

# AUTOMATIC IDENTIFICATION OF ANONYMOUS PROFILES ON SOCIAL NETWORKS

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"Technology... is a queer thing.

It brings you great gifts with one hand,  
and it stabs you in the back with the other."

Charles Percy Snow, New York Times, 1971

# Cooperation: Police + Academic

- In the last years the cooperation between LEAs and Academic institutions has been increased.
- Dealing with crime has become a race for the use of technology.
  - New technologies enable new ways of crime and requires new ways of fighting it.
- (Lack of) Security is global. LEAs have to tackle crime.
- In this context, LEAs need the support of Academia for being updated.
- From the Academic point of view, LEAs show real needs and provide information based on real threats.

# Dark Side of Technology



- ICTs are also tools for for extremist groups,
  - Those sheltering / hiding behind different social movements
  - Transnational terrorist groups.
- Used for:
  - Propaganda
  - Recruitment
  - Training
  - Radicalization

# The Dark Side of Social Networks



- Offenders have moved from websites (easier to identify) to Social Networks (greater anonymity).
  - These networks ensure a wider dissemination with no costs
- Radical groups can use SN for:
  - Recruitment of new followers
    - Frequently telling stories looking to arise feelings of (in)justice
  - Mobilizing some sectors of society or specific individuals
    - SNs are currently the best tool to obtain publicity and go viral
  - Giving legitimacy to radicalization

# The Dark Side of Social Networks

- Once recruitment and mobilization has been performed, it follows the ideological recruitment → transform the recruited individuals into individuals trained and ready to act.
- For these reasons, it is important to “patrol” these channels.
  - It is vital for collecting information and intelligence
- There is a need for (automatic) technological tools supporting this task
  - Better understanding and identification of information

# Example of Hate Crime on Twitter

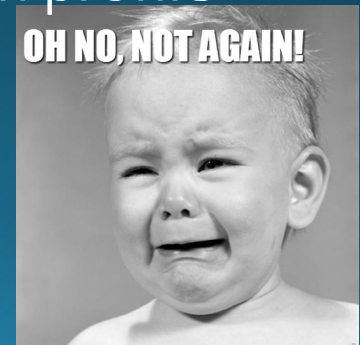
## Detienen a un tuitero en Palma de Mallorca por un delito de odio contra el colectivo LGTB

- El arrestado tenía una intensa actividad en la red social, con más de 4.000 mensajes publicados



# A “classical” problem

- One of the largest obstacles to LEAs is how easy results to create a new profile on this SN.
- It was observed, time and time again, that short time after closing a Twitter profile, similar ones came up:
  - Similar dogma and speech
  - Very difficult to identify individuals behind a given profile or publication
  - Police work should start all over again.





# The goal



- Designing and implementing a data analysis tool...
- ... Based on Semantic Analysis of written texts
- ... Intended to link an anonymous SN profile to a “public” one
  - That is, a profile with elements enabling an identification of the user: personal pictures, address, work place, previous knowledge, etc.
- The initial prototype was implemented to analyse Twitter profiles
- The proposed approach does not rely on any specific language.
  - For example, for counter terrorism activities it can be used in language as different as Spanish, English, Pasthum or Bangla.

# Similar to...



Tweets **3.122**   Siguiendo **797**   Seguidores **59,3 K**   Me gusta **6.131**   Listas **2**

**Siguiendo**

## Europol

@Europol

Official Twitter of the EU Agency for Law Enforcement Cooperation. Follow us on [facebook.com/europol](https://facebook.com/europol) & [instagram.com/europol.eu](https://instagram.com/europol.eu)

The Hague, The Netherlands

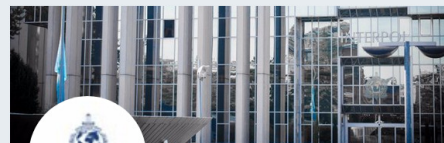
[europol.europa.eu](https://europol.europa.eu)

Se unió en diciembre de 2012

Twittear a Europol

87 Seguidores que conoces

## Recomendados



INTERPOL

Seguir

### INTERPOL

@INTERPOL\_HQ

All the latest news from the world's largest international police organization, with 192 member countries [#INTERPOL](#) [#Police](#) [#LawEnforcement](#)



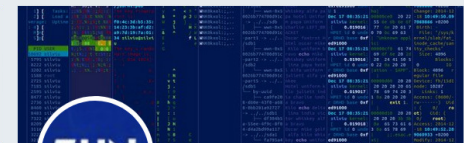
EC3

Seguir

### EC3

@EC3Europol

Official account of @Europol's European Cybercrime Centre ([#EC3Europol](#)), the EU Agency for Law Enforcement Cooperation. Our aim is to combat



THN

Seguir

### The Hacker News

@TheHackersNews

Popular, trusted, widely-read cyber security news source for everyone, including hackers, technologists, enthusiasts & IT nerds » [fb.com](https://fb.com)

# ... but more subtle



- What if the target profile wants to “hide” itself?

## How to Hide Your Identity on Twitter

*by Michelle Varsallona ; Updated September 28, 2017*

- Our hypothesis is that a given person will use similar speeches in all of them
  - The real person could be identified against profiles with real data that are similar to the one investigated.
- The message resemblance is evaluated based on both the topics (words) and the perspective (grammar constructions) used.

# Comparing in Twitter

- Analyzing tweet by tweet would not produce significant results.
  - Individual tweets are too short. (280 chars from Nov.7 !!).
- Our proposal is to combine all the tweets of a profile in a single document.
- So comparing Twitter profiles → comparing text documents.

# Comparing documents

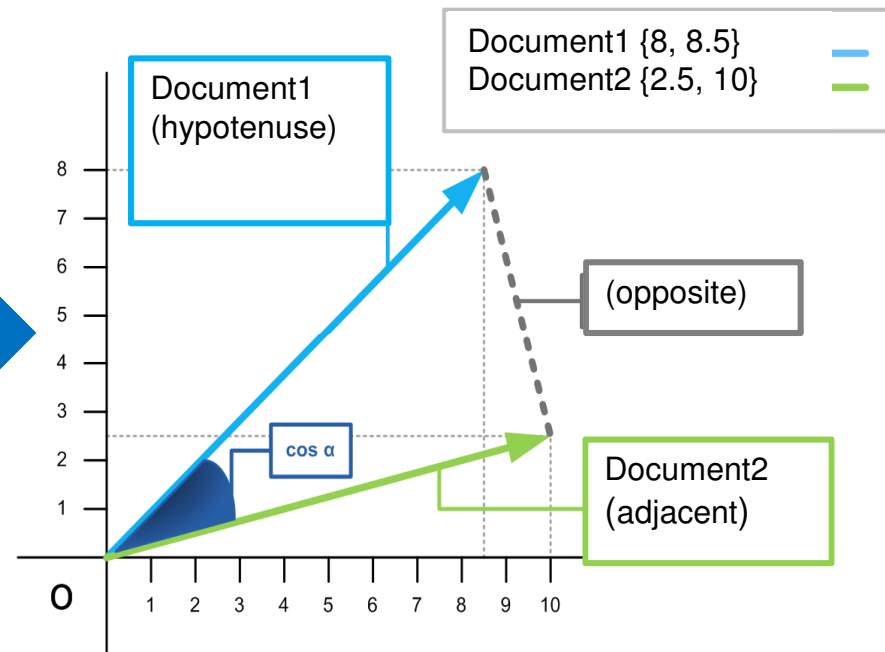
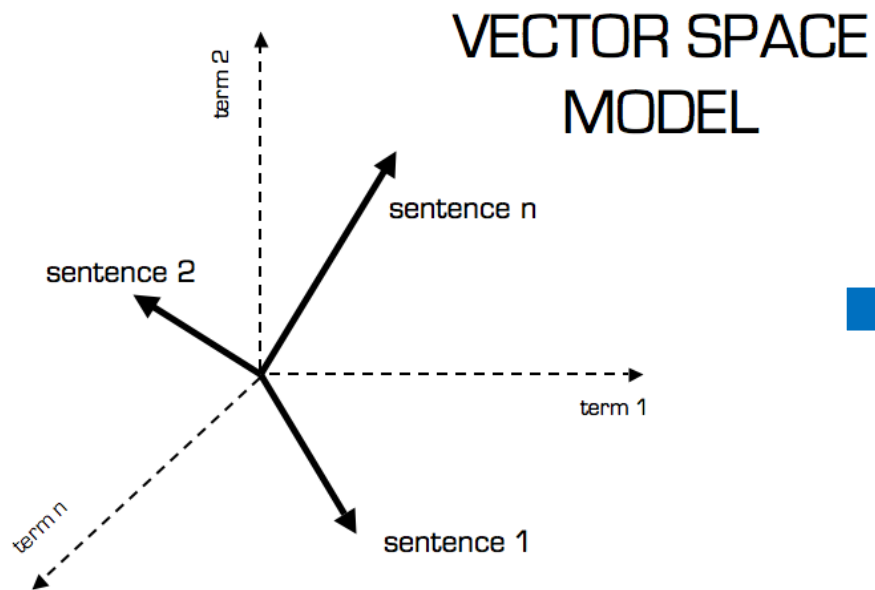
- Recurrent problem on the NLP field.
- It is needed a method for measuring document similarity

$$\frac{\text{words in common between documents}}{\text{total number of words on both documents}}$$



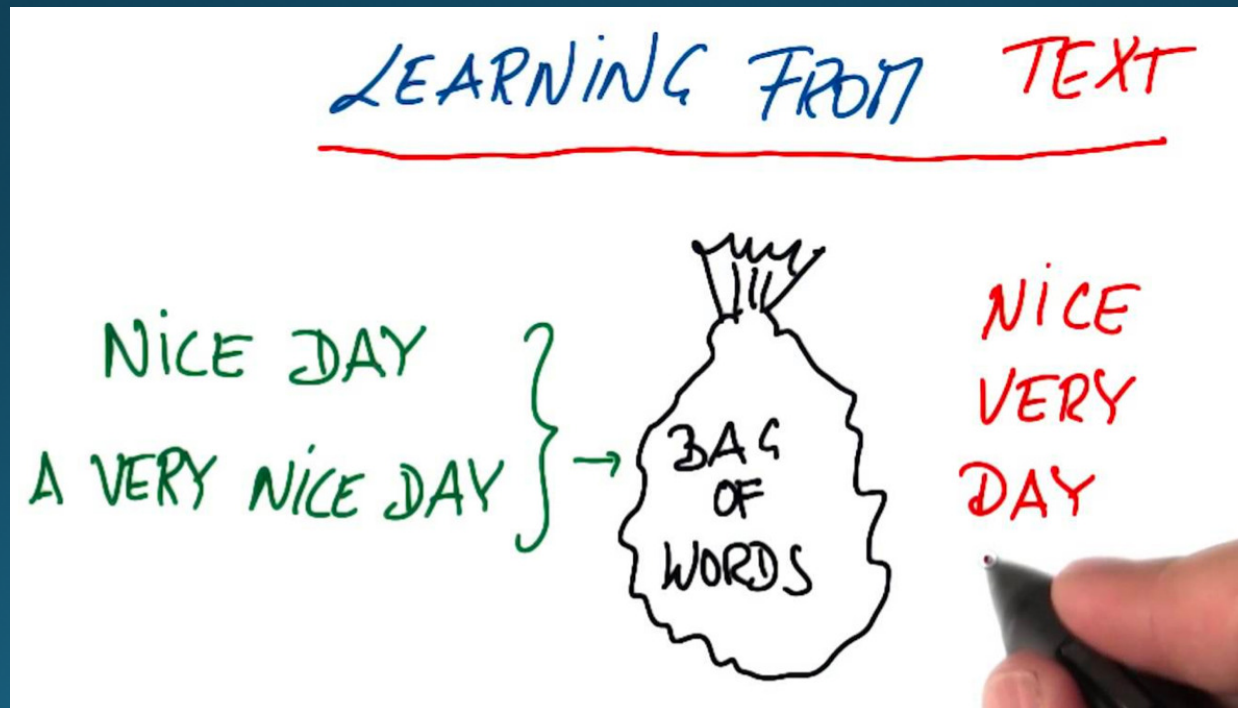
# Distance between vectors (docs)

- Frequently documents are transformed to vectors for a faster calculation



# Converting docs in vectors

- Original *Bag of Words* methods were used
  - They do not take into account the order or context of the words



# Bag of Words example

Either present/not present or counting number of times

## Document 1

The quick brown fox jumped over the lazy dog's back.

## Document 2

Now is the time for all good men to come to the aid of their party.

Term	Document 1	Document 2
aid	0	1
all	0	1
back	1	0
brown	1	0
come	0	1
dog	1	0
fox	1	0
good	0	1
jump	1	0
lazy	1	0
men	0	1
now	0	1
over	1	0
party	0	1
quick	1	0
their	0	1
time	0	1

## Stopword List

for
is
of
the
to

Most frequent words are discarded

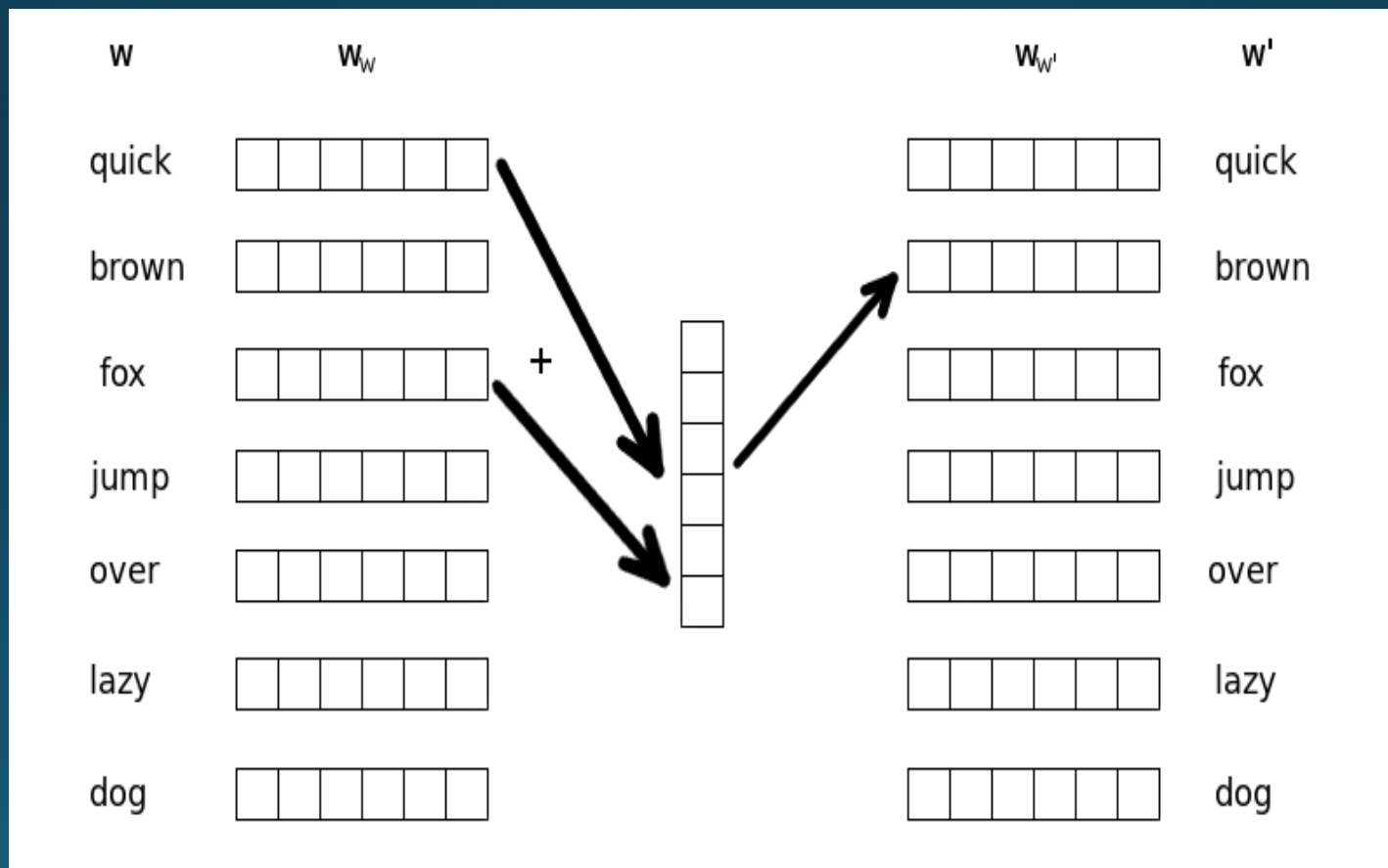


# Our proposal: to use Doc2Vec

- *Bag-of-words* methods have a fundamental limitation: because they do not consider word order, they ignore how sentences are constructed (semantic).
  - If the way of expressing is significant, the grammar constructions and verb tenses should be considered
- Doc2Vec is based on a **neural network** used to maximize the probability to predict a word based on a set of words and a given document vector.
- Depending on the goal StopWords are discarded and Steaming method used (**looking for similar topics**) or the original text is used (**looking for similar profiles**)

# Based on Word2Vec

- Predicting the next word given its context (semantic)



# Word2Vec at Work

**Input:**  
one document

Lorem ipsum dolor  
sit amet, consete-  
tur sadipscing elitr,  
sed diam nonumy  
eirmod tempor  
invidunt ut labore  
et dolore magna  
aliquyam erat, sed  
diam voluptua. At  
vero eos et



**Model:**

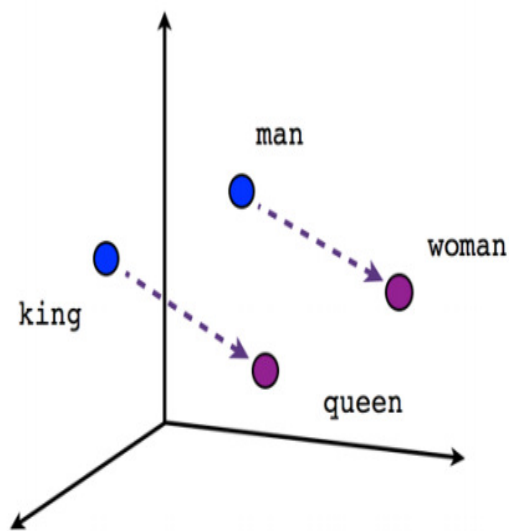


**most\_similar('france'):**

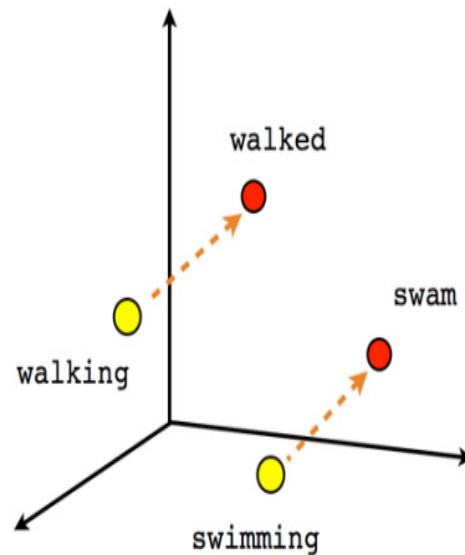
spain	0.678515
belgium	0.665923
netherlands	0.652428
italy	0.633130

highest cosine  
distance values  
in vector space  
of the nearest  
words

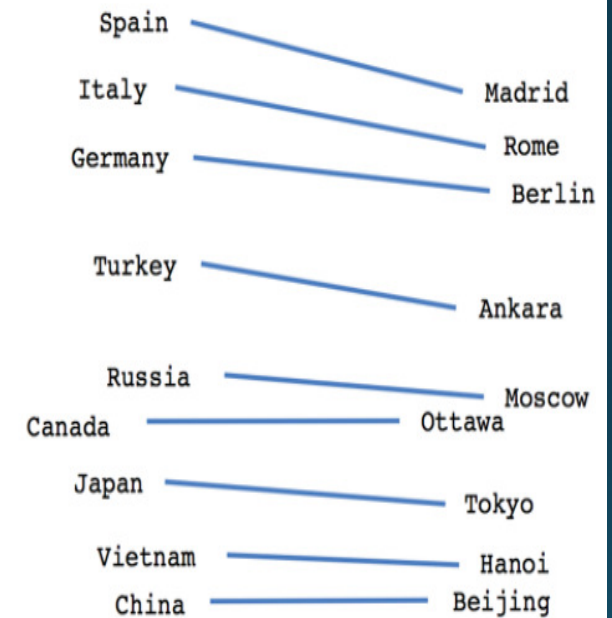
# Results of Word2Vec



Male-Female

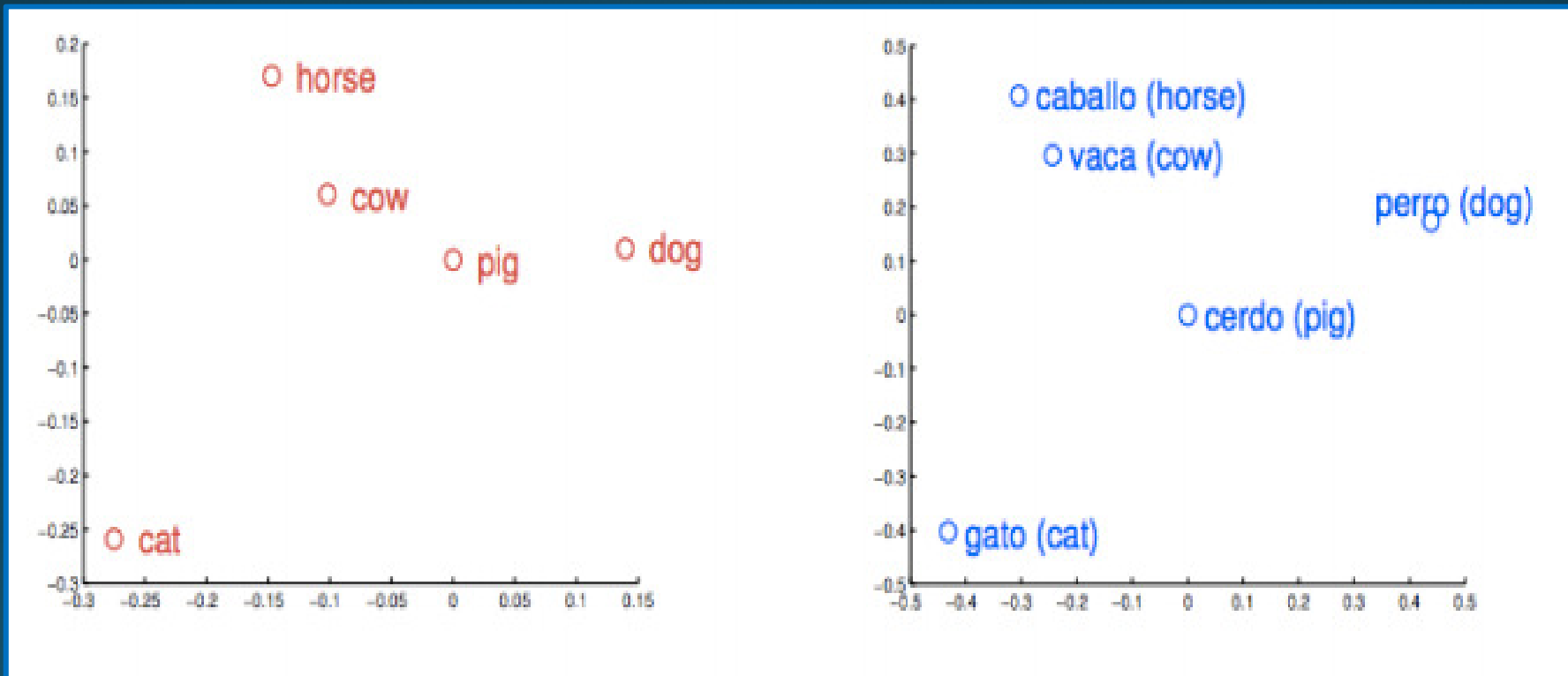


Verb tense

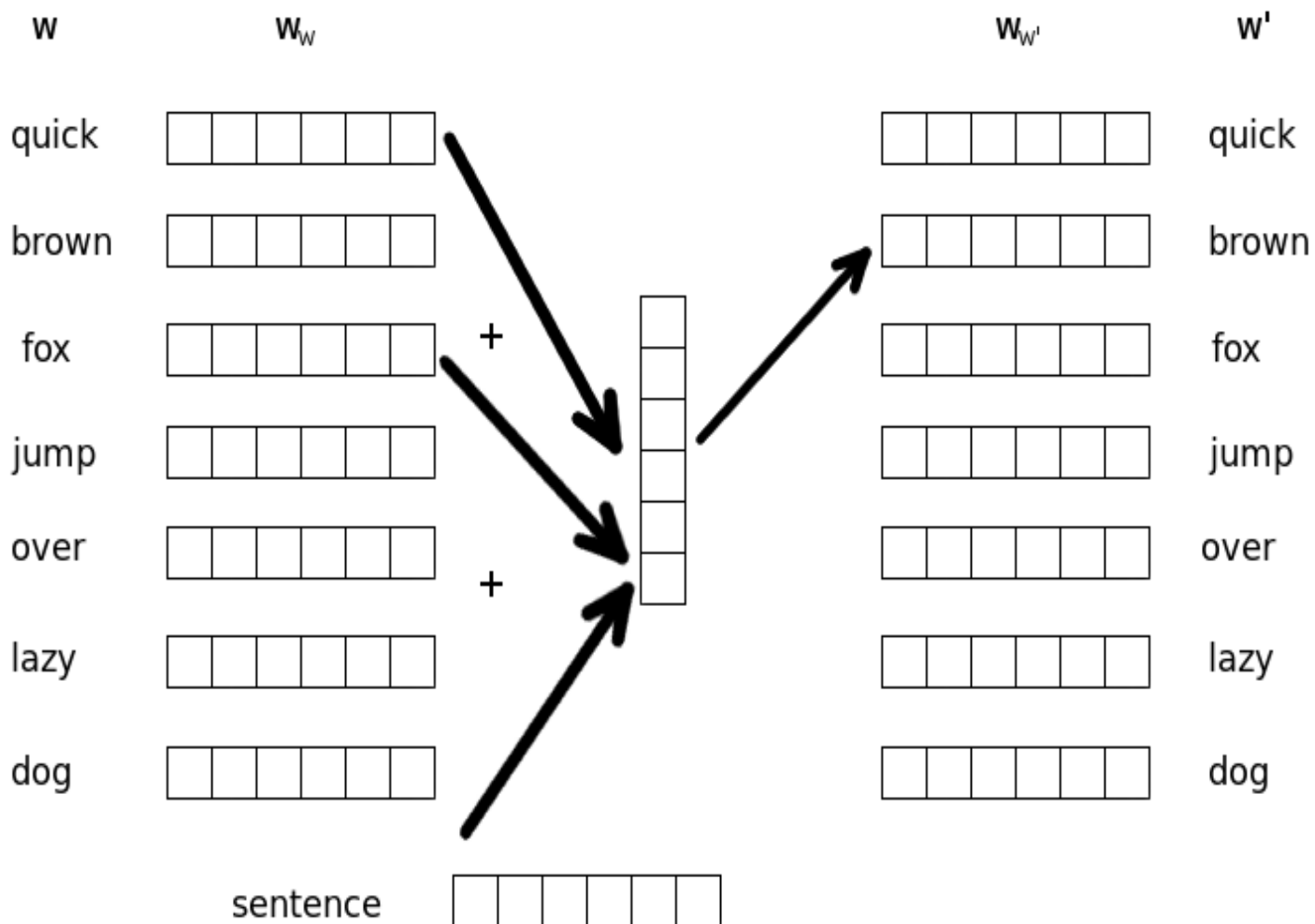


Country-Capital

# Results of Word2Vec



# Doc2Vec at Work



# Doc2Vec at Work

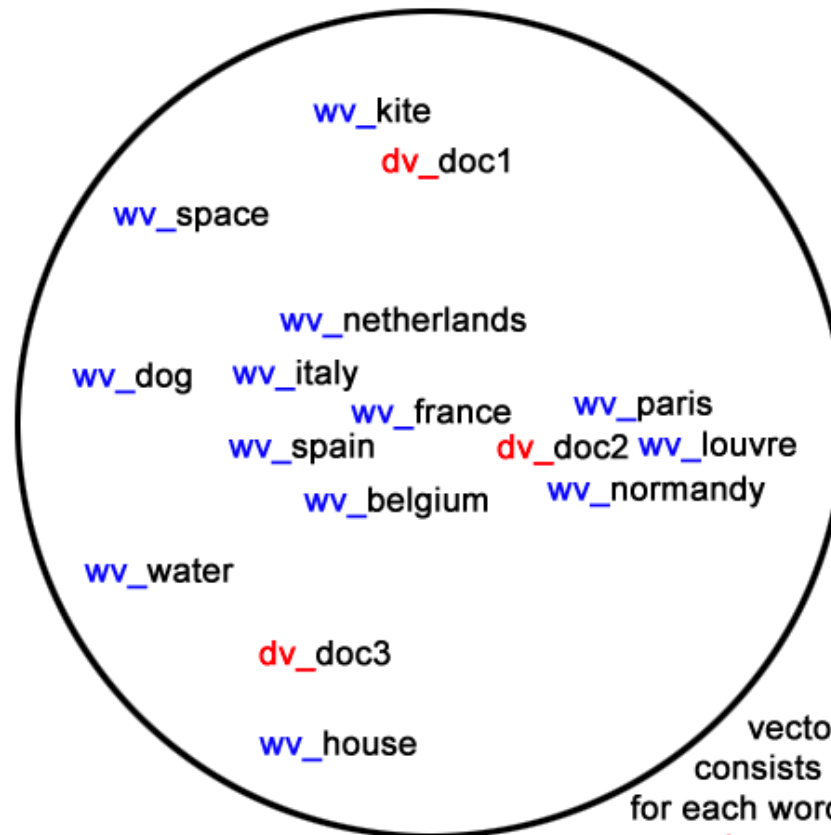
Input:  
many document



doc1,  
doc2,  
doc3 ...

training a word vector for each word and each document gets an ID/tag with a vector while training

Model:



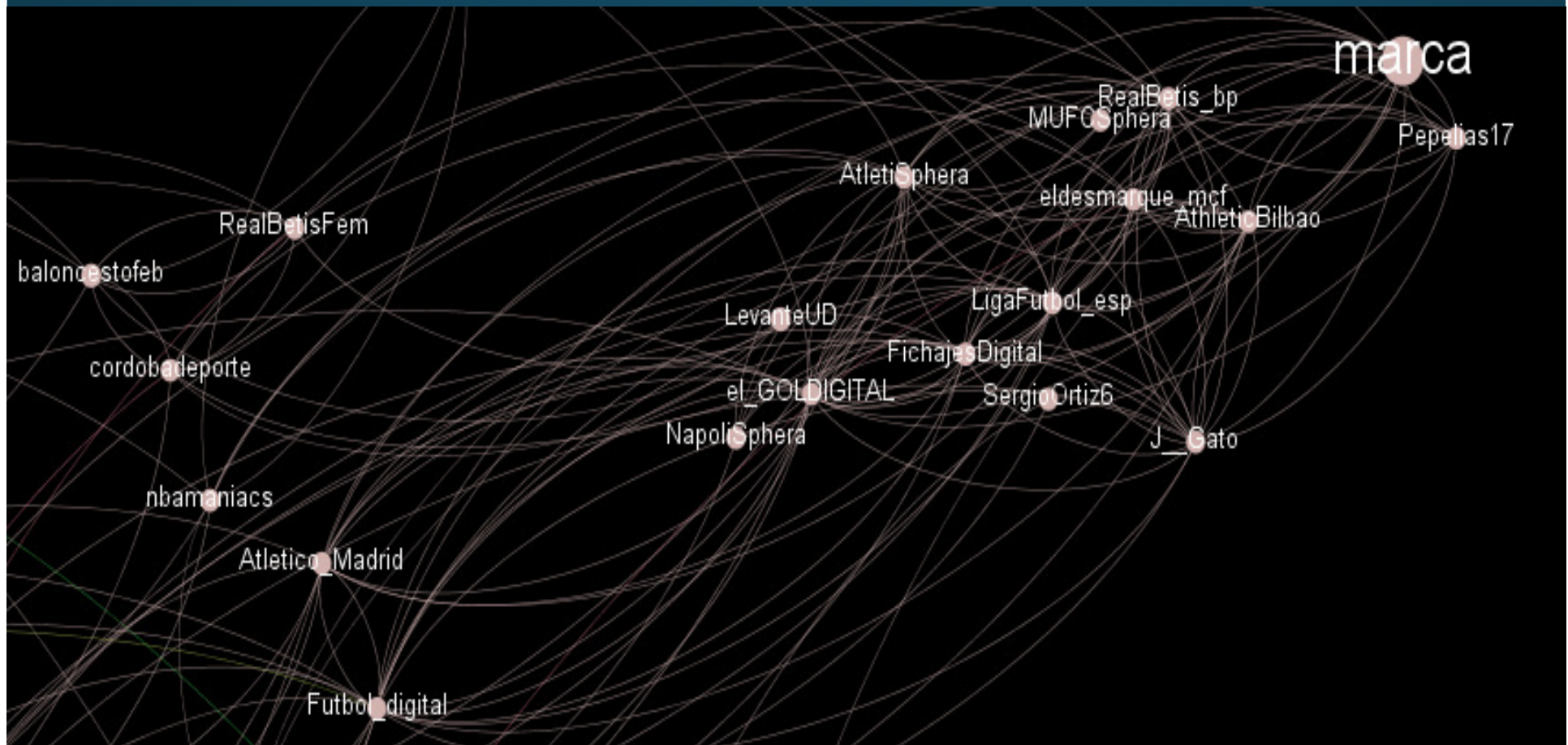
most\_similar('france'):

paris	0.876543
louvre	0.765432
normandy	0.654321
...	

highest cosine distance values in vector space with consideration of the document vectors

vector space:  
consists of **word vectors**  
for each word and additional  
**document vectors**

# Doc2Vec Results





# Example of Use: twin profiles



Buscar usuarios similares a:  ✓

Idioma: ES

Máx. Resultados: 100

Buscar similares

Buscar por:  
 Temas de los que habla  Forma de hablar

Buscar en:  
 Twitter global  Comunidad del usuario

Profile Name	Creation Date	Location	Following	Followers
@gdtguardiacivil	09/09/2010	España	62	159K
@guardiacivil	09/05/2011	España	166	545K
@polprsevilla	08/05/2013	Sevilla	1	4K
@delitosi	20/07/2012	Madrid	17K	27K
@cargongar72	02/12/2011	Una abulense en Madrid	272	3K
@aspo_sevilla	08/04/2013	Sevilla	33	...

Descargar como CSV

# The big data problem

- We need to search for profiles similar to one given among millions of candidates in a minimal time.
- We have used some of the latest technological development for dealing with large datasets
- The prototype requires some time for training (learning vectors), but then it can scan through millions of profiles in a few seconds.



# Use Cases

- The search of twin profiles can be done either vertically (within a given SN) or horizontally (looking for similar publications among different SN and forums).
- The method can also be used to link current profiles with profiles long inactive.
- Given a well-known and dangerous user, the method can be applied to look for similar users
  - That feature could enable to discover hidden relationships, even criminal networks.
- The method is indifferent to crime domain (radicalization, child pornography, hate speech, etc.) or language.

# Context of Use

- The tool would not replace the human work in any case; on the contrary, it was designed to assist police agents on their work.
  - It can save time by searching in a few minutes through millions of SN profiles, producing a small set of candidate profiles.
  - These profiles should be afterwards analyzed by experts in order to determine if any of them could correspond to the target.

# Future Work

- Algorithms for text classification can be improved, as new and more efficient versions are being continuously researched.
- Text analysis can be combined with traditional SNA analysis for measuring and looking for similarity.
- These analysis can be combined with personality profiling
  - Personality is an individual features extremely difficult to hide.
  - It can be used to risk prediction.
  - We have already made some progress on inferring personality through open source information.

Thank you very much!

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