M5. Minería de Texto + webscraping

Clase 6. Embeddings y clasificación de texto



Hipótesis distribucional

- "El significado deriva del uso de las palabras en el lenguaje" (Wittgenstein)
- "Conocerás a una palabra por su compañía" (Firth)
- Palabras cercanas tienen sentidos "cercanos"
- İtems lingüísticos con distribuciones similares tienen significados similares"
- Idea de co-ocurrencia => términos que ocurren juntos







Aplicaciones en Ciencias Sociales



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ARTICLE OPEN

Automated analysis of free speech predicts psychosis onset in high-risk youths

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BACKGROUND/OBJECTIVES: Psychiatry lacks the objective clinical tests routinely used in other specializations. Novel computerized methods to characterize complex behaviors such as speech could be used to identify and predict psychiatric illness in individuals.

AIMS: In this proof-of-principle study, our aim was to test automated speech analyses combined with Machine Learning to predict later psychosis onset in youths at clinical high-risk (CHR) for psychosis.

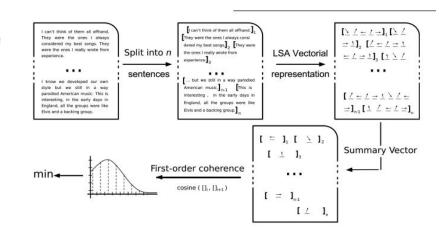
METHODS: Thirty-four CHR youths (11 females) had baseline interviews and were assessed quarterly for up to 2.5 years; five transitioned to psychosis. Using automated analysis, transcripts of interviews were evaluated for semantic and syntactic features predicting later psychosis onset. Speech features were fed into a convex hull classification algorithm with leave-one-subject-out cross-validation to assess their predictive value for psychosis outcome. The canonical correlation between the speech features and prodromal symptom ratings was computed.

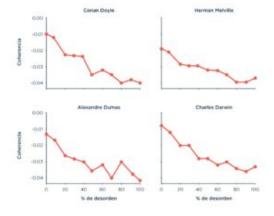
RESULTS: Derived speech features included a Latent Semantic Analysis measure of semantic coherence and two syntactic markers of speech complexity: maximum phrase length and use of determiners (e.g., *which*). These speech features predicted later psychosis development with 100% accuracy, outperforming classification from clinical interviews. Speech features were significantly correlated with prodromal symptoms.

CONCLUSIONS: Findings support the utility of automated speech analysis to measure subtle, clinically relevant mental state changes in emergent psychosis. Recent developments in computer science, including natural language processing, could provide the foundation for future development of objective clinical tests for psychiatry.

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The Geometry of Culture: Analyzing the Meanings of Class through Word Embeddings

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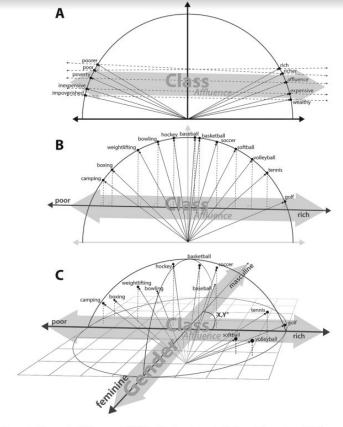
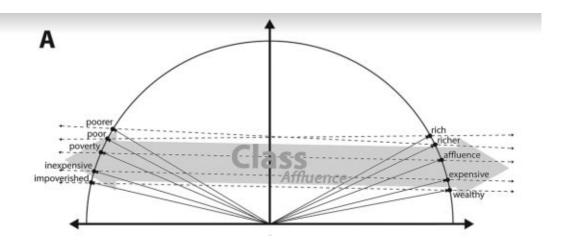


Figure 2. Conceptual Diagram of (A) the Construction of a Cultural Dimension; (B) the Projection of Words onto That Dimension; and (C) the Simultaneous Projection of Words onto Multiple Dimensions





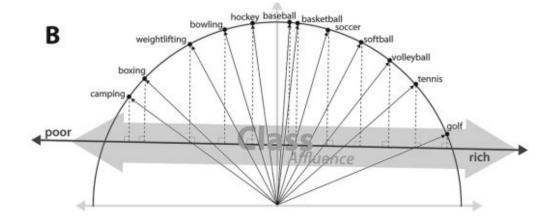
Measuring Cultural Dimensions

To identify cultural dimensions in word embedding models, we average numerous pairs of antonym words. Cultural dimensions are calculated by simply taking the mean of all word pair differences that approximate a

given dimension,
$$\frac{\sum_{p}^{|P|}\overrightarrow{p_1}-\overrightarrow{p_2}}{|P|}$$
, where p are

all antonym word pairs in relevant set P, and $\overline{p_1}$ and $\overline{p_2}$ are the first and second word vectors of each pair.¹⁷ The projection of a normalized word vector onto a cultural dimension is calculated with cosine similarity, as is the angle between cultural dimensions.







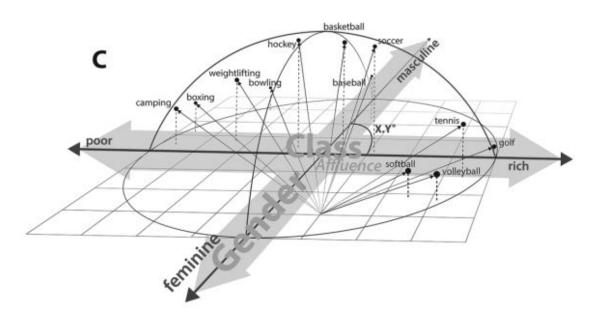


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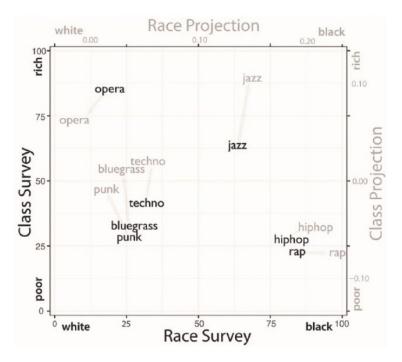


Figure 3. Projection of Music Genres onto Race and Class Dimensions of the Google News Word Embedding (Gray) and Average Survey Ratings for Race and Class Associations (Black)



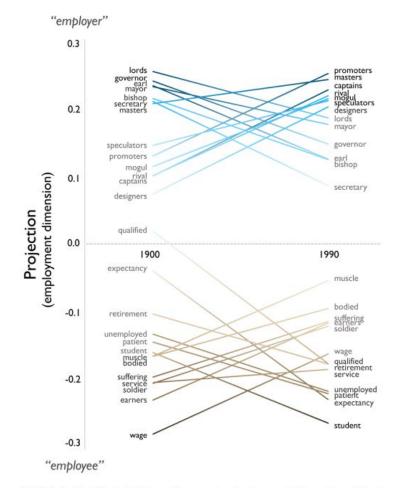


Figure 10. Words That Project High and Low on the Employment Dimension of Word Embedding Models Trained on Texts Published at the Beginning and End of the Twentieth Century; 1900–1919 and 1980–1999 Google Ngrams Corpus



Semantics derived automatically from language corpora contain human-like biases

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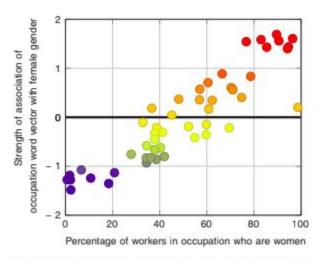


Fig. 1. Occupation-gender association. Pearson's correlation coefficient $\rho = 0.90$ with $P < 10^{-18}$.

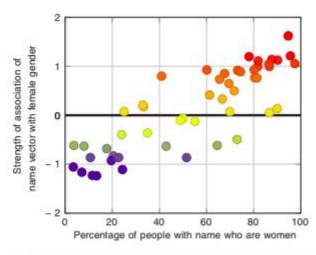
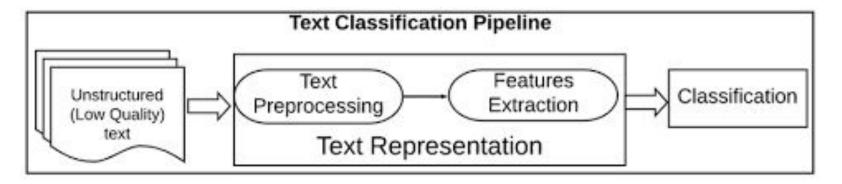
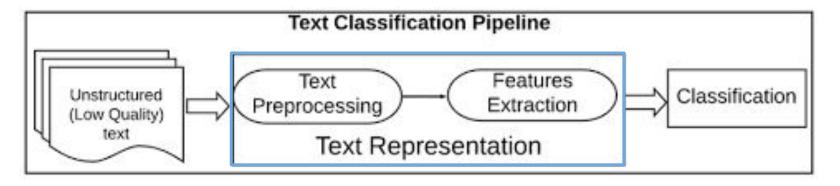


Fig. 2. Name-gender association. Pearson's correlation coefficient $\rho = 0.84$ with $P < 10^{-13}$.



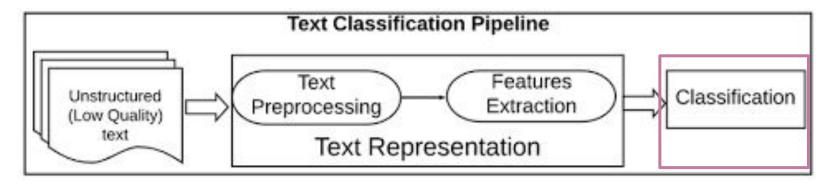






- Dos formas (que vimos):
 - Bag of Words => Term Frequency Matrix
 - Word Embeddings => Dense word Matrix





- Cualquiera de los métodos supervisados que vimos:
 - Regresión (¿lineal o logística?)
 - Random Forest
 - Gradient Boosting ...





Vamos al notebook...



Algunas cuestiones para cerrar...

- NLP "Del giro linguístico al giro (lingüístico) computacional".
- Posibilidades metodológicas para las ciencias sociales
- Discusiones "no metodológicas" que suscitan
 - Innatismo o no del lenguaje (discusión con Chomsky)
 - "Los límites de mi mundo son los de mi lenguaje".





