NLP and social media:

Language Modelling, Benchmarking and Temporal Challenges

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Manchester, 6 February 2024



Outline



- Specialized language models in social media
- > Evaluation
- > Temporal challenges and applications







- Specialized language models in social media
- > Evaluation
- Temporal challenges and applications





About me

- Professor at Cardiff University (Wales, UK)
 - **UKRI Future Leaders Fellow** (£1.4M funding, 4+ years)
 - Co-founder and head of the **Cardiff NLP group**.
- Areas of expertise: Semantics, resources, multilinguality, social media
 - Co-author of "Embeddings in NLP" book
 - General chair of *SEM-2024





Cardiff NLP



- > Young group (3 years old), growing fast (30+ lab members)
- ➤ Website: <u>cardiffnlp.github.io</u> ⊕
- > Activities: hybrid seminars, workshops, hackathons, etc.
- Twitter: <u>@Cardiff_NLP</u>
- > Open-source contributions



Cardiff NLP Workshop (1-2 July 2024)



`ARDIFF

UNIVERSITY

PRIFYSGOL

Specialized Language Models

Language models (LMs)



Specialized language models

- \succ LMs can be specialized by:
 - augmenting with external information
 - pretraining on domain-specific corpora

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Specialized language models

Pretraining on domain-specific corpora



image credit: Gu, Y., Tinn, R., Cheng, H., Lucas, M., Usuyama, N., Liu, X., ... & Poon, H. (2021).
Domain-specific language model pretraining for biomedical natural language processing. ACM Transactions on Computing for Healthcare (HEALTH), 3(1), 1-23.

 \succ Why?

- ➤ Why?
 - Informal grammar
 - **Multilingual** (code-switching, etc.)
 - Irregular vocabulary
 - Emoji 😀, abbreviations, typos, hashtags, mentions...
 - Tweets are often **not standalone messages**
 - RTs, mentions, replies, threads, pictures...
 - And because social media is **important**



Haaland! That was <mask>





TweetEval:

Language Models and Evaluation Benchmark



TweetEval, the language model (Barbieri et al. EMNLP Findings 2020)

- ≻ How?
 - RoBERTa architecture
 - Continue from RoBERTa checkpoint vs from scratch (BERTweet is from scratch)

(Barbieri et al. EMNLP Findings 2020)



\succ Why?



➤ Why?





➤ What?

Task	Lab	Train	Val	Test
Emoji prediction	20	45,000	5,000	50,000
Emotion rec.	4	3257	374	1421
Hate speech det.	2	9,000	1,000	2,970
Irony detection	2	2,862	955	784
Offensive lg. id.	2	11,916	1,324	860
Sent. analysis	3	45,389	2,000	11,906
Stance detection	3	2620	294	1249
Stance/Abortion	3	587	66	280
Stance/Atheism	3	461	52	220
Stance/Climate	3	355	40	169
Stance/Feminism	3	597	67	285
Stance/H. Clinton	3	620	69	295

(Antypas et al. EMNLP Findings 2023)

> What?

Task (Dataset)	Train	Valid.	Test
TWEETNER7	4,616	576	2,807
TWEETEMOTION	6,838	886	3,259
TWEETQG	9,489	1,086	1,203
TWEETNERD	20,164	4,100	20,075
TWEETSENTIMENT	26,632	4,000	12,379
TEMPOWIC	1,427	395	1,472
TWEETEMOJI100	50,000	5,000	50,000
TWEETINTIMACY	1,191	396	396
TWEETQA	9,489	1,086	1,203
TWEETTOPIC	4,585	573	1,679
TWEETHATE	5,019	716	1,433
TWEETSIM	450	100	450



An extended and more challenging benchmark in the age of LLMs!

(Antypas et al. EMNLP Findings 2023)

12 diverse NLP tasks



Task (Dataset)	Example Input	Example Output
NER (TWEETNER7)	Tweet: Winter solstice 2019 : A short day that 's long on ancient traditions url via @CNN_Travel	Winter solstice 2019: event @CNN_Travel: product
Emotion Classification (TWEETEMOTION)	Tweet: Whatever you decide to do make sure it makes you #happy.	joy, love, optimism
Question Generation (TWEETQG)	Tweet: 5 years in 5 seconds. Darren Booth (@darbooth) January 25, 2013 Context: vine	what site does the link take you to?
Name Entity Disambiguation (TWEETNERD) Tweet: hella excited for ios 15 because siri reads notifications out loud to you [] Target: siri Definition: intelligent personal assistant on various Apple devices		True
entiment Classification Tweet: #ArianaGrande Ari By Ariana Grande 80% Full url #Singer #Actress url TWEETSENTIMENT) Target: #ArianaGrande		negative or neutral
Meaning Shift Detection TEMPOWIC) Tweet 1: The minute I can walk well I'm going to delta pot Tweet 2: Then this new delta variant out im vaccinated but stillIll likeee' Target: delta		False
Emoji Classification (TweetEmon100)	Tweet: SpiderMAtS back at it	ð
Intimacy Analysis (TWEETINTIMACY)	Tweet: @user SKY scored 4 less runs just lol	1.20
Question Answering TWEETQA) Tweet: 5 years in 5 seconds. Darren Booth (@user) January 25, 2013 Question: which measurements of time are mentioned?		years and seconds
Topic Classification (TWEETTOPIC)	Tweet: Sweet, #IOWAvsISU is a nationally televised night game! Nebraska getting bumped to @FOX_Business is just a bonus.	film_tv_&_video, sports
Hate Speech Detection (TWEETHATE)	Tweet: Support Black Trans youth url	not_hate
Tweet Similarity (TWEETSIM)	Tweet 1: I wish kayvee all the best #bbnaija Tweet 2: Sammie about to cry to the housemates all night #bbnaija	2.33

(Antypas et al. EMNLP Findings 2023)

🤗 Already available at

huggingface.co/datasets/cardiffnlp/super_tweeteval

Includes generative, regression and classification tasks

Also tasks with **temporal splits**!

Results? Smaller specialized models with supervision still better than LLMs (including ChatGPT)

Specialized language models (+fine-tuned)

Models 200 Q

1↓ Sort: Most downloads

cardiffnlp/twitter-roberta-base-sentiment-latest 3 Text Classification • Updated May 28, 2023 • ± 43.8M • ♥ 300

cardiffnlp/twitter-roberta-base-sentiment ∰ Text Classification + Updated Jan 20, 2023 + ± 1.68M + ♥ 232

& cardiffnlp/twitter-roberta-base-offensive ﷺ Text Classification → Updated Nov 28, 2022 → 🖄 473k → ♡ 13

4 cardiffnlp/twitter-xlm-roberta-base-sentiment-multi... 🚓 Text Classification + Updated Dec 1, 2022 + ± 41.8k + ♡ 5

cardiffnlp/twitter-roberta-base-emotion

☆ Text Classification + Updated May 28, 2023 + ± 17.6k + ♥ 38

cardiffnlp/twitter-roberta-base-irony Text Classification + Updated Aug 2, 2023 + ± 10.6M + ♥ 16

cardiffnlp/twitter-xlm-roberta-base-sentiment a∰ Text Classification + Updated Jul 19, 2023 + ± 839k + ♡ 161

6 cardiffnlp/tweet-topic-21-multi # Text Classification + Updated May 28, 2023 + ± 52.4k + ♥ 54

cardiffnlp/twitter-xlm-roberta-base □ Fill-Mask + Updated Aug 31, 2023 + ± 19.5k + ♥ 12

cardiffnlp/twitter-roberta-base-hate all Text Classification → Updated Apr 19, 2023 → ± 5.96k → ♡ 12



XLM-T: Multilingual Language Model for Twitter

XLM-T: Multilingual Twitter LM

- ➤ Why?
 - Same as in TweetEval: no multilingual LMs, and no unified multilingual benchmarks

XLM-T: Multilingual Twitter LM

➤ What?



XLM-T: Multilingual Twitter LM

- ≻ How?
 - XLM-R architecture
 - Continue from XLM-R checkpoint

Updated large-size 2023 model now available!

TimeLMs: Diachronic Language Models

Temporal challenges in NLP

Language is **changing** all the time.

New terms being introduced (e.g. *COVID-19*) or terms acquired new meanings (e.g. *Karen*).

Popular models are **old** (e.g. BERT, 2018).

This is especially true in **social media**, which is very dynamic.

Temporal challenges in NLP

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Solution?

TimeLMs initiative

(Loureiro et al. ACL Demo 2022)

Commitment to train **new language models** every quarter (three months)

Twitter as our training corpus 🔰

RoBERTa models from 2019 until now already available

Language models improve over time

Models	2020-Q1	2020-Q2	2020-Q3	2020-Q4	2021-Q1	2021-Q2	2021-Q3	2021-Q4	Change
Barbieri et al., 2020	9.420	9.602	9.631	9.651	9.832	9.924	10.073	10.247	N/A
2019-90M	4.823	4.936	4.936	4.928	5.093	5.179	5.273	5.362	N/A
2020-Q1	4.521	4.625	4.699	4.692	4.862	4.952	5.043	5.140	-
2020-Q2	4.441	4.439	4.548	4.554	4.716	4.801	4.902	5.005	-4.01%
2020-Q3	4.534	4.525	4.450	4.487	4.652	4.738	4.831	4.945	-2.15%
2020-Q4	4.533	4.524	4.429	4.361	4.571	4.672	4.763	4.859	-2.81%
2021-Q1	4.509	4.499	4.399	4.334	4.439	4.574	4.668	4.767	-2.89%
2021-Q2	4.499	4.481	4.376	4.319	4.411	4.445	4.570	4.675	-2.83%
2021-Q3	4.471	4.455	4.335	4.280	4.366	4.394	4.422	4.565	-3.26%
2021-Q4	4.467	4.455	4.330	4.263	4.351	4.381	4.402	4.463	-2.24%
2021-124M	4.319	4.297	4.279	4.219	4.322	4.361	4.404	4.489	N/A

Perplexity-based evaluation

Language models improve over time

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TimeLMs

Sample predictions with models trained at different time periods.

Model	So glad I'm <mask> vaccinated.</mask>	I keep forgetting to bring a <mask>.</mask>	Looking forward to watching <mask> Game tonight!</mask>
1	not	bag	the
2020-Q1	getting	purse	The
	self	charger	this
() 	not	mask	The
2020-Q2	getting	bag	the
	fully	purse	End
	not	mask	the
2020-Q3	getting	bag	The
	fully	purse	End
2.	not	bag	the
2020-Q4	getting	purse	The
	fully	charger	End
8	getting	purse	the
2021-Q1	not	charger	The
	fully	bag	End
8	fully	bag	the
2021-Q2	getting	charger	The
	not	lighter	this
8	fully	charger	the
2021-Q3	getting	bag	The
	not	purse	This
2	fully	bag	Squid
2021-Q4	getting	lighter	the
	not	charger	The

TimeLMs

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2020-Q1	getting	purse	The
	self	charger	this
1	not	mask	The
2020-Q2	getting	bag	the
	fully	purse	End
	not	mask	the
2020-Q3	getting	bag	The
	fully	purse	End
25	not	bag	the
2020-Q4	getting	purse	The
	fully	charger	End
55 	getting	purse	the
2021-Q1	not	charger	The
	fully	bag	End
8	fully	bag	the
2021-Q2	getting	charger	The
	not	lighter	this
8	fully	charger	the
2021-Q3	getting	bag	The
	not	purse	This
2	fully	bag	Squid
2021-Q4	getting	lighter	the
	not	charger	The

NER and Topic Classification

(Antypas et al. COLING 2022; Ushio et al. AACL 2022)

Two datasets with temporal splits (i.e. training and test sets from different time periods):

- TweetNER7 (Ushio et al. 2022) for NER
- **TweetTopic** (Antypas and Ushio et al. 2022) for topic classification

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Conclusion: Performance on temporal test splits lower than when dates are shuffled.

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Conclusion: Performance on temporal test splits lower than when dates are shuffled.

LongEval series at CLEF to evaluate performance drop over time

Temporal challenges: Ongoing work

- Analysis of main sources of performance drop (ongoing):
 - pre-training data?
 - training data?
 - o nature of the domain/task?

Development of new methods to integrate contextual information or non-labeled data in predictive models



Ushio et al. (AACL 2022)

TweetNLP (<u>tweetnlp.org</u>)

		About	Get Started	Docs	News	Contact	Insights	Demo	
Get insights using cutting-edge NLP models specialised in social media! Sentiment Analysis Demo: Type a sentence or add a tweet link								Cardim Nup	
CARDIFF UNIVERSITY PRIFYSGOL (ARD)(P) Sum Inc	0		UNIVERSITY OF						

DE GRANADA

Snap Inc.

TweetNLP - the team grows!



Francesco Barbieri Contributor

Snap



Asahi Ushio Contributor Cardiff University



Luis Espinosa-Anke Contributor Cardiff University & Amplyfi





Kiamehr Rezaee Backend Developer

Cardiff University





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Leonardo Neves Contributor

Snap



Fangyu Liu Contributor

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Cambridge University



Joanne Boisson

Tester

Cardiff University

TweetNLP (Camacho-Collados et al. EMNLP Demo 2022)

A platform for **NLP specialised on social media**.

Integration of all resources with relatively small models.

NLP **applications** from sentiment analysis to hate speech detection and NER.

Demo, tutorials and Python API.



TweetNLP Python library

Includes pre-trained models, inference, fine-tuning, evaluation...

```
import tweetnlp
# ENGLISH MODEL
model = tweetnlp.load_model('sentiment') # Or `model = tweetnlp.Sentiment()`
model.sentiment("Yes, including Medicare and social security saving '' # Or
>>> {'label': 'positive'}
```





Sentiment analysis

Type a sentence or a tweet to get insights (tweet URLs are also accepted)

Predictions are based on an English or a multilingual model. Languages supported are:

Today is a lovely day! 😊

GO!

For Example: <u>Today is a lovely day!</u>, <u>I really don't like eating vegetables</u>, <u>https://twitter.com/Cardiff_NLP/status/1485518987807137792</u>





Word prediction

Sentence/Tweet Classification	Type a sentence or a tweet with a masked word (<mask>) to predict the most likely wo every three months to select from.</mask>	ord.
Hashtag Analysis	Haaland! That was <mask> For Example: Lkeep forgetting to bring a <mask>., COVID is a <mask>., So glad I'm <mask> vaccinated, Looking</mask></mask></mask></mask>	-57
Word Prediction	Latest June 2022 - English V GO!	
Sentence/Tweet similarity	fast (5%)	
Named Entity Recognition	quick (4%) it (3%)	
Question	close (3%)	ATR WAYS
Answering/Generation	amazing (2%)	

Word prediction (different years)

2020 model

2021 model



Looking forward to watching <mask> Game tonight!</mask>
For Example: <u>I keep forgetting to bring a <mask>.</mask></u> , COVID is a <mask>., So glad I'm <mask> vaccinated, Looking forward to watching <mask> Game tonight!</mask></mask></mask>
2021 - English ~
GO!
Squid (34%)
the (23%)
The (15%)
End (2%)
this (1%)



Topic classification

(Antypas and Ushio et al. COLING 2022)

Predictions are based on English (all tasks) or a multilingual model (continent). Languages supported are:	
reactions are based on English (all lasks) of a multilingual model (semiment). Languages supported are:	
https://twitter.com/livescore/status/1632652988228710402 For Example: Today is a lovely day!, I really don't like eating vegetables, https://twitter.com/Cardiff_NLP/status/1485518987807137792 Topic classification	gaming (2%) news & social concern (1%) celebrity & pop culture (1%) diaries & daily life (1%)
GO! GO!	

Named Entity Recognition (NER) (Ushio et al. AACL 2022)



Question answering & generation (Ushio et al. EMNLP 2022, ACL Findings 2023)

Jarry, who won yesterday in Cincinnati, withdrew from the tournament to follow the birth o

For Example: <u>Beyonce further expanded her acting career, starring as blues singer Etta James in the 2008 musical biopic, Cadillac Records.</u>

	who withdrew	from the	tournament?
--	--------------	----------	-------------

For Example: what did beyonce do in the musical biopic?

Output

jarry



Tweet Insights tweetnlp.org/insights



Tweet Insights (Loureiro et al. 2023)



Other social media applications

Politics, sentiment and virality



Dimosthenis Antypas



Alun Preece



Crime and Security Research Institute

Sefydliad Ymchwil Trosedd a Diogelwch

Negativity spreads faster: A large-scale multilingual twitter analysis on the role of sentiment in political communication (*Online Social Networks and Media* journal, 2023)

Politics, sentiment and virality

Collected a corpus of Twitter messages from MPs in **Greece, Spain and UK** (focus on 2021, 400K tweets)

Analysed the relation between **sentiment** (as provided by our Twitter-based models) and **virality** (measured by number of retweets and other metrics)

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Sentiment of MPs' tweets



Viral tweets are overwhelmingly negative

Sentiment of Spanish MP tweets





Popular tweets (top 5%)

Unpopular tweets (bottom 35%) Viral tweets (top 5%)



71% are negative!



Negative

Sentiment over time



Tweets by MPs are becoming more negative over time (UK)

Government vs. opposition (Spain)



MPs from the government party are **more positive and less negative** -> *this also holds in other countries with different ideologies*

Government vs. opposition (Spain)



MPs from the government party are **more positive and less negative** -> *this also holds in other countries with different ideologies*

Government vs. opposition (UK)





MPs from the government party are **more positive and less negative**

Government vs. opposition (Spain, 2021)



Pedro Sánchez Prime Minister 13% vs. 63% Negative tweets



Pablo Casado Leader of the opposition

Government vs. opposition (UK, 2021)



5% vs. 35% Negative tweets



Keir Starmer Leader of the opposition

Boris Johnson Prime Minister

Work in progress (other multidisciplinary collaborations)

- Analysing disinformation at scale (with a special focus on textual content)
- Early health care interventions: depression detection on social media (Twitter, Reddit)
- Analysing outbreaks and adherence/sentiment to health interventions (e.g. COVID) using social media

Conclusion

Social media entails many challenges.

Specialized language models are a good solution to domain-specific problems (plus: no need for huge models!)

Temporal adaptation is needed, and can only be partially solved with updated models.

Applications are endless, huge opportunities for NLPers.



Summary of resources TweetNLP





All models available in the Hugging Face hub:

https://huggingface.co/cardiffnlp 🤗

Thank you!

