver Topic 11: lignst wondering thought don wanted mean tell fine oh maybe Topic 12: does know anybody don let mean doesn
-	-
program commercial south year Topic 13:	help program appreciated Topic 13:
-	-
god jesus does people atheists believe atheism bible religious good	like sounds looks look sound lot things new doing sell
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NMF tf In this case the matrix is fed with sim- NMF tf-idf In this case the matrix uses term

frequency inverse document frequency:

ple term frequency:

Conclusions

Initial observations show that NMF presents coherent topic words, especially when combined with the tf-idf input matrix. Simple tf performs makes some serious mistakes, such as missing the "Middle East Politics" and misplacing many words; Topic 0 (firearms, floppy [disk]), Topic 2 (bits, gas), etc.

Although the results for NMF (tf-idf) and LDA (tf) are similar, the execution time of NMF however is much lower than LDA, and in certain situations might prove to be a determining factor in its use.

	Execution time
NMF	1.79s
LDA	18.01s

LDA produces very coherent topics even with simple term frequency data input.

Convergence time and topic modeling performance might improve in LDA when used in a semi-supervised learning scenario, with strong priors. This would be the next step.