

NLP for German CMC Texts

Tokenization, POS Tagging, and
a New Gold Standard for Lemmatization

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PHILOSOPHISCHE FAKULTÄT
UND FACHBEREICH THEOLOGIE

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 - Tokenization
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EmpiriST 2015 shared task

Beißwenger et al. (2016)

- Computer-mediated communication (CMC) poses challenges to NLP tools:
 - ▶ `Eveeventuell` suchte ich gerade das Spiel :D
 - ▶ Ein Distributor darf `zB` Verlag des gemeinsamen Besitzers auch nicht `ggüber` anderen `Verlägen` bevorzugen.
 - ▶ `@Montaaag #semibk` DieseFrage spiele ich zurück `indieRunde`: Zu `welchemHandlungsbereich` gehört `unsereKomm` hier? `Bildung?Freizeit?Mischung?`
- Goal: Encourage adaptation of NLP tools to CMC/web texts
 - ▶ Tokenization
 - ▶ POS tagging
- Dataset: 22,858 tokens
 - ▶ CMC part: Tweets, chats, Wiki talk pages, blog comments
 - ▶ Web part: Websites, blogs, Wikipedia articles, Wikinews

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EmpiriST tokenization guidelines in a nutshell

Beißwenger et al. (2015a)

- Usual treatment of whitespace and punctuation
- Whitespace errors usually not corrected (schona ber)
 - ▶ Except in emoticons and symbols:
:) → :), f - > d → f -> d
- Split CamelCase (deineMutter) if not proper name (MySpace)
- Abbreviations:
 - ▶ multidot abbreviations representing multiple words are split up:
d.h. → d. h., u.dgl. → u. dgl.
 - ▶ multidot abbreviations representing single words are not split up:
o.k. → o.k., Dipl.-Ing. → Dipl.-Ing.
- Contractions are not split up (⚡ Universal Dependencies):
machste → machste
- Single tokens:
 - ▶ HTML/XML tags, E-mail addresses, URLs, Filenames, Emoticons

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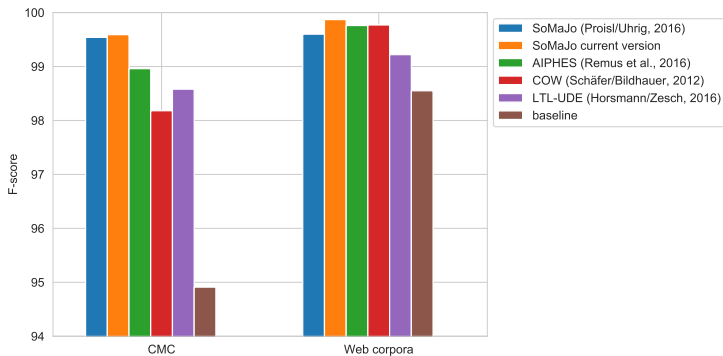
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Some of the problematic cases

- Abbreviations vs. actual words: `automat.`, `zum.`
- Cardinal number at end of sentence vs. ordinal number
- Section numbers vs. dates: `5.3.`
- Citations (`Storrer2007`) vs. proper names (`Blume2000`)
- Horizontal ellipsis:
 - ▶ `du bist echt ein Arm... → Arm...`
 - ▶ `zeig mir mal deinen Arm... → Arm ...`
- Confusion of hyphen (`-`) and en dash (`–`)

Results



SoMaJo

Proisl and Uhrig (2016)

- Rule-based: cascade of regular expressions
- Can output the token class for each token, e.g. number, emoticon, abbreviation, etc.
- Can also output additional information for each token that can help to reconstruct the original untokenized text

```
echo 'der beste Betreuer? - >ProfSmith! : )' | somajo-tokenizer -cet -
der      regular
beste    regular
Betreuer regular  SpaceAfter=No
?        symbol
->       symbol   SpaceAfter=No, OriginalSpelling="- >"
Prof     regular  SpaceAfter=No
Smith    regular  SpaceAfter=No
!        symbol
:)       emoticon  OriginalSpelling=": )"
```

POS tagging – New tags added to STTS (1)

Beißwenger et al. (2015b)

- Tags for CMC phenomena:
 - ▶ EMO(ASC|IMG) (ASCII/graphic emoticon): :-) ^^ 0.0 ☺ ☹
 - ▶ AKW (interaction word): *lach*, freu, grübel, *lol*
 - ▶ HST (hash tag): Kreta war super! #Urlaub
 - ▶ ADR (addressing term): @lothar: Wie isset so?
 - ▶ URL: <https://www.uni-bamberg.de/germ-ling/>
 - ▶ EML (e-mail address): thomas.proisl@fau.de
- Tags for phenomena in colloquial registers:
 - ▶ Contractions with pronoun: (VV|VM|VA|KOUS|PPER)PPER: schreibste, ichs, obse
 - ▶ Contractions with article: APPRART, ADVART: vorm, fürn, sone

POS tagging – New tags added to STTS (1)

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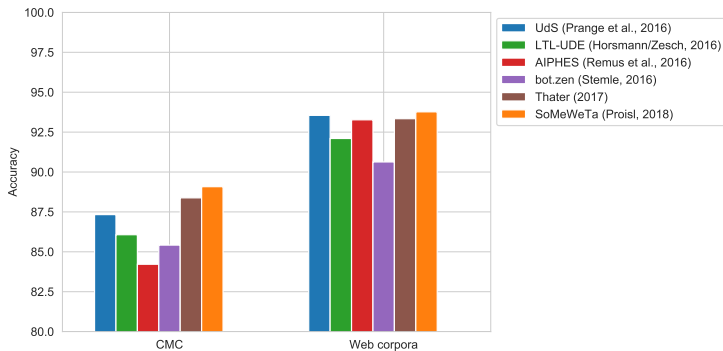
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POS tagging – New tags added to STTS (2)

Beißwenger et al. (2015b)

- Other refinements of STTS:
 - ▶ PTKIFG (intensifier/gradation particles): sehr schön, voll geil
 - ▶ PTKMA (modal particles and downtoners): Das ist ja doof
 - ▶ PTKMWL (particle as part of MW lexeme): keine mehr, noch mal
 - ▶ DM (discourse markers): weil, obwohl, nur
 - ▶ ONO (onomatopoeia): boing, miau, zisch
- PTK(IFG|MA|MWL) correspond to ADV or ADJD in standard STTS
- DM corresponds to KOUS or ADV

Results



SoMeWeTa

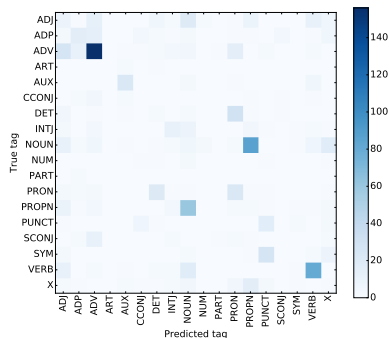
Proisl (2018)

- Averaged structured perceptron using beam search and early update strategy (Rosenblatt, 1958; Freund and Schapire, 1999; Collins, 2002; Collins and Roark, 2004)
- Domain adaptation (Chelba and Acero, 2004):
 - ▶ Use model trained on TIGER as prior on weights
- External resources:
 - ▶ Brown clusters (Brown et al., 1992) extracted from DECOW14 (Schäfer and Bildhauer, 2012; Schäfer, 2015)
 - ▶ Coarse-grained word class information from Morphy (Lezius, 2000)

Error analysis

SoMeWeTa (Proisl, 2018)

- Misclassifications related to new particle tags (PTKIFG, PTKMA, PTKMWL)
- Other major sources of errors:
 - ▶ confusion of NN and NE
 - ▶ misclassifications within verb tags



Lemmatization of German CMC data

Adding lemmatization to the EmpiriST corpus

- CMC-specific additions to TIGER morphology annotation scheme
- Two lemmatization strategies:
 - 1 Surface-oriented lemmatization (based on inflectional suffixes, retains non-standard orthographical features)
 - ★ Grigfe → Grigf
 - ★ Soooo → soooo
 - 2 Normalized lemmatization (correct obvious spelling errors, standard form of non-standard tokens)
 - ★ Grigfe → Griff
 - ★ Soooo → so

Guidelines

- For most of the new tags, the lemma is the surface form (EMOASC, EMOIMG, AKW, HST, ADR, URL, EML, PTKIFG, PTKMA, PTKMW, DM, ONO)
- Contractions: Lemma of contraction is lemma of first component (cf. APPRART)
 - ▶ schreibste → schreiben
 - ▶ isses → sein
 - ▶ obse → ob
 - ▶ ichs → ich
 - ▶ hamwas → haben
- The tricky part is the normalization!
 - ▶ das/dass/daß
 - ▶ hinstelt
 - ▶ Disko(theke) vs. Disku(ssion)
 - ▶ bißt
 - ▶ jedenfalls (tagged as NN)

Guidelines

What's in a name? – heuristics for proper (nick-)names in chat protocols

- Established proper name: lemma is capitalised (marc for user marc30 → Marc)
- Non-established proper name: lemma = surface form (sternchen for user Nudelsuppenstern → sternchen)
- Problematic cases: correspondences to lexicalised words, short forms (lanto for user Lantonie), creative nicknames/ inflectional phenomena (username stoeps is "pluralised" to stöps → stöps)

Not to be disregarded: motivational issues in the final phase (ausgeliefert → ausgeliefern)

Annotation

- Gold standard data independently lemmatized by four student annotators
- Unclear cases decided in group meetings with task organizers
- Agreement between annotators and gold standard (accuracy)
 - ▶ Drop from inter-annotator to annotator-gold due to post-annotation change in guidelines w.r.t. proper names

	Surface-oriented lemmata				Normalized lemmata			
	AJ	DW	EH	LR	AJ	DW	EH	LR
gold	93.64	92.87	93.73	93.67	93.10	92.82	93.80	93.46
AJ		96.08	96.54	96.50		96.00	96.28	95.92
DW			96.21	96.55			96.33	96.19
EH				96.89				96.70

Baseline systems

- Do-nothing baseline: Always return the word form
- Weak baseline: Given a word form and a POS tag, return the most frequent lemma from TIGER \cup EmpiriST training data
 - ▶ Fallback 1: Ignore case
 - ▶ Fallback 2: Return word form
- Strong baseline: Apache OpenNLP maximum entropy lemmatizer

Baseline results (accuracy, ignoring case)

Baseline	surface-oriented	normalized
Do-nothing	71.63	70.73
Weak	83.90	83.19
Strong	87.50	85.97

→ Room for improvement!

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And now for something completely different.

What's Reddit?

- Social news aggregation, discussion and micro-blogging platform
- Founded in 2005
- One of the most popular websites in the US (Alexa rank: 6¹, SimilarWeb rank: 10²)
- Increasingly popular in Germany (Alexa rank: 9³, SimilarWeb rank: 33⁴)

¹<https://www.alexa.com/topsites/countries/US>, 16/9/2019

²<https://www.similarweb.com/top-websites/united-states>,
16/9/2019

³<https://www.alexa.com/topsites/countries/DE>, 16/9/2019

⁴<https://www.similarweb.com/top-websites/germany>, 16/9/2019

What's Reddit?

- Self-proclaimed “front page of the internet”:
 - ▶ Users submit content (e.g. links, images, text)
 - ▶ Submissions are voted up or down
 - ▶ Submissions with the most upvotes make it to the front page
- Submissions can be commented on, comments can also be voted up or down
- Nested conversation threading
- Since 2008, users (“redditors”) can create “subreddits” (categories or communities) with their own rules

Who uses Reddit?

- While more women than men use social media in general (at least in the US⁵, Reddit is more popular among men than among women⁶
- Users are mostly young⁷, with a high share of people interested in technology

⁵<https://www.statista.com/statistics/471345/us-adults-who-use-social-networks-gender/>

⁶<https://www.statista.com/statistics/261765/share-of-us-internet-users-who-use-reddit-by-gender/>

⁷<https://www.statista.com/statistics/261766/share-of-us-internet-users-who-use-reddit-by-age-group/>

The German Reddit Sphere

- Subreddits can be very heterogenous
- Parts of Reddit can be very weird to outsiders, the same goes for large parts of the German Reddit sphere
- Special emoticons: :) → Ü, :o → Ö, :< → Ä
- Quite common: word-for-word translations of English words and phrases:
 - ▶ *pfofstieren*
 - ▶ *ausgelöst*
 - ▶ *Maimai*
 - ▶ *Fixierte das für dich.*
 - ▶ *schuldiges Vergnügen*
 - ▶ *Lases, Unterlases*
 - ▶ *kantig, Kantenfürst*

Building a Reddit Corpus

- In an ongoing effort, Jason Baumgartner collects every Reddit submission and comment, publicly accessible via <https://files.pushshift.io/reddit/>⁸ (some caveats apply, see Gaffney and Matias, 2018)
- 2015: first attempt at extracting German comments (Barbaresi, 2015), resulting corpus still relatively small (97505 comments, 566362 tokens)
- Since then, activity on Reddit has increased greatly
- We adapt Barbaresi's approach of using a two-tiered filter to detect German comments in the vast dataset:
 - 1 spell checking
 - 2 language identification

⁸see also https://www.reddit.com/r/pushshift/comments/bcxguf/new_to_pushshift_read_this_faq/

Raw data (1)

```
'archived': False,  
'author': 'Mythic_Emperor',  
'author_created_utc': 1521765799,  
'author_flair_background_color': None,  
'author_flair_css_class': None,  
'author_flair_richtext': [],  
'author_flair_template_id': None,  
'author_flair_text': None,  
'author_flair_text_color': None,  
'author_flair_type': 'text',  
'author_fullname': 't2_12w6sr28',  
'author_patreon_flair': False,  
'body': 'Es geht mir Gut. Und dir?',  
'can_gild': True,  
'can_mod_post': False,  
'collapsed': False,  
'collapsed_reason': None,  
'controversiality': 0,  
'created_utc': 1541030413,
```

Raw data (2)

```
'distinguished': None,  
'edited': False,  
'gilded': 0,  
'gildings': {'gid_1': 0, 'gid_2': 0, 'gid_3': 0},  
'id': 'e8tkilo',  
'is_submitter': False,  
'link_id': 't3_9t23an',  
'no_follow': True,  
'parent_id': 't1_e8tfeac',  
'permalink': '/r/HistoryMemes/comments/9t23an/are_you_just_gonna_scroll',  
'removal_reason': None,  
'retrieved_on': 1544848339,  
'score': 2,  
'send_replies': True,  
'stickied': False,  
'subreddit': 'HistoryMemes',  
'subreddit_id': 't5_2v2cd',  
'subreddit_name_prefixed': 'r/HistoryMemes',  
'subreddit_type': 'public'
```

Detecting German posts – the strategy

Barbaresi (2015)

- ❶ Sanitizing
 - ▶ Ignore newlines and URLs
 - ▶ Don't consider very short posts or posts which were deleted
- ❷ Spell checking
 - ▶ Check every token (`re.findall(\w+)`) against both a German and an English dictionary
 - ▶ If not in respective dictionary: count an error
 - ▶ Classify as potentially German if English error rate higher than 30% and German error rate less than 70%
- ❸ Language identification
 - ▶ Run `langid` (Lui and Baldwin, 2014) on potentially German posts

Evaluation of the Two-tiered Filter

- Random sample of 1618 comments from 2016
- Looking at the first 565 comments (20020 lines), we see many of them (148, or 26%) are not actually in German (1423 lines → 7%)
- Longer comments are apparently recognized quite reliably as German, while shorter ones are often misclassified
- Also problematic: comments which only contain links to subreddits, lots of markup or proper names

Corpus Cleanup

- Top 10 subreddits, 2016-2018 (out of 33144, containing 6784837 “German” comments):

Subreddit	“German”	Total	Fraction
de	3644481	4364440	0.835
Austria	389041	483125	0.805
rocketbeans	142787	164699	0.867
AskReddit	125411	183147454	0.000685
edefreiheit	121370	147150	0.825
de_IAmA	58172	65768	0.885
Finanzen	48033	55009	0.873
Fireteams	38042	2125695	0.0179
soccer	33727	22918204	0.00147
rbtvcirclejerk	32493	38945	0.834

- Comments in subreddits with low fractions of “German” comments are probably not German after all

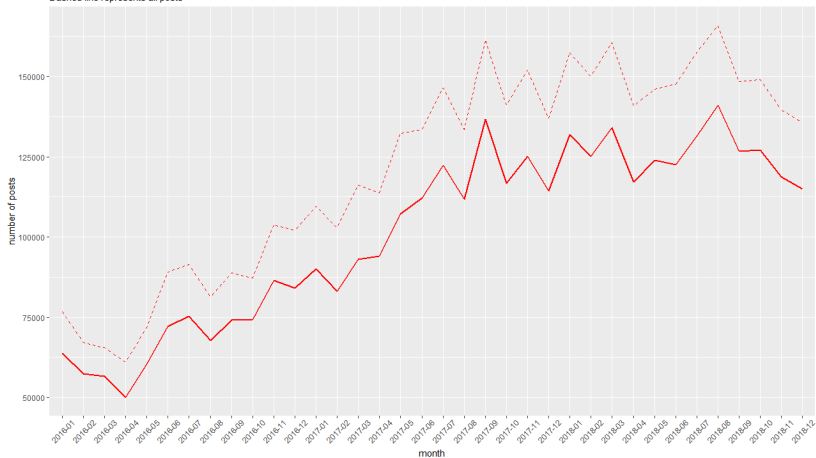
Corpus Cleanup

- By filtering out subreddits with less than 10% and/or less than ten “German” comments, we can remove 27% of all comments classified as German
- Remaining comments: 4926924
- However, for those subreddits which are predominantly German, shouldn't we include *all* comments in our corpus?

Top 10 Subreddits

Posts classified as German in /r/de, 2016-2018

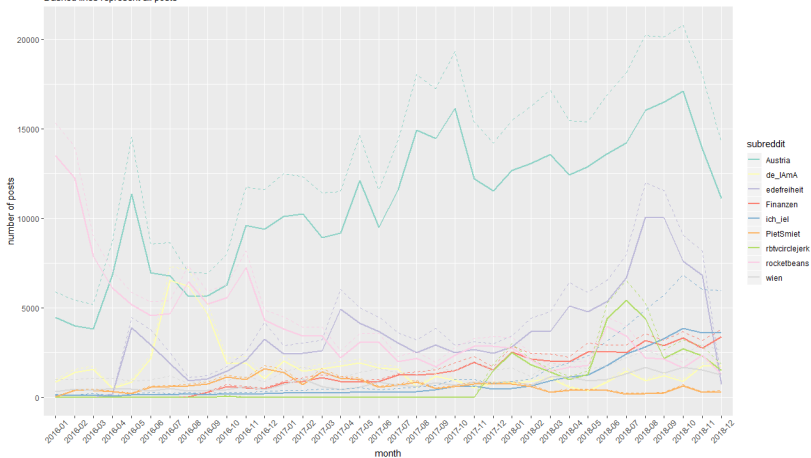
Dashed line represents all posts



Top 10 Subreddits

Subreddits with the most posts in German (after /r/de), 2016-2018

Dashed lines represent all posts



Pre-processing and Tokenization

- Reddit comments can contain Markdown markup⁹, so they have to be pre-processed to avoid curious tokenization results:
 - ▶ *~~der gute~~ Ketchup gehört in den ~~Reis oder aufs Brot~~ Müll.*
 - ▶ *Ich würde dir die [~~Sokratische Methode~~](https://de.wikipedia.org/wiki/Sokratische_Methode) [Mäeutik](<https://de.wikipedia.org/wiki/M%C3%A4eutik>) empfehlen.*
- Short form links to subreddits and user pages have to be tokenized properly (*r/oldschoolcool*, */r/de*, */r/de/about/traffic*, */de*, */u/username*, */u/username*, ...)
- In the absence of punctuation, line breaks (and sometimes, emoticons) can mark sentence boundaries

⁹see <https://www.reddit.com/wiki/markdown>

Example: Emoticon as Punctuation

Token	TreeTagger	Lemma	SoMeWeTa	Corrected POS
<s>				
So	ADV	so	ADV	DM
bin	VAFIN	sein	VAFIN	VAFIN
endlich	ADV	endlich	ADJD	ADV
zu	APPR	zu	APPR	APPR
was	PIS	was	PIS	PIS
gekommen	ADJD	kommen	VVPP	VVPP
:D	ADJD	:D	EMOASC	EMOASC
Habe	VAFIN	haben	VAFIN	VAFIN
jetzt	ADV	jetzt	ADV	ADV
...				

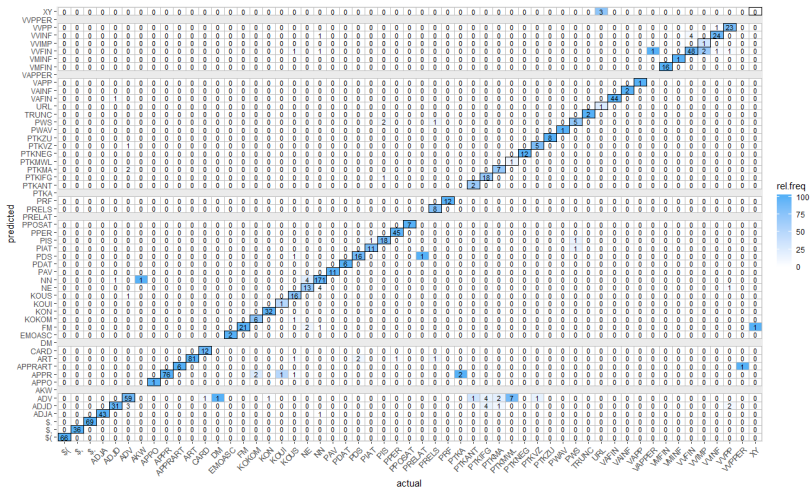
POS Tagging

- Random sample tagged with SoMeWeTa (trained on EmpiriST corpus)
- Corrected POS tags in a small number of comments (1186 tokens – so far) to evaluate performance
- Accuracy: 92.6%

Some Notes on STTS-IBK

- Some very fine-grained categories like DM (e.g. “epistemisches *weil*”), PTKIFG, PTKMWL which are difficult to annotate manually, but no differentiation between some more obvious ones, e.g. *sein* as an auxiliary or as a full verb, definite and indefinite articles, ...
- Tags for some contracted forms but not all (STTS-IBK guidelines state to just annotate POS for the first word in these cases)
- Catch-all category \$(→ no differentiation between opening and closing brackets or quotation marks, dashes, slashes etc. (asterisks to mark “Aktionswörter” also fall into this category)
- What to do with acronyms like *scnr* (*sorry, could not resist*) or *imho* (*in my humble opinion*)? Do we need new tags?
- TRUNC nur für Kompositionserst-, nicht aber für -zweitglieder

Confusion Matrix



Conclusion

What we've learned

- Do not blindly use others' methods
- Pay close attention to tokenization and sentence splitting
- Adapt it if necessary
- STTS-IBK is a little weird (and we might need to discuss its rules and categories)

Reddit Corpus

- More corpus cleaning necessary
- Tokenization rules have to be updated
- POS-tagging and lemmatization further down the pipeline

Links

Tools

- SoMaJo:
<https://github.com/tsproisl/SoMaJo>
- SoMeWeTa:
<https://github.com/tsproisl/SoMeWeTa>

Data

- EmpiriST 2015:
<https://sites.google.com/site/empirist2015/>
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Thanks for listening.
Questions?