

Exploring Multi-level Distributional Semantics for Cross-lingual Entity Discovery and Linking

Boliang Zhang, Xiaoman Pan, Lifu Huang, Ying Lin, Heng Ji

jih@rpi.edu



Rensselaer

Noisy Training Data Acquisition 1: Chinese Room

⌚ Elapsed time: 00:00:25

Latin Mode

1 Mooshinkii ka dhanka ahaa hogaamiyaha aqlabiyada baarlamaanka [PER Aden Ducale] oo laga tanaasulay

Delete:

Type:

Unclear: Translation: Entity:

Mark as:

Previous Annotation:

Type: Translation:

Parallel A motion against the Majority Leader, Aden Duale, was abandoned

2 Xildhibaannada [Loc gobolka bartamaha Kenya] ayaa go'aansaday in ay ka tanaasulaan qorsheynta mooshin kalsooni kala noqosho ah oo ka dhan ahaa

Delete:

hogaamiyaha aqlabiyada Baarlamaanka ahna xildhibaanka [GPE degmada Garissa] Mudane [PER Aden Barre Ducale] .

Parallel Lawmakers in central Kenya have decided to abandon plans for a no-confidence motion against the Majority Leader and the Garissa District Councilor, Mr.

Aden Ducale *Aden Ducale*

Rule

Gazetteer

- Teatro ducale (Parma, Italy): Teatro regio (Parma, Italy)

0.25 wiki_org

Alignment

- Mooshinkii ka dhanka ahaa hogaamiyaha aqlabiyada baarlamaanka Aden Ducale oo laga tanaasulay

A motion against the Majority Leader, Aden Duale, was abandoned

- Xildhibaannada gobolka bartamaha Kenya ayaa go'aansaday in ay ka tanaasulaan qorsheynta mooshin kalsooni kala noqosho ah oo ka dhan ahaa hogaamiyaha aqlabiyada Baarlamaanka ahna xildhibaanka degmada Garissa Mudane Aden Barre Ducale.

Rule	Designator	Affix	Name	Non-name	Lexicon	Other
<input type="button" value="designator"/>	degmada	GPE	district			yinglin
<input type="button" value="designator"/>	magaalada	GPE	town			yinglin
<input type="button" value="designator"/>	dalka	GPE	country			yinglin
<input type="button" value="designator"/>	deegaanka	GPE	area			yinglin

Noisy Training Data Acquisition 1: Chinese Room

The screenshot shows the ELISA IE Annotation Tool interface. At the top, it says "ELISA IE | Annotation Tool" and "Logout". Below that, there's a timer "Elapsed time: 00:00:25" and a "Latin Mode" button with a "Submit" button. The main text area contains a sentence in Somali: "Mooshinkii ka dhanka ahaa hogaamiyaha aqlabiyada baarlamaanka [Aden Ducale] oo laga tanaasulay". The name "Aden Ducale" is highlighted in red. Below the text, there's an annotation panel with fields for "Type" (PER, ORG, LOC, GPE), "Unclear" (checkbox), "Translation" (Aden Ducale), and "Entity" (NIL). There are also buttons for "Apply", "Apply all", and "Delete". Below this, there's a "Previous Annotation" section showing a similar entry. A "Parallel" section shows a corresponding English sentence: "Parallel: A motion against the Majority Leader, Aden Duale, was abandoned". At the bottom, there's a list of suggestions for the highlighted name, including "designator degmada GPE district", "designator magaalada GPE town", "designator dalka GPE country", and "designator deegaanka GPE area".

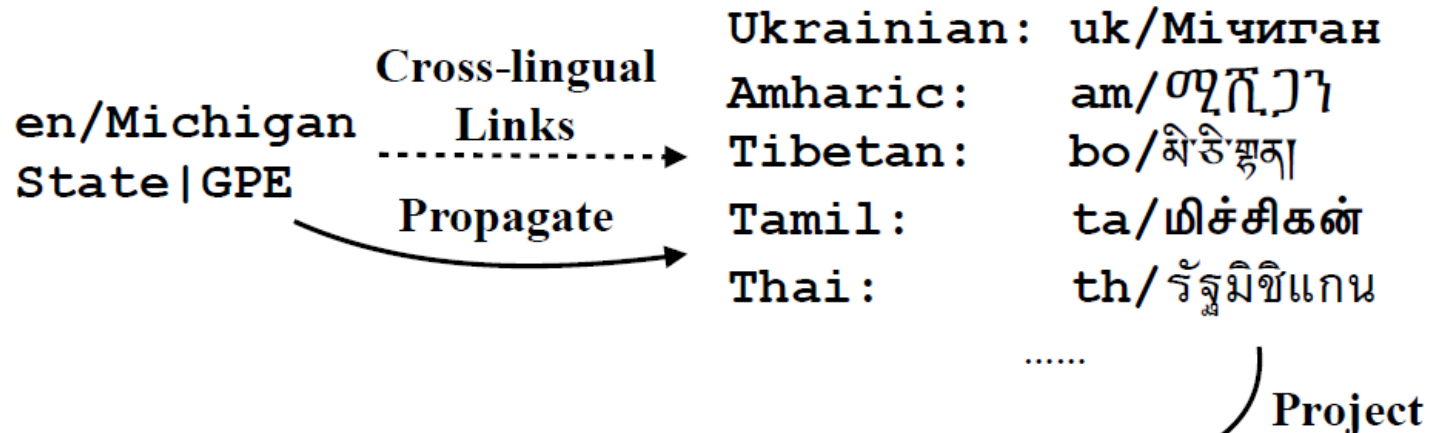
- Annotation for entity type, KB id, translation
- Call Google translation API.
- Mark name as gazetteers.
- Hints from other annotators' annotation.
- Parallel sentence if available.
- Display the converted Latin script if it is not a Latin language.
- Designator hint
- Remove uncertain sentences from the annotation.

0.25 wiki_org

Designator	Affix	Name	Non-name	Lexicon	Other
designator		degmada	GPE district		yinglin
designator		magaalada	GPE town		yinglin
designator		dalka	GPE country		yinglin
designator		deegaanka	GPE area		yinglin

• [Source] Xildhibaanada gobolka bartamaha Kenya ayaa go'aansaday in ay ka tanaasulaan qorsheynta mooshin kalsooni kala noqosho ah oo ka dhan ahaa hogaamiyaha aqlabiyada Baarlamaanka ahna xildhibaan ka degmada Garissa Mudane Aden Barre Ducale.

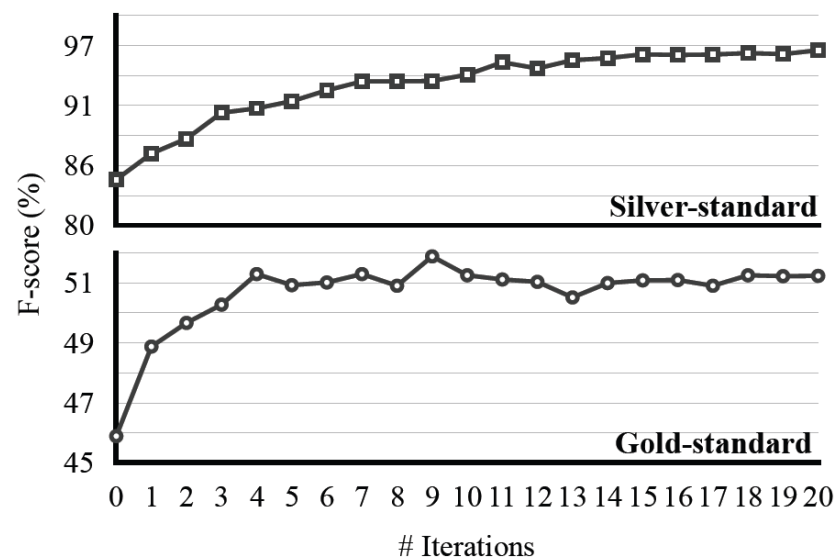
Noisy Training Data Acquisition 2: Wikipedia Mining



[[Мітт Ромні]] Politician|PER народився в [[Детройт]] City|GPE, [[Мічиган]] State|GPE. Закінчив [[Гарвардський університет]] University|ORG.

(Mitt Romney was born in Detroit, Michigan. He graduated from Harvard University.)

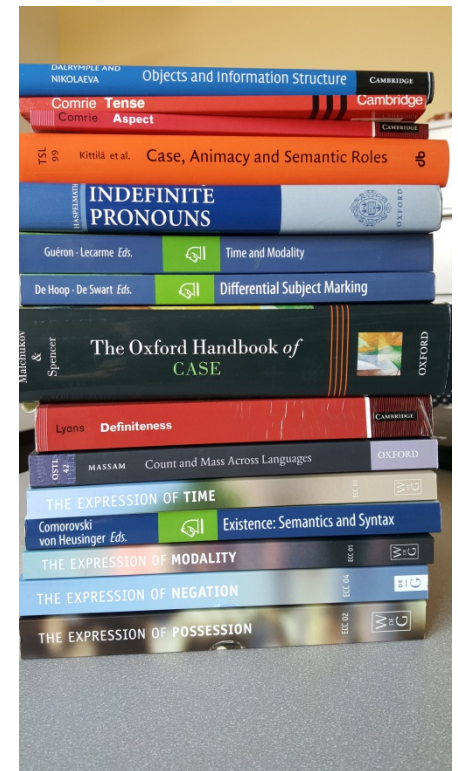
- Generate “silver-standard” training data automatically
- Apply self-training to make training data for complete and consistent



Exploit Non-traditional Universal Linguistic Resources

- Grammar books from Lori Levin's bookshelf and CIA Names from DARPA PM's bookshelf
- Unicode Common Locale Data Repository, Wiktionary, Panlex, Multilingual WordNet, GeoNames, JRC Names, phrase pairs mined from Wikipedia
- Phrase Books from Language Survival Kits and Elicitation Corpus
- Ignored by NLP community

English	Romanization	Target Language
<input type="checkbox"/> You are badly hurt.	<input type="checkbox"/> voo zet blesey seRyEzm/e/	Vous êtes blessé sérieusement
<input type="checkbox"/> You are very sick.	<input type="checkbox"/> voo zet tRey malad	Vous êtes très malade
<input type="checkbox"/> We need to take you to surgery.	<input type="checkbox"/> /o/ dwa voo zemney ah la sal dopeRasy/o/	On doit vous emmener à la salle d'operation
<input type="checkbox"/> We need to remove this.	<input type="checkbox"/> /o/ dwa voo Reteey sa	On doit vous retirer ça
<input type="checkbox"/> We need to repair this.	<input type="checkbox"/> /o/ dwa RepaRey sa	On doit réparer ça



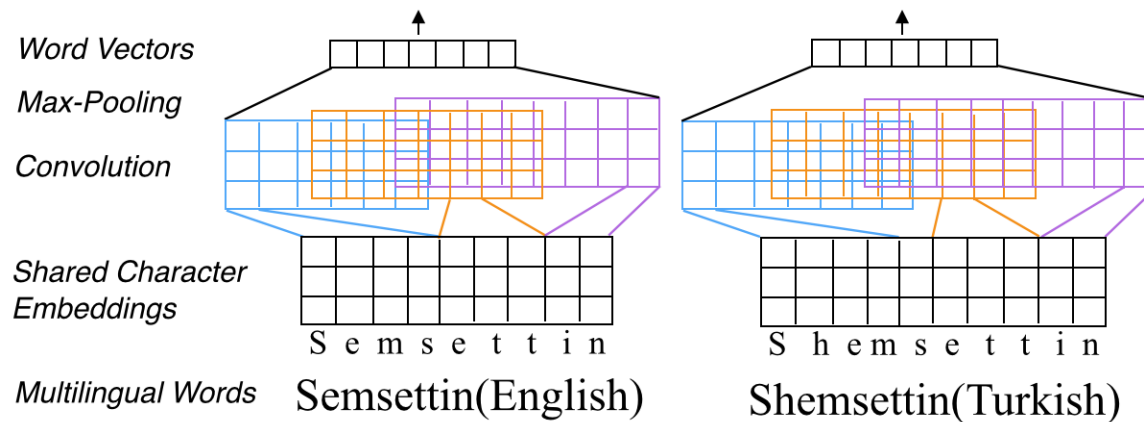
Linguistic Structure from WALS database and Syntactic Structures of the World's Languages

Languages	Categories	Description	Name Related Characteristics	
Tagalog	Subject, Verb, Object Order	VS, VO, VSO	the word at the beginning of a sentence is unlikely to be a name	
Turkish	Negation	Suffix V-Neg indicates negations	not a name	
Bengali	Animacy	-ta is a case that indicates inanimacy		
Japanese	Associative Plural Pattern	Tanaka-tachi (Tanaka and his associates)		
Thai	Nested Structure	Name	Order and special delimiter between modifier and head of a nested name. e.g., [ORG กระทรวงต่างประเทศ] ของ[LOC อินโดนีเซีย] ([ORG Foreign Ministry] of [LOC Indonesia])	Name boundary
Tamil	Conjunction Structure	Name1- yum Name2- yum (Name1 and Name2)	Name type consistency	

- Universal Morphology Analyzer based on Wikipedia Markups
 - *Kıta Fransı, güneyde [[Akdeniz]]den kuzeyde [[Manş Denizi]]ve [[Kuzey Denizi]]ne, doğuda [[Ren Nehri]]nden batıda [[Atlas Okyanusu]]na kadar yayılan topraklarda yer alır. (Continental France is located in the south [[Mediterranean Sea]] in the north [[English Sea]] and [[North Sea]] in the east [[Rhine River]] to the west [[Atlantic Ocean]].)*

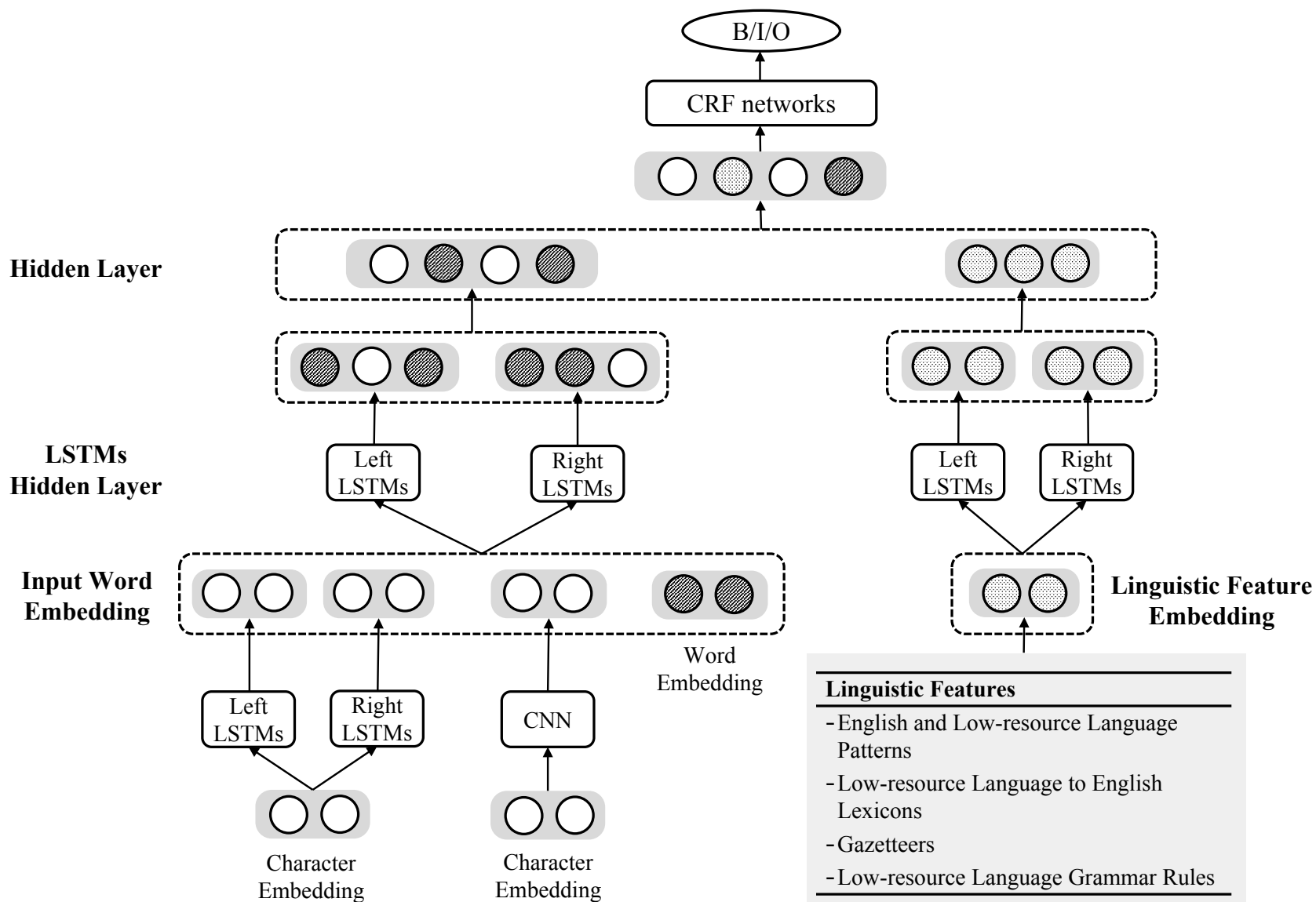
Character-Aware Word Embeddings

- Motivation: mentions of the same concept across languages may share a set of similar characters, e.g., Semsettin Gunaltay (English) = Şemsettin Günaltay (Turkish) = Semsetin Ganoltey (Somali)
- Compose word embeddings from shared character embeddings using Convolutional Neural networks

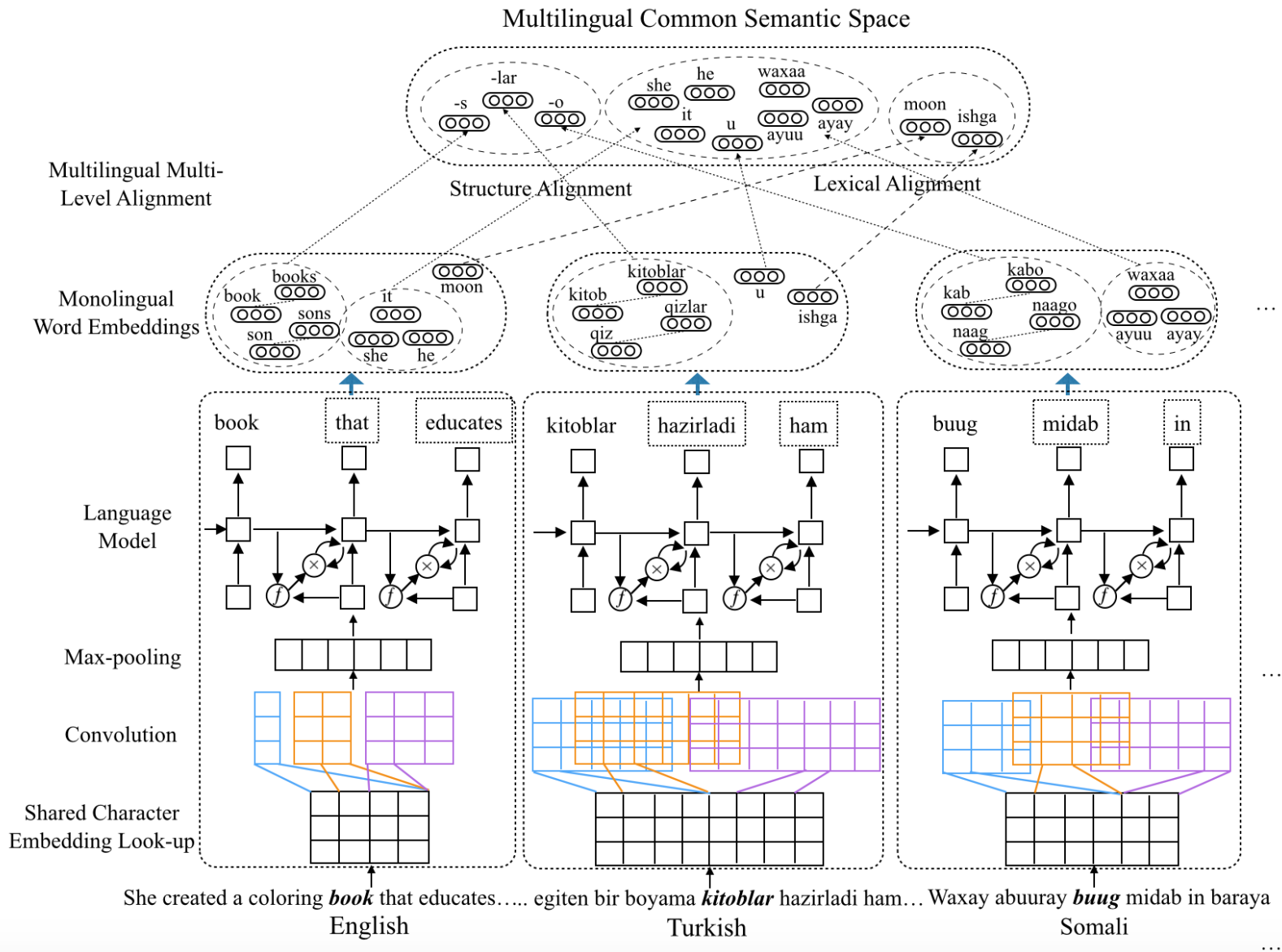


- Further optimized by language model based on Recurrent Neural Networks
 - maximize the prediction of the current word based on previous words

Feed Non-traditional Linguistic Resources into DNN



Common Semantic Space Construction



Construct a Common Semantic Space for Thousands of Languages

- Motivations
 - There are 3000+ languages with electronic record
 - NLP training data only available for several dominant languages
- Goals
 - Build a common semantic space across thousands of languages for resource sharing and richer semantic continuous representation for words, concepts and entities
- Limitations of Previous Attempts (e.g., Upadhyay et al., 2016, Cho et al., 2017)
 - Mostly English-anchored, cannot capture all linguistic phenomena
 - Heavily relied on bilingual dictionaries and parallel data which are not always available
 - Only limited to dozens of languages

Multi-Level Multi-lingual Alignment

- When bilingual word dictionaries are not available, back-off to shared linguistic structures
 - e.g., apposition, conjunction, plural suffix (English (-s / -es), Turkish (-lar / -ler), Somali (-o))
 - Generalized from language universal resources such as *WALS database* and *Syntactic Structures of the World's Languages*
 - Classify languages according to a large number of topological properties (phonological, lexical, grammatical)
 - 2,676 languages, 58,000+ (language, feature, feature value) tuples, e.g., (English, canonical word order, SVO)
- Project monolingual word embeddings into a common semantic space, and align both representations of words and linguistic-structures in the common space

Model Training

- Model training
 - Language model prediction loss

$$Loss_1 = - \sum_{l \in L} \sum_{t=1}^{|C_l|} \log Pr(w_t = W | w_{1:t})$$

$l \in L$: language l in the language set L
 $C_l = [w_1, w_2, \dots, w_T]$: the sequence of words in the corpus for language l
 W : the real t^{th} word in C_l

- Multilingual alignment loss:

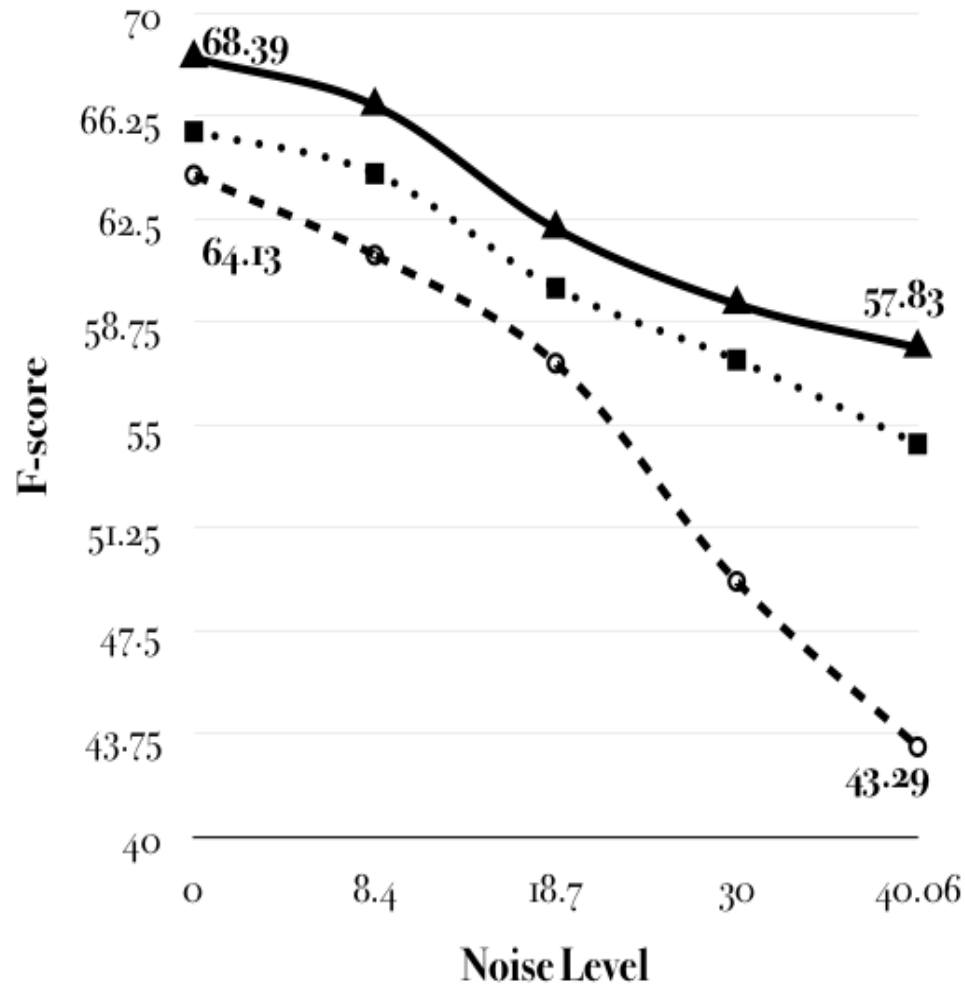
$$Loss_2 = \frac{1}{2} \sum_{i,j \in L} \sum_{(v_m^i, v_m^j) \in Align^{ij}} \|M_i \cdot v_m^i - M_j \cdot v_m^j\|$$

$Align^{ij}$: the word/structure alignment for language i and j
 (v_m^i, v_m^j) : the vector representation of the m^{th} alignment pair in $Align^{ij}$
 M_i, M_j : the projection matrix for language i and j from original semantic space to the common semantic space

- Overall loss:

$$Loss = Loss_1 + Loss_2$$

Linguistic Features Matter: More Robust to Noise



Uzbek (Zhang et al., 2017)

- Embedding Features
- Embedding+Traditional Linguistic Features
- ▲ Embedding+Traditional+Non-traditional Linguistic Features

Impact of Character-Aware Word Embeddings

- Name Tagging F-Score (%)

Models	Chinese	English	Spanish
Before	64.1	67.4	64.6
After	68.0	70.9	68.9

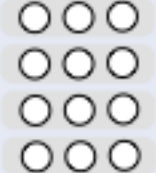
Impact of Common Semantic Space

- Chechen Name Tagging

Models	P (%)	R (%)	F (%)
Randomly initialized	46.3	45.31	45.8
Pre-trained	54.8	41.3	47.1
+ Common semantic space word embedding	62.1	50.1	55.4

Something Old: Hierarchical Brown Clustering

Brown Clustering

Word	Cluster Bit String	Features	Feature <u>Embeddings</u>
Obama	01101 11010 11011 11001	01101 01101 11010 01101 11010 11011 01101 11010 11011 11001	
Clinton	01101 11010 11011 11000
people	01101 11010 11000 00000		

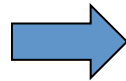
Languages	w/o BC (%)	with BC (%)	Languages	w/o BC (%)	with BC (%)
Albanian	72.4	74.6	Northern Sotho	90.2	90.8
Chechen	53.1	55.4	Polish	49.6	53.2
Chinese	66.3	68.0	Somali	76.9	78.5
English	69.5	70.9	Spanish	67.1	68.9
Kannada	51.9	56.0	Swahili	64.3	67.8
Kikuyu	84.2	88.7	Yoruba	46.1	49.5
Nepali	41.6	43.9			

Joint Learning of Word and Entity Embeddings from Wikipedia

- Consider all Wikipedia anchor links as entity annotations, a training corpus can be created by replacing anchor links with unique entity IDs.

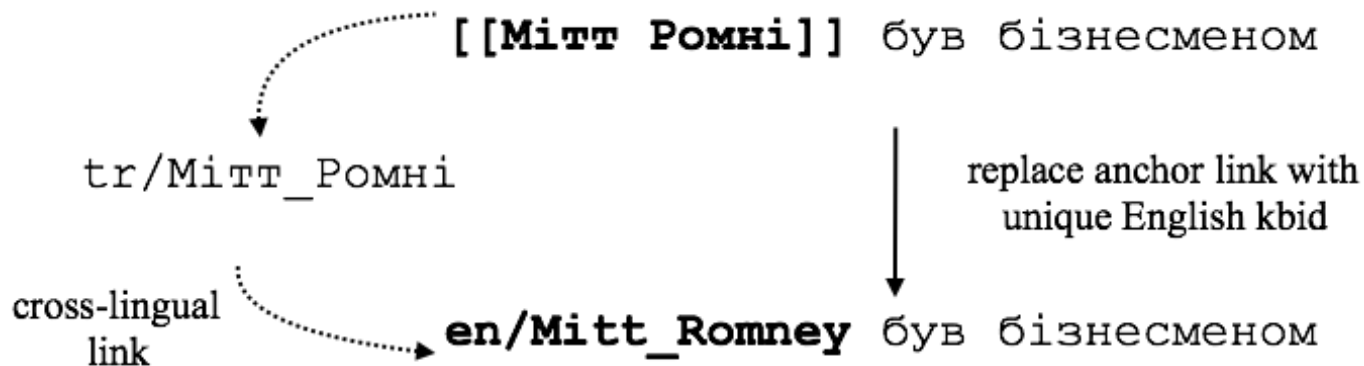
e.g., `[[en/Apple|apple]]` is a fruit
`[[en/Apple_Inc.|apple]]` is a company

apple is a fruit
apple is a company

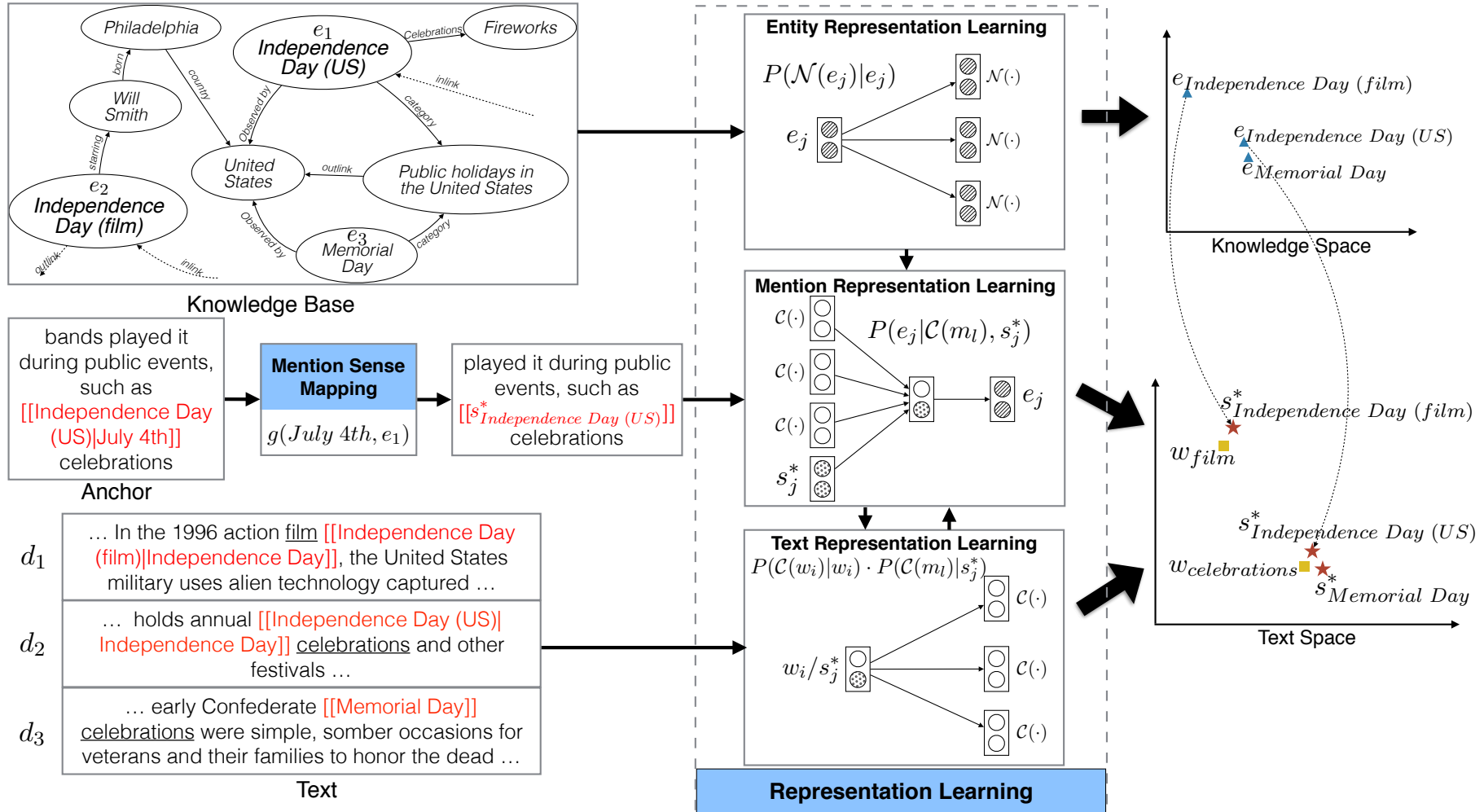


en/Apple is a fruit
en/Apple_Inc. is a company

- Multi-lingual



Joint Learning of Word and Entity Embeddings from Wikipedia



Learning Entity Embeddings from DBpedia

- Construct a weighted undirected graph $G = (E, D)$ from DBpedia, where E is a set of all entities in DBpedia and $d_{ij} \in D$ indicates that two entities e_i and e_j share some DBpedia properties. The weight of d_{ij} , w_{ij} is computed as:

$$w_{ij} = \frac{|p_i \cap p_j|}{\max(|p_i|, |p_j|)}$$

where p_i, p_j are the sets of DBpedia properties of e_i and e_j respectively.

- Apply the graph embedding framework proposed by (Tang et al., 2015) to generate knowledge representations for all entities

Impact of Joint Embeddings on Entity Linking

- Unsupervised entity linking based on salience, similarity and coherence
- Tested on EDL16 perfect English NAM mentions

	CEAFm P	CEAFm R	CEAFm F1
Baseline	0.762	0.843	0.801
+ Joint word and entity embeddings from Wikipedia	0.791	0.875	0.831
+ Entity embedding from DBpedia	0.812	0.897	0.852

Resources and Demos

Systems, Data and Resources Publicly Available

- Re-trainable Systems:
 - http://blender02.cs.rpi.edu:3300/elisa_ie/api
 - Source code base available for government users upon requests
 - Tri-lingual EDL is being integrated into CoreNLP and hope to release in 2017
- Data and Resources:
 - <http://nlp.cs.rpi.edu/wikiann/>
- Demos:
 - http://blender02.cs.rpi.edu:3300/elisa_ie
 - http://blender02.cs.rpi.edu:3300/elisa_ie/heatmap

Demo 1: Cross-lingual Entity Discovery and Linking for 282 Languages

Demo APIs Development Human In Loop Annotation Tool Heat Map Name Transliteration

Incident Language IE Demo

Choose a language:

Abkhazian

Three examples are provided. You could also enter your own text.

example 1 example 2 example 3 enter

```
# REDIRECT Аԥхәимҭаԥа
' ' ' Рацибуж ' ' ' ( ' ' ' ' ) Полша ақалақь .
' ' ' Лос-Анцелес ' ' ' ( ' ' Los Angeles ' ' ) -
Гәдоута араион .
* Монтевидео ( проектируется )
* Лыхны
' ' ' Владимир ' ' ' - Урыстәыла ақалақь .
# Очамчыра араион
```

Browse ▾

Abkhazian	Achinese	Adyghe	Afrikaans	Akan	Albanian
Alemannisch	Amharic	Arabic	Aragonese	Aramaic	Armenian
Aromanian	Arpitan	Assamese	Asturian	Avaric	Aymara
Azerbaijani	Bambara	Bangla	Banjar	Basa Banyumasan	Bashkir
Basque	Bavarian	Belarusian	Belarusian (Tara...	Bihari	Bikol Central
Bishnupriya	Bislama	Bosnian	Breton	Buginese	Bulgarian
Burmese	Catalan	Cebuano	Central Kurdish	Chamorro	Chavacano de Za...
Chechen	Cherokee	Cheyenne	Chinese	Chinese (Min Nan)	Church Slavic
Chuvash	Colognian	Cornish	Corsican	Crimean Turkish	Croatian
Czech	Danish	Divehi	Dutch	Dzongkha	Eastern Mari

- http://blender02.cs.rpi.edu:3300/elisa_ie
- http://blender02.cs.rpi.edu:3300/elisa_ie/heatmap

Cross-lingual Entity Discovery and Linking for 282 Languages (Cont')

DNN

Submit

Tagging result:

demo

Total: 13 NAM: 13 NOM: 0
PER: 5 ORG: 1 CPE: 0 LOC: 7 FAC: 0 TTL: 0

Gold

Mention ID: demo-6
Mention Str: Югославії 300:308
Reference KB: [Socialist Federal Republic of Yugoslavia](#)
Entity Type: LOC
Mention Type: NAM
Confidence Value: 1.0
Images:
Not Found

Засновником саме ренесансної архітектури став [PER Філіппо Брунеллескі]
"Примітка : таблиця складена за даними сайту [ORG Верховної Ради України]
* Автори сценарію : Поль Андреотті , [PER Олександр Галич]
* [PER Район Янпу] (楊浦区 Yángpǔ Qū)
Після завершення війни використовувалась арміями [LOC Австрії] , [LOC Чехословаччини] , [LOC Югославії] .
* [PER Грехем Чепмен] — Браян Коен / Біггус Діккус (товариш Пилата) ;
Муніципалітет розташований на відстані близько 220 км на схід від [LOC Парижа] , 80 км на південний захід від [LOC Меца] , 13 км на південний схід від [LOC Бар] –ле–[LOC Дюка] .

Mention ID: demo-3
Mention Str: போலந்து 132:138
Reference KB: [Poland](#)
Entity Type: LOC
Mention Type: NAM
Confidence Value: 1.0
Images:
Not Found

புது [ORG ததாய தேசய நெடுஞ்சாலையில்] இருந்து 3 கி.மீ தொலைவில் உள்ளது .
" [LOC போலந்து] "
[ORG மார்வெல் காமிக்ஸ்] நிறுவனத்தின் வரைகதைகளில் தோன்றும் [PER அயன் மேன்] என்ற கதாபாத்திரத்தை மையமாகக் கொண்டு எடுக்கப்பட்டது .
வழிமாற்று [ORG என்] . எம். [LOC சோசி]
* [LOC நோர்போக் த்வ] - 2,302
வழிமாற்று [LOC சேமக்கலம் (பொறியியல்)]

Cross-lingual Entity Discovery and Linking for 282 Languages (Cont')

NW_CRI_HAU_006015_20140827

Total: 9 NAM: 9 NOM: 0

PER: 3 ORG: 0 GPE: 0 LOC: 6 FAC: 0 TTL: 0

Gold

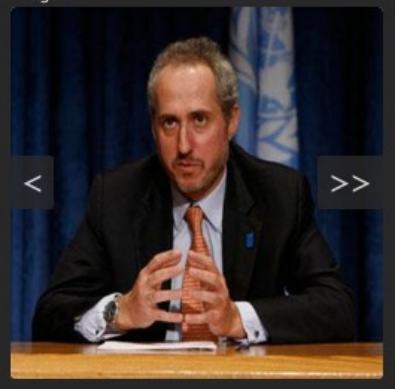
Mutane 3 sun rasu sakamakon faduwar wani jirgin sama a [LOC Sudan ta kudu] Kakakin babban magatakardan MDD [PER Stephane Dujarric] ya ce wani jirgin sama, dauke da mambobin tawagar musamman ta MDD a [LOC Sudan ta kudu] ya fadi a birnin [LOC Bentiu] dake [LOC Jihar Unity] a arewacin kasar, lamarin da ya yi sanadin mutuwar ma'aikatan jirgin 3 yayin da wani guda kuma ya jikkata. A wani taron manema labaru da aka yi a wannan rana, [PER Stephane Dujarric] ya ce jirgin saman mai saukar ungulu wanda aka yi hayar sa daga [LOC Rasha], ya yi hadari ne yayin da ya ke jigilar kayayyaki, kuma dukkanin wadanda suka rasu su 3 'yan kasar [LOC Rasha] ne. [PER Stephane Dujarric] ya kara da cewa tuni aka tura ma'aikatan rundunar zuwa wurin da hadarin ya auku, domin binciken dalilan aukuwar hadarin. (Amina)

System

Missing

Spurious

Mention ID: Boliang_NE_tagger_00010
 Mention Str: Stephane Dujarric 601:617
 Reference KB: St%C3%A9phane_Dujarric
 Entity Type: PER
 Mention Type: NAM
 Confidence Value: 1.0
 Images:

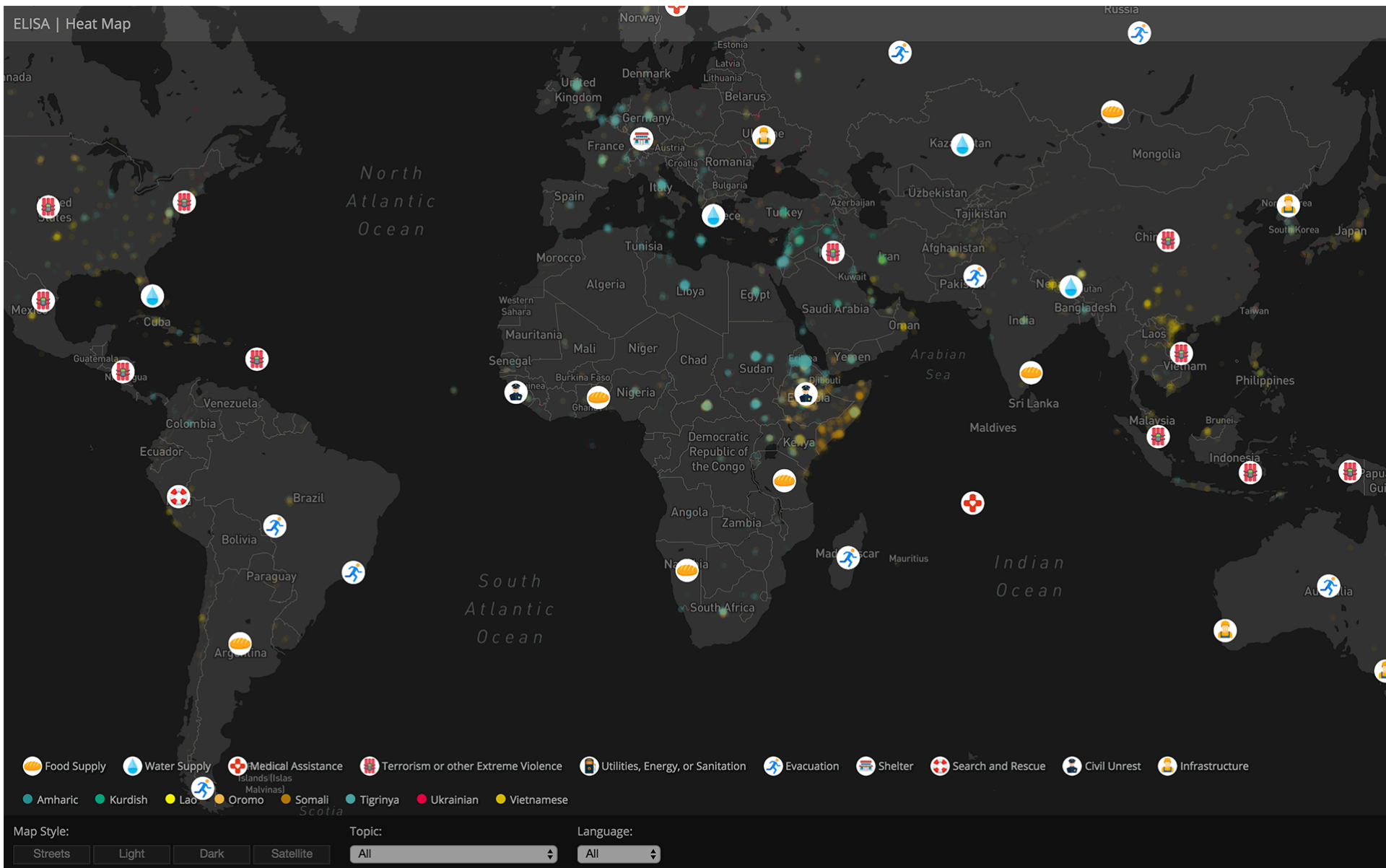


Type

[n] ta kudu Kakakin babban magatakardan [ORG MDD] [PER Stephane Dujarric] ya ce wani jirgin sama, [LOC Sudan ta kudu] ya fadi a [GPE birnin Bentiu] dake [GPE jihar Unity] a arewacin kasar, lamarin da ya yi sanadin mutuwar ma'aikatan jirgin 3 yayin da wani guda kuma ya jikkata. A wani taron manema labaru da aka yi a wannan rana, [PER Stephane Dujarric] ya ce jirgin saman mai saukar ungulu wanda aka yi hayar sa daga [GPE kasar Rasha], ya yi hadari ne yayin da ya ke jigilar kayayyaki, kuma dukkanin wadanda suka rasu su 3 'yan [GPE kasar Rasha] ne. [PER Stephane Dujarric] ya kara da cewa tuni aka tura ma'aikatan rundunar zuwa wurin da hadarin ya auku, domin binciken dalilan aukuwar hadarin. ([PER Amina])

Name	Country
1 Bentiu Bantiyo, Bentiu, Bentiu, Kilwal, Meshra' Bentiu, Meshra' Bentiu, ban ti wu, bantyyw, bntyyw, Бентиу, Бентију, بٔٔ...	South Sudan , Northern Liech
2 Bentiu Town	South Sudan , Unity
3 Bentiu Hospital	South Sudan , Unity
4 Bentiu Stadium	South Sudan , Unity
5 Bentiu Graveyard	South Sudan , Unity

IE Application Example: Disaster Relief



Cross-lingual Entity Discovery and Linking for 282 Languages (Cont')

