

TEXT MINING THE HISTORY OF IDEAS

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GENERIC “TOY” QUESTIONS

Some generic questions that may interest intellectual historians:

- **Concept prevalence**: how prevalent is concept X ?
- **Concept association**: which concepts are typically associated with X ?
- **Topic discovery**: which topics are frequently discussed?
- **Topic prevalence**: is there a correlation between the prevalence of different topics?

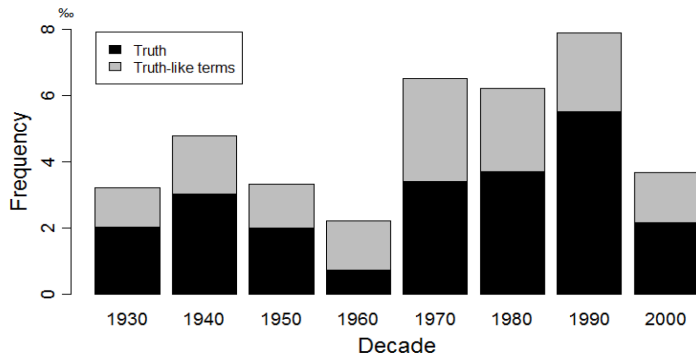
Intellectual historians typically assess such questions **qualitatively**.

I will illustrate how to approach them **quantitatively**.

TEST CORPUS

- Sample of 324 philosophical articles from 1930-2010:
 - The Journal of Philosophy: 118 articles
 - The Philosophical Review: 206 articles
- Subcorpora:
 - 1930s: 38 articles
 - 1940s: 37 articles
 - ⋮
 - 2010s: 53 articles

PREVALENCE OF “TRUTH-LIKE” CONCEPTS (1)



PREVALENCE OF “TRUTH-LIKE” CONCEPTS (2)

```

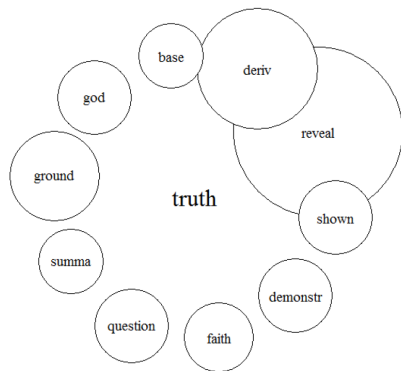
myscript.R* x
Source on Save

146
147 ### Relative frequency of set of words over decades ###
148
149 # Function: words+decade --> rel. freq. of words in decade
150 words.freq.dec.f <- function(dtm,words.v,decade){
151   dec.art.m <- dtm[which(dtm[,1] %in% c(decade:(decade+9))),]
152   total <- sum(as.numeric(dec.art.m[,3:ncol(dec.art.m)]))
153   raw.words.freq <- sum(as.numeric(dec.art.m[,which(dimnames(dec.art.m)[[2]] %in% words.v)]))
154   words.freq.dec <- raw.words.freq/total
155   return(words.freq.dec)}
156
157 # Vector: rel. freq. of words over decades
158 words.freq.f <- function(dtm,words.v){
159   words.freq.l <- list()
160   for (i in 1:length(decades)){
161     words.freq.l <- c(words.freq.l,words.freq.dec.f(dtm,words.v,decades[[i]]))
162   }
163   words.freq.v <- unlist(words.freq.l)
164   return(words.freq.v)}
165
166 # Plot: rel. freq. of words against decades
167 truth.v <- c("truth")
168 truth.phrases.v <- c("false","falsity","falsehood","truths","untrue","truthful")
169 barplot(words.freq.f(cor.sparse.md.m,truth.phrases.v),names.arg = decades,
170         main = "Frequency of 'truth/falsity' terms",xlab = "Decade",ylab = "Frequency")
171 barplot(words.freq.f(cor.sparse.md.m,truth.v),names.arg = decades,
172         main = "Frequency of 'truth/falsity' terms",xlab = "Decade",ylab = "Frequency",
173         legend("topright",legend = c("a", "b"),fill = c("black","grey"))

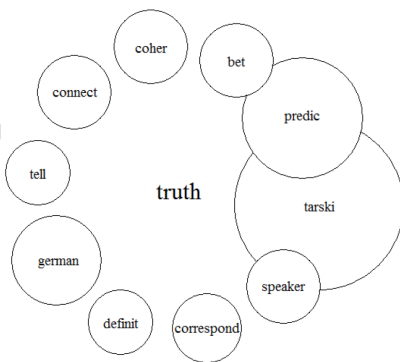
```

WORDS ASSOCIATED WITH “TRUTH” (1)

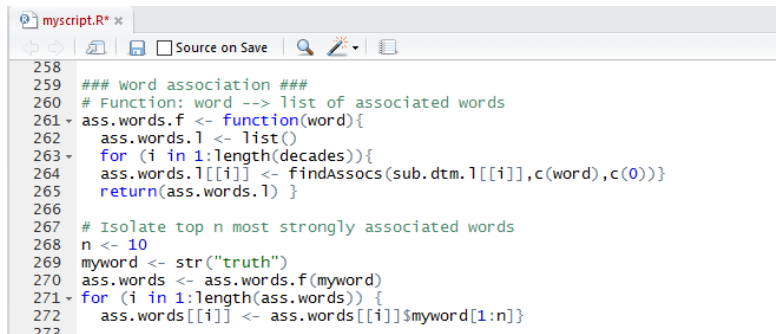
1940s



1990s



WORDS ASSOCIATED WITH “TRUTH” (2)



```
258
259 ### word association ###
260 # Function: word --> list of associated words
261 ass.words.f <- function(word){
262   ass.words.l <- list()
263   for (i in 1:length(decades)){
264     ass.words.l[[i]] <- findAssocs(sub.dtm.l[[i]],c(word),c(0))
265     return(ass.words.l) }
266
267 # Isolate top n most strongly associated words
268 n <- 10
269 myword <- str("truth")
270 ass.words <- ass.words.f(myword)
271 for (i in 1:length(ass.words)) {
272   ass.words[[i]] <- ass.words[[i]]$myword[1:n]}
273
```

TOPIC DISCOVERY (1)

Topic models are algorithms that allow us to identify hidden thematic patterns in (sets of) texts.

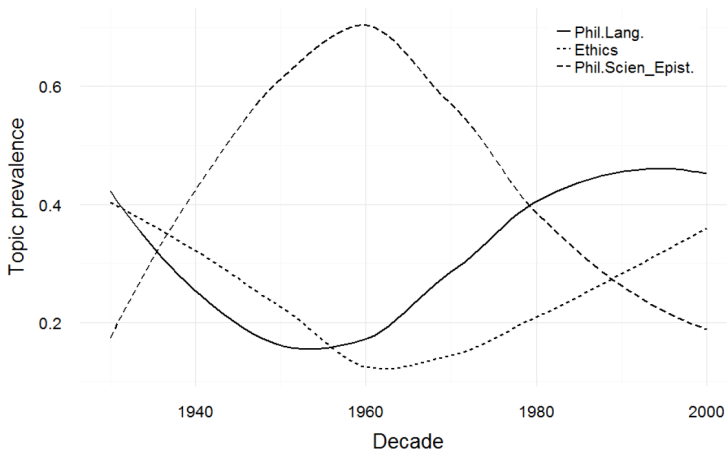
- A **topic** is treated as a distribution over **words**.
- A **document** is treated as a distribution over **topics**.



TOPIC DISCOVERY (2)

```
myscript.R* x
Source on Save
310
311 ### Topic prevalence ###
312
313 # Number of topics
314 n <- 3
315
316 # Run LDA using Gibbs sampling
317 ldaOut <- LDA(cor.sparse.dtm,n,method="Gibbs",control = list(seed=seed))
318
319 # Posterior probability of each word in each topic
320 ldaOutpost.1 <- posterior(ldaOut, cor.sparse.dtm)
321
322 # List of topics
323 top.wc.1 <- list()
324 for (i in 1:n){
325   top.wc.1[[i]] <- sort(ldaOutpost.1$terms[i,], decreasing = T)
326 }
327
328 # word cloud of topic
329 top.nr <- 2
330 word.nr <- 25
331 greyscale <- brewer.pal(8,"Greys")
332 wordcloud(names(top.wc.1[[top.nr]][1:word.nr]),top.wc.1[[top.nr]][1:word.nr],
333           scale=c(8,.2), random.order=FALSE, rot.per=.15,colors = greyscale)
334
```

TOPIC PREVALENCE



Thanks for your attention!