



Entity-Level Event Impact Analytics

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Entity-Level Event Impact Analytics

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Objective:

Automatically Predict the Event Diffusion into Foreign Language Communities

Conceptual Approach

"Discovery of semantic connections to languages from named entities in the article"

ELEVATE Framework [1]

- Recursive exploration of  relations
- Country centric:**
 - `<isCitizenOf>`, `<diedIn>`,
 - `<isLocatedIn>`, `<isLeaderOf>`,
 - `<isPoliticianOf>`, `<wasBornIn>`, `<livesIn>`
- Organization centric:**
 - `<owns>`, `<created>`, `<worksAt>`
- Linking of entities to languages

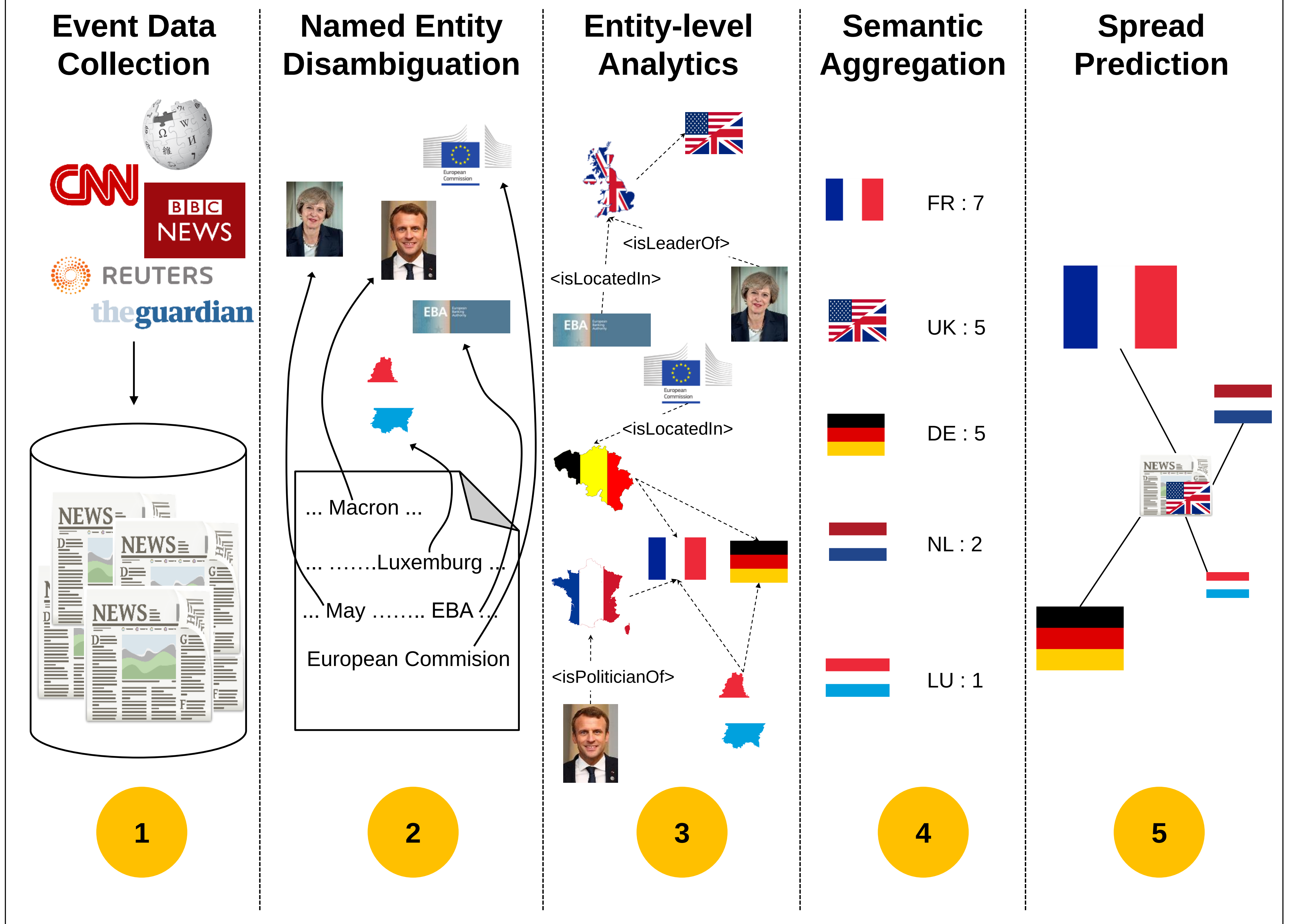
Breadth-first-search (BFS):

Stopping the exploration after the discovery of first language for a named entity

Depth-first-search (DFS):

Revealing all languages associated with a named entity exhaustively

ELEVATE Pipeline



Spread Prediction

Task: Pick the best candidates from all the scored languages

Adjusted Thresholding

- Threshold(θ) = average spread in the ground truth
- k-fold cross-validation
- Risk of picking the irrelevant languages

Multi-label Classification

- Output labels as the languages in event spread
- Candidate language scores as feature vectors
- Classifiers decide the spread

Other Research Works

ELEVATE-Live [3]

- Online news article virality prediction to countries
- Extension of ELEVATE framework
- Available at: <https://elevate.greyc.fr>

Semantic Fingerprinting [2]

- Fine-grained entity-level Web content classification
- Concise semantic representation of documents based on their entities

Short Biography



3rd year PhD student
University of Caen, France
Advisor: Prof. Marc Spaniol

Research Interests

- Entity-level analytics
- Data aggregation via LOD
- Deep learning for noisy data analytics

Education

- Master degree in Maths and Computing from Indian Institute of Technology (IIT) Patna (India)
- Bachelor degree in computer science from Guru Jambheshwar University of Science and Technology Hisar (India)

Publications

- [1] Govind, Spaniol M. ELEVATE: A Framework for Entity-level Event Diffusion Prediction into Foreign Language Communities. In Proceedings of the 9th ACM on Web Science Conference 2017 Jun 25 (pp. 111-120). ACM.
- [2] Govind, Alec C, Spaniol M. Semantic Fingerprinting: A Novel Method for Entity-Level Content Classification. In Proceedings of the 18th International Conference on Web Engineering 2018 Jun 5 (pp. 279-287). Springer.
- [3] Govind, Alec C, Spaniol M. ELEVATE-Live: Assessment and Visualization of Online News Virality via Entity-Level Analytics. In Proceedings of the 18th International Conference on Web Engineering 2018 Jun 5 (pp. 482-486). Springer.

Experimental Results

Method	5 days				10 days				20 days			
	Precision	Recall	F1	#PC	Precision	Recall	F1	#PC	Precision	Recall	F1	#PC
Random	0.07736	0.07691	0.07714	11288	0.07736	0.07691	0.07714	11288	0.07736	0.07691	0.07714	11288
Inlinks _{CNC}	0.23571	0.21023	0.22224	5045	0.25685	0.24077	0.24855	5875	0.28541	0.27803	0.28167	6607
Inlinks _{Adamic}	0.23553	0.21011	0.22210	5045	0.25685	0.24151	0.24894	5875	0.28470	0.27854	0.28159	6607
Outlinks _{CNC}	0.33301	0.42227	0.37236	10568	0.34840	0.45237	0.39363	10845	0.36049	0.46988	0.40798	11049
Outlinks _{Adamic}	0.33230	0.42473	0.37287	10568	0.34928	0.45498	0.39518	10845	0.36182	0.47766	0.41175	11049
DirectMapping	0.35574	0.11457	0.17332	1571	0.36171	0.12032	0.18058	1677	0.36565	0.12337	0.18449	1770
Entities _{DFS}	0.48845	0.22548	0.30854	4612	0.48580	0.23545	0.31718	4886	0.48444	0.24076	0.32166	5087
Entities _{BFS}	0.53199	0.19554	0.28597	2842	0.53082	0.20154	0.29215	3068	0.53368	0.20823	0.29958	3234

Macro-average scores for the adjusted threshold based models (#PC: number of predictions)

Method	5 days				10 days				20 days			
	Precision	Recall	F1	#PC	Precision	Recall	F1	#PC	Precision	Recall	F1	#PC
Inlinks _{CNC}	0.18634	0.20129	0.19353	7055	0.20312	0.22926	0.21540	8336	0.23301	0.25911	0.24537	9220
Inlinks _{Adamic}	0.21763	0.18534	0.20020	6028	0.22731	0.20412	0.21509	7011	0.25686	0.23574	0.24585	7687
Outlinks _{CNC}	0.27909	0.38017	0.32188	14169	0.28725	0.39046	0.33100	14320	0.28945	0.41396	0.34069	14533
Outlinks _{Adamic}	0.30896	0.37136	0.33730	13315	0.30479	0.40219	0.34678	14153	0.30112	0.40503	0.34543	14282
DirectMapping	0.31113	0.14553	0.19830	3538	0.31941	0.14587	0.20028	3529	0.32077	0.14912	0.20359	3755
Entities _{DFS}	0.49333	0.27747	0.35517	7179	0.47678	0.28490	0.35667	7509	0.47549	0.28420	0.35576	8210
Entities _{BFS}	0.45635	0.23128	0.30698	5413	0.46679	0.22763	0.30603	5408	0.46420	0.23649	0.31334	5702

Macro-average scores for the machine learning approach (#PC: number of predictions)