# Translation and the Media

Alexandra Birch









Jamal Khashoggi

# Challenges facing the Media

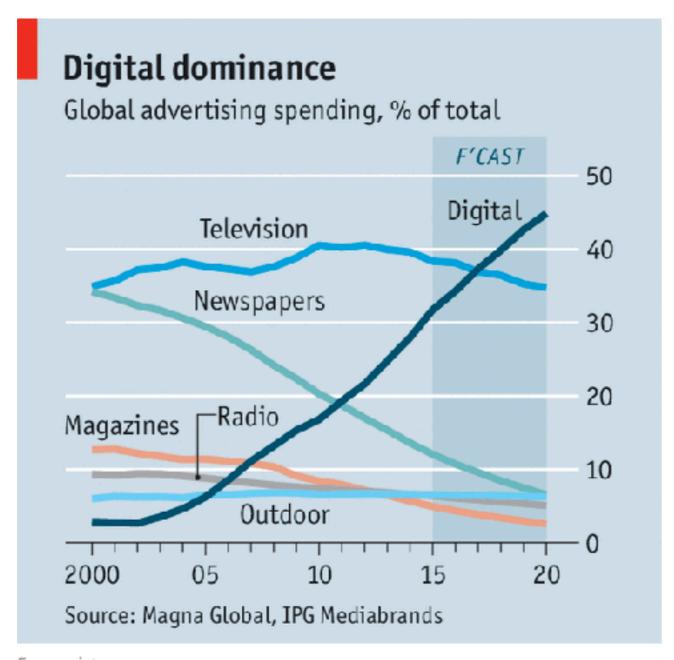








#### Challenges facing the Media



Economist.com

Source: Economist





















Speech Recognition, text normalisation



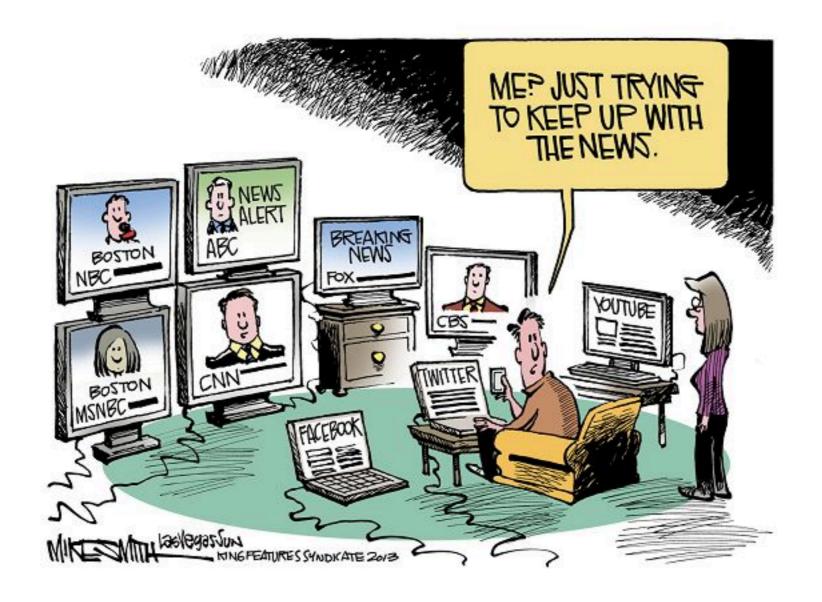




#### **Machine Translation**



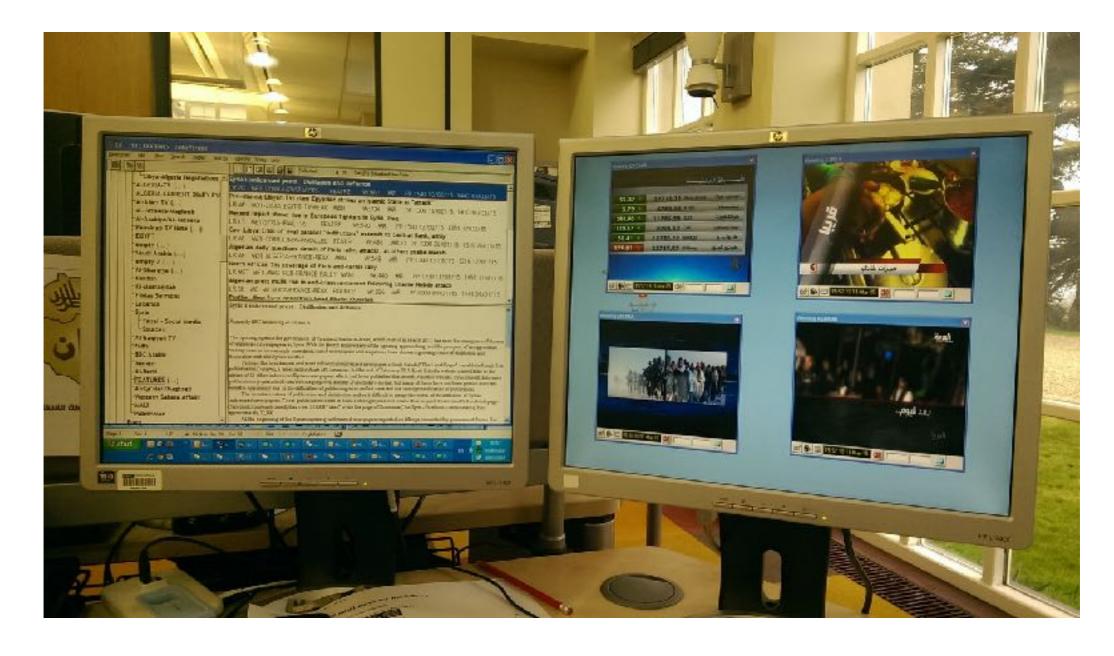




#### Natural Language Processing







# BBC Media Monitoring





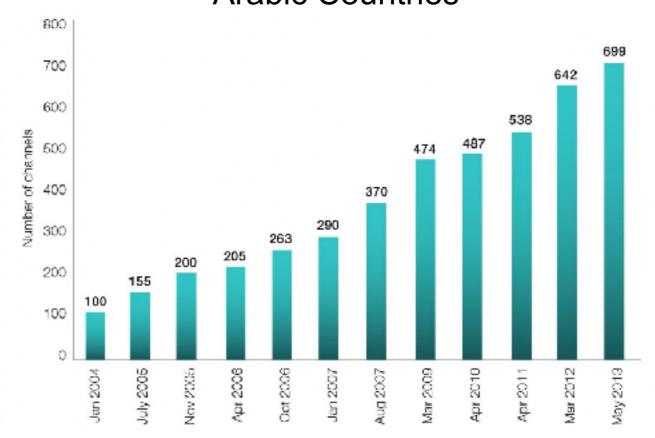
#### Scale of Media Sources



#### **BBC Monitoring 2015**

- 13,500 sources
- 1,500 Television broadcasters
- 1,350 Radio stations

# Free To Air Channels in Arabic Countries



78% in Arabic

Source: Arab Advisors Group research





#### Questions:



- How can NLP help the Media?
  - Unlocking more content from different media types/languages
  - Making sense of deluge of data
  - Reach a wider audience
- How can we deal with multilingual content?
  - + Translation
  - Multilingual NLP
- Context of EU Projects:
  - SUMMA: Automated Media Monitoring Platform
  - Gournette Tension



# Scalable Understanding of Multilingual Media





















H2020: 1 February 2016 - 31 January 2019



#### BBC Monitoring Use Case













#### Aims

Robo eyes and ears

Find stories quicker

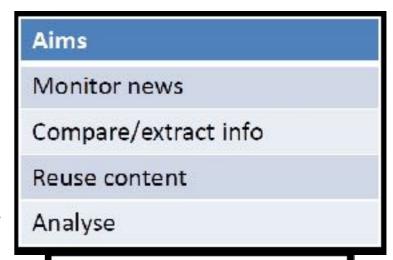
Alerts

Data journalism

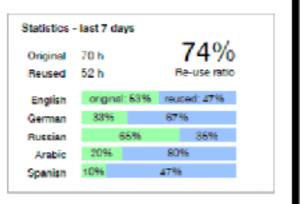
# Deutsche Welle Internal Monitoring Use Case









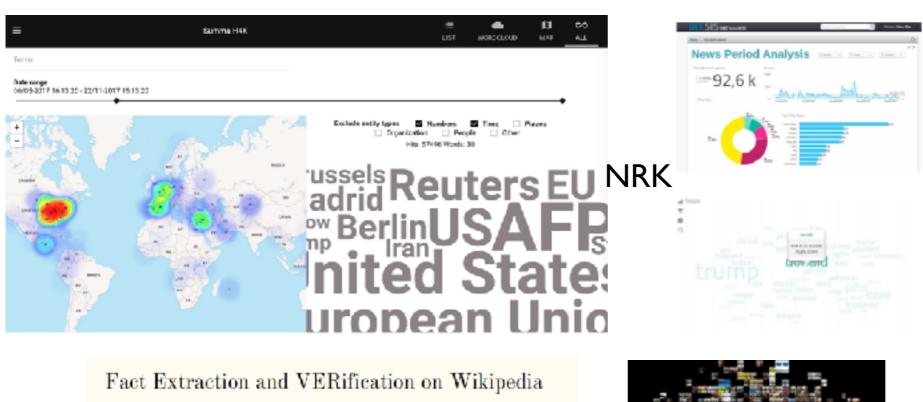






#### Paul Bradshaw:

Nose for news + scale and range of digital data



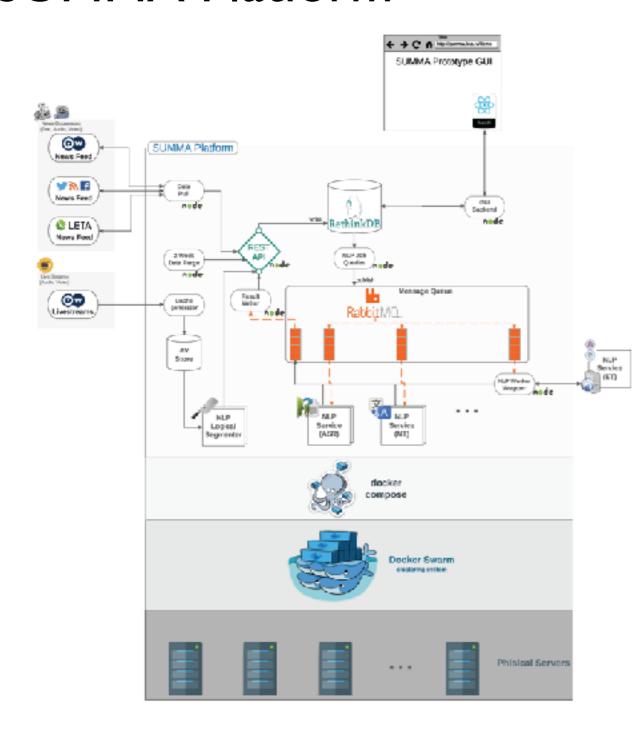
Decsis





#### **SUMMA Platform**

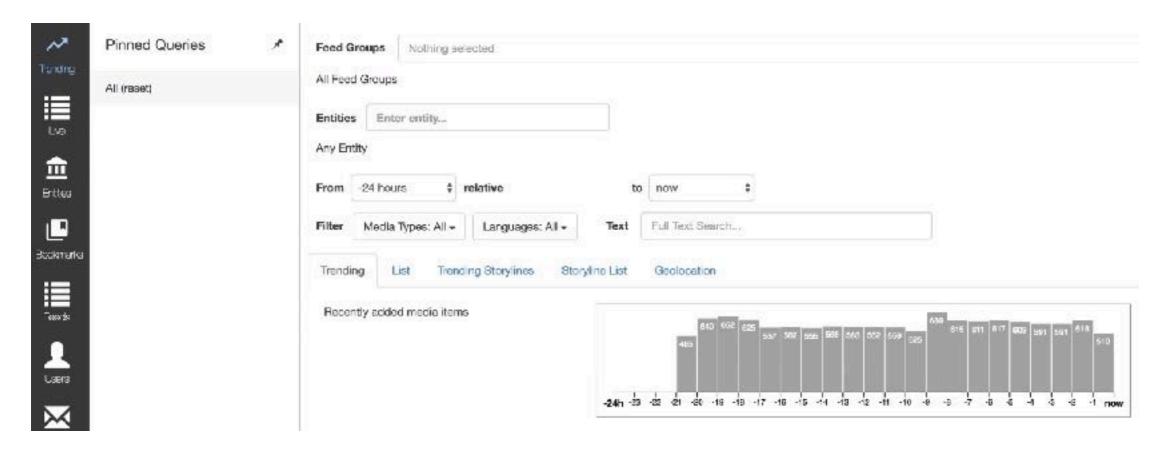




- Microservices architecture
- All services are Docker containers
- Scalability is achieved by launching as many Docker container instances per task as required
- Open source platform and open source NLP Modules <a href="https://github.com/summa-platform">https://github.com/summa-platform</a>



#### SUMMA Platform: Scaleability test



- Last week tested 24 hours of processing of 62 broadcast channels
- One AWS p3.8xlarge (244.0 GiB RAM, 32 vCPUs, 4 Tesla V100) to serve the data in HLS format
- Three AWS m5.24xlarge (384.0 GiB RAM, 96 vCPUs) to process it

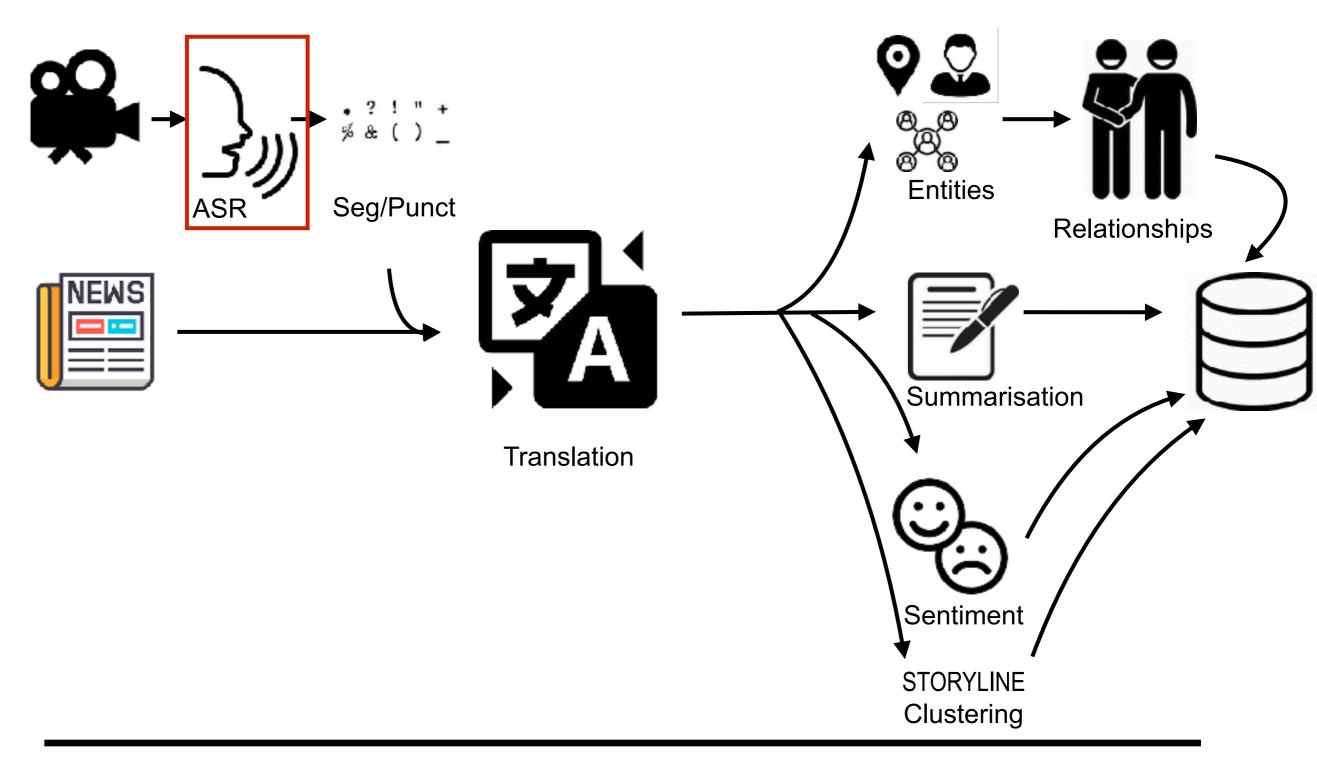
#### **SUMMA Channels**



Country	Name	Language
Qatar	Al-Jazeera English	English
Iran	Al-Alam TV	Arabic
United Arab Emirates	Al-Arabiya TV	Arabic
Algeria	Algerian TV	Arabic
Algeria	Algerian TV Channel 3 (A3C)	Arabic
Iraq	Iraqia News TV	Arabic
Iraq	Al-Sharqiyah News TV	Arabic
Germany	German ARD TV	German
Korea (South)	Arirang TV	English
United Kingdom	BBC1	English
United Kingdom	BBC News Channel	English
United Kingdom	BBC Parliament TV channel	English
United Kingdom	BBC World Service	English
United Kingdom	BBC World TV	
•		English
Belarus	Belarus 24 TV	Russian
Bahrain	Bahrain TV	Arabic
Egypt	Al-Misriyah TV	Arabic
China	China Global Television Network (CGTN, formerly CCTV)	English
Russia	Channel One TV	Russian
Russia	Channel One TV Worldwide (for Europe)	Russian
Algeria	Ennahar TV	Arabic
France	France 24 English Television	English
Lebanon	Future News TV	Arabic
Saudi Arabia	Al-Ikhbariyah TV	Arabic
Iran	Islamic Republic of Iran News Network	Persian
Iran	Vision of the Islamic Republic of Iran Network 1	Persian
Iran	Vision of the Islamic Republic of Iran Network 2	Persian
Kuwait	Kuwait TV 1	Arabic
Lebanon	LBC Sat TV	Arabic
Libya	Libya 218 TV	Arabic
Libya	Libya's Channel TV	Arabic
Libya	Libya al-Hadath TV	Arabic
Libya	Al-Rasmiyah (Official) TV	Arabic
Libya	Libya al-Mostakbal TV	Arabic
Morocco	2M Monde	Arabic
Lebanon	Al-Manar TV	Arabic
Mauritania	Mauritanian TV	Arabic
Lebanon	Al-Mayadeen TV	Arabic
Morocco	Al Aoula TV	Arabic
Libya	Al-Naba TV	Arabic
Egypt	Nile News TV	Arabic
Russia	NTV Mir	Russian
Oman	Oman Television	Arabic
Ukraine	One Plus One TV	Ukrainian
Palestinian Territory, Occupied	Palestinian Satellite Channel TV	Arabic
Iran	Press TV	English
Qatar	Al-Jazeera TV	Arabic
Qatar	Qatar TV	Arabic
Russia	RTR Planeta TV	Russian
Russia	RT	English
Russia	Rusiya al-Yawm	Arabic
Yemen	Yemen TV	Arabic
Yemen	Republic of Yemen TV	Arabic
United Arab Emirates	Sky News Arabia	Arabic
Syria	Al-lkhbariyah al-Suriyah TV	Arabic
Syria	Syria TV	Arabic
Tunisia	National Tunisian TV	Arabic
Libya	Libya (Al-Ahrar) TV	Arabic
,	ICTV television	Ukrainian
Ukraine		Jan an null
Ukraine Ukraine	5 Kanal TV	Ukrainian
Ukraine Ukraine Russia	5 Kanal TV Rossiya 24 news channel	Ukrainian Russian

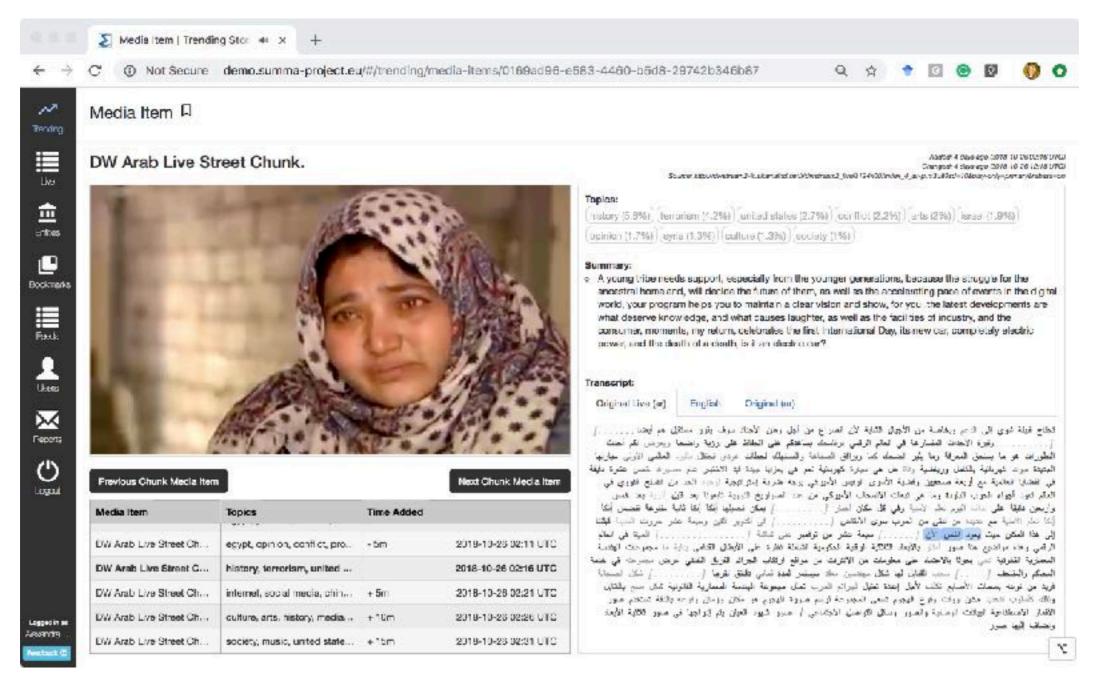
# **SUMMA** Pipeline













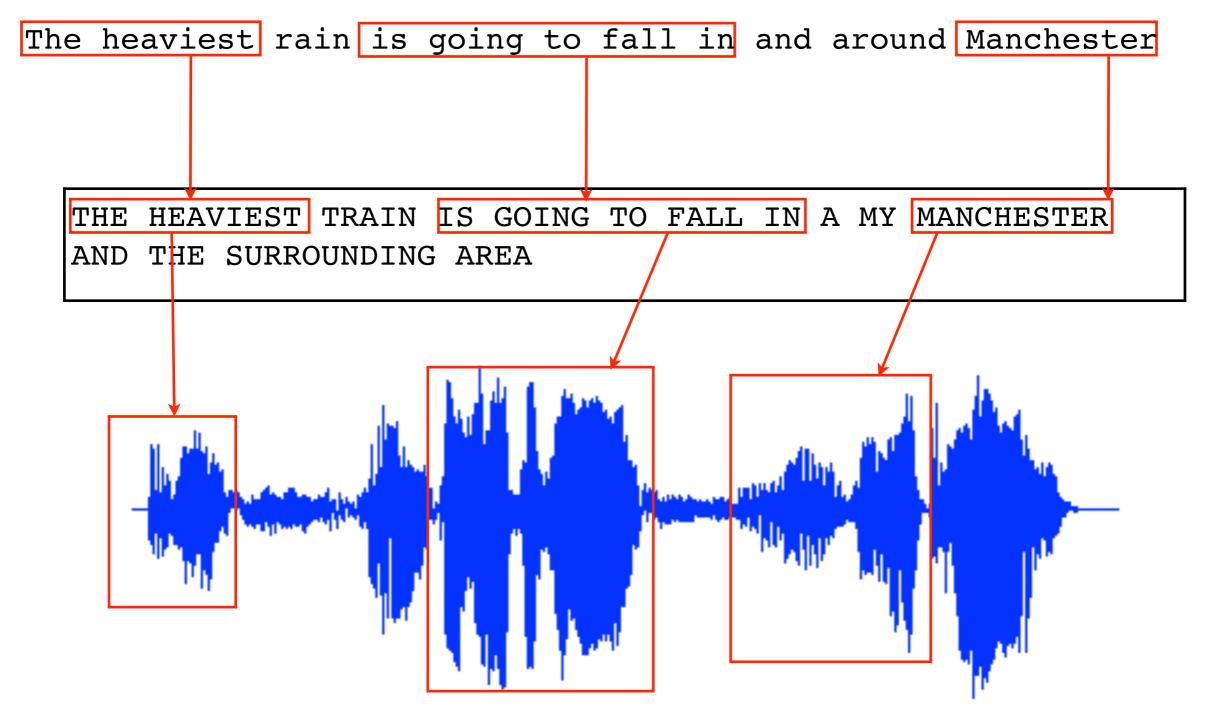
#### ASR across many languages

Language	Train Set	Train Size	WER Indomain	WER SUMMA
English	Multi-Genre Broadcast Challenge	1600 hours	26.1	
Arabic	Arabic Multi-Dialect Broadcast Challenge	1200 hours	14.7	
German	BNC radio news	160 hours	9.9	34.6
Spanish	Global Phone	17 hours	13.2	
Russian	Euronews	60 hours	14.4	

Added: Latvian, Ukranian, Farsi, and Portuguese



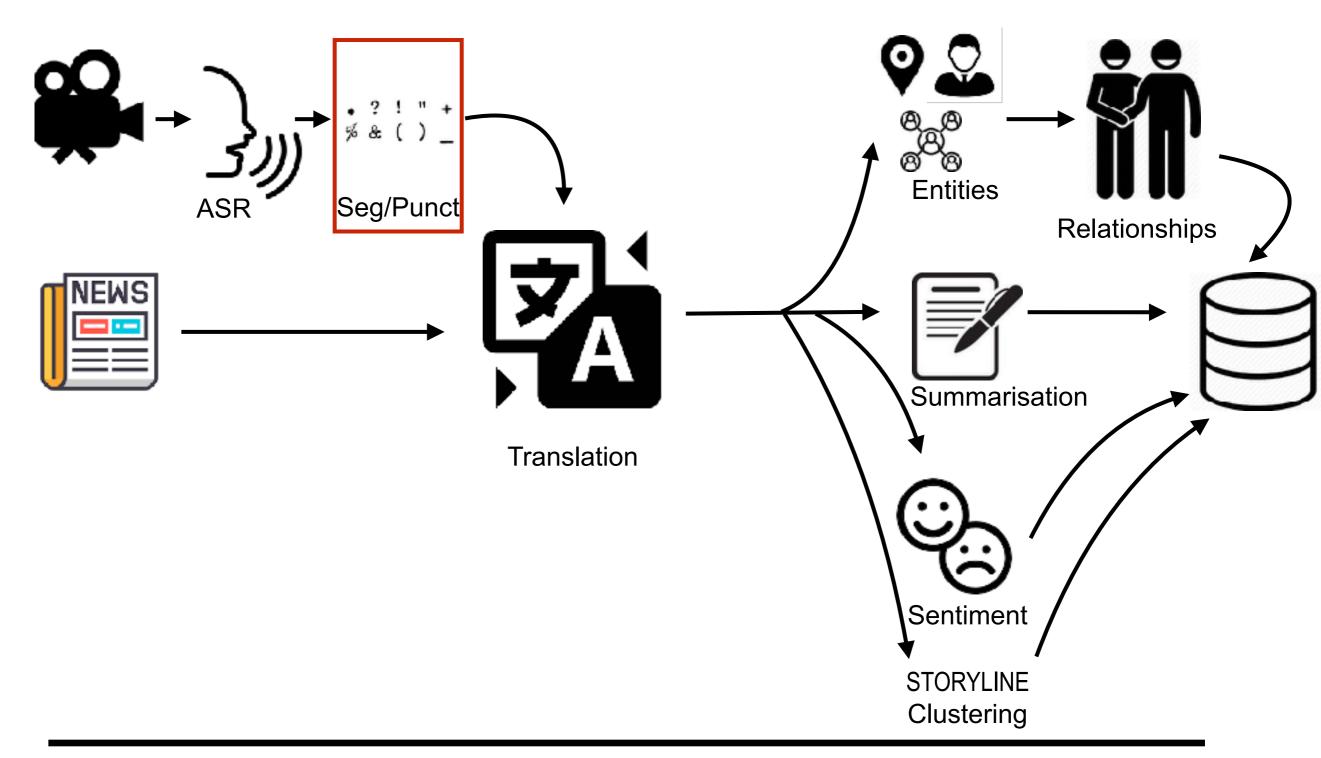
# Training ASR Using Subtitles Light supervision



Matching Error Rate used as cut-off

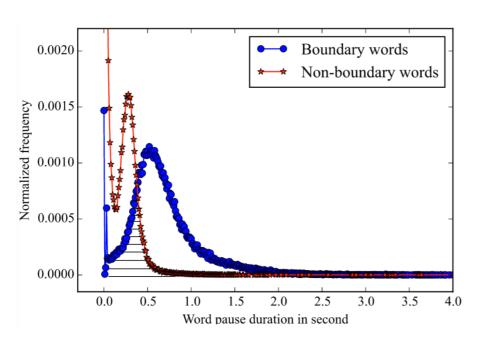
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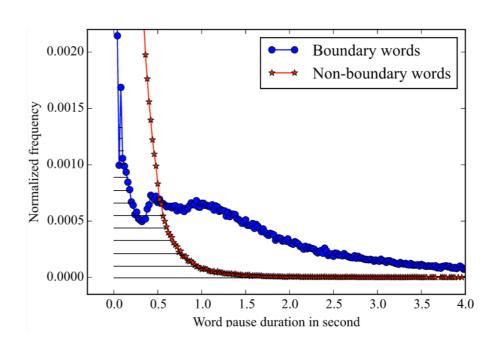












Pause duration distribution on German BCN corpus (left) and English MGB corpus (right)

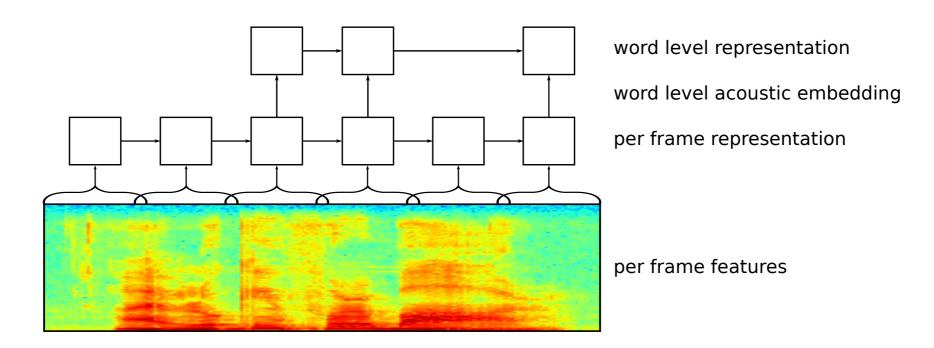
Yang Wang, Alexandre Nanchen, Alexandros Lazaridis, David Imseng, and Philip N.Garner. Comparative study on sentence boundary prediction for german and english broadcast news. Idiap-RR Idiap-RR-18-2017, Idiap, 7 2017











Hierarchical encoder

#### In practice:

- use pause segmentation model to segment on long pauses

transcription combining lexical and acoustic features. ICASP, April 2017



Ondrej Klejch, Peter Bell, and Steve Renals. Sequence-to-sequence models for punctuated



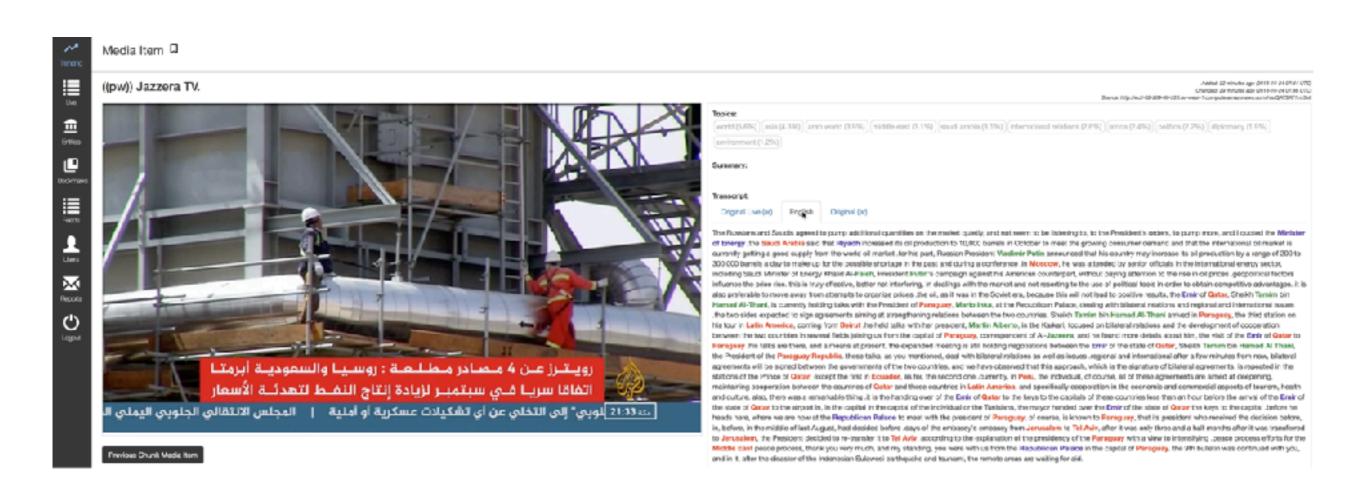
# SUMMA NLP Pipeline





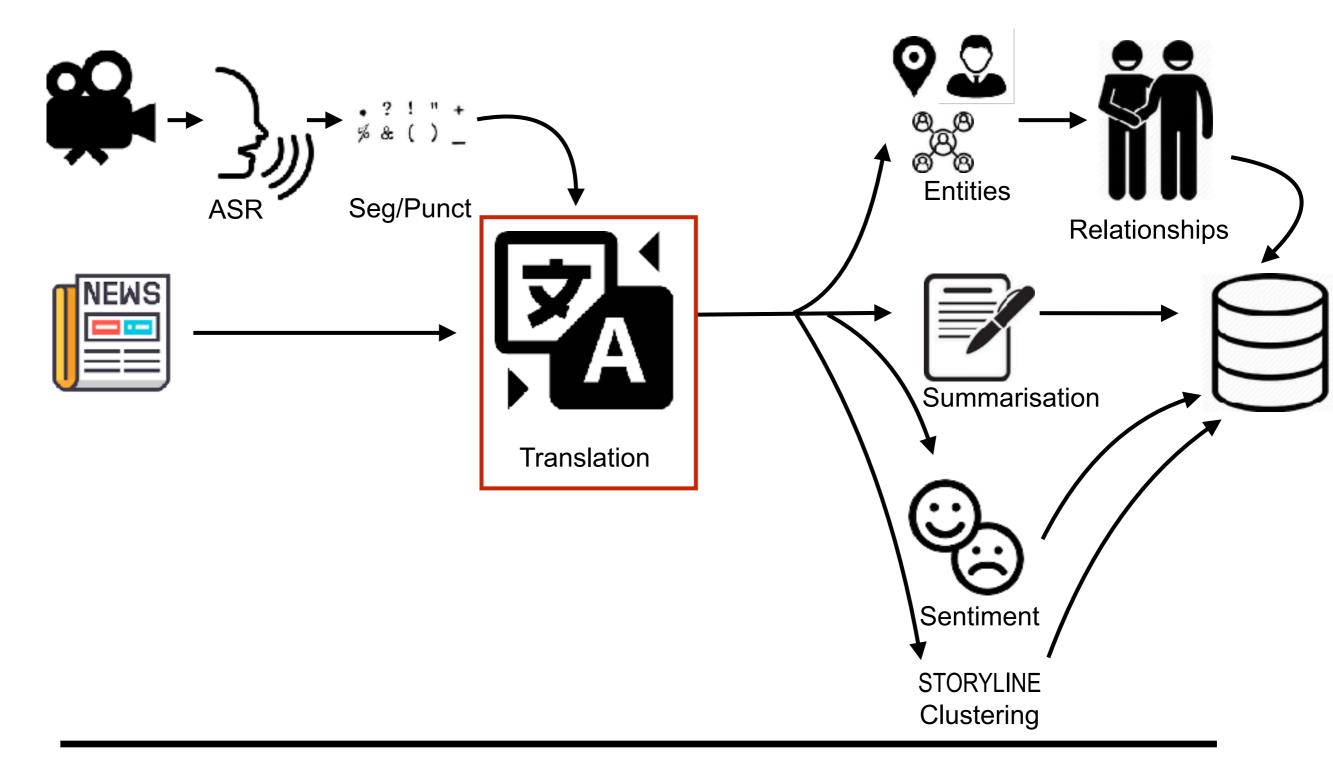
#### SUMMA NLP Pipeline





# **SUMMA** Pipeline









- Efficient, fast translation of large volumes
- High quality translation
- Story level translation
- Robust to different dialects
- Evaluation of for gisting





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#### Marian



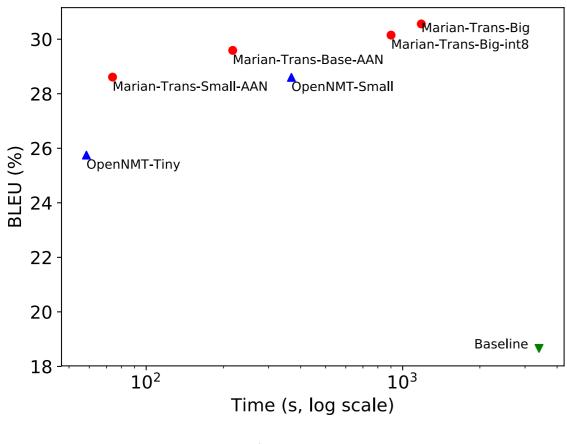
- Portable C++11 code with minimal dependencies (only CUDA and Boost)
- Custom auto-diff engine with dynamic graphs (similar to DyNet)
- Optimized towards NMT
- Multi-device, multi-node training, and decoding (GPU and CPU)
- Websites:
  - http://marian-nmt.github.io and
  - https://github.com/marian-nmt/marian

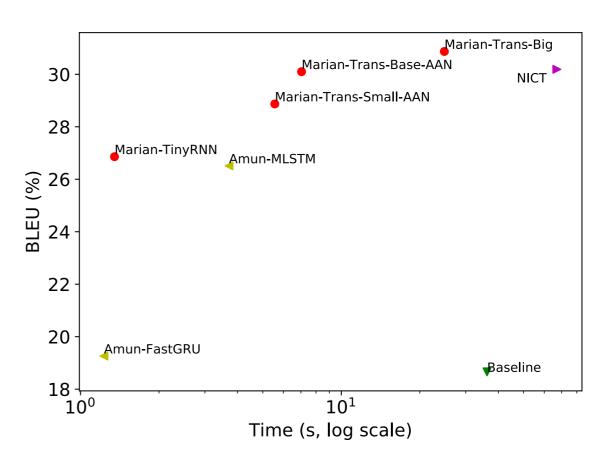
Marcin Junczys-Dowmunt, Roman Grundkiewicz, Tomasz Dwojak, Hieu Hoang, Kenneth Heafield, Tom Neckermann, Frank Seide, Ulrich Germann, Alham Fikri Aji, Nikolay Bogoychev, André F. T. Martins, and Alexandra Birch

Marian: Fast Neural Machine Translation in C++. ACL 2018



#### WNMT 2017: Shared task on efficiency





(a) CPU Time vs. Accuracy

(b) GPU Time vs. Accuracy

Alexandra Birch, Andrew Finch, Graham Neubig, and Yusuke Oda (2018). Findings of the Second Workshop on Neural Machine Translation and Generation. ACL





- Marian 169
- OpenNMT 89
- Tensor2tensor (+T2T) 59
- Sockeye 47
- Nematus 42
- Fairseq 8

Source: Marcin Junczys-Dowmunt



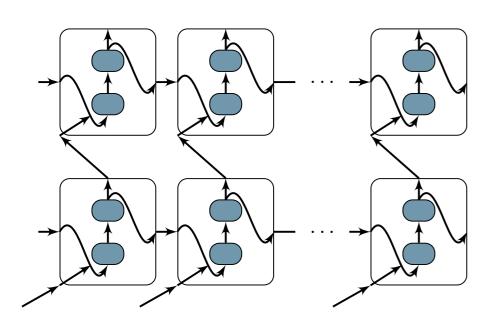


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#### Deeper Models



#### 2017 WMT News Task



Bi-deep

Constrained news task, including ties:

	From English	lo English
Turkish	Edinburgh	Edinburgh
Czech	Edinburgh	Edinburgh
Chinese	Edinburgh	Edinburgh
German	Munich	Edinburgh
Russian	Edinburgh	National Research Council
Latvian	Tilde	Edinburgh
Finnish	Helsinki	UPC

(Edinburgh did not participate in Finnish)

Antonio Valerio Miceli Barone, Jindrich Helcle, Rico Sennrich, Barry Haddow and Alexandra Birch. "Deep Architectures for Neural Machine Translation". In Proceedings of WMT, 2017

Rico Sennrich, Alexandra Birch, Anna Currey, Ulrich Germann, Barry Haddow, Kenneth Heafield, Antonio Valerio Miceli Barone, Philip Williams (2017). The University of Edinburgh's Neural MT Systems for WMT17





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#### **Source:**

context: Oh, I hate **flies**. Look, there's another one!

current sent.: Don't worry, I'll kill it for you.

#### **Target:**

1 context: Ô je déteste les **mouches**. Regarde, il y en a une autre!

correct: T'inquiète, je **la** tuerai pour toi. incorrect: T'inquiète, je **le** tuerai pour toi.

meorreet. I inquiete, je te tuerar pour tor.

2 context: Ô je déteste les **moucherons**. Regarde, il y en a un autre!

correct: T'inquiète, je **le** tuerai pour toi. incorrect: T'inquiète, je **la** tuerai pour toi.

Rachel Bawden, Rico Sennrich, Alexandra Birch, and Barry Haddow (2018). Evaluating Discourse Phenomena in Neural Machine Translation. NAACL

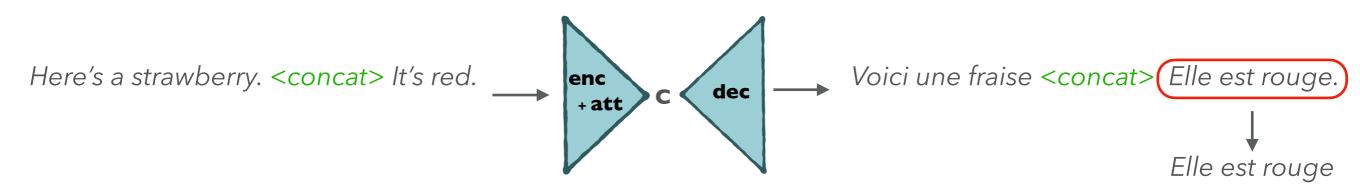
Alexandra Birch

MT and the Media

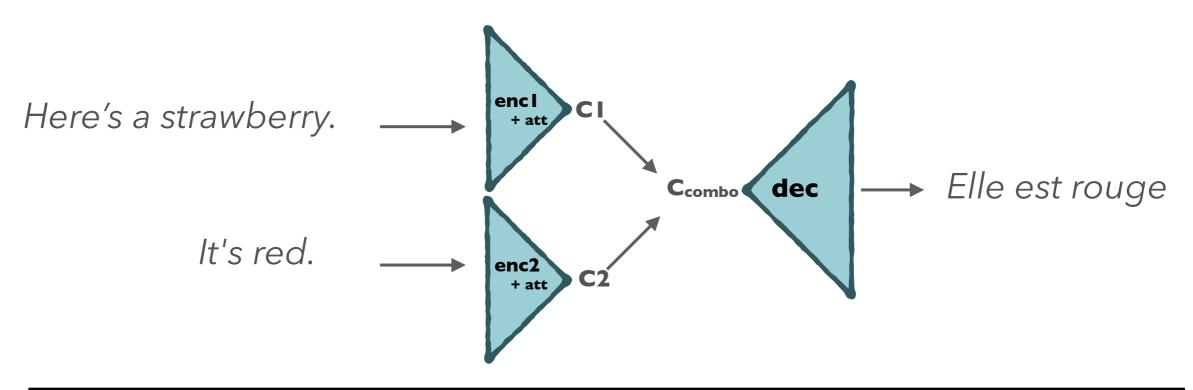


## Modelling Context

#### **Concatenated Model:**



#### Multi-source Model:







	BLEU	Co-reference	<b>Ambiguity</b>
Baseline	19.52	50.0	50.0
Multis. (prev. source)	20.22	50.0	53.0
Multis. (prev. target)	17.89	47.0	50.5
Concat. $(src + tgt)$	20.09	63.5	52.0
Multis. and Concat. (tgt)	20.85	72.5	57.0

Rachel Bawden, Rico Sennrich, Alexandra Birch, and Barry Haddow (2018). Evaluating Discourse Phenomena in Neural Machine Translation. NAACL

Alexandra Birch MT and the Media 40

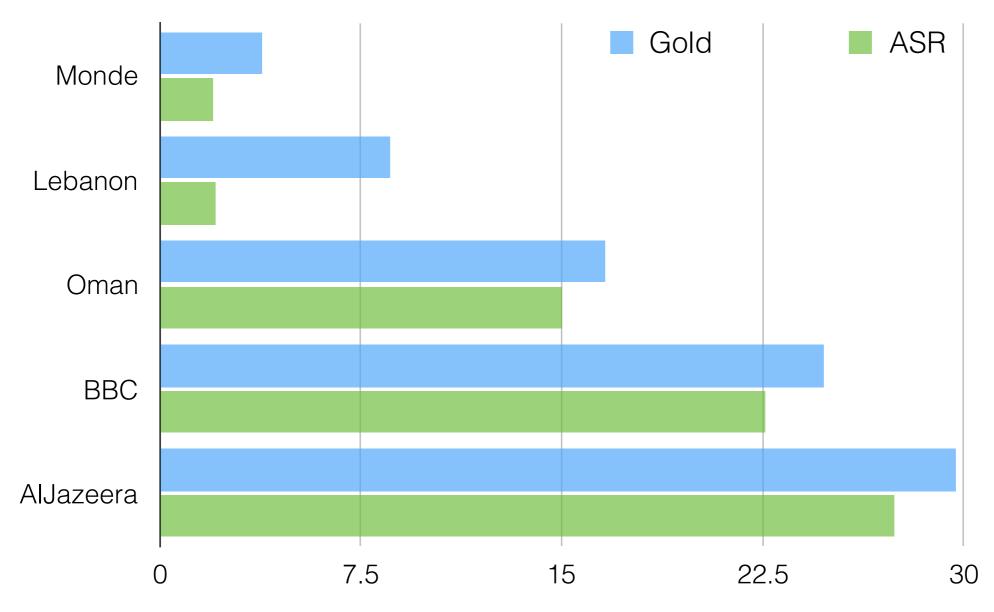




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### Dialectal Translation Arabic



BLEU Score: Arabic-English SUMMA Test Set





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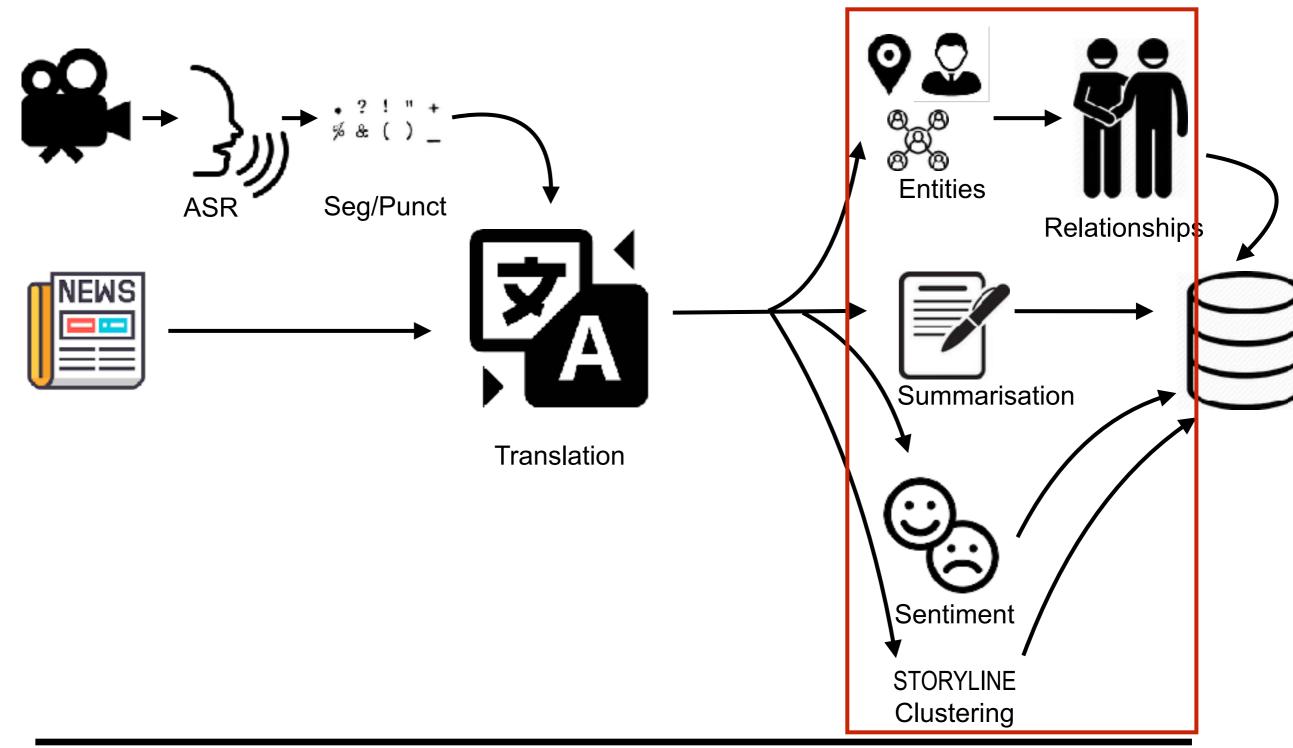


		t the bottom with the most fitting <b>single word</b> , using only infor	matio
from the <i>hint text</i> (if there	e is one).		
Hint text: (you might ne	ed to scroll to find some highlighted text)		
,,		conomic boom, which was the economic basis for a stable	
democracy.		•	
_	atic Republic the socialist one-party dic	tatorship of the SED and the socialist planned economy h	ave
been introduced at the	same time.		
Until 1989, the GDR had	d therefore great economic problems.		
The consequences had	a major impact on life in the GDR.		,
The consequences had	a major impact on life in the GDR.		<b>,</b>
4		epublic , the socialist one-party dictatorship of the SED and	, }
4		epublic , the socialist one-party dictatorship of the SED and were introduced .	, }

Mikel Forcada, Carolina Scarton, Lucia Specia, Barry Haddow, and Alexandra Birch. Exploring Gap Filling as a Cheaper Alternative to Reading Comprehension Questionnaires when Evaluating Machine Translation for Gisting, WMT 2018

# SUMMA Pipeline





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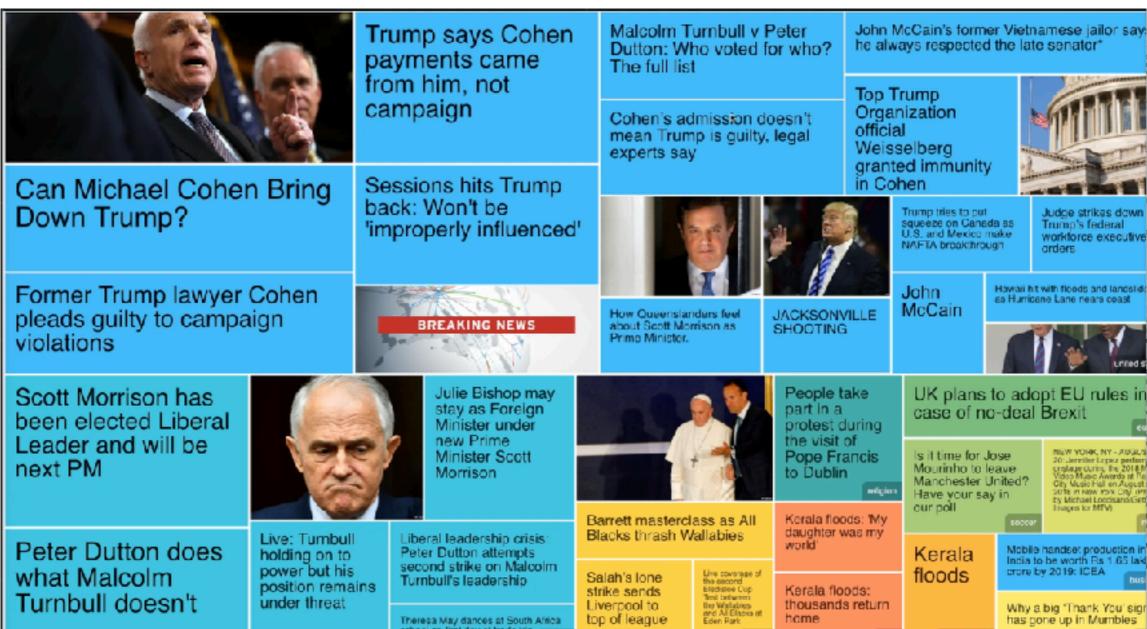
## Multilingual Clustering

under threat



Why a big 'Thank You' sign

has gone up in Mumbles



Miranda, S., Znotinš, A., Cohen, S. B., & Barzdins, G. (2018). Multilingual Clustering of Streaming News. EMNLP 2018

thousands return

Theresa May dances at South Africa

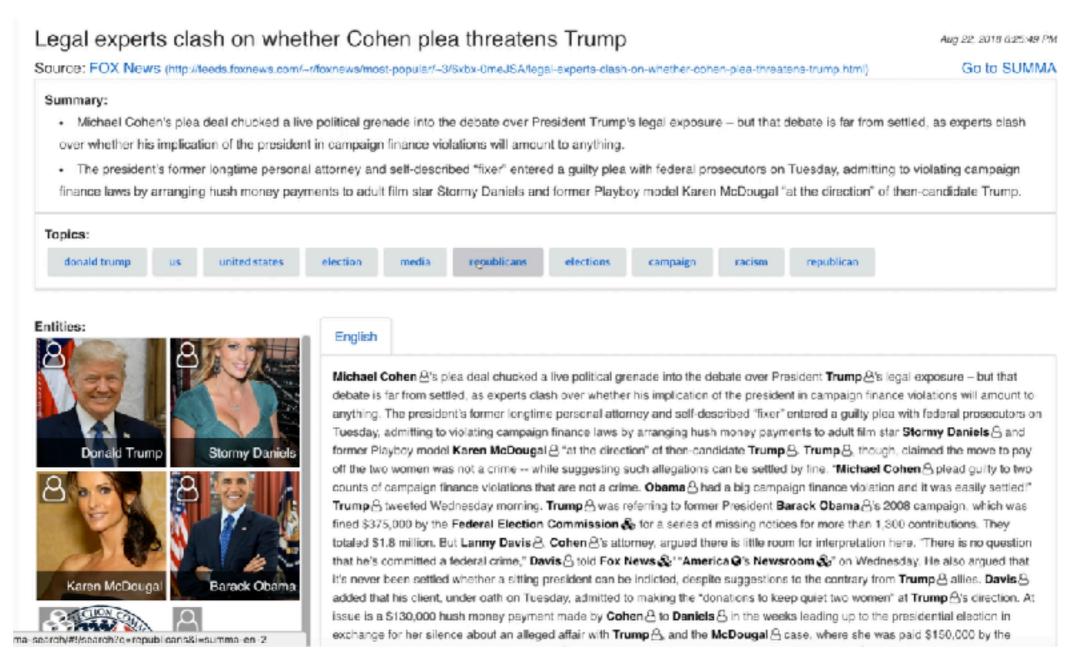
school on first day of trade trip

Liverpool to

top of league

## Multilingual Named Entity detection





Afonso Mendes, David Nogueira, Samuel Broscheit, Filipe Aleixo, Pedro Balage, Rui Martins, Sebastiao Miranda, and Mariana S. C. Almeida SUMMA at TAC Knowledge Base Population Task 2017

### Ethics of SUMMA



- Legal Compliance GDPR (EU General Data Protection Regulation) May 2018
- Ethics Board:
  - → Professor Mireille Hildebrandt, Vrije Universiteit Brussels
  - → Dr Els Kindt, KU Leuven Centre for IT & IP Law (CiTiP)
  - ◆ Dr Julia Powles, Cornell Tech / University of Cambridge
  - → Professor Lorna Woods, Essex University
- Protection of Personal Data and Privacy by Design: Social Media
  - Public vs private figures
  - Minimize risk: no videos or images
  - Aggregation of posts
  - ◆ No transfer of tweets between partners, only links

### **Broader Ethical Issues**



- Open Source Project:
  - ◆ Dual Use: Military application
  - → Research tool: Politics, Social Sciences
- Job security and automation:
  - → More resilient to existing economic pressures
  - Augmentation not automation
- Media Bubble:
  - Allows us to access global stories and viewpoints

### **GoURMET**













1 January 2019 – 31 December 2021





### **BBC World Service**

2017: Launched 12 new languages - Afaan Oromo, Amharic, Gujarati, Igbo, Korean, Marathi, Pidgin, Punjabi, Serbian, Telugu, Tigrinya, and Yoruba.

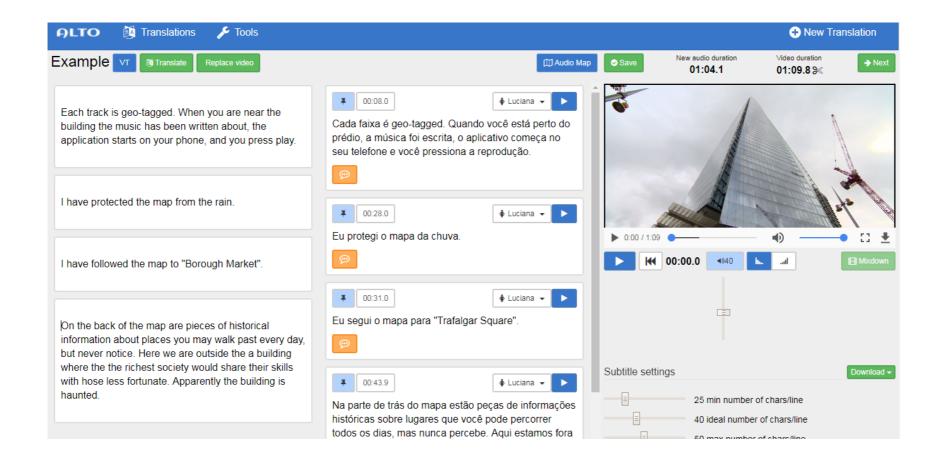


Goal: Reach 500 Million people by 2022





#### Efficient content creation



Research problem: low-resource machine translation



## Summary

- How can NLP help the Media?
  - Unlocking more content from different media types/languages
  - Making sense of deluge of data
  - Reach a wider audience
- How can we deal with multilingual content?
  - Translation: Faster, deeper, with context
  - Multilingual NLP
- Context of EU Projects:
  - SUMMA: Automated Media Monitoring Platform
  - Gournett: Low-resource Machine Translation



## Thank you

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