

01 Introduction

As the official language of 22 countries spread across the Middle-East North Africa (MENA) region, <u>Arabic is the 4th most used language on the Internet</u>. <u>Statistics from 2018</u> show 164 million internet users in the Middle East and 121 million internet users in North Africa.

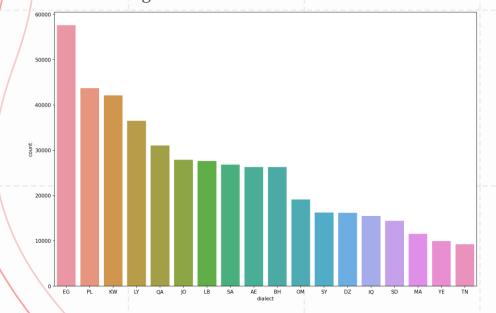
As a language, Arabic has complex morphology and various dialects. The complexity increases significantly when considering the informal nature of social-media text and the distinction between Modern Standard Arabic (MSA) and Dialectical Arabic (DA). Further complicating matters, there are numerous dialects, for example the Egyptian dialect is different from the Levantine dialect which is used in Palestine, Jordan, Syria, and Lebanon. Both of these dialects are also distinct from the Gulf dialect used in Kuwait, Bahrain, Qatar, and the United Arab Emirates.



https://towardsdatascience.com/machine-learning-advancements-in-arabic-nlp-c6982b2f602b

02 DataSet

Our Dataset contains 458197 rows where labels are imbalanced. "EG" is the most major label, Our as "TN" is least label in occurrence. With a distribution as shown in the figure.



	id	dialect	text						
0	1175358310087892992	IQ	. لكن بالنهاية ينتفض يغير Nw8ieJUwaCAAreT @						
1	1175416117793349632	IQ	…يعني هذا محسوب على البشر ح 7zNqXP0yrODdRjK®						
2	1175450108898565888	IQ	مبين من كلامه خليجي KanaanRema@						
3	1175471073770573824	IQ	🎎 يسلملي مرورك وروحك الحلوه HAIDER76128900@						
4	1175496913145217024	IQ	🍨 🏗 وين هل الغيبه اخ محمد hmo2406@						
458192	1019484980282580992	ВН	😇 مبسوطين منك اللي باسطانا Al_mhbaa_7@						
458193	1021083283709407232	ВН	والله ماينده ابش يختي Zzainabali @P_ameerah@						
458194	1017477537889431552	ВН	شو عملنا لك حنا تهربي مننا احنا مس Al_mhbaa_7@						
458195	1022430374696239232	ВН	@haneenalmwla الله يبارك فيها وبالعافيه 🥞 😂 🥹						
458196	1022409931029458944	ВН	🥰 😅 السحله ضيفي ي بتطلع لك سحليه jolnar121@						
458197 rows × 3 columns									

O3 Proposed Methodology

1. PreProcessing "text" which includes:

- Remove URLs
- Remove RT and cc
- Remove @username
- Remove new line in both Windows or Apple
- Clean hashtags
- Clean emojis where translating native emojis (such as: ':)', ':(', ...etc to words), and remove emojis belongs to emojis library or other utf-8 format.
- Remove punctuations
- For Arabic normalizations: remove stopwords, normalize Arabic letters (such as: transform ε to ε , transform ε to ε , etc...), Using pyarabic Package to remove Tashkel, and finally remove longation.

	id	dialect	text	word_count	char_count	avg_char_per_word	stopwords	emoji_count
0	1175358310087892992	IQ	بالنهايه ينتفض يغير	3	19	5.666667	0	0
1	1175416117793349632	IQ	يعني محسوب البشر حيونه ووحشيه وتطلبون الغرب يحترمكم ويءمن بدينكم ولاينعتكم بالارهاب	12	83	6.000000	0	0
2	1175450108898565888	IQ	مبين كلامه خليجي	3	16	4.666667	0	0
3	1175471073770573824	IQ	يسلملي مرورك وروحك الحلوه	4	25	5.500000	0	0
4	1175496913145217024	IQ	وين الغيبه اخ محمد	4	18	3.750000	0	0

03 Proposed Methodology

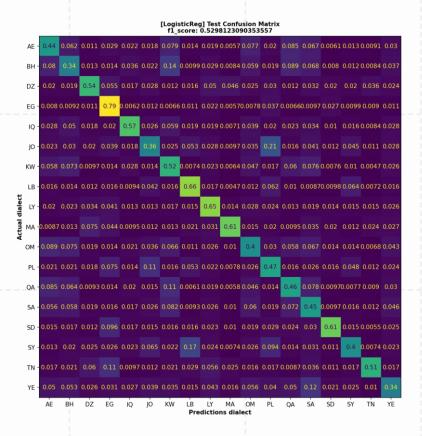
2. For ML Models:

- TF-IDF Vectorizer with unigram word analyzer extraction.
- A Stratified Train test split is considered to be more fit for unbalanced data for more generalized validation score.
- A Lasso Logistic Regression were tested along the TF-IDF Vectorizer on the cleaned data and with class weight to overcome unbalance data
- Complement Naive Bayes classifier were tested too. It is particularly suited for imbalanced data sets.
- Stochastic gradient descent Classifier (SGDClassifier) with Hinge loss as parameter that performs as Linear SVM were tested with class weights setted.

3. For Deep Learning Model:

 An AraBERT version 2 model were tested from "aubmindlab/bert-base-arabertv02" without FarasaSegmenter over the original data with "ArabertPreprocessor()" object, offered Transformers Package over Pytorch.

04 Metrics and Results



[AraBERT] Test Confusion Matrix f1 score: 0.5262330859886513 0.0068 0.029 0.011 0.038 0.19 0.0053 0.016 0.008 0.018 0.043 0.1 0.049 0.0011 0.0061 0.0049 0.004 0.0042 0.014 0.021 0.047 0.24 0.0042 0.024 0.0053 0.01 0.051 0.062 0.062 0.003 0.0049 0.0042 0.0042 0.013 0.024 0.011 0.032 0.032 0.0068 0.057 0.077 0.0056 0.091 0.0056 0.019 0.0025 0.011 0.045 0.004 0.0019 0.011 0.0052 0.0012 0.009 0.0087 0.0016 0.015 0.0043 0.0012 0.067 0.0021 0.0043 0.01 0.0014 0.005 0.0016 0.0077 0.0097 0.0084 0.0032 0.061 0.029 0.026 0.0026 0.014 0.0045 0.004 0.061 0.016 0.017 0.0061 0.0025 0.0086 0.061 0.017 0.035 0.0079 0.028 0.032 0.00036 0.017 0.0047 0.0033 0.0026 0.016 0.0059 0.0028 0.025 0.047 0.047 0.0014 0.0033 0.0024 0.003 0.004 0.019 0.013 0.016 0.0069 0.063 0.021 0.016 0.01 0.00072 0.21 .0063 0.039 0.035 0.052 0.0055 0.021 0.04 0.0063 0.62 0.018 0.0074 0.081 0.009 0.016 0.0047 0.0068 0.029 0.0058 0.011 0.023 0.094 0.037 0.0061 0.016 0.02 0.0078 0.031 0.0071 0.029 0.01 0.067 0.0043 0.091 0.025 0.02 0.018 0.0064 0.0055 10094 0.011 0.0027 0.019 0.0037 0.002 0.13 0.0045 0.014 0.0055 0.031 0.22 0.0029 0.0093 0.0051 0.0084 0.023 0.022 0.097 0.016 0.02 0.01 0.041 0.2 0.0022 0.017 0.0078 0.01 0.036 0.081 0.0049 0.037 0.012 0.11 0.0049 0.024 0.028 0.0035 0.025 0.011 0.0035 0.11 0.017 0.009 0.0074 0.031 0.02 0.03 0.014 0.095 0.034 0.13 0.018 0.011 0.0043 0.25 0.0068 0.029 0.0037 0.022 0.13 0.022 0.053 0.015 0.062 0.096 0.012 0.042 0.015 0.019 LB DZ KW SD

Predictions dialect

05 Deployment

In this section, 5 main libraries were used:

- 1. scikit-learn for tfidf vectorizer and Logistic regression Model
- 2. Pyarabic for striping tashkeel
- 3. Nltk for Arabic stop words
- 4. Transformers from Hugging Face to add our AraBERT model to the pipeline to predict dialect of text.
- 5. Fast API used to implement our webserver app





HUGGING FACE



default

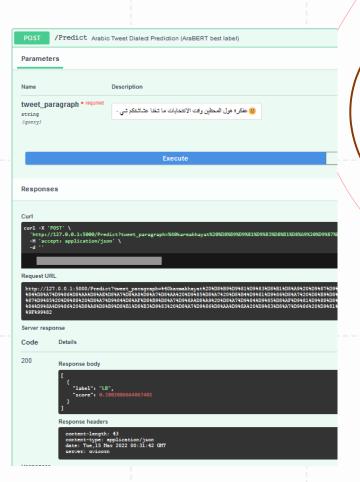
POST / Predict_ml Arabic Tweet Dialect Prediction (Logistic Regression Model)

POST / Predict Arabic Tweet Dialect Prediction (AraBERT best label)

POST / Predict_all_labels Arabic Tweet Dialect Prediction with (AraBERT all labels)

05 Deployment

```
/Predict ml Arabic Tweet Dialect Prediction (Logistic Regression Model)
Parameters
Name
                                   Description
tweet_paragraph * required
                                     🥶 عفكرة هول المحتلين وقت الإنتخابات ما شفدا عشاشتكم شي ،
                                                      Execute
Responses
Curl
   http://127.0.0.1:5000/Predict_ml?tweet_paragraph=%40karmakhayat%20%D8%89%D9%81%D9%83%D8%81%D8%A9%20%D9%
   -H 'accept: application/json' \
Request URL
  http://127.0.0.1:5000/Predict ml?tweet paragraph=440karmakhayat4204D84B94D94814D94834D84B14D84A94204D94874D1
4844D84A74D94864D84A4D84AE84D84A74D84A74D84AA4204D94854D84A74204D84B44D94814D94864D84A74204D84B4B94D84D
  $874D9$854204D9$884204D8$A74D9$844D8$AF4D8$B94D8$A74D9$8A4D8$A94204D8$A74D9$844D9$854D8$AF4D9$814D9$884D88E
  $84$D9$8A$D9$86$20$D8$A8$D8$B9$D8$B1$D8$B3$D9$83$20$D8$A7$D9$86$D8$AA$D9$8A$20$D9$83$D8$A7$D9$86$20$D9$81$D
  %9F%99%82
Server response
              Details
Code
200
              Response body
                   "label": "LB",
                    "score": 0.535316545582301
```



THANKS! Do you have any questions?