# Integration of heterogeneous datasets

Arthur Imbert, Inria, June 21, 2017

https://gitlab.inria.fr/parietal/arthur\_imbert/



# Introduction

# Problems related to data integration

### • Record linkage

### **Hospital A**





Patient Name: Vivian Christensen

Visit ID: 837720

Date of Birth: 3/20/1953

SSN: 000-86-6628

MRN: 9968427

Dx: 414.00

### **Hospital B**





Patient Name: Viv Christensen

Visit ID: 483005

Date of Birth: 3/20/1953

SSN: 000-68-6628 MRN: 0099523461

Dx: 493.01

# Problems related to data integration

### Record linkage

# Hospital A (B) Printing

Patient Name: Vivian Christensen

Visit ID: 837720

Date of Birth: 3/20/1953

SSN: 000-86-6628

MRN: 9968427 Dx: 414.00 **Hospital B** 





Patient Name: Viv Christensen

Visit ID: 483005

Date of Birth: 3/20/1953

SSN: 000-68-6628 MRN: 0099523461

Dx: 493.01

Foreign key discovery

# Problems related to data integration

### Record linkage

# 

Patient Name: Vivian Christensen

Visit ID: 837720

Date of Birth: 3/20/1953

SSN: 000-86-6628

MRN: 9968427

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Foreign key discovery

• Semantic analysis

# Hospital B





Patient Name: Viv Christensen

Visit ID: 483005

Date of Birth: 3/20/1953

SSN: 000-68-6628 MRN: 0099523461

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# What is Open data?

"Open data is the idea that some data should be **freely available** to everyone to use and republish as they wish, **without restrictions** from copyright, patents or other mechanisms of control"

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Volume and heterogeneity!

### Reuses

1711 reuses using one or several open datasets are listed

- Map of public services
- Visualization of Presidential election results
- Statistical analysis crossing socio-economic data
- ...

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### New task: Predict reuses

Closer datasets are in a file embedding, more relevant a reuse should be

# Build a corpus of datasets

55138 downloaded files (348GB)

23112 cleaned files (72GB)

# 'Tablizing' everything

### An example of file easy to clean

D	E	F	
ods_adresse	Code_postal	Libelle_commune	S
23 RUE DOMREMY	75013	Paris 13e Arrondissement	Si
SAINT VICTOR SUR LOIRE		Saint-Étienne	Si
LD LE GRAND HAMEAU	14520	Sainte-Honorine-des-Pertes	Ţŗ
8 Rue de Vaugirard	75006	Paris 6e Arrondissement	Si
Rue de la Mécanique		Louviers	Ţŗ
CELE	20127	Bollogordo	T

- a tabular form
- a first row as header
- consistency of the data below the header

### ... still with some limitations

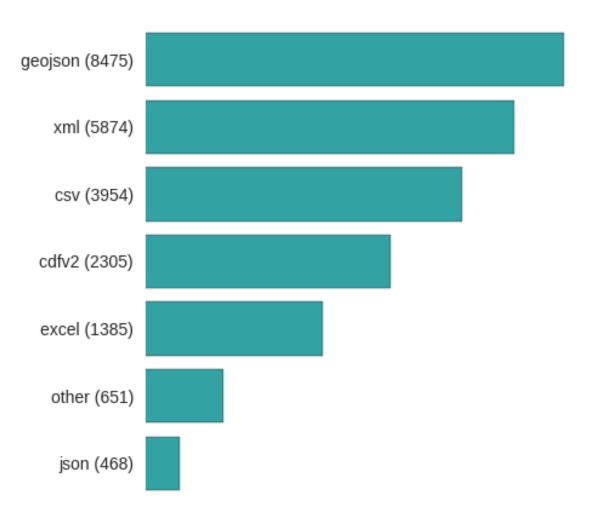
- no Uniforme Ressource Identifier for the adresses
- no Uniforme Ressource Identifier for the city names

# 'Tablizing' everything

### An example of 'dirty data'

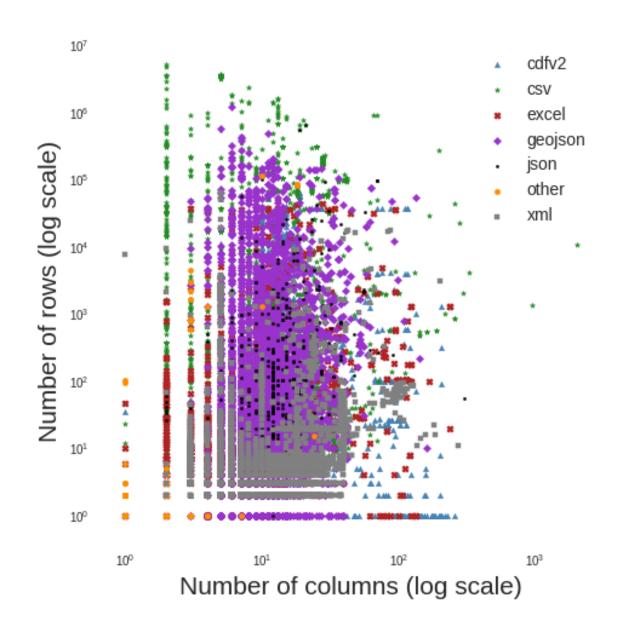
	A	В	C	D	E
	T79JNAIS : Répartition quotidienne des naissar	ces vivante	s		
	CHAMP : France métropolitaine, territoire au 31 décembre 201				
	,				
		JOUR"	01	02	03
	ANNÉE DE 1968 À 2013	MOIS •			
2		Septembre	1,844	1,741	2,237
		Octobre	2,393	2,471	2,327
3 4		Novembre	1,949	2,264	1,889
5		Décembre	1,948	1,768	2,197
6		Janvier	1,745	2,187	2,151
7		Février	2,173	1,879	1,778
8		Mars	2,222	1,872	1,712
9		Avril	1,741	2,129	2,167
0		Mai	1,890	2,281	2,291
1	2013	Juin	1,887	1,723	2,081
2	2013	Juillet	2,298	2,337	2,370
3		Août	2,441	2,426	2,008
4		Septembre	1,818	2,138	2,327
3 4 5 6		Octobre	2,428	2,385	2,333
6		Novembre	1,830	1,834	1,868
7		Décembre	1,871	2,220	2,284
8					
9	Source : Insee, statistiques de l'état civil				
0					
d					

# Extensions of the cleaned files



Number of files (log scale)

# Size of the cleaned files



# Build a file embedding

TF-IDF

Non-negative Matrix Factorization

Metric learning

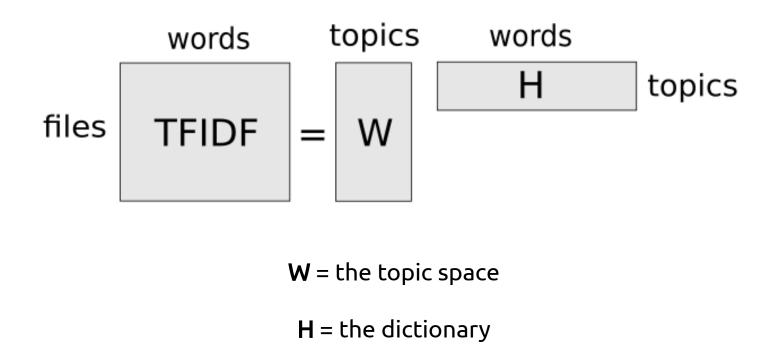
Cross-validation

# TF-IDF and NMF

Weight each word frequency by its inverse document frequency

# TF-IDF and NMF

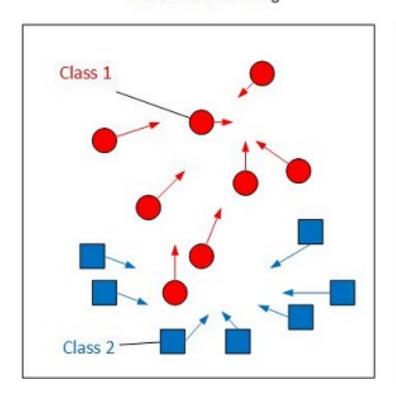
Weight each word frequency by its inverse document frequency



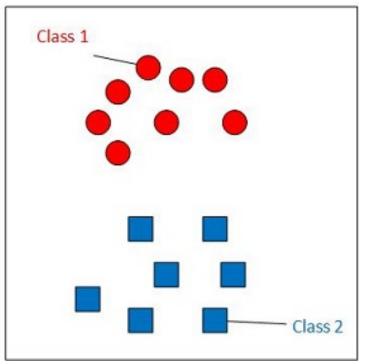
# Metric learning

X and y built from W

Before Metric Learning



After Metric Learning

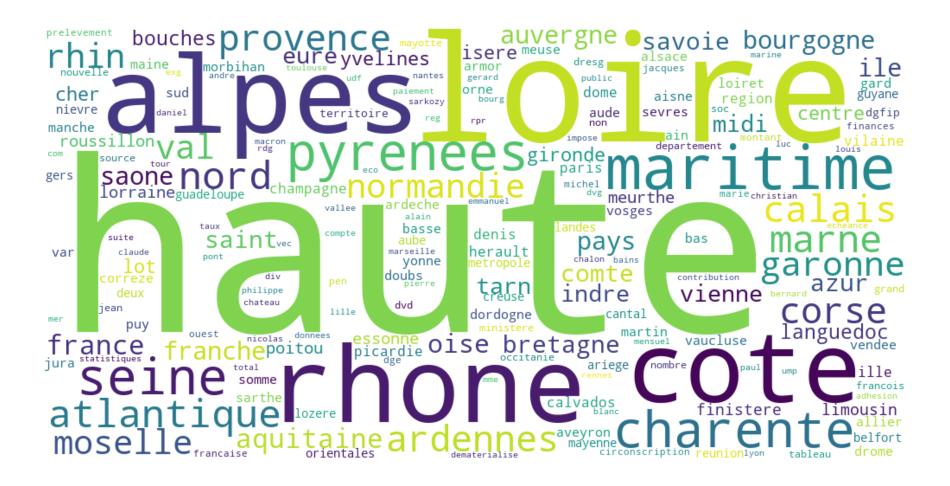


# Cross-validation results

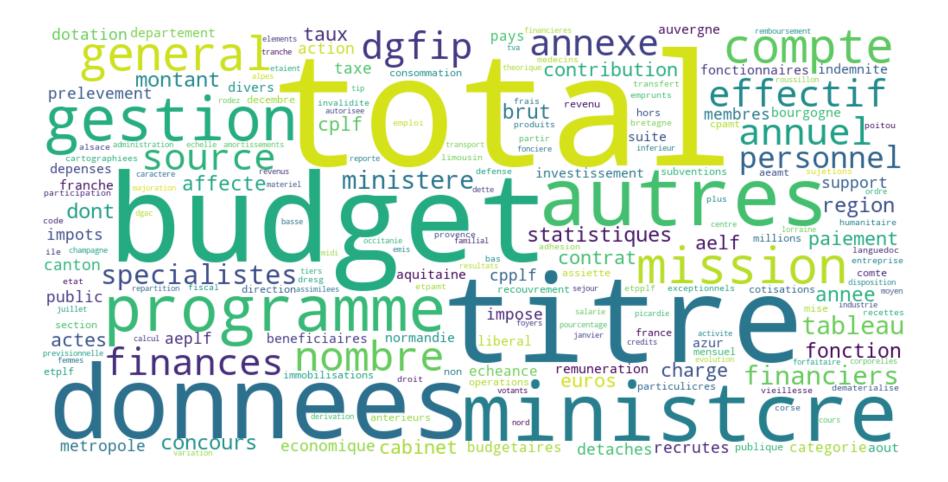
# Results

# Cosine distance in the embedding

# Region topic



# Budget topic



# Criminality topic



# Limits

- Heterogeneous data and cleaning failures
- Not enough reuses
- Undersampling of non reused pairs
- Reuse = similarity?

## Future work

- Use geopandas to infer geographical information from GEOJSON
- Infer temporal information
- Perform foreign key discovery and record linkage over the closest files and merge them
- Share the results with <a href="https://www.data.gouv.fr/fr/">https://www.data.gouv.fr/fr/</a>

