Online citizen participation & Natural language processing

Auditing the urban planning process in Decidim Barcelona

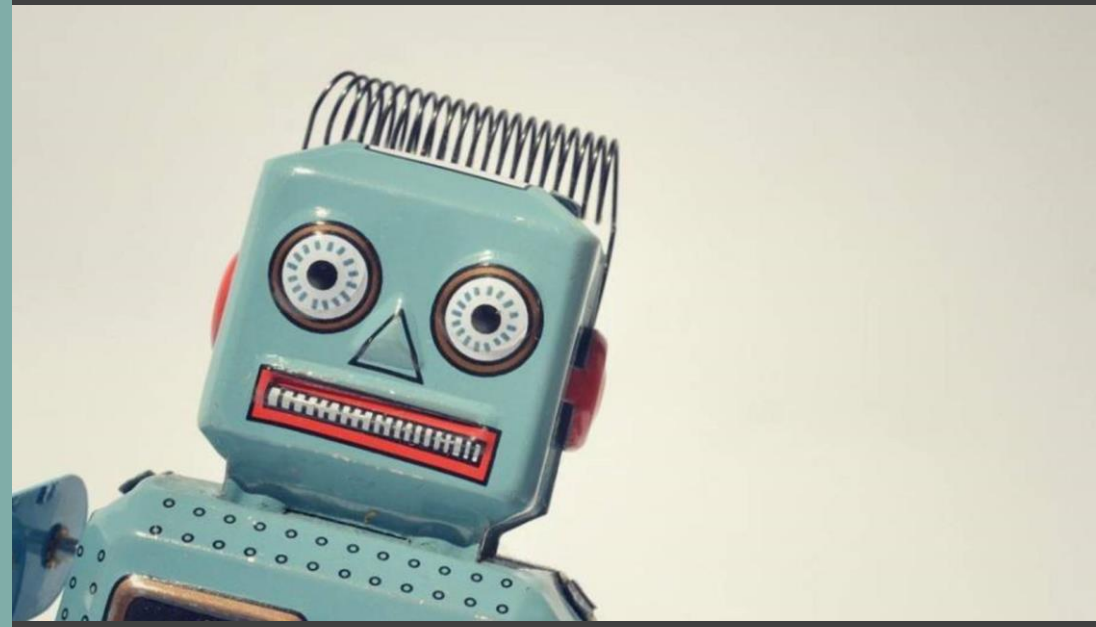
Ana Valdivia
Data Scientist

| *PyData Salamanca, 28 de Noviembre de 2019*

ROAD MAP

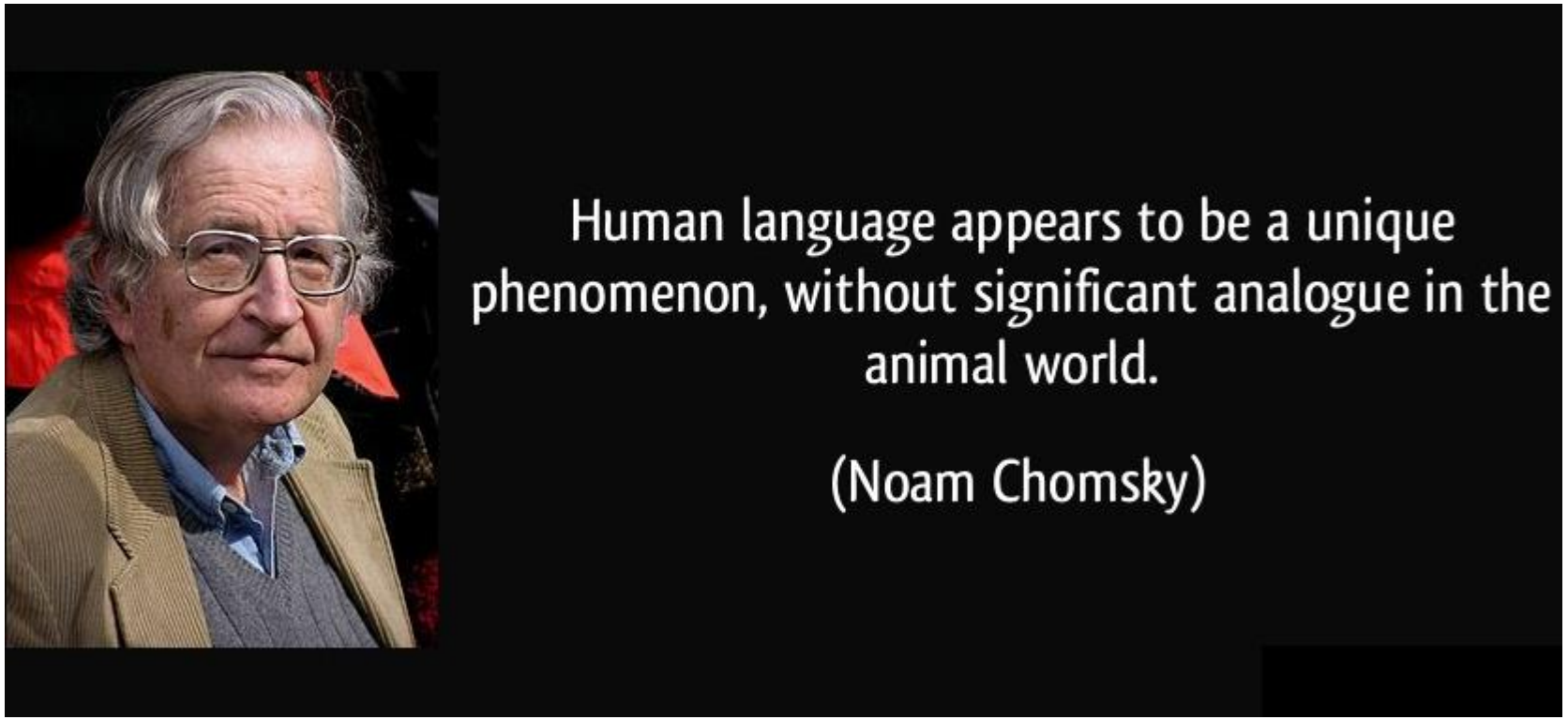
1. Intro to NLP
2. Word embeddings
3. Decidim Barcelona

Natural Language Processing



N L P

Human Language



N

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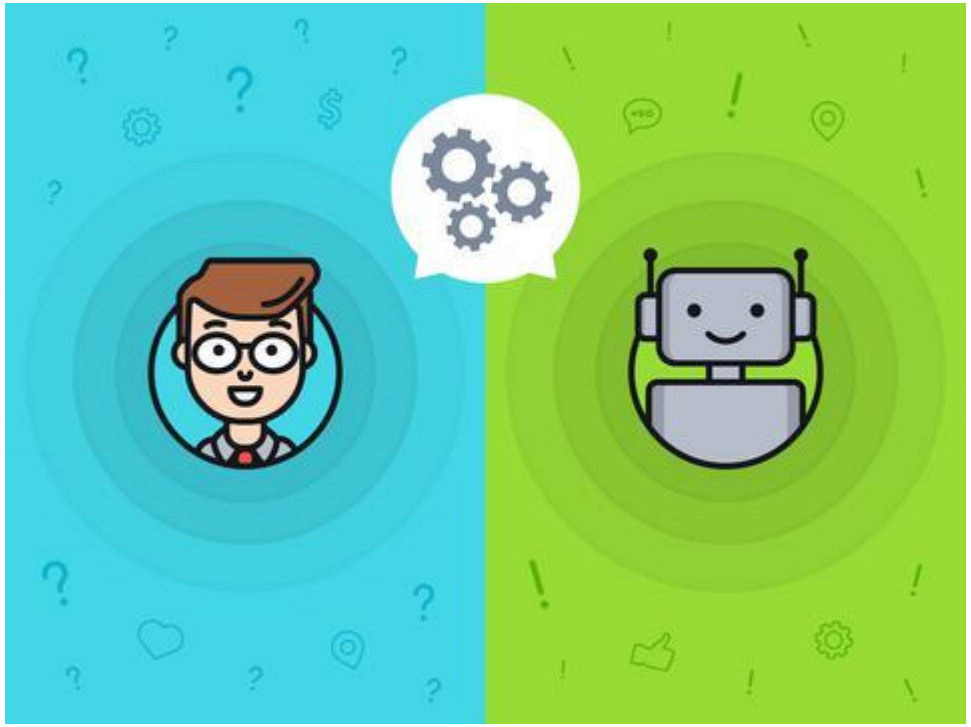
Natural Language Processing (NLP)

Natural Language Processing *(NLP)* is a sub-field of ***Artificial Intelligence*** that is focused on enabling computers to "understand", process and generate human languages, to get computers closer to a human-level understanding of language.



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Tasks:

- github.com/Kyubyong/nlp_tasks
- nlp.lsi.upc.edu/freeling

LOW

- Part of Speech Tagging
- Named Entity Recognition
- Spam Detection

MEDIUM

- Syntactic Parsing
- Word Sense Disambiguation
- Sentiment Analysis
- Topic Modeling

HIGH

- Machine Translation
- **Text Generation**
- Automatic Summarization
- Question Answering

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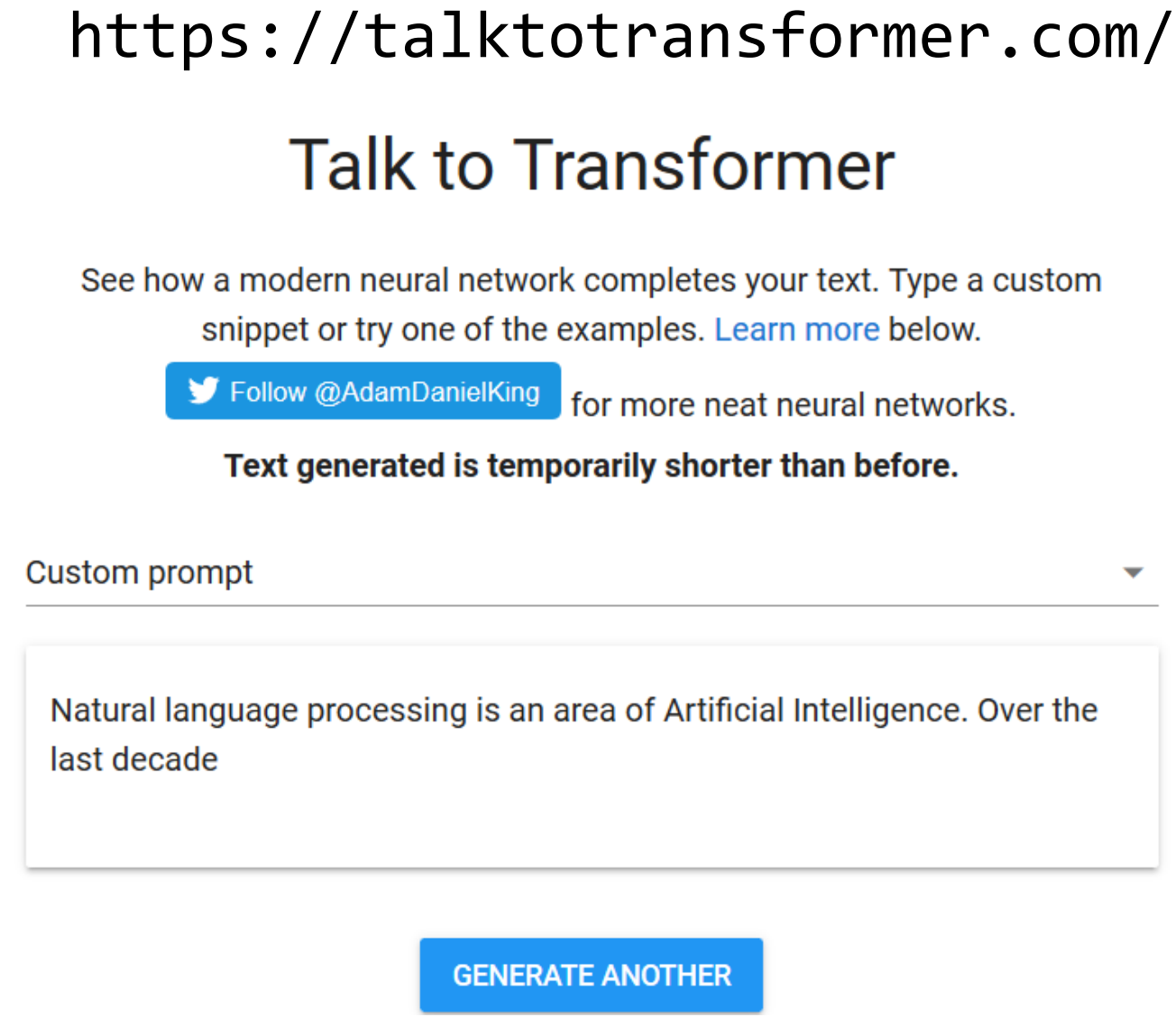
HIGH: Text Generation

https://talktotransformer.com/



N
L
P

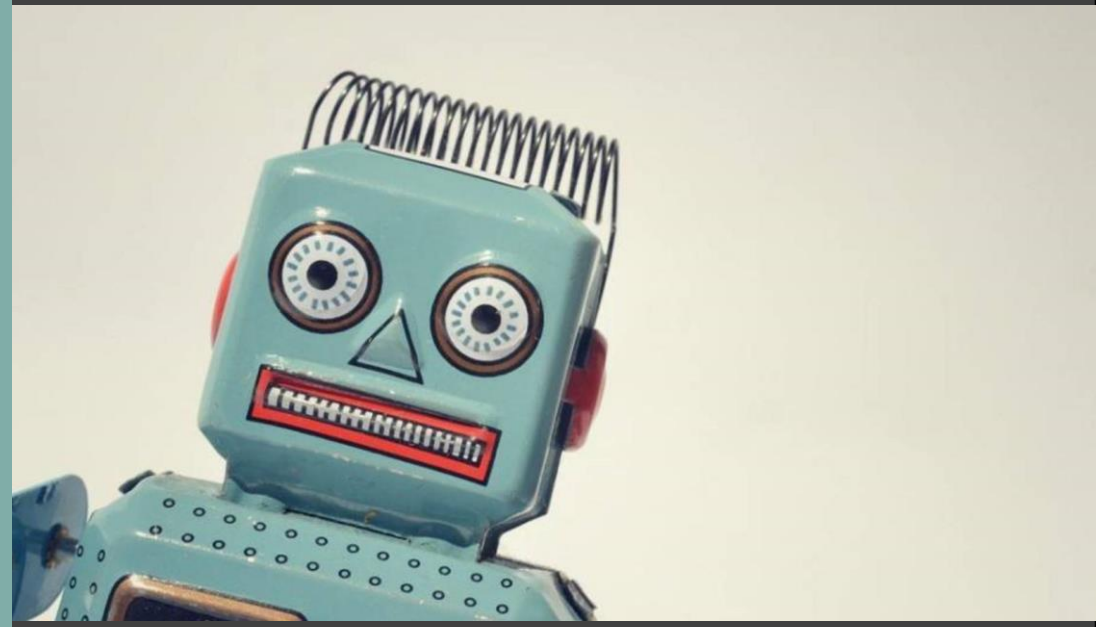
HIGH: Text Generation

https://talktotransformer.com/

Completion

Natural language processing is an area of Artificial Intelligence. Over the last decade the core component of most applications is called a neural network. Neural networks were originally developed as a general purpose statistical learning and the magic of making sense out of the large quantities of data often used in real-time human-to-computer and robot interaction. In essence, they are a class of artificial neural networks that offer more accuracy in classification and discrimination than traditional neural networks that have been used for over a decade.
Most natural language processing requires adding and subtracting features from a large collection of small chunks of text.

Word Embeddings



How to transform a word into a number?

How to transform a *word* into a *number*?

one-hot encoding

```
dog = [ 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 ]

cat = [ 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 ]
```

How to transform a *word* into a *number*?

Lack of information:

How a model will know that these two words are related/similar?

```
airplane = [ 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 ]
flight   = [ 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 ]
```

How to transform a *word* into a *number*?

word embeddings

Word embedding is the collective name for a set of language modeling and feature learning techniques in NLP where words or phrases from the vocabulary are mapped to vectors of real numbers.

CONTEXT.

- "What time is your **airplane** scheduled?"
- "The pilot marked the cruise speed on our **airplane**'s flight."
- "The engine of an **airplane** uses the propulsion force to take off."
- "Many passengers are afraid to fly, even though the **airplane** is the safest mode of transportation."

- **Airplane** is related with **scheduled, cruise speed, pilot, fligh, take off, passengers**, etc.

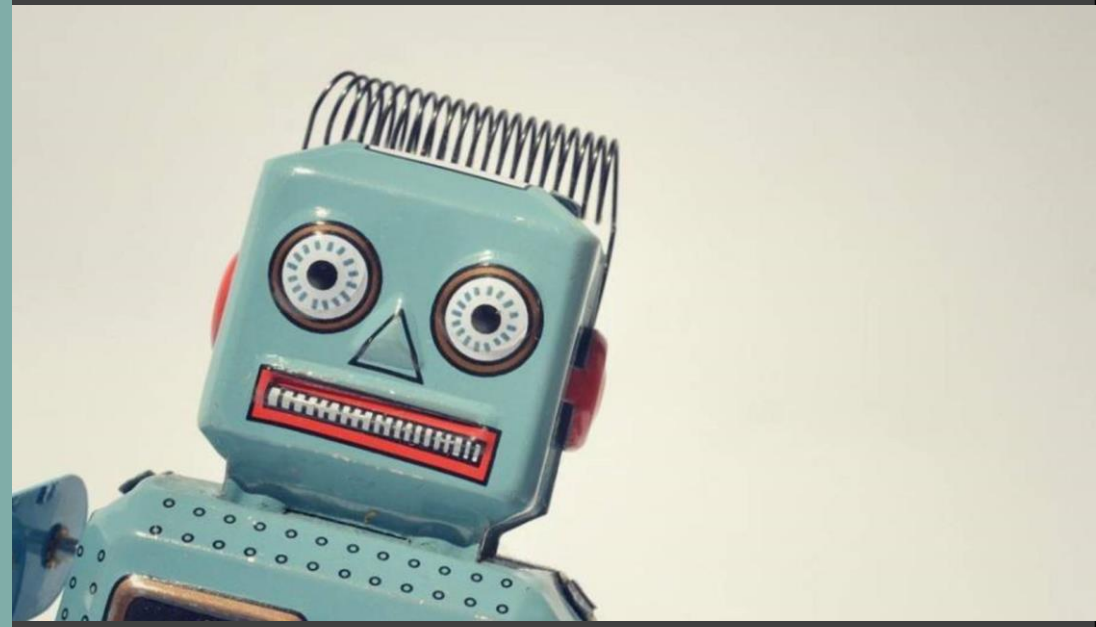
WORD EMBEDDINGS

How to transform a *word* into a *number*?

word embeddings

- Matrix representations: LDA, GloVe
- Neural networks: **word2vec**, ELMo

word2vec

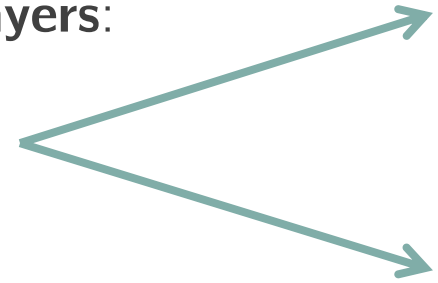


w o r d 2 v e c

How to transform a *word* into a *number*?

word2vec (Mikolov et. al. 2013)

- Two **neural networks** with two **layers**:
  - *input*: one-hot vectors.
  - *hidden* layer: lineal.
  - *output*: softmax function.

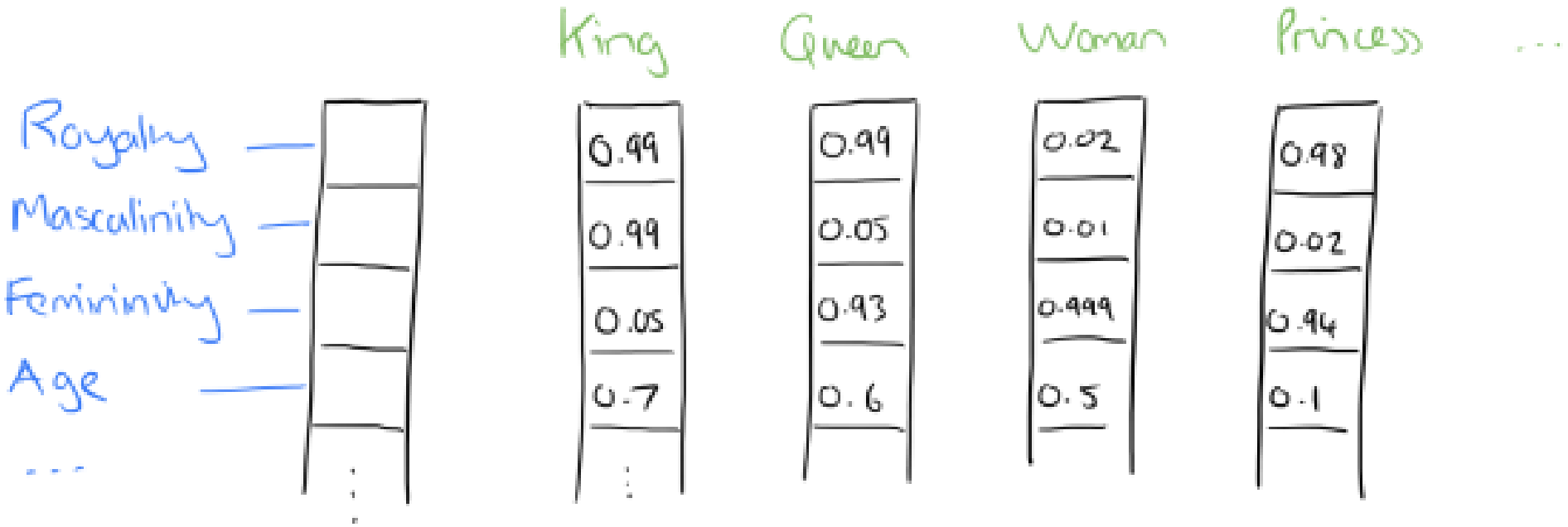


- **CBOW** (Conditional Bag of Words): given the **context** $\{w_{t-C} \dots, w_{t-1}, w_{t+1}, \dots, w_{t+C}\}$, predict the **central word** w_t.
- **Skip-gram**: given the **central word** w_t, predict the **context** $\{w_{t-C} \dots, w_{t-1}, w_{t+1}, \dots, w_{t+C}\}$.

- **Weights** of the hidden layer are the embeddings representations.
- Its performance depends on the size of the corpus (the bigger, the better).

How to transform a *word* into a *number*?

word2vec (Mikolov et. al. 2013)



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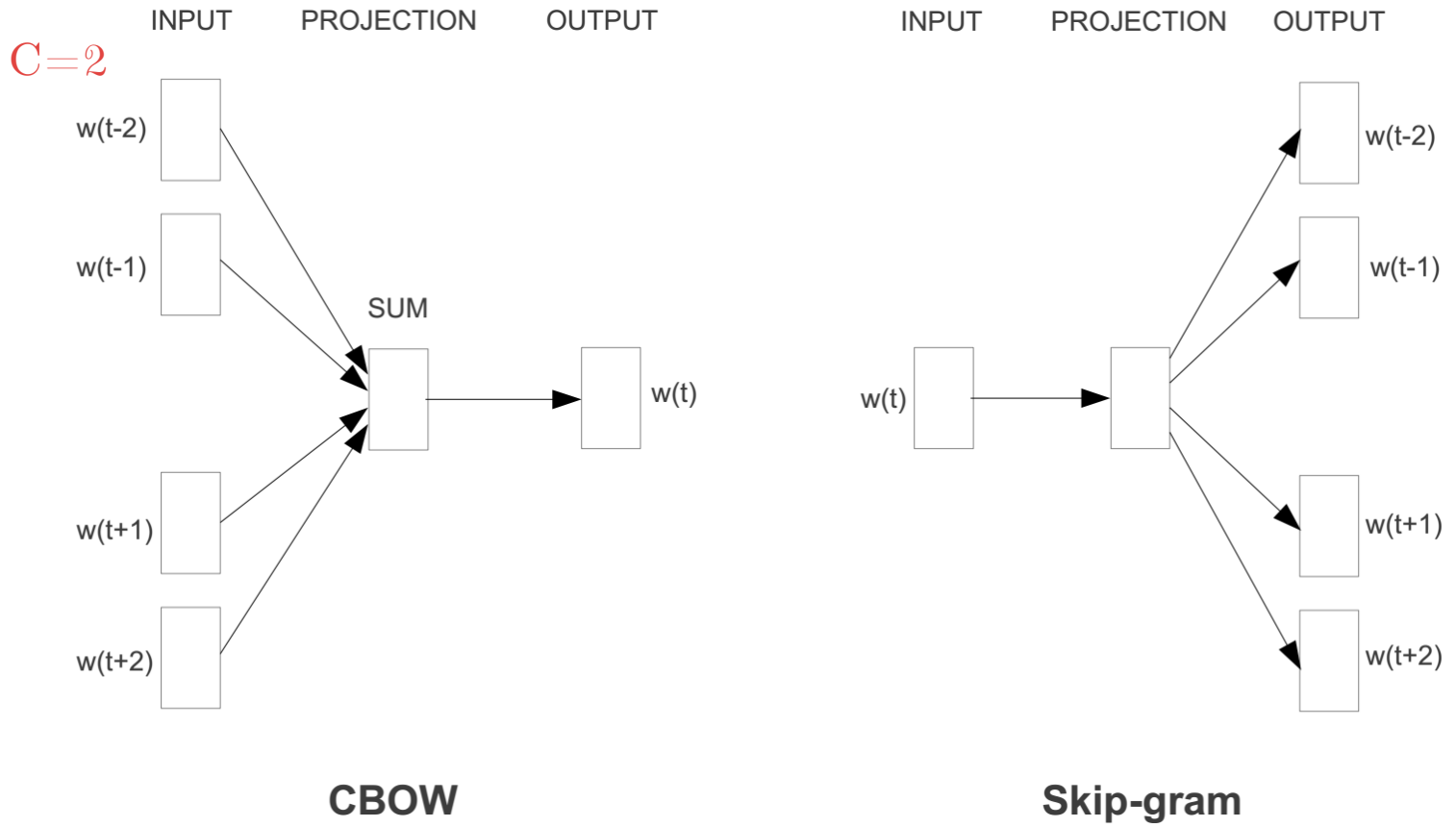

word2vec

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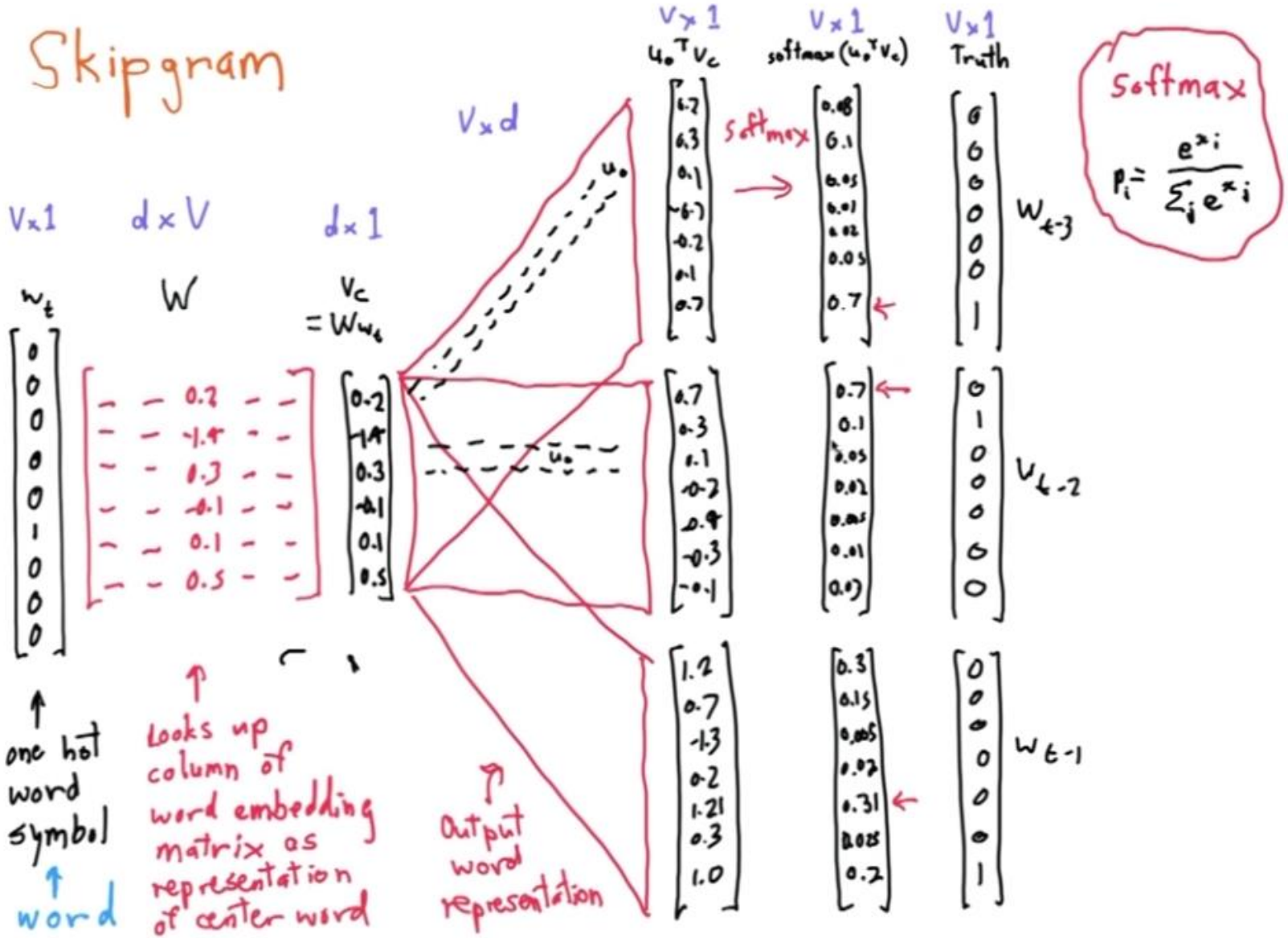
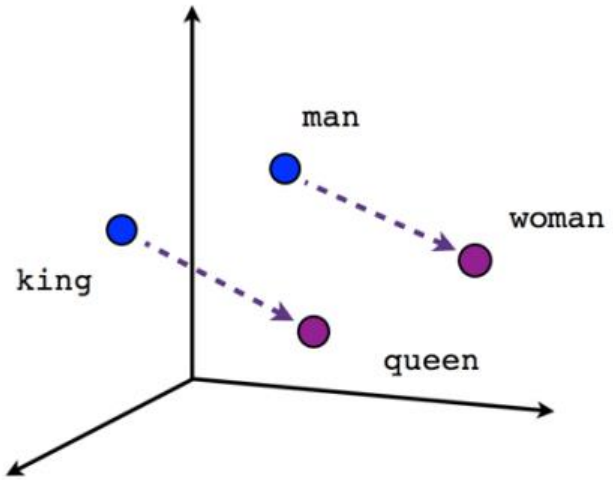
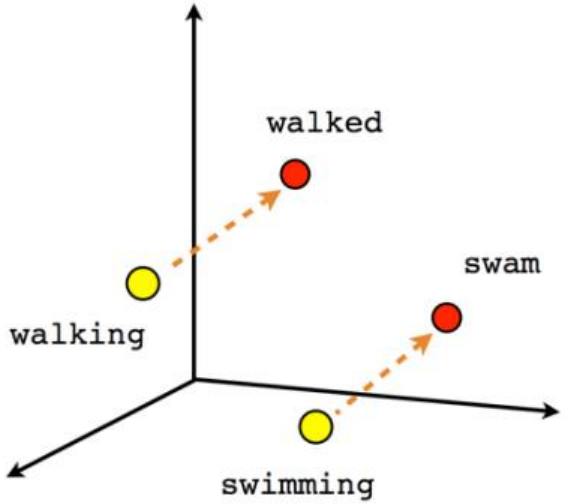


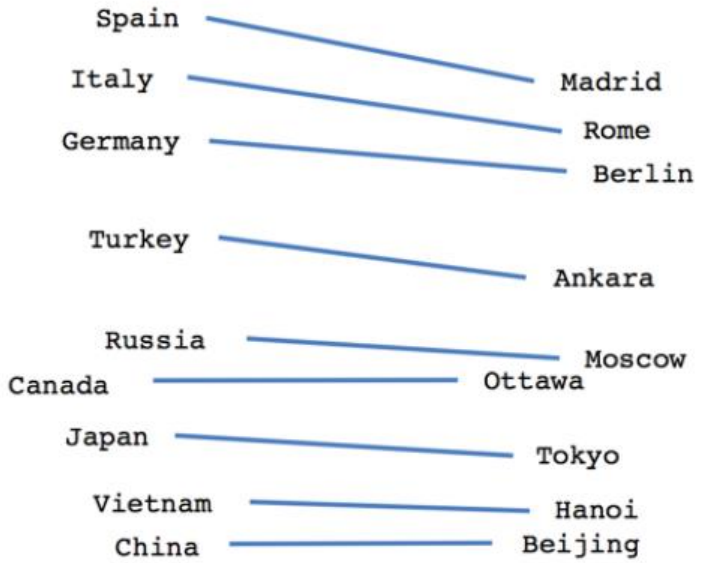
word2vec



Male-Female

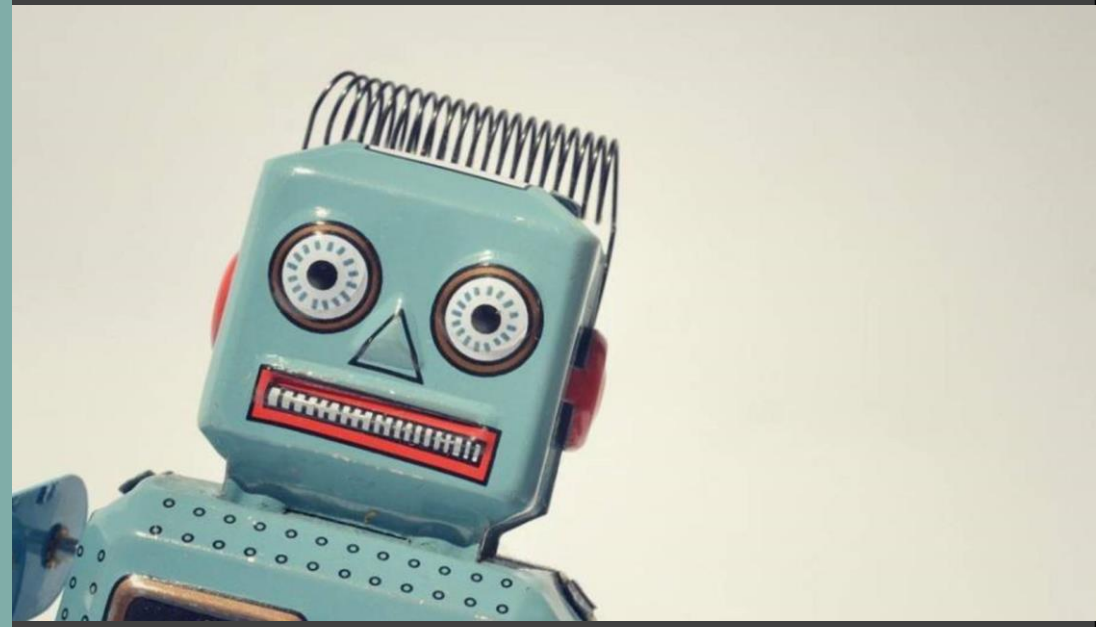


Verb tense



Country-Capital

Decidim Barcelona



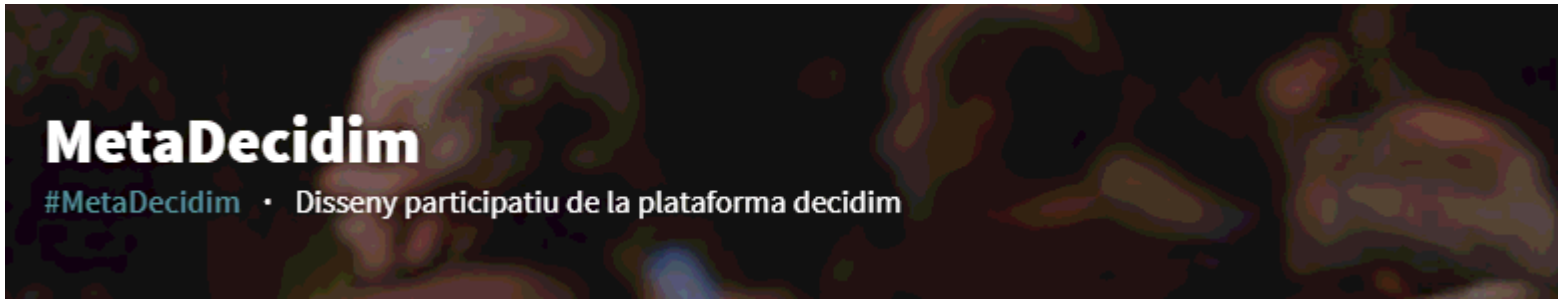
Decidim BCN


Decidim Barcelona
Buscar
Castellano
Regístrate · Entra
Inicio
Procesos
Órganos
Iniciativas
Rendición de cuentas
Ayuda
Decidamos la Barcelona que queremos
PROCESOS EN MARCHA
Bienvenido/a a la plataforma de participación de Barcelona. Construyamos una ciudad más abierta, transparente y colaborativa. Entra, participa y decide.

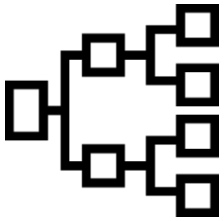
Decidim BCN

Concepts

PAM* Flow



1. Citizens, organizations and the city council write **proposals.**

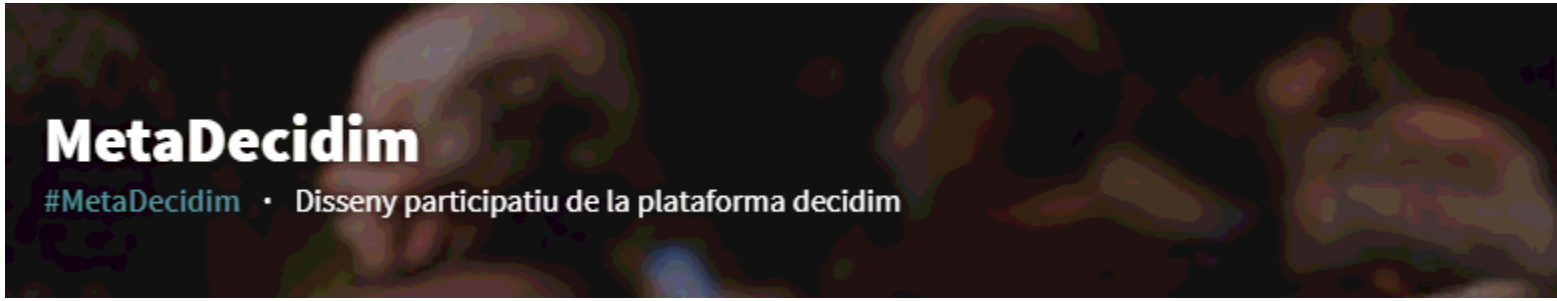


2. Similar proposals are clustered into **actions.**

3. Some **actions** are approved and they go on to be executed.

*PAM is a strategic plan that establishes the actions that the municipal government must implement during the corresponding political term.

Research questions



1. Do actions clearly **reflect** proposals ideas without considering authorship?

2. Do citizens **write** the same way as the Administration?

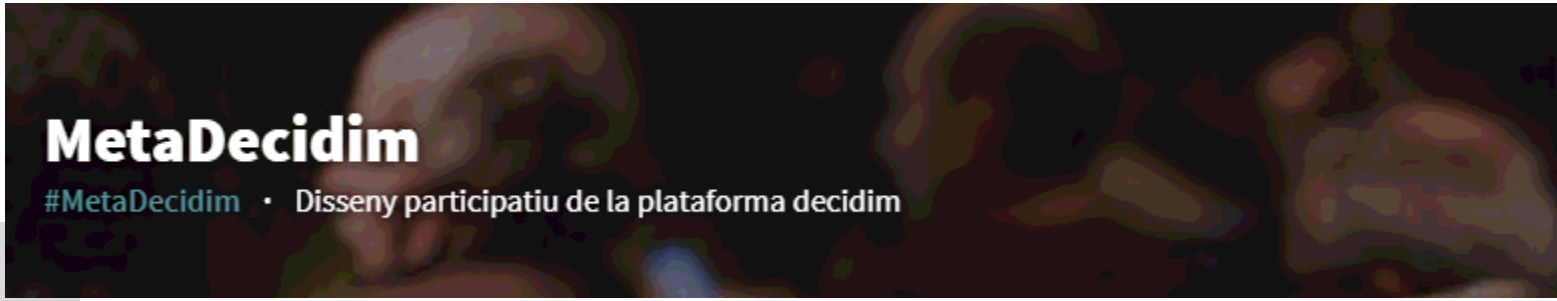
Can Machine Learning and Deep Learning answer those questions within a bilingual context?

Decidim BCN

Decidim BCN

First step

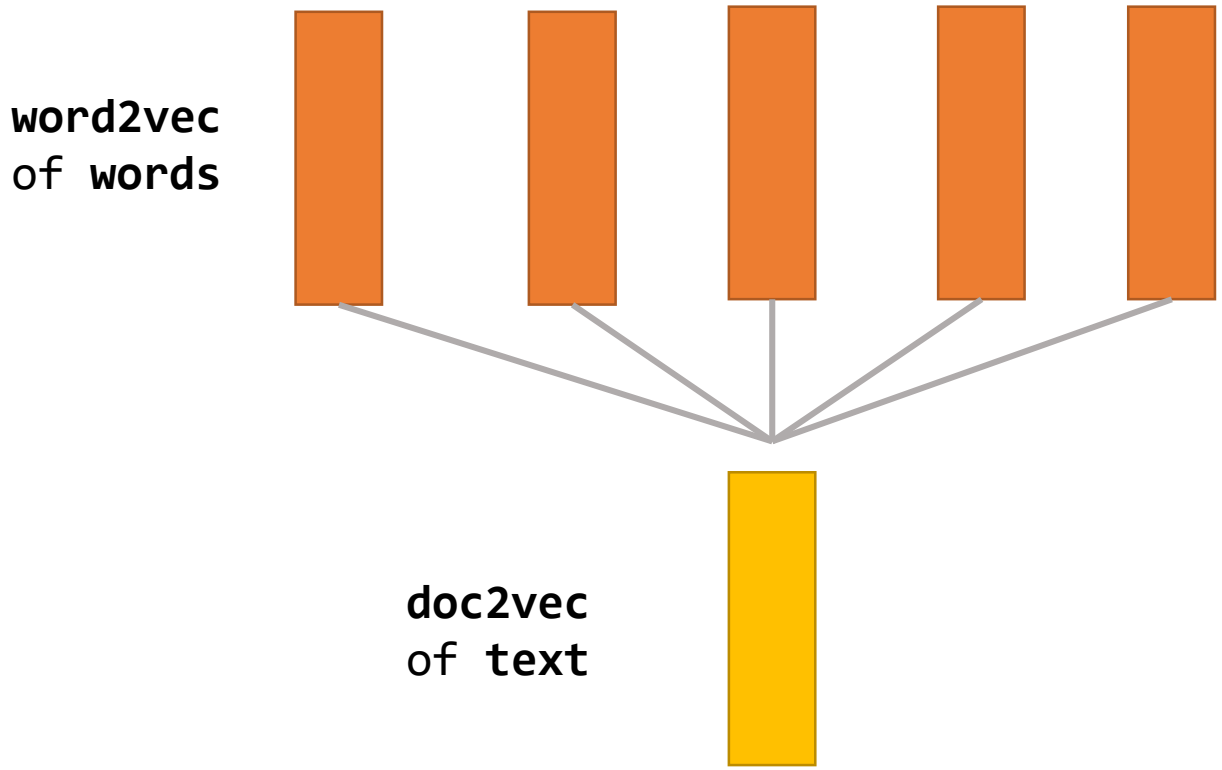
From text to word embeddings



doc2vec

Mean of all word2vec of words within a text.

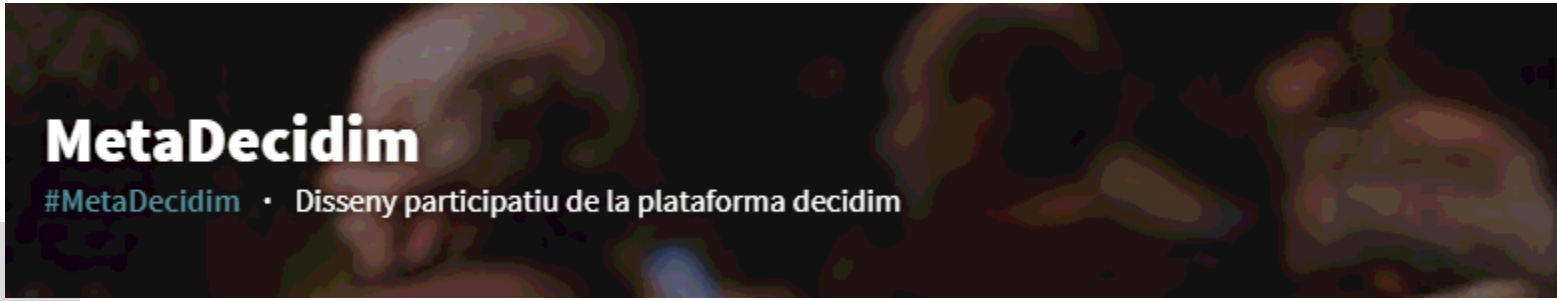
"Crear espacios nuevos para perros."



Decidim BCN

Second step

Analyze distances



When documents are represented as vectors, several distance functions can be used to reflect the degree of closeness between two of them.

Cosine distance. When documents are represented as vectors the correlation can be measured as the cosine of the angel between them. It is also one of the most popular distance metrics applied to text documents. Given two documents represented as u and v embeddings, their cosine distances is:

$$\cos_{\text{dist}}(u, v) = 1 - \frac{u \cdot v}{||u||_2||v||_2}$$

Euclidean distance. Euclidean distance quantifies the minimum distance among two vectors. The formal definition is:

$$\text{eucl}_{\text{dist}}(u, v) = \sum_i (u_i - v_i)^2$$

Manhattan distance. The Manhattan distance is defined as the distance between two objects measured along axes at right angles.

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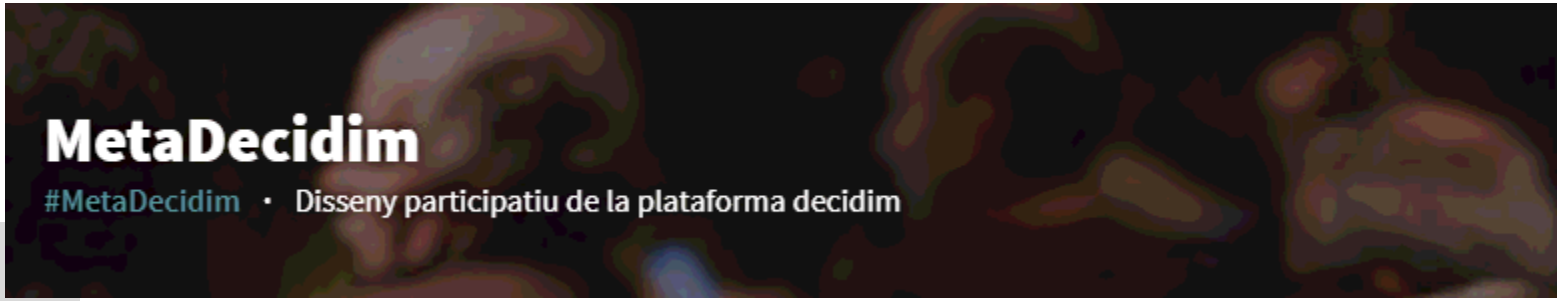

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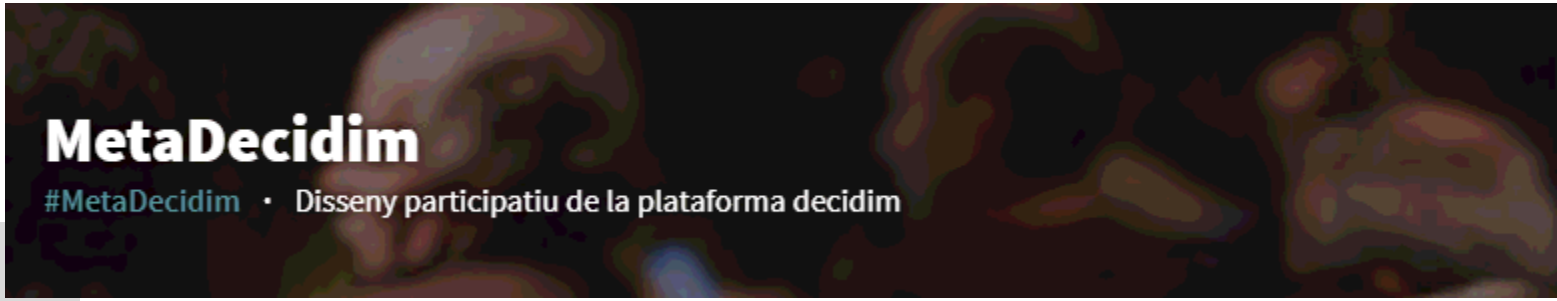

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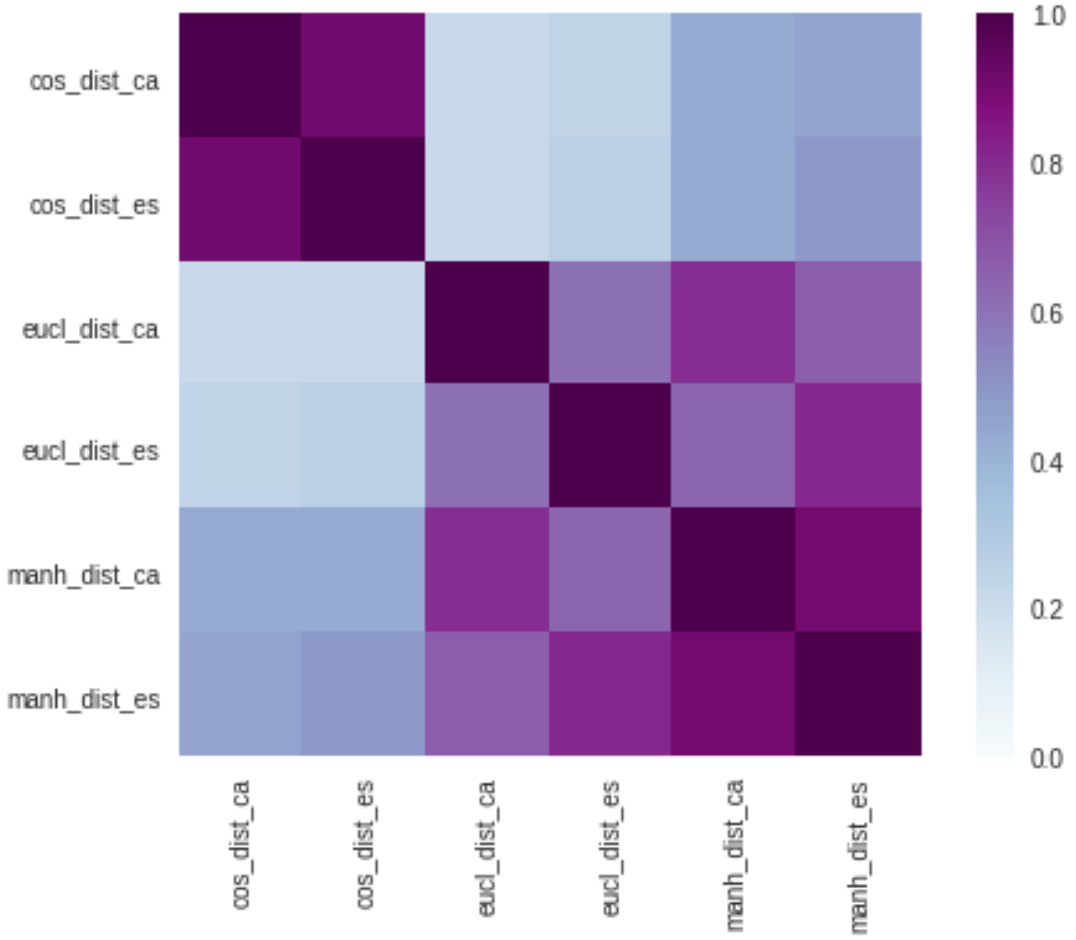

Decidim BCN

Second step

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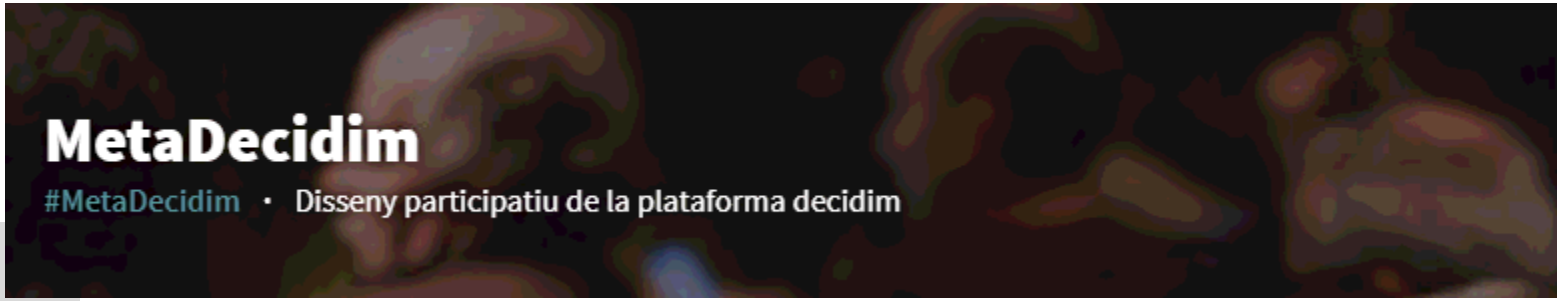
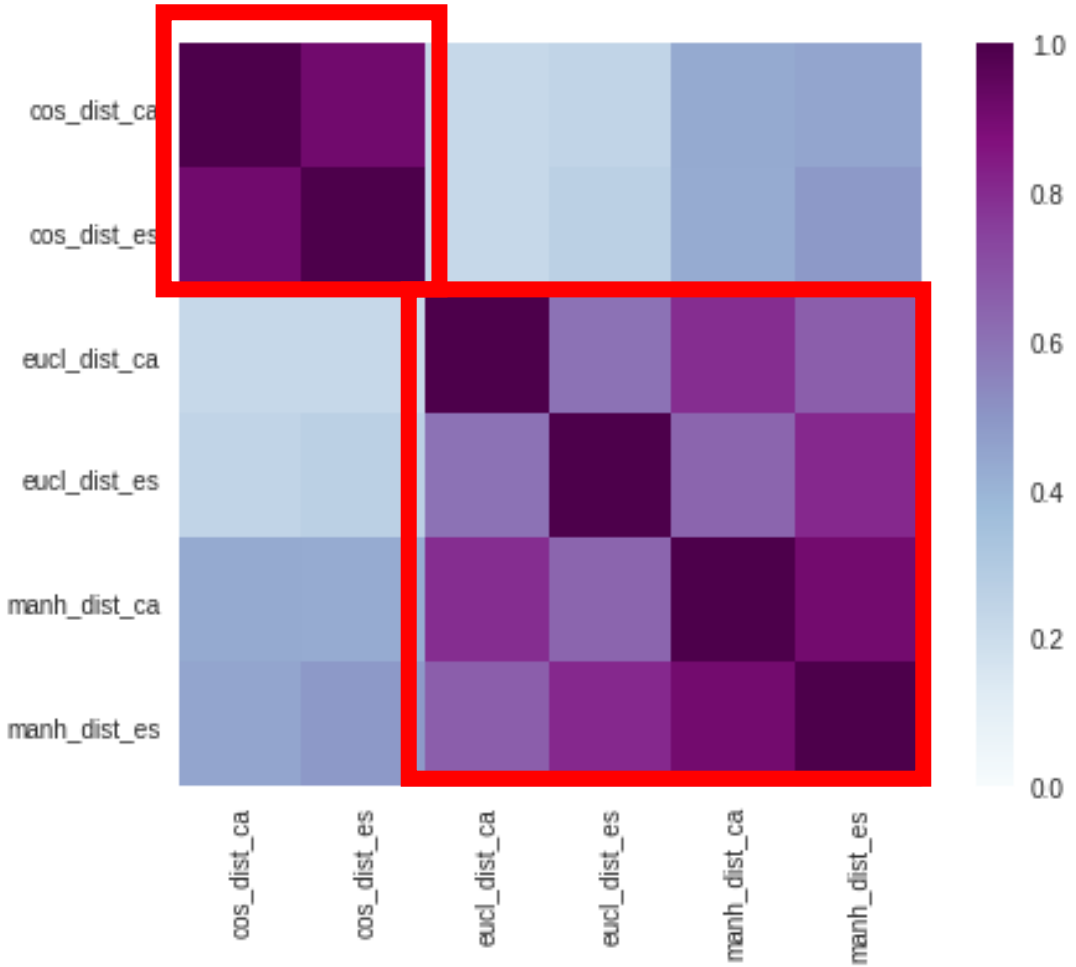
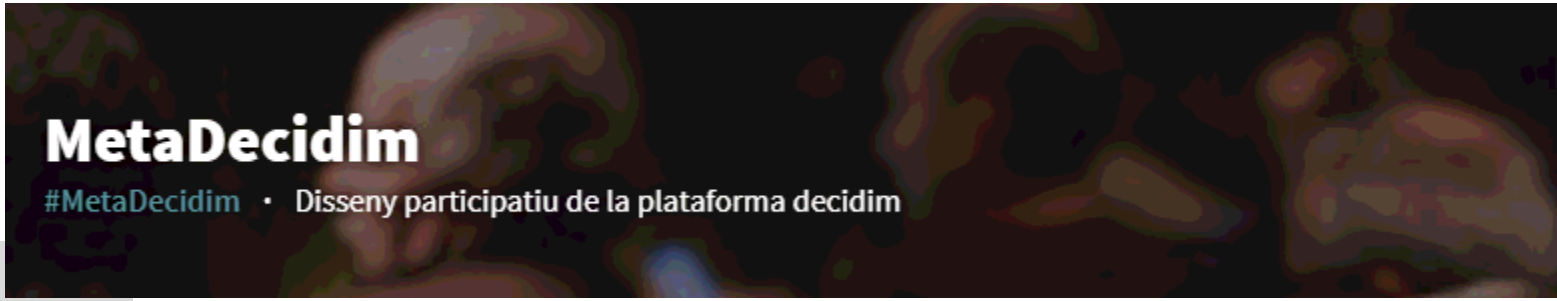


Same distance functions are highly correlated considering both languages, which implies that **doc2vec representations are equivalent** either in Catalan and Spanish



Decidim BCN

Third step



Bias!

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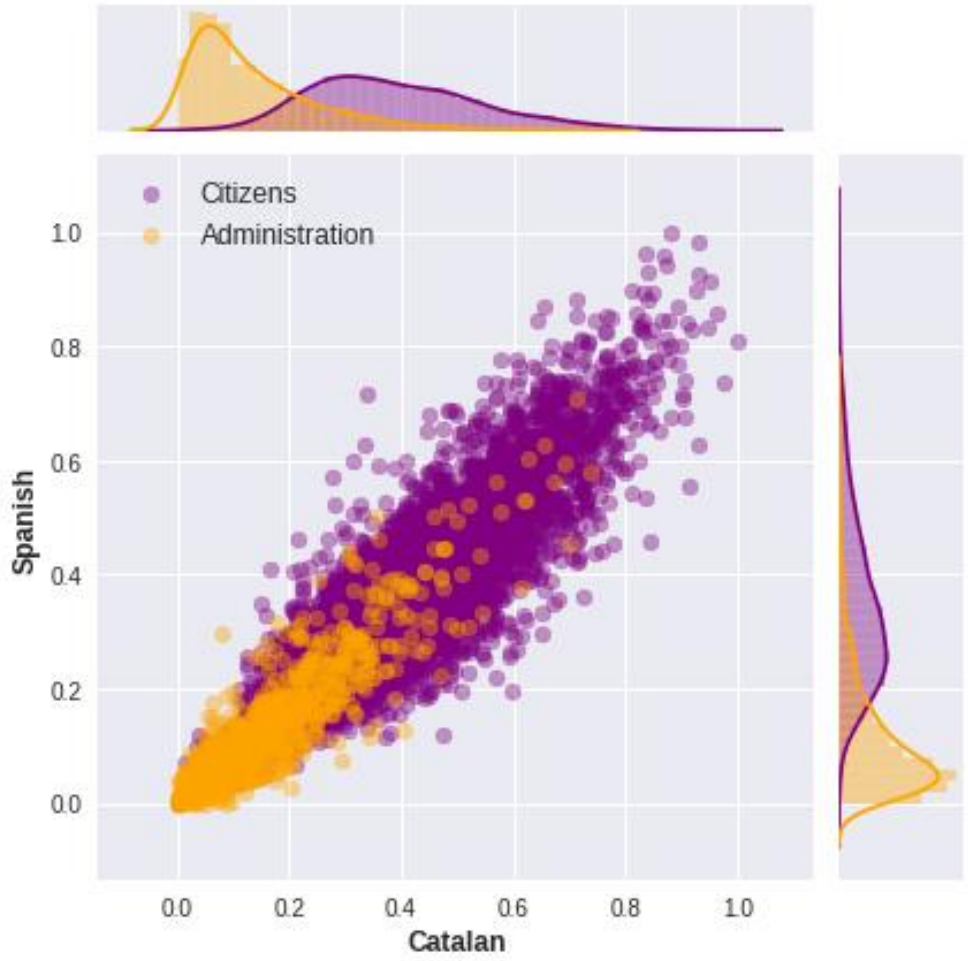

Decidim BCN

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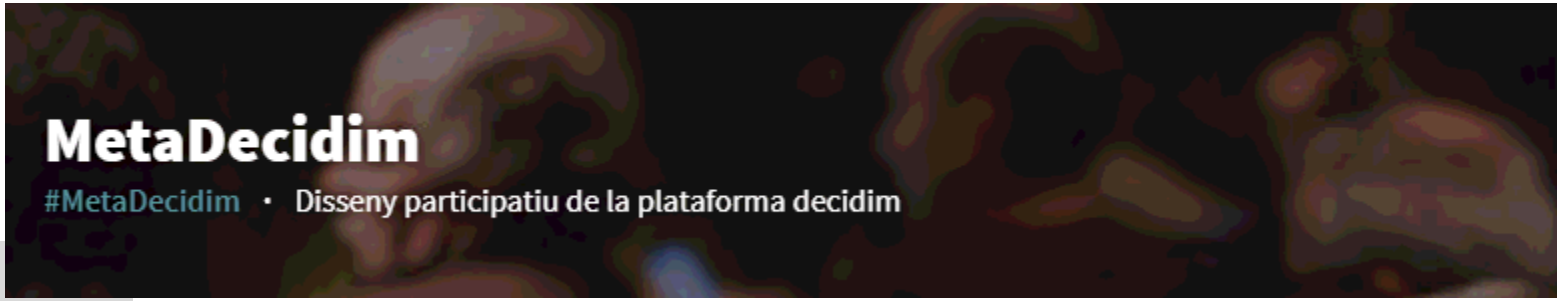

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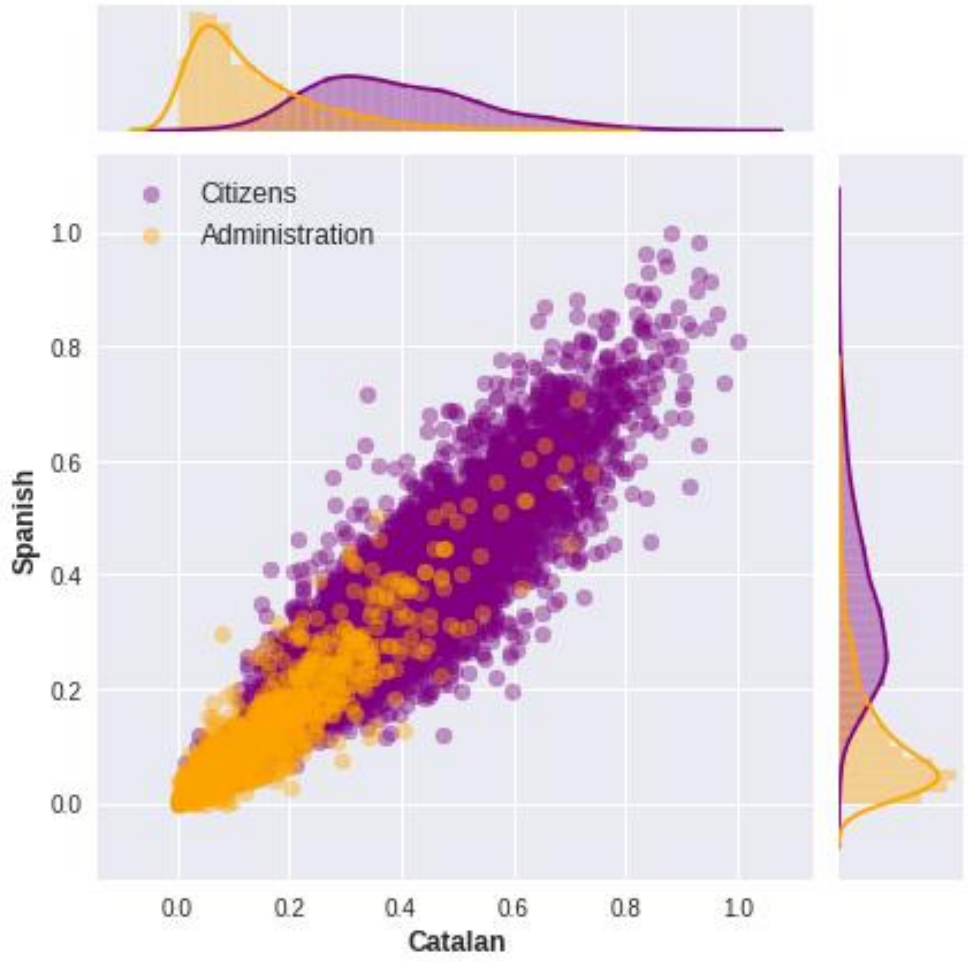
No!



Decidim BCN

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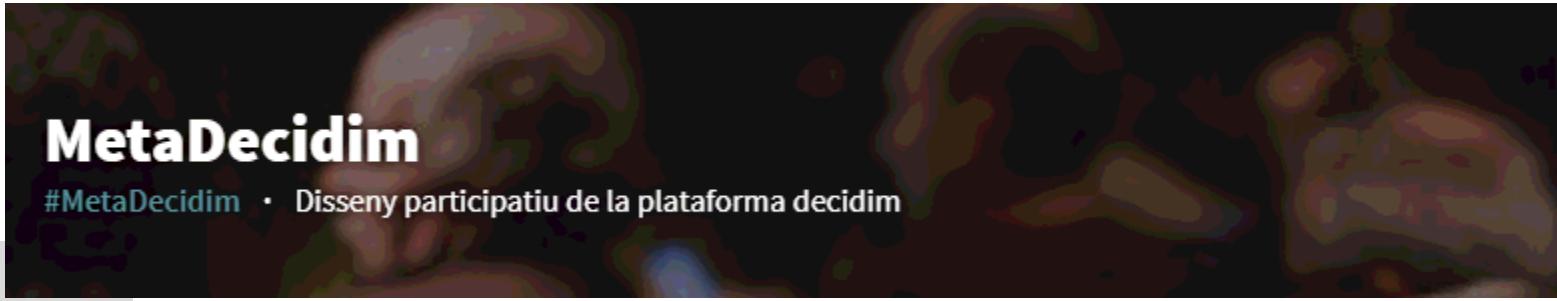

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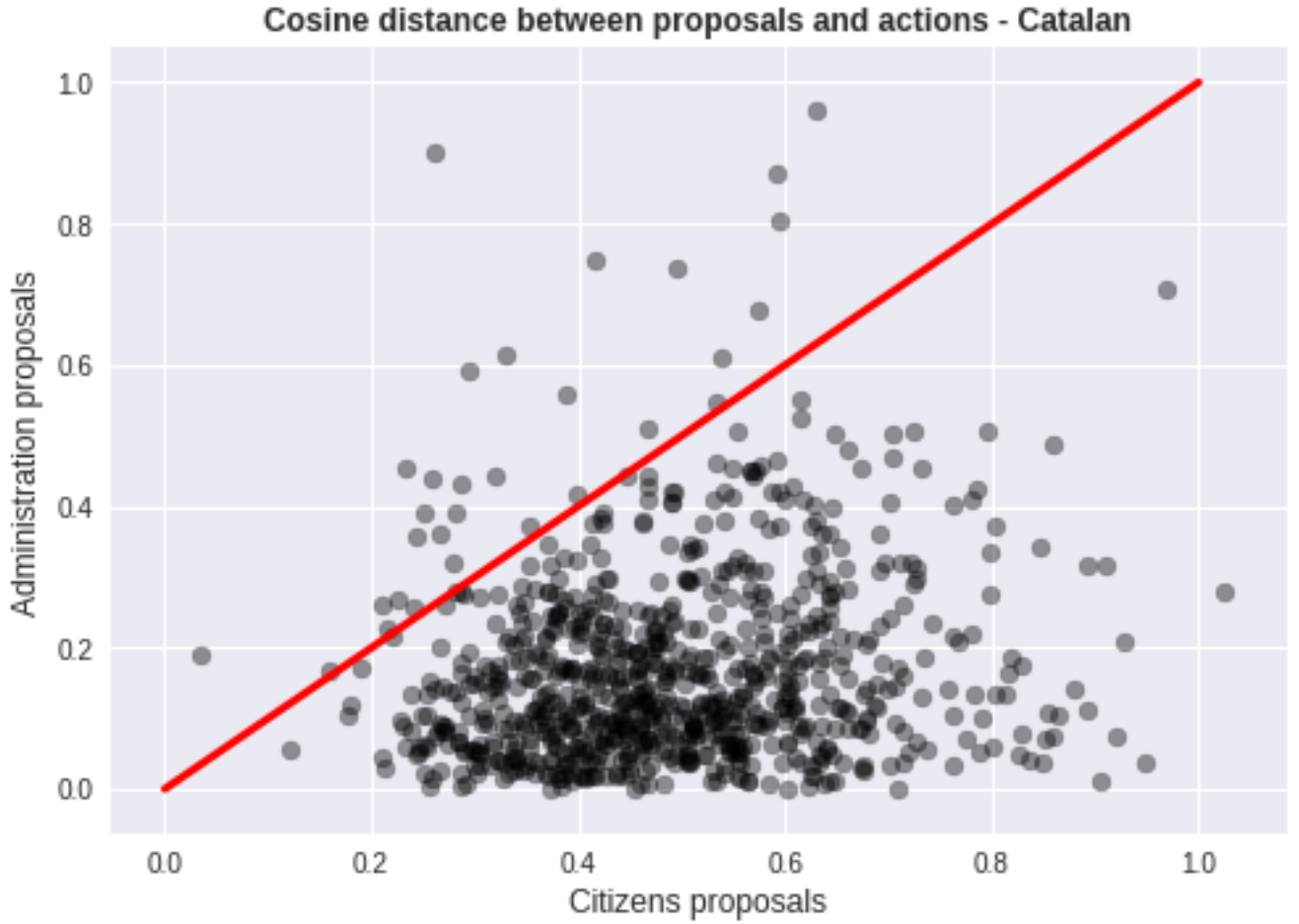

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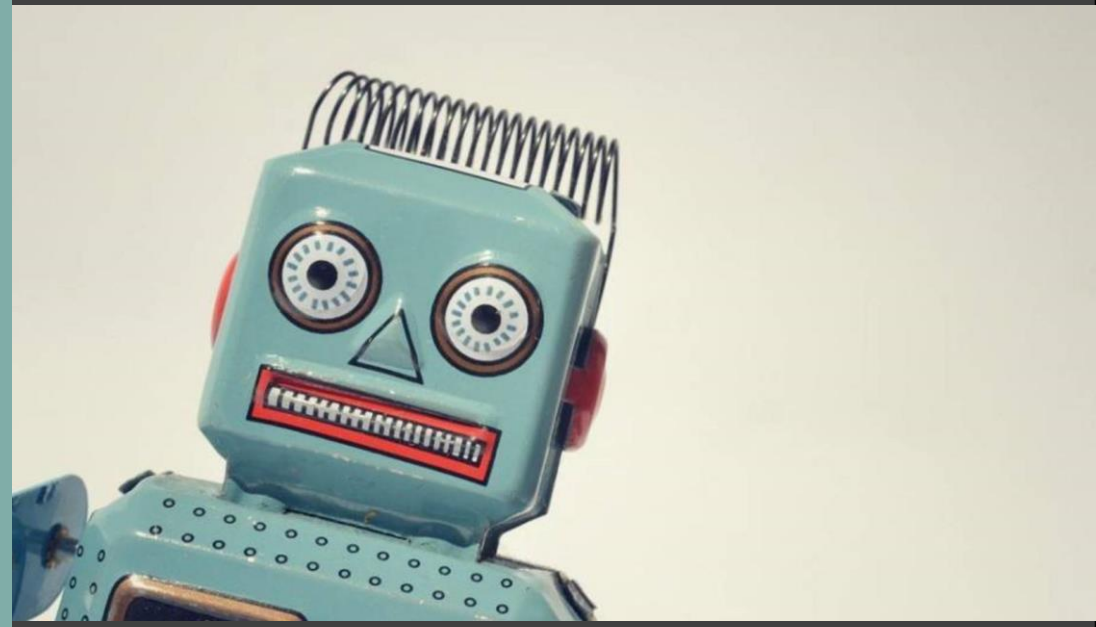

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- https://github.com/genimarca/caepia2018_tutorial_nlp_sa/blob/master/2018_caepia_tutorial_nlp.pdf