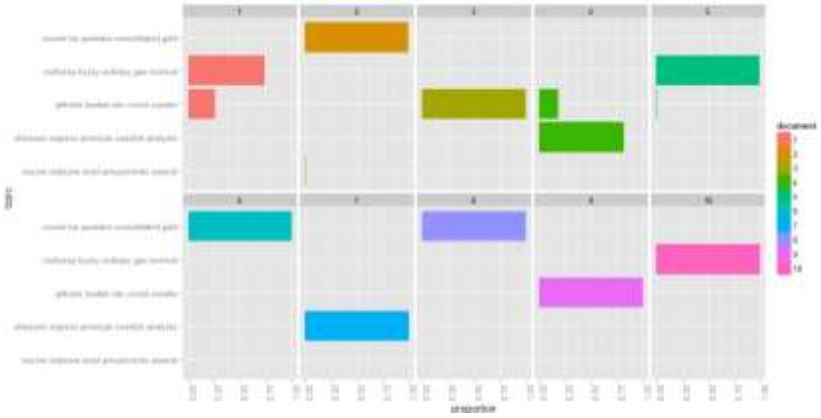
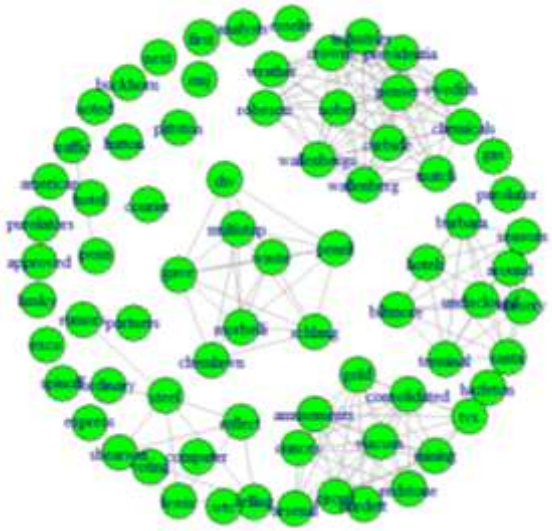
Text Mining Lab: Visualization of Topic Modelling

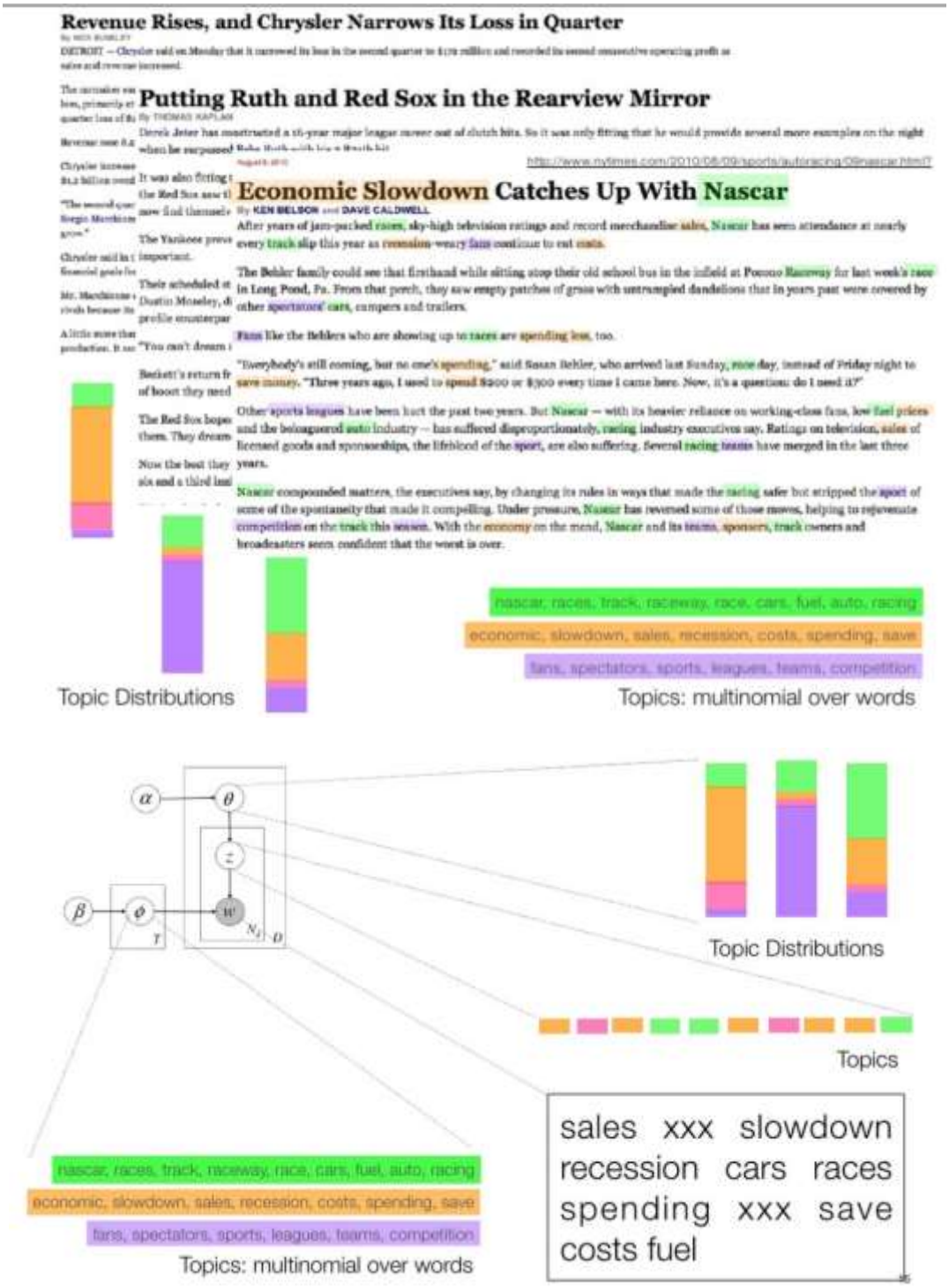
You will be able to create following visualization:





What is the "Topic Modeling"?

Probabilistic models for uncovering the underlying semantic structure of a document collection based on hierarchical Bayesian analysis of the original texts.



LDA Illustration (Yohan&Alice, 2010)

Tools:

"Gensim" ,"Tethne" and "Mallet" on Python: http://devo-evo.lab.asu.edu/?q=node/474

"lda" and "topicmodels" on R: http://cran.r-project.org/web/packages/lda/ (or /topicmodels)

Data :

data "acq" from "tm" package: a list of 50 Reuters articles

Preprocess data for topic modeling

#Pre-process data:

>install.packages("tm")

> require("tm")

 # Load the corpus of Reuters articles from the package of "tm"

>data(acq)

>reuters<-Corpus(VectorSource(as.vector(acq)))

#Convert to lower case

> reuters <- tm_map(reuters, tolower)

#Remove stop words

> reuters <- tm_map(reuters, removeWords, stopwords("english"))

 # Creat an term-document matrix

> tdm <- TermDocumentMatrix(reuters, control = list(removePunctuation = TRUE, removeNumbers = TRUE))

#Inspect() to check part of the term-document matrix

#Further Thought: How to stem words and typos?

reuters <- tm_map(reuters,stemDocument) #May need to install "SnowballC"

Construct LDA model and visualize topics in documents

#LDA model

```
> install.packages("topicmodels")

> require("topicmodels")
```

#Choose number of topics (we arbitrarily choose 5 here) and type of methods(VEM or Gibbs)

```
>lda_model <- LDA(tdm, method="VEM", control = list(alpha = 0.1), k = 5)

>lda_inf <- posterior(lda_model, tdm)
```

 # lda_inf contains two matrices: topics and terms

#Choose top five possible words to represent each topic

#Use top.topic.words()function from lda package to do so

```
>install.packages("lda")

>require("lda")

> top.words <- top.topic.words(t(lda_inf$topics), 5, by.score=TRUE)
```

#Get topic proportion in each documents

```
>topic.proportions <- t(lda_inf$terms) / colSums(lda_inf$terms)
```

#Assign topic as the top five words of each topic

```
> colnames(topic.proportions) <- apply(top.words, 2, paste, collapse=" ")
```

#Use melt() function from "reshape2" to transform data for the plot purpose

```
>install.packages("reshape2")

>require(reshape2)

> topic.proportions.df <- melt(cbind(data.frame(topic.proportions), document=factor(1:50)),
variable.name="topic",id.vars ="document")
```

#Visualize topics in documents:

```
>require(ggplot2)

> qplot(topic, value, fill=document, ylab="proportion", data=topic.proportions.df,
geom="bar")+opts(axis.text.x = theme_text(angle=90, hjust=1))+coord_flip()+facet_wrap(~
document, ncol=5)
```

Visualize words in topics

#Visualize words in topics

#Visualize top 15 words in each topic

```
>top.words <- top.topic.words(t(lda_inf$topics), 15, by.score=TRUE)
```

#Search for the top 15 words

```
> index<-c()

> for(element in top.words)   index<-c(index,which(rownames(lda_inf$topics)==element))
```

#Use heatmap() to plot

```
>hm<- heatmap(lda_inf$topics[index,], Rowv=NA, Colv=NA, col = cm.colors(256), scale="column", margins=c(5,10))
```

#Create another heatmap sorted by alphabetic order

```
> sub_matrix<-lda_inf$topics[index,]
```

#Use order() function to sort the matrix

```
>sorted_matrix<-sub_matrix[order(rownames(sub_matrix)),]
```

#Assign column names

```
>colnames(sorted_matrix)<-c("Topic1","Topic2","Topic3","Topic4","Topic5")

> hm2 <- heatmap(sorted_matrix, Rowv=NA, Colv=NA, col = cm.colors(256), scale="column", margins=c(5,10))
```

Visualize a word-network

#Visualize a network of words :

```
> correlation<-cor(t(lda_inf$topics))
```

#Choose top 15 words of each topics from above to get a subset of correlation matrix

```
> index<-c()
> for(element in top.words)   index<-c(index,which(rownames(correlation)==element))
```

#Get a subset of correlation matrix

```
> adjacency<-correlation[index,index]
>diag(adjacency)<-0
```

#Use functions from "igraph" package to construct a network object and then plot it.

```
>install.packages("igraph")
> require("igraph")
```

#Convert adjacency matrix to a network object

```
>g<-graph.adjacency(adjacency,mode="undirected")
> plot(g,layout=layout.kamada.kawai,vertex.color="green")
```

After Class Exercise:

Use the "get.edgelist()" function from the "igraph" package and import edgelist to a .csv file using write.csv() function. Reconstruct a nicer word-network on Gephi.

Reference:

Blei, D. M., Ng, A. Y., & Jordan, M. I. (2003). Latent dirichlet allocation. *the Journal of machine Learning research*, *3*, 993-1022.

Yohan Jo& Alice Oh (2010)[Slides Show], Aspect and Sentiment Unification Model, *AMC Websearch and Data Mining* https://vialogues.com/vialogues/play/14351