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Adapting a resource-light highly multilingual Named Entity Recognition system to Arabic

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Objective

- Adapt an existing language-independent named entity recognition (NER) system to cover Arabic.
- · Use it for daily large-scale information extraction and media monitoring within the Europe Media Monitor family of applications.

Europe Media Monitor (EMM) -Multilingual news analysis

- · The EMM family of applications was developed at the European Commission's Joint Research Centre.
- · Freely accessible at: http://emm.newsbrief.eu/overview.html;
- EMM gathers an average of 100,000 news articles per day in 50 languages.
- EMM clusters and classifies the articles, follows topics over time (topic detection and tracking) and detects trends (alerting)
- EMM-NewsExplorer applies text mining tools to 20 languages. Tools include:
 - · Recognition and disambiguation (grounding) of persons, organisations, locations;
 - · Name variant matching, including across languages and scripts e.g. Javier Solana, Кhavier Solana, خافییر سولانا, Хавьер Солана, ...;
 - · Recognition of quotations by and about people;
 - · Linking related news across languages, for all language pairs;
 - · Produce social networks based on multilingual media information:
 - · Gather and aggregate multilingual information about people.

Multilingual NER rules in EMM

- · Manually produced NER rules are languageindependent, but make reference to language-specific parameter files.
- The language-specific parameter files contain long lists of trigger words: titles, professions, ethnic groups, religious groups, modifiers, determiners, stop words, quotation verbs, etc.
- These trigger word lists were produced semi-automatically, using machine learning and boot-strapping methods.
- Light-weight: no use of parsers, POS taggers, large-scale dictionaries.
- Sample NER rules:
- PERSON_TRIGGER+\bUppercaseWord\b (1)UppercaseWord Swiss world champion Roger Federer
- UppercaseWord(\bMODIFIER)*\b (2)PERSON TRIGGER+ Hamid Karzai, the newly elected Afghan president
- About 600 new names are detected every day and added to a list of over 1 million known names. Known names are then recognised through a simple lookup.



- death of former Prime Minister Rafik Hariri, blamed by many opposition en es asesinato del exprimer ministro Rafic al-Hariri, que la oposición atribuyó
- l'assassinat de l'ex-dirigeant Rafie Hariri et le départ du chef de la diplom fr
- na de moord op oud-premier Rafiq al-Hariri gingen gisteren bijna een nl
- libanesischen Regierungschef Rafik Hariri vor einem Monat wichtige B de
- danjega libanonskega premiera Rafika Haririja. Libanonska opozicija si sl
- et möödumisele ekspeaminister Rafik al-Hariri surma põhjustanud pommipl
- ar اغتيال رئيس الوزراء السابق رفيق الحريري بأياد يهودية وما حدث سابقا
- ru Бывший премьер-министр Ливана Рафик Харири, который

Figure: NER examples for 9 out of the 20 languages covered

Pervez Musharraf

Names	- 34	Key Titles and Phrases	66	External resources
Reverse Nuclearies (Eq., v_1) General Brevese Nuclearies $a \rightarrow a \lambda \mu (\mathbf{r})$ Gen Hutharraf (e_1) Brevese Nuclearies (r_2) Brevese Nuclearies (r_3) Brevese Nuclearies (r_4) Brevese Nuclearies (r_4) Brevese Nuclearies (r_4) Gen Brevese Nuclearies (r_4) Brevese Nuclearies (r_4) Brevese Nuclearies (r_4) Devese Nuclearies (r_4) Prevese Nuclearies (r_4) Prevese Nuclearies (r_4) Prevese Nuclearies (r_4) Brevese Nuclearies (r_4) Breves	0 11 10 10 10 10 10 10 10 10 10 10 10 10	president particler and the second se	<	- Andrew State

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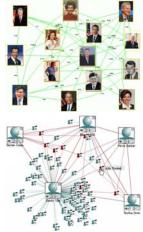


Figure: 2 different social networks generated from information found in multilingual news.

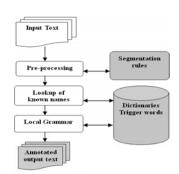


Figure: Architecture of the system

Major challenge for adapting EMM rules to Arabic

- Many language-independent NER rules rely on case information, but Arabic does not distinguish upper and lower case, making it difficult to identify the name boundaries.
- · Arabic commonly uses prefixes and suffixes
- Some titles are also common first names. e.g. امير (Sultan), امير (Prince)

How we solved these issues

- Language-specific Arabic NER rules were added to the language-specific parameter file
- Longer lists of ~20,000 known name parts were used.
- Name stop words help detect the end of the name.
- · Light-weight morphological processing strips prefixes and suffixes from the word stems before applying the grammar
 - (and for the president). (for the president). (president).
- Online demo: Select Arabic on the page: http://emm.newsexplorer.eu/

Arabic-specific NER rules

- (3) KNOWN NAME+\b(\w+)\bNAME INFIX*\b KNOWN NAME
- (4)(\w+)\bNAME_INFIX+\b(\w+) (Mohammed ali ben Halima) محمد على بن حليمة
- (5) PERSON TRIGGER+\b(\w+)\bKNOWN NAME (Mr. Issa Ahmed – Ahmed is known) السيد عيسى أحمد
- (6) NAME_STOP_WORDS\b(\w+)(\bMODIFIER)*\b PERSON_TRIGGER+ وقال حامد كرزاي الرئيس الأفغاني المنتخب الجديد

(And said Hamid Karzai, the newly elected Afghani president)

Evaluation

Test corpus: 35 manually tagged online news articles (34,000 tokens) from the newspapers Assabah (Tunisia) and Alanwar (Lebanon)

Category	Number	Precision	Recall	F-measure
Person	804	87 %	66.54 %	75.40 %
Organization	514	69.96 %	35.79	47.35 %
Location	433	91.52 %	74.82 %	82.33 %
Date and time	54	96.13 %	94.11 %	95.10 %
Numeric expression	46	93.29 %	89.47 %	91.34
Overall	1851	87,17 %	65,74 %	74,95 %

Table: Results obtained for the various NE types

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Figure: Part of the multilingual historical information hered by EMM-NewsExplorer about over 1 million entition 1 million entities / walilra'iis/ وللرئيس /lilra'iis/للرئيس

/ra'iis/رئيس

