Topic Modeling

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Abstract

We will analyze the State of the Union speeches corpus, and try to visualize how main topics of the speeches have shifted over time in relation to historical events using Topic Modeling.

1 Introduction

We start-off by first preprocessing the data and understanding it and then move on to topic modelling. Then we will visualize the main topics of the speeches and give them a title wherever possible. We will then use a method to study how topics changed each decade during the 20th and 21st century.

1.1 Pre-processing and generating TF-IDF scores

Firstly we check out any empty rows in the file and observe that some years have empty rows for eg the years 1829,1830,etc. We remove such empty rows and also some years have more than one state of the union addresses, for such years we combine all the addresses for that particular year.

Next we need to process the text of the speech delivered. We remove any special characters present in the text and punctuatuions as those are not of use to us and then tokenise the words. This is done mainly using preprocessing module of nltk. Now each speech is a series of words.

We use the stopwords from both scikit learn and nltk libraries as they both have a different set of stopwords and thus combining both we can remove almost all the non-useful words. Now we have a series of words for each year and all those words add some meaning to the speeches.

We generate the TF-IDF(Term Frequency- Inverse Data Frequency) scores for these speeches using gensim. Formula for non-normalized weight of term i in document j in a corpus of D documents is given by

 $weight_{i,j} = wlocal(frequency_{i,j}) * wglobal(document_frequency_{i,}D)$

1.2 LSI Topic Modeling

We now have TF-ID scores and now will do topic modeling using LSI. LSI or Latent Semantic Indexing is a method where SVD is first performed on the term document matrix as the spare matrix is very large and it learns latent topics through this process. LSI is faster compared to LDA.

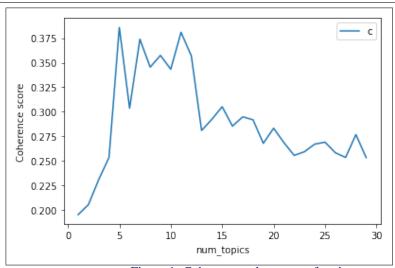


Figure 1: Coherence value vs no of topics

Deciding the number of topics:

This is a very crucial step while performing LSI topic modeling. We use the cohernece metric to determine what number of topics lead to the best human interpretability. Coherence metric measures how semantically similar are the high scoring topics and therefore helps us discard topic which received higher score just because of more occurences. We measure coherence and visualize it using a line plot(Figure 1) with number of topics ranging from 1 to 30. Maximum coherence is found for number of topic equal to 5 and then almost similar for number of topics equal to 12. We choose 12 as assignment required us to study atleast 10 topics.

Annotations:

Now we sort the keywords in each topic by their scores and plot top 12 words and their score using a bar histogram plot. Topic name has been provided manually(figure 2). Topic name is clearly visible in some of the topics but for some it was harder to decide and the decision for name wasn't always perfect. It is just indicative. We can observe that tonight and program are the two words occuring everywhere and contributing heavily. It is because they weren't in the stopwords list and are present across every speech and also multiple times in a single speech. Therefore there TF-IDF score was high leading to this event. If we talk about similarity among the topics then we can summarise all these in two words War and Economy. These are the two things spoken about in almost every State of the Union address.

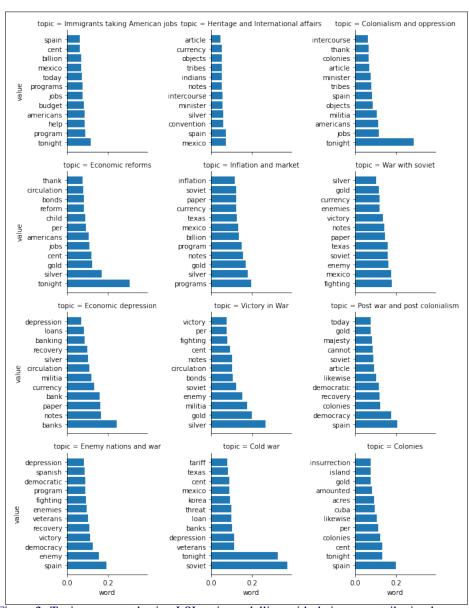


Figure 2: Topics generated using LSI topic modelling with their top contributing keywords. Annotations are done manually and written at the top of each plot.

4 :

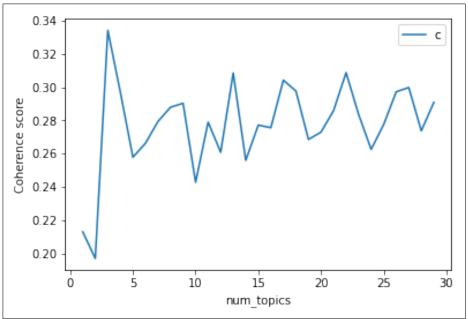


Figure 3: Coherence value vs no of topics for LDA topic modeling

1.3 LDA Topic Modeling:

LDA is an unsupervised topic modelling method and assumes Dirichlet prior over latent topics. LDA is said to have more accuracy than LSI in general but is also slower as compared to LSI. Now we will see if it holds true.

Determining number of topics: As we did in LSI now we again use the coherence metric to pick the no of topics which offers maximum human interpretability. Figure 3 shows the coherence vs no of topics plot. At a first glance we can see the maximum coherence is for no of topic equal to 4 and is less than the max coherence value of LSI topic modeling. However we continue with no of topics greater than 10 and max coherence and thus we get 13 as the optimum number.

Annotations:

Annotating the topics generated by LDA topic modeling was easier as compared to LSI except a few outliers. We employed a similar strategy for plotting as in LSI. The topics seemed more relevant and top keywords conveyed more meaning as compared to LSI although the coherence was lower compared to LSI. LSI is a linear algebraic method as compared to LDA which is a probabilistic model. For looking at topics within a speech and across speeches, according to the results, LDA performs better looking at inter speech topics.

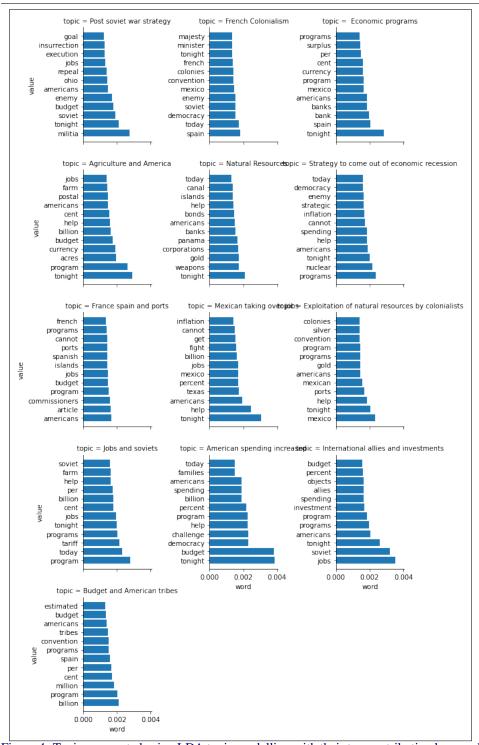


Figure 4: Topics generated using LDA topic modelling with their top contributing keywords. Annotations are done manually and written at the top of each plot.

1.4 How speech topics differ over time :

Here we use LDA and generate top 3 topics during each decade and see how they change over time due to various events throughout history. For this purpose topic names are a combination of top 10 contributing keywords for that topic. For generating decade wise data we first combined Speeches over 10 year periods and then used LDA over them. Results are shown in Figure 5. There are two ways to do this, first to directly combine the speeches in one text and use Ida or keep them distinct and use LDA so that cross speech frequency can help get better results.

First we make broad inferences and observe that before the 1950s the speeches were comprising more legislative and fiscal terms and it changed into more simpler terms that would make sense to the general public immediately. It is majorly because 1950-60s was the time of boom of Television in America and therefore the population watching the SOTU had changed. Later we see a shift to more war related terms majorly because of a rising spirit of nationalism and violence as the modern era begins. Modern era speeches also comprised more of healthcare and college policies as they made direct impact on the public which has now widened due to various technical platforms.

Another way to infer is how in decades which were near economic recession such as 1920's, 1940-50's, 1970s and the late 2000s we see more economic reforms and jobs related terms/topics as these were the topics of interest over the short run. Then we can observe how during times of war such as 1910-1920 and 1940-1950 and the cold war in 70s, the topics revolved around military, appraisal of the military, economy(post war years), allies, enemy nations, etc.

We can also observe how topics like Islam, Afghanistan, Jihad, etc are used more frequently

We can also observe how topics like Islam, Afghanistan, Jihad, etc are used more frequently during decades that USA was attacked by Terrorists.

Topics found via LDA in between years: 1790 to 1800 united country peace treaty time general provision session senate indians states government citizens measures war gentlemen house necessary great union public state shall representatives national congress present law commerce laws Topics found via LDA in between years: 1800 to 1810 country citizens united government war present laws course subject law public peace time necessary year commerce debt new session sea Topic : 3 shall states congress millions state great nations vessels force nation Topics found via LDA in between years: 1810 to 1820 war great public congress force national present enemy treaty peace united british spain millions citizens year american general view treasury states government state country commerce subject time consideration effect prope Topics found via LDA in between years: 1820 to 1830 government public year state session treaty people revenue general report united great present duties commerce citizens years nation time condition states congress act th war country union power powers nations Topics found via LDA in between years: 1830 to 1840 states united congress people citizens year bank new interests banks state subject general time treaty shall attention session important means government public country present great power treasury necessary duty duties Topics found via LDA in between years: 1840 to 1850

Topics found via LDA in between years: 1840 to 1850 united mexico congress country people time state treasury citizens treaty states government public war texas peace th territory constitution necessary r act shall year present policy revenue mexican laws Topics found via LDA in between years: 1850 to 1860 states government congress country shall power union treasury necessary mexico great state constitution citizens subject people act law territory th united year public present time general treaty war new condition Topics found via LDA in between years: 1860 to 1870 ates congress constitution national citizens subject laws secretary peace power country great union public shall time general law new foreign government united people war year state power present service department Topics found via LDA in between years: 1870 to 1880 states government congress vear country citizens time report present general service subject claims secretary years consideration fiscal right relations power ropic. Summited great public people law department act legislation commission foreign Topics found via LDA in between years: 1880 to 1890 congress people public present general service american subject laws attention states country citizens treaty time silver consideration necessary state rights government united year law foreign department report secretary countries new Topics found via LDA in between years: 1890 to 1900

(b)

(a)

Figure 5: Speech topics over time

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Topics found via LDA in between years: 1890 to 1906
states great new public number cent spain trade notes free
government congress time american country law secretary act service war
united year general people gold work increase treasury legislation foreign
Topics found via LDA in between years: 1900 to 1910
Topics found via LDA in between years:
great public country work men nation good man present make
        ment states law people national business service power state year
congress united american time war far shall conditions general laws
Topics found via LDA in between years: 1910 to 1920
country time war men new purpose action public interests nations
states united great foreign year department present people shall necessary
Topic : 3
government congress american law make world state peace general international Topics found via LDA in between years: 1920 to 1930
government congress public national people war states great time american
Topic : 2 country law service year world years new work order department
Tederal present necessary power legislation agriculture ought make large state Topics found via LDA in between years: 1930 to 1940
congress national nation new work business country peace great banks
       ment world year time economic nations shall united relief agriculture
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people federal public states power action employment large income years Topics found via LDA in between years: 1940 to 1950

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Topics found via LDA in between years: 1940 to 1950
dollars national program nations peace great economic federal american security
Topic : 2
war states new legislation billion public men labor business work
year world government united congress people million production nation fiscal
Topics found via LDA in between years:
world peace shall years states progress make need legislation programs
economic year military new federal program defense strength united great
free government people nations congress security nation freedom power effort
Topics found via LDA in between years: 1960 to 1970
years nations billion free america program federal make states government
Topic :
new world year people help national million war freedom americans
congress nation american peace time united great shall tax tonight Topics found via LDA in between years: 1970 to 1980
\label{topic:1} \mbox{Topic}: \mbox{\bf 1} \\ \mbox{world america years nation great states americans energy time state}
new people congress year make programs inflation economy national major
   vernment american federal peace president union today united economic program
Topics found via LDA in between years: 1980 to 1990
new congress years world administration economic national programs peace freedom
people federal american states program soviet policy budget future make
america government year nation time work security help let energy
Topics found via LDA in between years: 1990 to 2000
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(d)

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congress nation american peace time united great shall tax tonight
Topics found via LDA in between years: 1970 to 1980
world america years nation great states americans energy time state
new people congress year make programs inflation economy national major
Topic: 3
government american federal peace president union today united economic program
Topics found via LDA in between years: 1980 to 1990
new congress years world administration economic national programs peace freedom
Topic: 2
people federal american states program soviet policy budget future make
america government year nation time work security help let energy
Topics found via LDA in between years: 1990 to 2000
Topic : 1
people year american years know americans let care say ask
Topic: 2
america new work world make government nation budget need welfare
children congress time country help tonight health jobs security way
Topics found via LDA in between years: 2000 to 2010
Topic: 1
new world country help years security health iraq freedom care
Topic: 2
people congress make children tax government reform women peace future
america american americans nation year work tonight economy terrorists great
Topics found via LDA in between years: 2010 to 2020
Topic: 1
new america american year americans energy know world business help
people work make let tax right country like tonight nation
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(e)

jobs years time need government businesses economy come companies education

Figure 7: Speech topics over time

1.5 Topic Modeling on AP wire stories dataset

The data in this dataset is much more diverse as the stories are not related to each other directly and are very diverse. There are a lot of names of people places, objects, etc. As compared to SOTU dataset which mainly comprised of war, economy, culture, etc, we here observe a diverse array of topics and the coherence increases as we increase number of topics upto some extent. LDA has been used for topic modelling for this AP wire stories dataset as it offers more accuracy. Results are shown in figure 8. The same process was followed and while annotating it was observed that naming the topics felt easier in the sense that this time we had a lot of options to choose from whereas in the SOTU case there were limited choices due to the content being almost the same. The topics could be for example topic 7 in figure 8 is religions, topic 5 could be christianity and disasters, topic 3 could be related to economy.

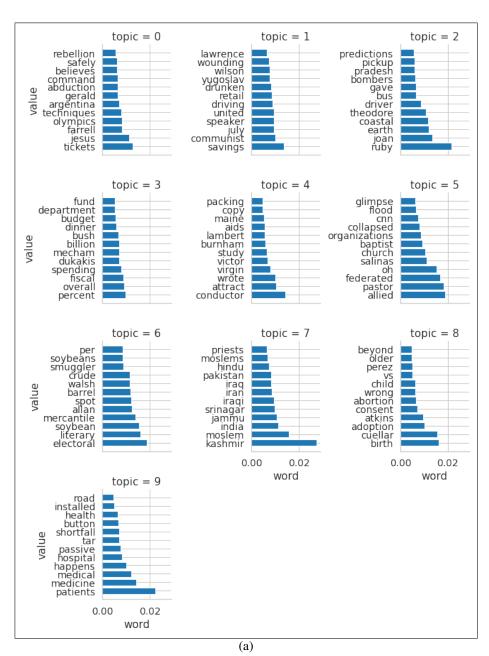


Figure 8: Top 12 contributing words for 10 selected topics after LDA topic modelling on AP Wire stories dataset

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1.6 Conclusion

To summarise Topic modelling worked better on AP wire stories due to diversity in topics. If we limit the number of topics in SOTU data then there will be no problem seen in annotating as the coherence is also maximum and that is actually the no of topics seen if topics are considered to be broad. Also in comparison to LSI, LDA gave better results although it took more time.

1.7 References

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