









MAC: An open and free Moroccan Arabic corpus for sentiment analysis

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Motivation

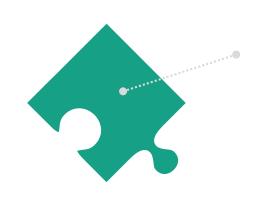
With the advent of the web 2.0 and the explosion of data sources such as review platforms, blogs and microblogs, there has been a need to analyze millions of posts, tweets or reviews in order to find out what internet users think.





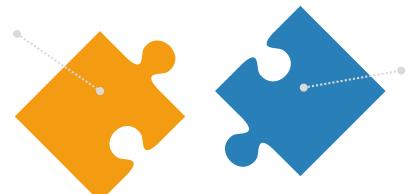
Motivation

The number of active social media users in Morocco has increased by **4M**¹ users over the past year, reaching the number of **22 million** social media users.



1- The research carried out on the analysis of the sentiment of tweets in Arabic is very limited, in particular Moroccan Arabic compared to other languages.

3- Morocco is thus ranked 9th among Arab countries with the highest number of users. .



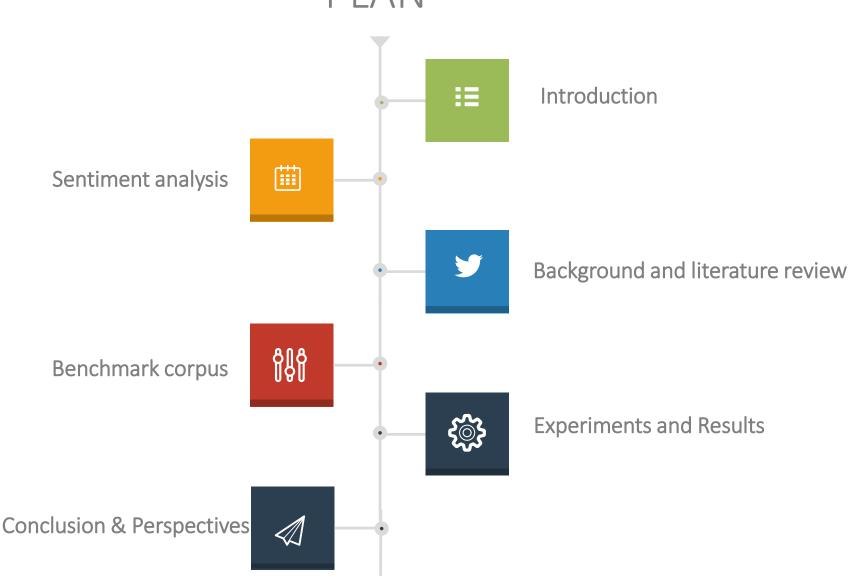
2- The total lack of additional resources for Moroccan Arabic.



¹ https://www.statista.com/statistics/1172771/number-of-social-media-users-morocco/









Introduction



Social media

Facebook, Twitter, Instagram, LinkedIn, these social platforms are now part of everyday life. The data aspect of these social media, such as Twitter messages, generates a rich wealth of data about who is involved in communication.

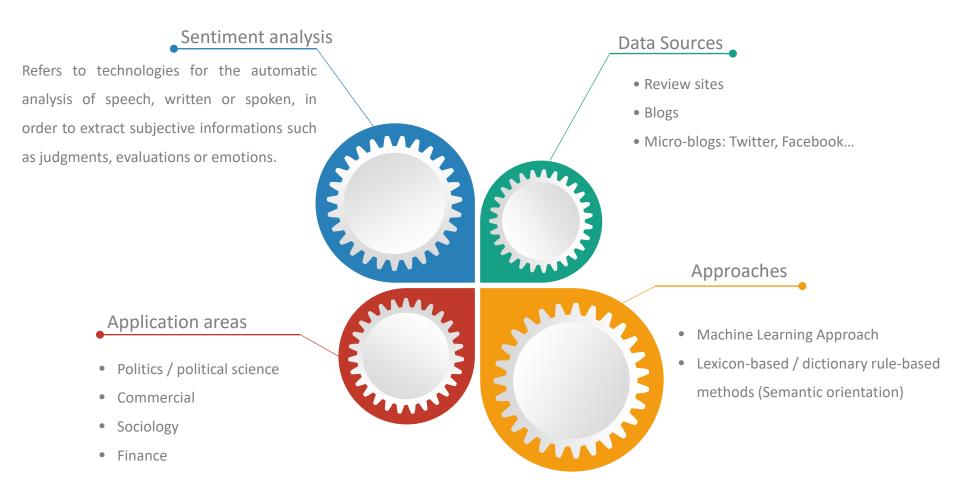


This data plays an important role in decision making for many people and organizations.





Sentiment Analysis







State of art

Maghrebian corpora

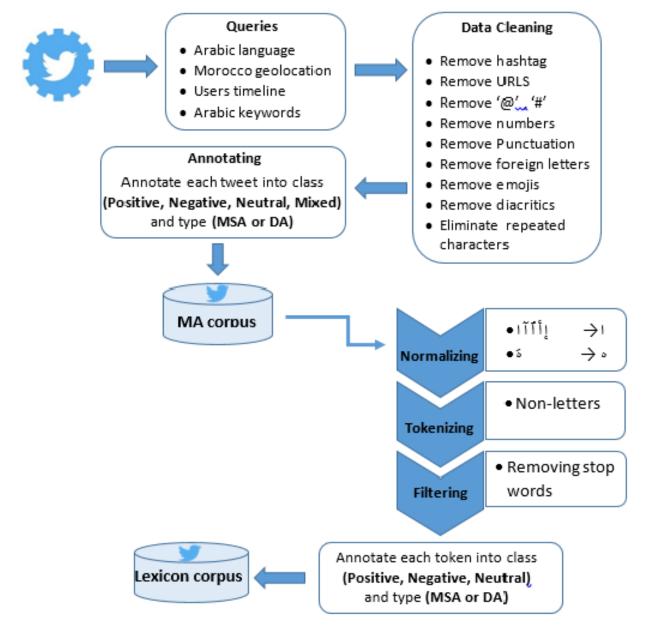
Dataset	Size	Arabic	Classes	Source	Year	Publicly Available
[15]	10006	Egyptian	4	Twitter	2015	√
[13]	6m	Tunisian	2	Twitter	2017	√
[1]	49864	Algerian	2	Facebook	2019	✓
[3]	930	Moroccan	2	Twitter	2017	×
[9]	10254	Moroccan	2	Facebook	2017	×
[10]	2000	Moroccan	2	Twitter	2019	✓
[4]	12K	Moroccan	4	Twitter	2020	×

Table 1: Arabic corpora for sentiment analysis.





Benchmark corpus



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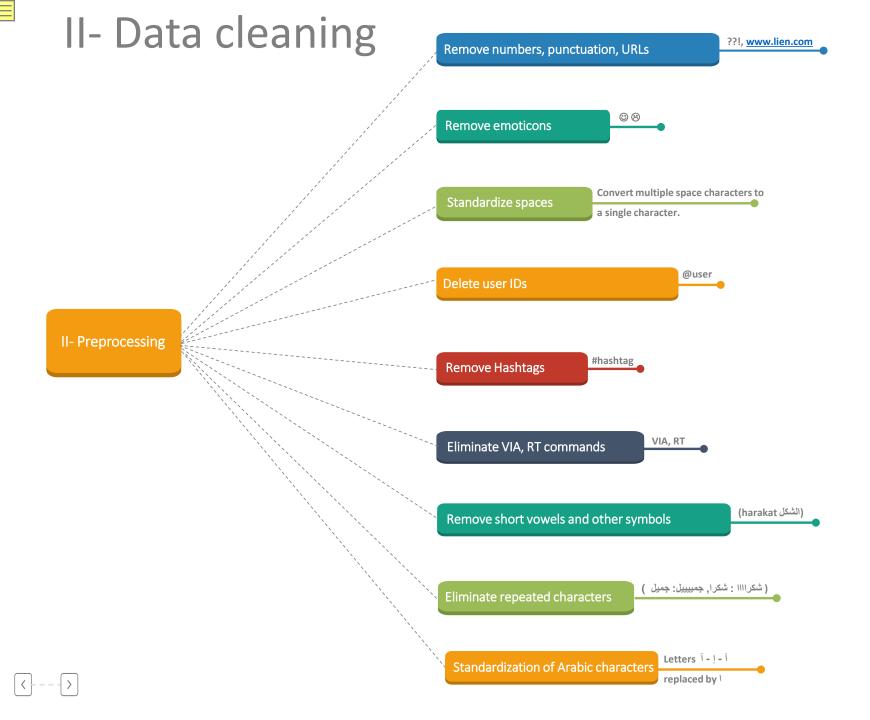
I- Data collection

Final corpus

• The corpus consists of the total of 18000 valid tweets based on 36,114 tweets collected

Number of tweets collected	18000	
Number of valid tweets	8360	
Number of retweets	9640	

Table 2: Statistics on the final corpus.





III- Annotation

- The corpus was labeled by ourselves, our task is to determine the polarity (Positive, Negative, Neutral, Mixed) and the language of the tweets (SA or MD).
- The annotation was done through a web application

Tweet	Туре	Class
Ar : توقع الخير و افتح صباحك بالتفاؤل و الأمل صباح النور En: Expect the good things and start your day with optimism and hope	Positive	SA
Ar : من المؤسف ان هذا حالنا الذي نعيشه الآن En: Unfortunately, this is our current situation	Negative	SA
Ar : تابعیني باش نقدر ندخلك En: Subscribe so that I can add you	Neutral	MD
رغم الصعوبات لي قاتلاني والمشاكل لي كنمر منها كنحاول نضحك ونقول الحمد لله: Ar En: Despite the difficulties and problems I have I try to laugh and thank God	Mixed	MD

Table 3: Example of annotated tweets



III- Annotation

The distribution of data according to their class and sentiment is shown in the following table:

SA	MD	Total
9 640	8360	18000

Table 4: Statistics on the corpus.

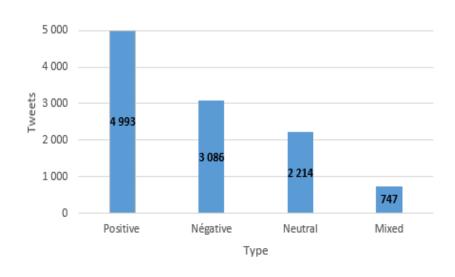


Figure 1: Distribution of feelings expressed in the SA corpus.

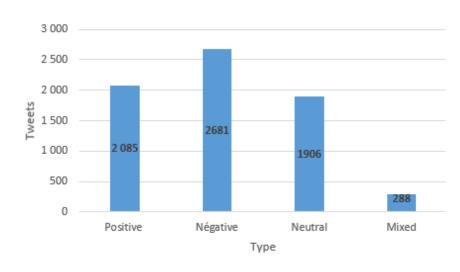


Figure 2: Distribution of feelings expressed in the MD corpus.



Lexicon construction

Statistics on the built dictionary:

Positif	Négatif	Neutre	Total
2 630	2 057	13 995	18 683

Table 5: Lexicon extracted from the SA database.

Positif	Négatif	Neutre	Total
1 291	702	8 902	10 895

Table 6: Lexicon extracted from the MD database.



V- Classification

Classifiers used

- 1. Convolutional Neural Networks (CNN)
- 2. Short-term long-term memory networks (LSTMs)
- 3. Support Vector Machine (SVM)
- 4. Logistic regression (LR)



Analysis evaluation

Task 1: Language identification

Model	Features	Stop words	Accuracy
LSTM	Word embeddings	0	91.27
		1	89.23
0.11	Word embeddings	0	89.78
CNN		1	89.16
6) (0.4	TF-IDF	0	89.13
SVM		1	86.30
La sistia Da sussaisus	TF-IDF	0	88.64
Logistic Regression		1	87.08

Table 6: Results of identification of the used language.





Analysis evaluation

Task 2 : Sentiment analysis

Model	Approach	Accuracy			
		AS	DM	AS_DM	
CNN	Corpus	91.78	84.17	90.87	
	Lexicon	90.85	85.42	89.25	
LSTM	Corpus	92.09	83.36	93.24	
	Lexicon	90.88	84.53	89.62	
SVM	Corpus	84.75	67.80	88.05	
	Lexicon	82.04	74.14	78.11	
Logistic Regression	Corpus	82.23	65.78	79.88	
	Lexicon	81.08	71.77	77.96	

Table 7: Evaluation results of the second task.



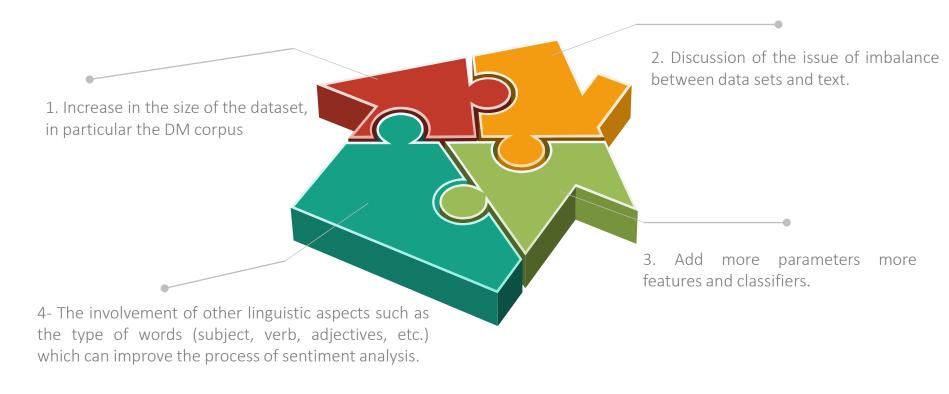
Conclusion





Perspectives &

The next planned steps include:





THANK YOU FOR YOUR ATTENTION

To your questions













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