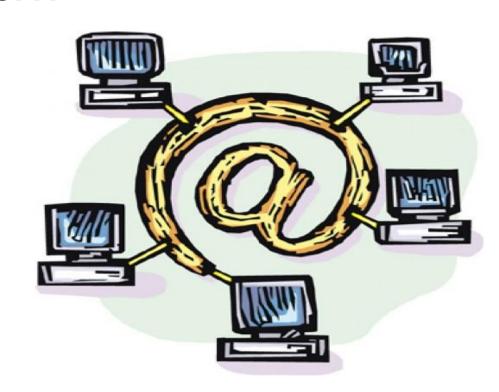


Can we extract the topics from a mail box?

- **3000** e-mails from the 20Newsgroups dataset
- 3 newsgroups :
 - o comp.graphics
 - rec.cars
 - o talk.misc.religion
- 2 modeling attempts:
 - Dimensionality reduction through LSA + Clustering through K-Means
 - Latent Dirichlet Analysis
- 1.5 Goal(s):
 - Automatic categorization of each e-mail to one of the 3 newsgroups main topic
 - Find the presence of other topics



Example mail - Clean Up

I agree, which is why I've asked for help with it.

The reason I'm working on this list is because I've recently had one too many Christians tell me "the Bible contains no contradictions whatsoever." They believe that it's true, and that it describes reality perfectly, and even predicts history before it happens.

Before I can carry on any sort of meaningful conversation with these people, I've got to SHOW them, with concrete evidence, that the Bible is not nearly as airtight as they thought. I hope to do that with this list.

Specifically: when I bring up the fact that Genesis contains two contradictory creation stories, I usually get blank stares or flat denials. I've never had a fundamentalist acknowledge that there are indeed two different accounts of creation.



Footer

Example mail - (Possibly) Relevant Tokens for Religion Topic

I agree, which is why I've asked for help with it.

The reason I'm working on this list is because I've recently had one too many Christians tell me "the Bible contains no contradictions whatsoever." They believe that it's true, and that it describes reality perfectly, and even predicts history before it happens.

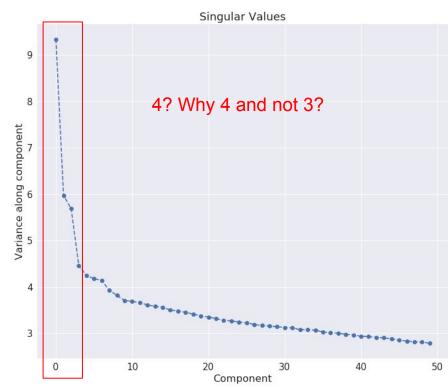
Before I can carry on any sort of meaningful conversation with these people, I've got to SHOW them, with concrete evidence, that the Bible is not nearly as airtight as they thought. I hope to do that with this list.

Specifically: when I bring up the fact that Genesis contains two contradictory creation stories, I usually get blank stares or flat denials. I've never had a fundamentalist acknowledge that there are indeed two different accounts of creation.

LSA

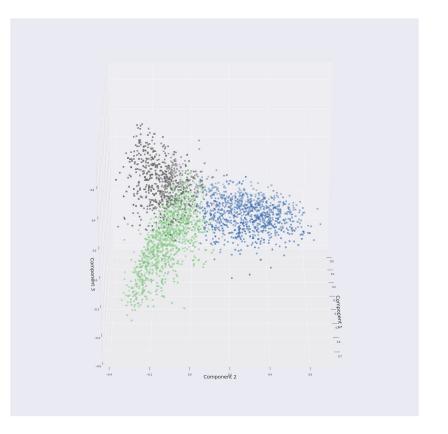
The majority of variance is explained by the first 4 components !!!

- 3000 Mail
- Each mail becomes a point in a 1000-dimensional space (Tfldf matrix)
- Reduction to 50-dimensional space(principal components) using truncated SVD
- Variance along component
- Singular values distribution imply high separation potential



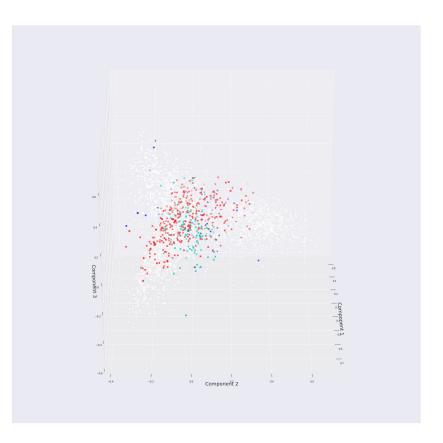
K-means Clusters

- Look for 3 clusters over 50 dimensional space (projected on the first 3 components)
- Most frequent terms per cluster:
 - Cluster 0: people god think like say know life make good jesus
 - Cluster 1: thanks graphics image know program files looking like file use
 - Cluster 2: car cars engine new like good speed ford dealer problem

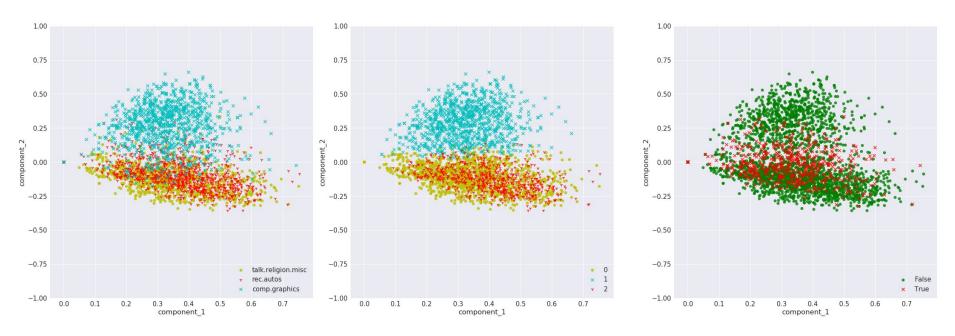


K-means Clusters -results

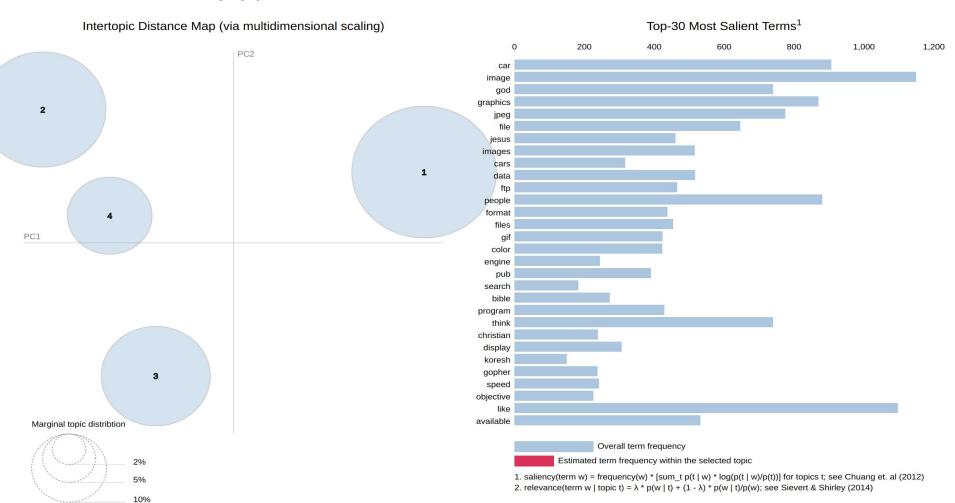
- 513 Misclassification over 3000 e-mails:
 - Messages from rec.autos not in Cluster 2: 35.3 % (red)
 - Messages from comp.graphics not in Cluster 1: 12.6% (cyan)
 - Messages from talk.religion.misc
 not in Cluster 0: 3.4% (blue)
 - Message in white correctly classified



Clusters vs Category on first 2 principal components



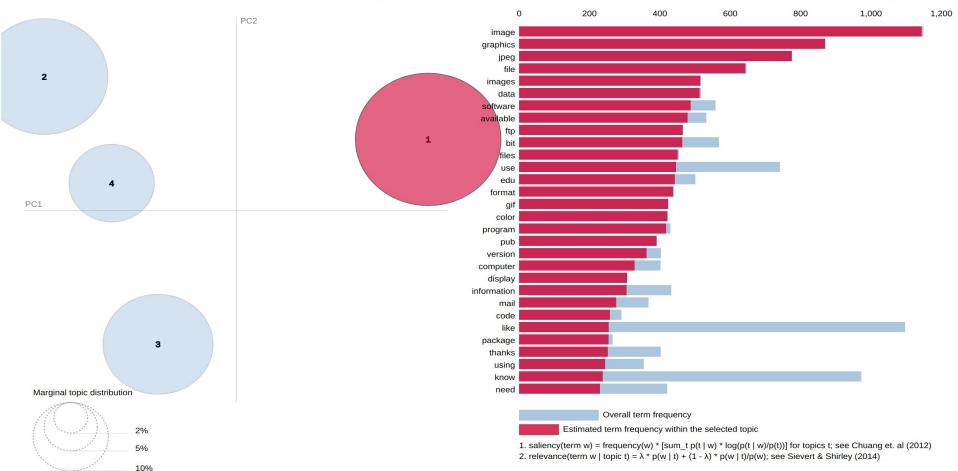
LDA - Overall



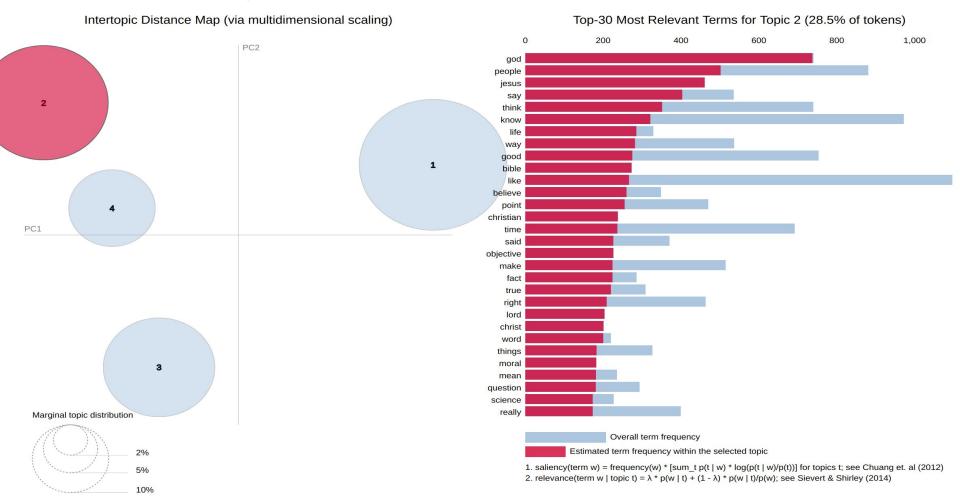
LDA - Graphics

Intertopic Distance Map (via multidimensional scaling)

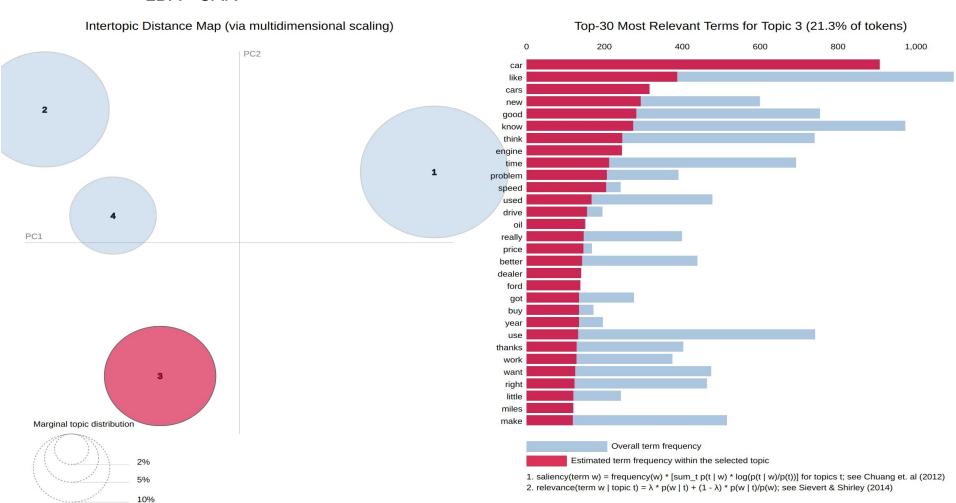
Top-30 Most Relevant Terms for Topic 1 (37.4% of tokens)



LDA - Religion



LDA - CAR

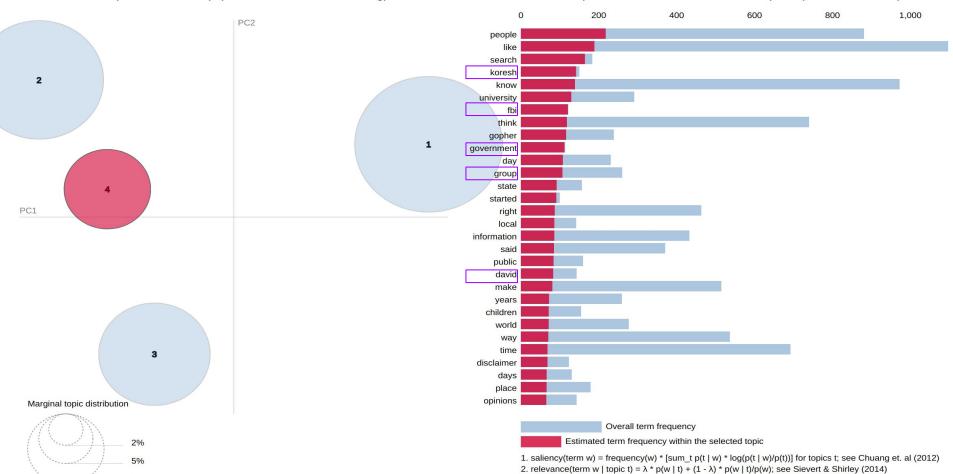


LDA - Unaspected topic

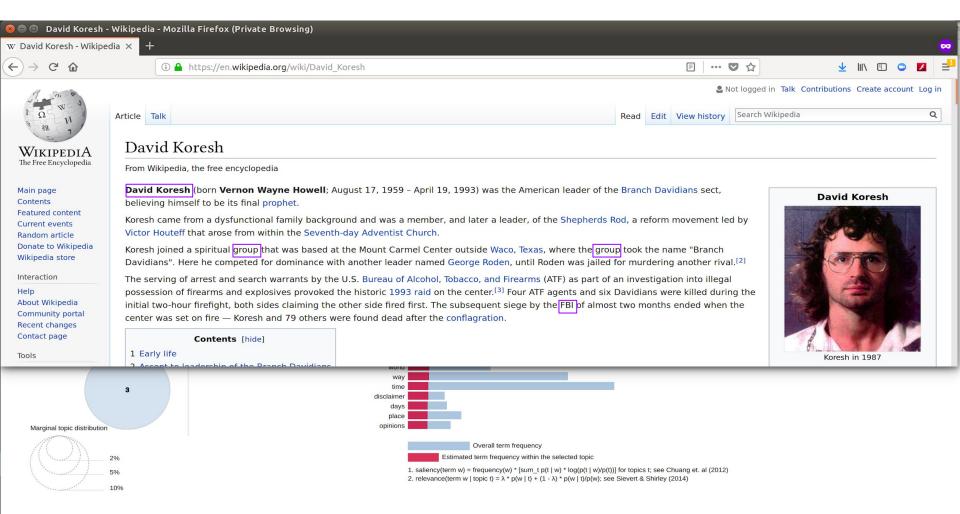
Intertopic Distance Map (via multidimensional scaling)

10%

Top-30 Most Relevant Terms for Topic 4 (12.8% of tokens)



LDA - it is about David Koresh !!!



Conclusions

- Dimensionality reduction through SVD improves the quality of clustering techniques for topic modeling
- Comparing the value of the components conveys informations about the topic distribution
- Extending the number of clusters beyond the number of expected categories by looking at the singular values can lead to the identification of new topics!
- Future :
 - Extend number of newsgroups
 - Train deep learning model on each clusters

