

Tagging Text in Money Transfers: A Use-Case of Spark in Banking

Luis Peinado Fuentes, BBVA Data & Analytics Jose A. Rodriguez Serrano, BBVA Data & Analytics

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#EUds7 @BBVAData

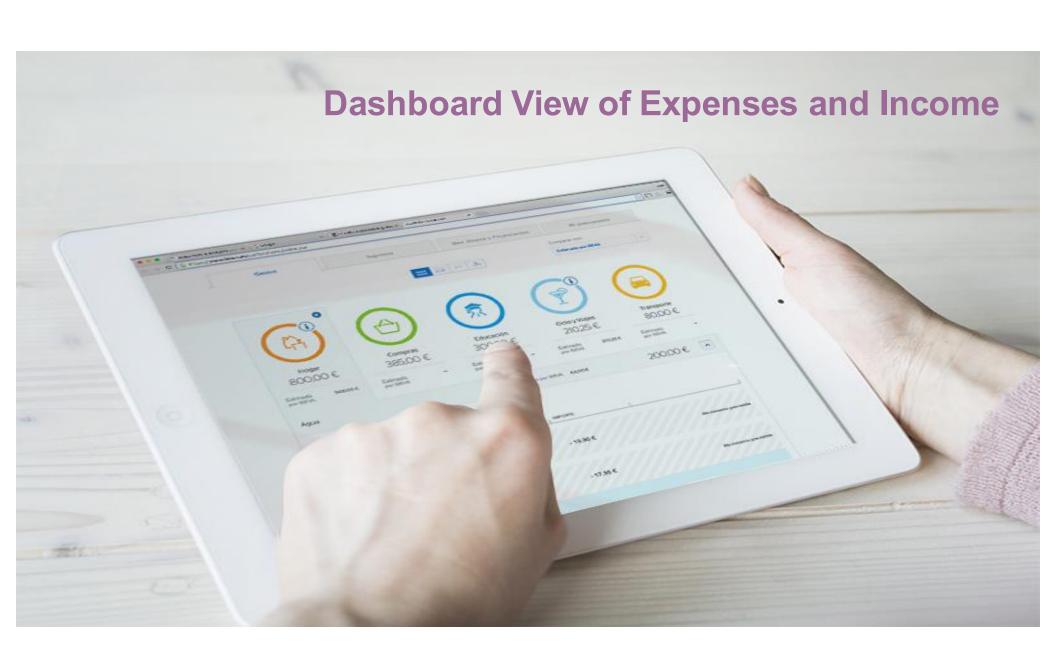
DATA & ANALYTICS

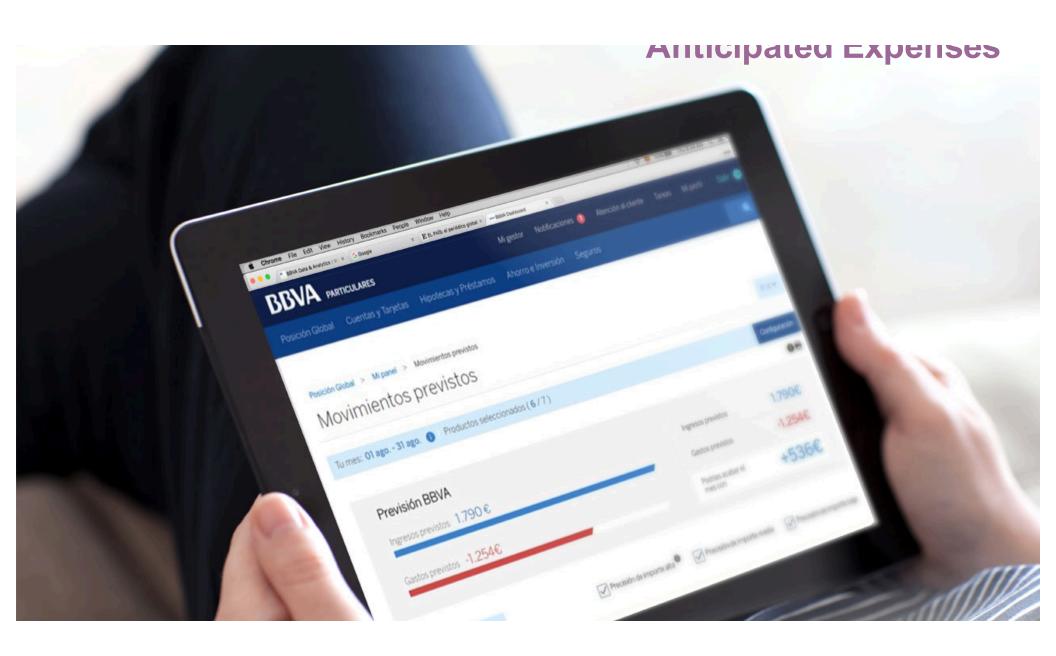
Our Goal:

Improving User Experience in Retail Banking Products with Machine Learning

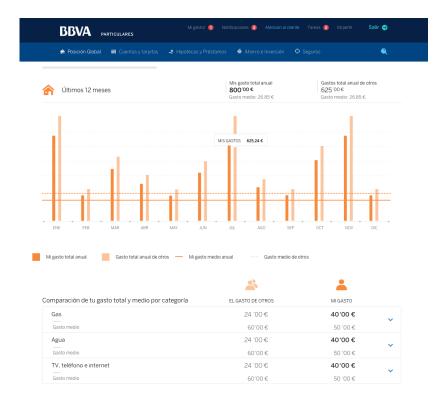
(Using Spark)







Customer Comparison Engine

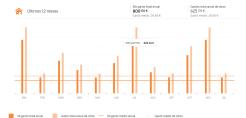




All these examples require categorized bank transactions



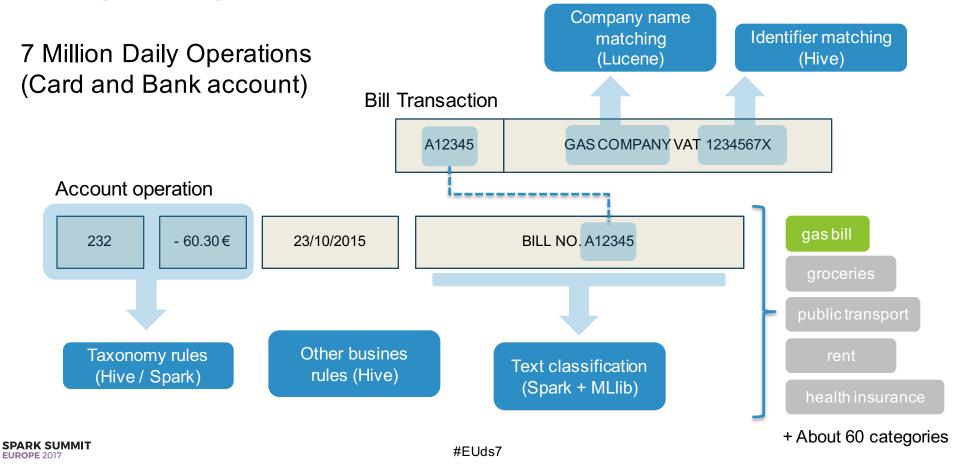




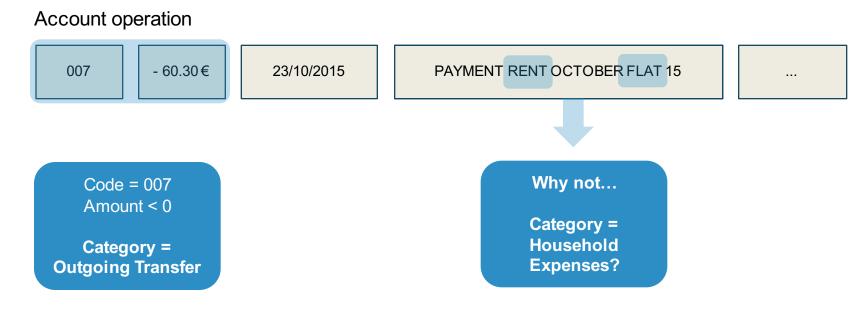
| | 2 | <u> </u> | |
|---|-------------------|----------|---|
| Comparación de tu gasto total y medio por categoría | EL GASTO DE OTROS | MI GASTO | |
| Gas | 24 '00 € | 40'00 € | ~ |
| Gasto medio | 60'00 € | 50 '00 € | |
| Agua | 24 '00 € | 40'00 € | ~ |
| Gasto medio | 60.00 € | 50 '00 € | |
| TV, teléfono e internet | 24 '00 € | 40'00 € | ~ |
| Gasto medio | 60'00 € | 50 '00 € | |



Categorizing Bank Transactions



The Problem: Categorizing Transfers



700,000 Daily Operations (Transfers)

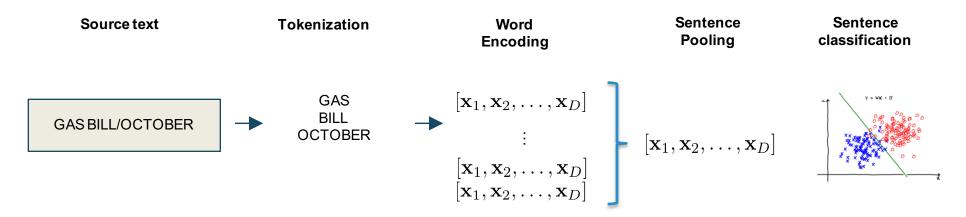


Data Science Challenges

- We did not know the data source in advance
- We did not have a labeled set
- A fraction of texts is useless ("detection" rather than classification)
- Distribution of categories is imbalanced
- Prefer false negatives over false positives
- Very short text, language not even syntactically correct



Basic Pipeline (in Pro since 2016)

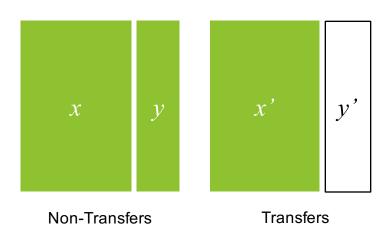


- First implementation: TF-IDF features + linear classifier (98% precision, 21% recall
- Further tests with word2vec + Vector of Locally Aggregated Descriptors (VLAD)
- Implemented in Spark/Scala, using MLlib classes
- · Own classes implemented for Multi-class Logistic Regression, VLAD
- Scala dependency injection useful to quickly setup variants of the above steps



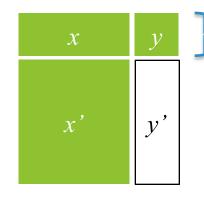
Where Do We Get a Dataset From?

First Attempt: Domain adaptation



But: $p(x, y) \neq p(x', y')$

Second Attempt: "Assisted Dataset creation"



Transfers

$$p(x, y) = p(x', y')$$

Assisted annotation with tagging interface.

Tags suggested by rules, incomplete training, similar-string matches, and confirmed manually.

> 45K Transfers Tagged



Exploring better embeddings/poolings: w2v + VLAD

Word2vec "Synonyms"

val word2vec = new Word2Vec()
val model = word2vec.fit(input)
val synonyms =
model.findSynonyms("alquiler", 10)

viaje → billete vuelo hotel reserva billetes gastos boda regalo casa

alquiler → piso cdad alq comunidad local garaje mes prop calle

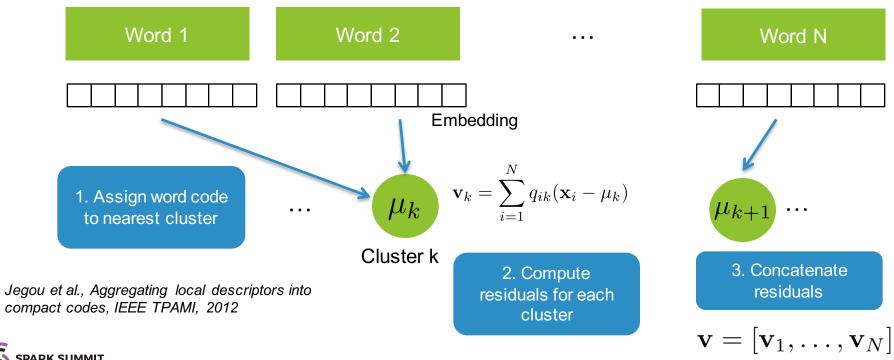
pago

factura fra fras fact 14 resto fac 50 facturas



Vector of Locally Aggregated Descriptors (VLAD)

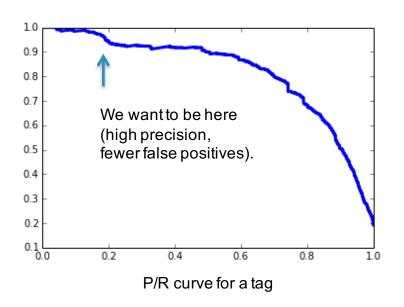
Own implementation in Spark/Scala





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Results of the Experiment



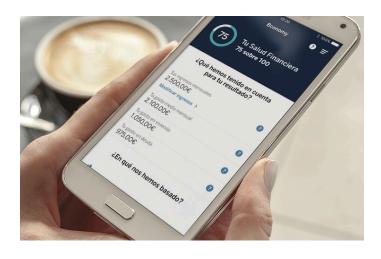
| Method | Recall @ Prec=98% |
|--------------------------------|-------------------|
| TF-IDF | 21.0 % |
| Word2vec (avg pool) | 24.5 % |
| Word2vec (avg pool)+ Amount | 26.9 % |
| Word2vec (VLAD pool) | 37.3 % |

- The combination of w2v + VLAD is not in production
- Currently working on own multi-word embedding which yields similar results

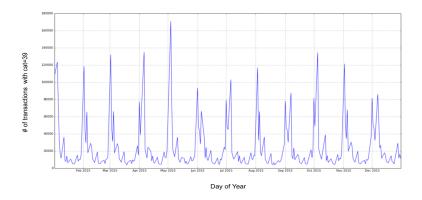


Enablers

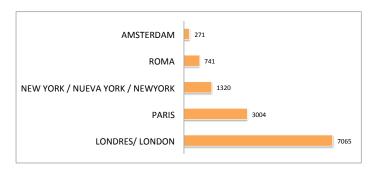
Bosonomy – Identifying Rental Expenses



When do household expenses happen?



Aggregated statistics about trip destinations





Other Challenges

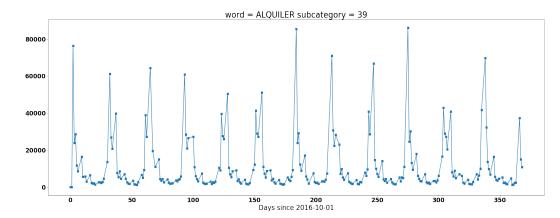
The system goes to the wild...

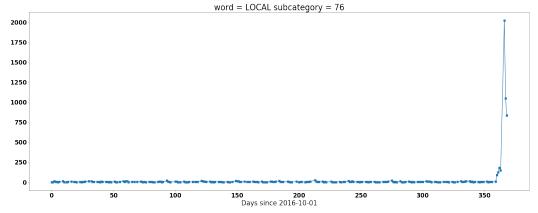
 Failure cases stay within the tolerated 2%

Open system

- Word black-list needed
- Proposal for daily quality checks (plots: daily evolution of number of word-tag pairs)

M. Zinkevich, Rules of ML: Best Practices for ML Engineering









Conclusions

How Spark Helped Us

- Spark (+MLlib) was crucial in this use case.
- We complemented with own implementations of Multi-Class LR and Vector of Locally aggregated descriptors.
- We took advantage of Scala's code injection mechanisms to frame the classification process as a set of standardized steps.

Data Science Experience

- Running text classifier in a real production system
- Improving the standard components through experiments on word2vec and VLAD
- Ongoing work: own embedding method







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Thanks to:

- Roberto Maestre and Advisory Team @ BBVA D&A for insightful comments
- Everyone at BBVA who made using Spark possible and disseminated Spark and Scala knowledge



DATA & ANALYTICS

(Just in case)

Appendix



Exploring better embeddings/poolings: w2v + VLAD

Word2vec "Synonyms"

```
val word2vec = new Word2Vec()
val model = word2vec.fit(input)
val synonyms =
model.findSynonyms("alquiler", 10)
```

```
viaje
        → billete vuelo hotel reserva billetes gastos boda regalo casa
alquiler →
            piso cdad alg comunidad local garaje mes prop calle
jose
            antonio juan manuel francisco luis maria miquel carmen jesus
            pis pagament quota comunitat jordi josep despeses parking joan
lloquer
            2013 06 07 08 04 05 02 09 03
2014
madrid
            viaje hotel barcelona sevilla malaga la san sl reserva
            factura fra fras fact 14 resto fac 50 facturas
pago
             febrero marzo abril mayo septiembre noviembre junio agosto
enero
```



References

Scientific literature

- Jegou et al., Aggregating local descriptors into compact codes, IEEE Trans. on Pattern Analysis and Machine Intelligence, 2012.
- Joachims, Text Categorization with Support Vector Machines: Learning with Many Relevant Features, ECML 1998.
- Mikolov et al., Distributed representation of words and phrases and their compositionality, NIPS 2013.
- Clinchant and Perronnin, Aggregating Continuous Word Embeddings for Information Retrieval, Workshop on Continuous Vector Space Models and Their Compositionality, 2013.

Docs and posts

- Feature Extraction and Transformation using the RDD API
- Word Embeddings in 2017: Trends and Future Directions
- M. Zinkevich, Rules of ML: Best Practices for ML Engineering

