Text Analysis for Social Media Cybersecurity: the AMiCA Project

Els Lefever

Language and Translation Technology Team (LT³) Ghent University, Belgium





LT³, LANGUAGE AND TRANSLATION TECHNOLOGY TEAM

LT³



•Dpt of Translation, Interpreting and Communication, Faculty of Arts and Philosophy, Ghent University

•fundamental and applied research in **language and** translation technology

•expertise in using **machine learning** for language technology problems (PoS-tagging and lemmatization, anaphora resolution, WSD, NER)

•Headed by Prof. Véronique Hoste





3 main research lines:

- Terminology & computational semantics
- Translation Technology
- Sentiment analysis and subjectivity detection



Terminology / computational semantics



- •Lead: Prof. Els Lefever
- •Automatic terminology extraction from monolingual, bilingual and comparable corpora (Ayla Rigouts Terryn)
- Automatic hypernym and synonym detection (Els Lefever)
- •Term ambiguity in interdisciplinary research (Julie Mennes)
- •Use of term extraction for translating documentaries (Sabien Hanoulle)





Translation Technology



•Lead: Prof. Lieve Macken

comparison of different methods of translation: human vs. post-editing, human vs. CAT (Joke Daems)
translation quality assessment and confidence estimation for machine translation (Arda Tezcan)







Sentiment Analysis and Subjectivity detection



- •Lead: Prof. Véronique Hoste
- •automatic detection of cyberbullying (Cynthia Van Hee)
 •suicide detection (Bart Desmet)
- •Aspect-based sentiment Analysis (Orphée De Clercq)
- detection of subjectivity in annual reports (Nils Smeuninx)
- Irony detection (Cynthia Van Hee)
- •Sentiment Analysis for economic events (Gilles Jacobs)













50	ŏŏ
)Ŏ(ÕÕ



Outline



- The context and goals of the AMiCA project
- Text normalization
- 3 Use cases:
 - 1. Detecting cyberbullying
 - 2. Suicide detection
 - 3. Age and gender profiling for detecting grooming

www.amicaproject.be

- IWT-SBO project, coordinated by CLiPS (UA)
- Partners:
 - CLiPS (text mining, UA)
 - MIOS (sociology, UA)
 - LT3 (text mining, UGent)
 - IBCN (software development, UGent)
 - VISICS (image processing, KUL)
- Combine text analytics, image and video analysis, and data mining





translation technology







Goals



- Detect situations that are harmful or threatening to young people in social networks
 - Cyberbullying
 - Sexually transgressive behaviour (for example grooming by paedophiles)
 - Depression and suicide announcement
- Facilitate efficient action by moderators, police, parents, peer group, social services, ...
- Objective measurement, monitoring, trend analysis, ...



User Committee







How urgent is the problem?

- European "Kids online" study (EU, 2011)
 - Motivation for the project
 - Age 9-16 in 25 European countries
 - Results
 - Children are 90 minutes per day online
 - Half of them in their bedroom
 - 33% added strangers as friends
 - 15% shared personal information with strangers (Including photographs)
 - 12% felt they experienced harm

www.eukidsonline.net

How urgent is the problem?

- European "Kids online" study: update in 2014
 - Age 9-16 in 25 European countries
 - Results since 2010 study, 9 to 16 year olds
 - Significant rise of use of social media
 - Rise of 23% to 43% of having contact with someone not met IRL before
 - Rise of 10% to 23% of having seen sexual images
 - Rise of 9% to 20% of having received sexual images
 - Rise of 13% to 17% are upset by something seen online
 - Rise of 13% to 20% of being exposed to hate messages
 - Rise of 7% to 11% of being exposed to self-harm sites
 - Rise of 7% to 12% of being exposed to cyberbullying

www.eukidsonline.net

Quick poll



 Who is in favor of software monitoring automatically your interactions in social media for risks and threats?

Should we do something about it?

- Majority of experts and adolescents is in favor of automatic monitoring
 - but only for situations they perceive as uncontrollable
 - with respect for privacy and with suitable followup, not involving too many parties, and giving control to the victim
- Mixed opinions with the parents depending on (negative) previous experience and level of trust in their children

Workflow





Crawl: example





Zwijg stomme trut! Gij hebt geen leven tot op je begravenis!!!

(English: Shut up stupid cow! You don't have a life see you at your funeral!!!)

Crawl: example



Django administration

Welcome, nlpapp - Recent Actions -

Home / NIp / Tweets / 2015-09-29 22:52:47+00:00: Zwijg stomme trut! Gij hebt geen leven tot op je b...

Change tweet

History

Fields in **bold** are required.

Tweet url:

https://twitter.com/SonicStef/statuses/6213

Timestamp:

 Date:
 2015-09-29
 Today |

 Time:
 22:52:47
 Now |
 🕥

Note: You are 2 hours ahead of server time.

Text:

Zwijg stomme trut! Gij hebt geen leven tot o

User name:

Pestertje1998

PREPROCESSING / NORMALISATION OF USER-GENERATED TEXT



User Generated Content



Social media: blogs and microblogs (Twitter: 190 million tweets/day), wikis, podcasts, social networks (Facebook: 70 billion shares/month)

 \Rightarrow Enormous amount of UGC



Properties of chat language



- Omission of words / characters (spoke spoken)
- Abbreviations, acronyms (LOL laughing out loud)
- Deviations from standard spelling (luv love, you iz you are)
- Expression of emotion:
 - Flooding (looooooooove)
 - Emoticons (:p)
 - Capitalized letters (STUPID)
- Dutch-specific:
 - Concatenation of tokens (khou ik hou)
 - Elimination of clitics and pronouns (edde heb je)
 - Lot of dialects!

Example



	Example of Dutch SMS language
Original	Oguz ! Edde me Jana gesproke ? En ze flipt lyk omdak ghsmoord heb !
Normalized	Oh gods ! Heb je met Jana gesproken ? En ze flipt gelijk omdat ik gesmoord heb !
Translated	Oh god ! Did you speak to Jana ? And she's flipping because I smoked !



Problem for Text Analysis Tools

- Most NLP tools are developed for or trained on standard language
- They fail miserably on UGC
- Solutions
 - Develop new tools
 - E.g. Tweet NLP (CMU): <u>http://www.cs.cmu.edu/~ark/TweetNLP/</u>
 - Normalize the 'non-standard' language
- On the positive side, non-standard language makes some analytics tasks easier!



Normalization Approaches

- Three dominant approaches
 - <u>Machine Translation</u>: Source Language = nonstandard and Target Language = standard
 - <u>Spell Checking</u>: Correct the incorrect words (statistical or dictionary-based)
 - <u>Speech Recognition</u>: Non-standard language = speech that has to be converted to text (HMMs)

=> We choose to follow an SMT approach and also go to the character-level



Ensemble Approach



Sarah Schulz, Guy De Pauw, Orphée De Clercq, Bart Desmet, Véronique Hoste, Walter Daelemans, and Lieve Macken. 2016. Multimodular text normalization of Dutch user-generated content. *ACM Trans. Intell. Syst. Technol.* 7, 4, (July 2016), 22 pages. DOI: <u>http://dx.doi.org/10.1145/2850422</u>

Modules



- <u>Preprocessing</u>
 - Tokenization and sentence splitting
 - includes emoticons, emojis etc.
 - Character floooooooding
- <u>Token-based modules</u>
 - Abbreviations
 - Expansion dictionary (~ 350 abbrevs)
 - Spell checker
 - Levenshtein on dictionary (~ 2.3 million words)
 - Compound Module
 - Checks if a pair of words is actually one word
 - Word Splitter
 - 'misje' = 'mis je' (miss you)

Modules



- <u>Context-based modules</u>
 - SMT
 - Token-unigram, character unigram, character-bigram and combinations
 - Transliteration (supervised ML)
 - supervised ML, memory-based learning style
 - +da+_n i ++_ged -> iet
 - WAYS (Write As You Speak): G2P + P2G (memory-based learning)
 - ni (niet*, not*)
 - kem (ik heb, *I have*)
- <u>"Original" Module</u>
 - Many words are correct

Modules



- Decision Module
 - Moses decoder (SMT), dynamic search among the suggestions of the component modules
 - Uses (5-gram) language model and phrase table (dev. Set)

Evaluation



- Three types of UGC
 - Chat (Netlog)
 - SMS (Sonar corpus)
 - Microblog (Twitter)
- Train (60%) Development (20%) Test (20%)
- Total: 70,000 tokens, manually annotated
 - insertions, deletions, substitutions, transpositions
 - near-perfect annotator agreement
- Background corpora for language modeling

CGN (Spoken Dutch Corpus)	6,765,336
SoNaR (Balanced text corpus)	3,581,182
Open Subtitles Dutch (OSD)	90,147,315
Training set (TS)	56,523

Results



- Module level evaluation:
 - SMT and Transliterate modules perform best
 - Especially compounding and splitting problems remain
- Ensemble evaluation:
 - Best ensemble system: 92.9
- Extrinsic and Portability Evaluation
 - Tested on Ask FM for NLP tasks (with and without normalizing)
 - POS (+12%), LEM (+13%), NER (+8%)
- Problems remain especially in tokens with multiple normalization problems

USE CASE 1: CYBERBULLYING DETECTION



Research Motivation

- ± 20-40% of all youth have been victimized online (Tokunaga, 2010)
- Anonymity, lack of supervision and impact make social media a convenient way for cyberbullies to target their victim (Hinduja & Patchin, 2006)
- Information overload on the Web has made manual monitoring unfeasible



more likely to be exposed to pro-anorexia sites

more likely to be exposed to self-harm sites

> more likely to be exposed to cyberbullying



13%

13%

9%

to



European 9- to 16-yearolds say they are now: more likely to say they were **Upset** by something seen online in 2014

Research Motivation



- Automatic detection systems allow for large-scale social media monitoring
- Goal => reduce manual monitoring efforts on social media

Related Research



- NLP applications for automatic cyberbullying prevention and detection
 - Cyberbullying detection (Yin et al., 2009; Reynolds et al., 2011; Nahar et al., 2013)
 - Sensitive topic identification (sexuality, race) (Dinakar et al., 2012)
 - Detection of bully profiles on social networks (Dadvar et al., 2013)

BUT:

- Focus on posts from harassers
- No distinction between different types of cyberbullying
- Datasets do not always follow a real-world distribution
Data set construction

- We need large data sets to train machine learning systems
- Data collection for Dutch and English
 - Data from relevant social media
 - BUT: few / private data
 - Media campaign for donating examples of cyberbullying messages
 - BUT: sensitive data!

- Cyberbullying simulations







Data set construction: media campaign



RESULT:

± 30 reactions

± 368 messages (FB messages, hate pages, Netlog, mail, chat, etc.)



ver deze oproep of ons project kan u steeds ter cynthia.vanhee@ugent.be

iwī

Antwerpen

Nieuw systeem spoort online pestgedrag automatisch op

SAH FEYS	mm. Verschillende onderzoekers van de universiteiten van Gent.	ook ingrijpen als jongeren met sul cichte ordachten karmon."
	Leuven en Antwerpen slaan	Het onderzoekstearn is daaron
Belgische onderzoekers	AMICA, een systeem datautoma-	pesters die ooit beledigende smsje
willen cyberpesten		
tegengaan door pestgedrag		
op Facebook en Twitter		
automatisch op te sporen.	"We willen zowel tekst als beeld	daarvan kunnen we accurate com
Het systeem moet op	op sociale media analyseren om 20	putermodelien bouwen die kunnen
termijn zens ingrijpen	volaummatisch abernsignalen te	alleiden warneer het om cyberper
oil zennioor ogenachten.	ation" Lost Verentrustende stu-	ten gaat. Hoe meer buildhare voor
Cybernesten is vandaag nog ere		
		den."
slachtoffers er zelden voor witho-	andrasion notofile coursess must	Debefoallter is om uiteindelti

		r	
'Elk bericht op Facebook zal een label meekrijgen'	gebrincht. Die kan dan de ernst van de situatie inschatten en eventueel de politie inschatten en eventueel De onderzoekers werken samen met onder meer Netlog, Child Focus	onmogelijk", weet H gewoen te wel input." Het Belgische onde speek internationaal e	
VÉRONIQUE HOSTE ONDERZOEKER LIGENT	en met de VRT, dat dagelijks tien- duizenden kinderen naar ketzet be ziet komen. "Het is voor om heel	ren van pestgedrag. O uiteindelijk zal slagen deels afhangen van d	
elk nieuw bericht op Facebook of Twitter te toetsen aan dat model. "Elk bericht zal een label meekop- ren", werduidelijt floate. "Als be	wellige ongewing vormt workinder run", beseft VRT-woordwoerster Anneke Ernon, "We zetten sterkin on one fillforeiten sterkin	king van Twitter Facebook, waar veel be openboor zijn.	
mendes berichten van een gebrui- ker rode lampjes gaan branden, sal een moderator, of sitebeheerder automatisch op de hoogte werden	leest ook nog eens alle berichten na die de kinderen online plaatan." "Op sittes als Facebook is volle- dige matuele controle echter	Cyberpestberichten vie data@amicaproju worden doorgestuur het onderzoeksteam	



GEZOCHT: CYBERPESTBERICHTEN

GEZOCHT: CYBERPESTBERICHTEN

Veilig internetten is belangrijk, maar niet altijd eve gemakkelijk. Heb jij al een keer rare of onveilige dinger gemaakt op het internet?

dene studies geven aan dat één Vlaa Venicheroene studies geven aan dat één Vlaamse jongere op tien recent het slachtoffer was van cyberpeaten. De impact van dergelijk sluisies is vaak erg groot doordat kwetsende berichten soms langere tijd online bijven staan en slachtoffers alleen met dit probleem zitten omdat ze er niet over willen of durven te

SAMEN STERK

In het wetenschappelijk project AMiCA slaan een aant and weinterstateppengic project AMECA staan een al anderzoeksgroepen van de universiteiten van Antwerpen, Leuven en Gent de handen in elkaar om een systeem te ontwikkelen dat automatisch kritieke situaties zoals cyberpestgedrag en seksueel grensoverschrijdend gedrag herkent op socia erksites om zo een veilige internetomgeving s offizio over jongeren.

Om te kunnen bepalen wat cyberpesten is en hoe het kan worden herkend, zijn voldoende en unnen bepalen wat cyberpesen is en hoe het kan worden nerkend, ogn verweinse even overe an deze online berichten. Daarom doet AMCA een oproep aan u als outer om zoveel mogelijk in aan ons beschikbear te stellen waarin cyberpesten merkbaar is.

Als Um Kind gehuige of slachtoffer is geweest van cyterporten en als hiervan bewijsmateriaal beschikbaa dan zijn deze berichten meer dan welkom. Die berichten kunnen tijvoorbeeld e-maila zijn, smijsk untgesperinden of berichten van sociale media. Die data worden anoelem behandeld en alleen gehruidt chatgesterinden berichten van sociale media. Die data worden anoelem behandeld en alleen gehruidt

erennen on benomen ver sociale mennen over som worredn annalen ver In onderzoek, berichten worden in geen geval doorgegeven aan derden.

Will je contact opnemen met AMICA (Automatic Monitoring for Cyberspace Applications)? Stuur dan een ve je vounet oprømen met vær.Ca (automatic konnoring for Cyberspace Applications)? Stuar dan eer malige naar data@amicaproject.be en blif niet bij de pakken zitten. Meer informate vind je op de website www.amicaprojectbe.

Dataset Construction: simulation experiments



- Role playing in secondary schools on social media platform: FB-like social network, scenarios, profile cards (roles), debriefing
- Additional goal: education (prevention)

AMICA	
Nieuwsoverzicht [Beri	chten]
	Dominique Verhaegen Jupdates Info Vrienden (5) Image: Solution of the state of the
Zend Bericht Blokkeer Gebruiker	Joni Claes Vind je jezelf nu beter dan mij nu je dit allemaal zegt? Zoek een leven en scheld niet met kanker, dat is onrespectvol. 9 mei · Vind ik leuk · 1 vindt dit leuk Laura Van Boom ja wa is u probleem?
Admin Settings	 9 mei · Vind ik ieuk Dominique Verhaegen wa moeide gij u nu weer! ga terug zuigen aan u tampons kankerhoer 9 mei · Vind ik ieuk
GEMEENSCHAPPELIJKE VRIENDEN	Emma Dewaele GJ ZIJT EEN DEBIELE KIND DOMINIQUE 9 mei · Vind ik leuk Laura Van Boom je echt een achterlijk kind 9 mei · Vind ik leuk

Data Annotation



- Brat rapid annotation tool (Stenetorp et al., 2012)
- Two annotation levels (Van Hee et al., 2015)
 - Post level
 - Cyberbullying -vs- non-cyberbullying

textual content that is published online by an individual and that is aggressive or hurtful against a victim.

- Harmfulness score
 - 0 \rightarrow the post does <u>not</u> contain indications of cyberbullying
 - 1 → the post contains <u>indications</u> of cyberbullying, although they are <u>not severe</u>
 - 2 \rightarrow the post contains <u>serious indications</u> of cyberbullying
- Author's role
 - Harasser
 Bystander-defender
 - Victim
 Bystander-assistant

Data Annotation



- (Sub)sentence level: identification of finegrained text categories related to cyberbullying
 - Threat/blackmail
 - Insult
 - Curse/exclusion
 - Defamation
 - Sexual talk
 - Defense
 - Encouragements (to the harasser)

<u>Guidelines for the fine-grained analysis of cyberbullying, version 1.0</u> (2015) Van Hee, C., Verhoeven, B., Lefever, E., De Pauw, G., Daelemans, W., & Hoste, V.

Data Annotation



Category	Brat annotation example	Translation
Threat/blackmail Expressions contain- ing physical or psychological threats, or indications of blackmail.	2_Har Threat or Blackmail ¶ als ik u tegen kom zieke rak op u gezicht x	I'll smash you in the face when I see you x
Insult Expressions containing abusive, degrading or offensive language that are meant to insult the addressee.	I Har General insult General insult I HAHAHAHA LOSER GIJ: (X AARDAPPELKOP	HAHAHAHA YOU LOSER :(X POTATO HEAD
Curse/exclusion Expressions of a wish that some form of adversity or misfortune will befall the victim and expressions that exclude the victim from a conversation or a social group.	2_Har Curse or Exclusion General insult Image: Pleeg zelfmoord internand vindt u geestig Pleeg zelfmoord internand vindt u geestig	Just commit suicide, nobody thinks you're funny
Defamation Expressions that reveal confident, embarrassing or defamatory information about the victim to a large public.	I_Har Defamation I u mama versiert andere mannen hahahaha	Your mom is flirting with other men haha- haha
Sexual talk Expressions with a sexual meaning that are possibly harmful.	1 Har Sexual harassment Image: Stuur my u naaktfoto, nu!!	Send me a naked picture of yourself, now!!
Defense Expressions in support of the victim, expressed by the victim himself or by a bystander.	1_Bystander_defender General victim defense General victim defense ¶ Meid, koppie omhoog he! Laat je ni doen door die domme anoniempjes	Cheer up girl, don't let those stupid anons make you feel bad
Encouragements to the harasser Expressions in support of the harasser.	2_Bystander_assistant Encouraging harasser ¶ inderdaad ze is geen leven waard !!	Indeed, she shouldn't be alive !!

Ask.fm preliminary experiments

- <u>Class</u>
 - Binary (bullying or non-bullying)
 - Binary (for each fine-grained class)
- <u>Features</u>
 - Word unigrams and bigrams
 - Character trigrams
 - Sentiment features
- <u>Classifier</u>: SVM (Pattern) with linear kernel
- <u>Data</u>: ~85,000 posts
- <u>Annotation agreement</u> (kappa) 60-65%
- Very <u>skewed data</u>, scarce positive data (~10%)

Van Hee, C., Lefever, E., Verhoeven, B., Mennes, J., Desmet, B., De Pauw, G., Daelemans, W. & Hoste, V. (2015). Detection and fine-grained classification of cyberbullying events. Proceedings of RANLP, 672–680. Hissar, Bulgaria.

Results



	Precision	recall	F1-score
NL	76%	56%	65%
EN	74%	55%	63%

BUT:

Ambiguity

"Hi bitches, anyone in for a movie tonight?" "Shut up, you bitch!"

Implicit realizations of cyberbullying

"You make my fists itch..."

Data sparseness



Results (Van Hee et al. 2015)



Monitoring desirable?



- Follow-up is needed
- Privacy of youngsters should be respected
- Technical feasibility?

(Van Royen et al., 2014)

More info?



Cynthia Van Hee: cynthia.vanhee@ugent.be



	90

USE CASE 2: SUICIDE DETECTION

Alarming figures Flemish adolescents



• Self-mutilation:

- Every year by 7% at the age of 14-17
- 2/3 through cutting & scratching (Van Rijsselberghe et al., 2009)
- Suicidal behaviour:
 - 15-20% (age of 18) have thoughts of suicide (more than once) (Hublet et al., 2010)

Online self-harm behaviour



Kheb het al 3 keer geprobeerd, ma kloop ier nog altijd rond... soms zeg ik spijtig genoeg, soms ben ik ook blij dat ik nog leef.



AMiCA technology: image analysis

- Automatic classification of images
- Object recognition in images
- Tekst recognition in images + OCR







If I jump now

who will catch me?

AMiCA technology: text analysis



Machine learning system **analyses** every message (word sequences, topic models, sentiment analysis, ...) and **answers two questions:**

- Is the message about suicide?
- Is there a serious suicidal threat?



I never thought about cutting or suicide, because it leaves scars ...



I already tried 3 times, but I'm still alive

Sometimes I feel bad, sometimes I'm glad I'm still alive

Text analysis: results



Experiments carried out on a data set of 10,000 messages, of which 851 are relevant and 257 are serious:

- Is the message about suicide? => recall: 9/10, 3% noise
- Is there a serious suicidal threat? => recall: 2/3, 25% noise

Does it work in practice?



What is the impact of the automatic detection system in a moderator setting?

<u>Simulation</u> of high work load of moderators:

- task: identify alarming messages that need a response (75)
- Lots of messages (1000)
- Limited moderation time (1 hour)
- Collaboration with CPZ (Flemish centre for suicide prevention) and moderators of the website "Wel Jong Niet Hetero" (LGBT web site)
- 1 group with / 1 group without system aid

Valorisation: interface

System	×			
C i suicide	-prevention.lt3.ugent.be/nlp/system/	☆	0	s 🔤 🖯 🥕
Annot	ations for suicide prevention			log out
Marked as re	All + System All	÷		
Messag	JES 1000 Page 1 of 20.			Next →
Date	Message Reaction		Sys	stem
2013-10-17 13:55	Heb je tips? Ikzelf droom om gitaar te kunnen spelen Er staat hier thuis een gitaar, maar behalve enkele domme dingskes lukt spelen niet echt Nog zo'n droom is drummen, dat lijkt me zooo zalig !	,	No	
2007-07-01 15:07	maar ik ken u ni, ze. (ik zat daar met mijn heteromaat in een hoekje. hij wou ni echt veel Yes No		No	
2015-06-26 15:46	Zal ik zeker doen! Yes No Ja, super nieuws tussen al die aanslagen!		No	
2015-06-26	Van alle homofobe reacties op de legalisering van het homohuwelijk in Amerika, vind ik Yes No	>	No	













More info?



Bart Desmet: bart.desmet@ugent.be



USE CASE 3: PROFILING FOR DETECTING PEDOPHILE GROOMING



Motivation





Motivation



- Survey: ±1000 youngsters about the frequency, nature and appropriateness of sexual messages on social media
- Especially on Facebook
- Who?
 - 32% strangers
 - 29% friends IRL
 - 19% online friends

67% didn't like the message + 11% reported the incident





Profiling



- AMiCA profiler
 - Based on Chris Emmery's OMESA
 - <u>https://github.com/cmry/omesa</u>
- Age and Gender
 - Finding dubious SN profiles
 - Computed age and gender does not match given information
 - Optimizing recall (for moderator application)
 - Adapting to binary classification
 - Legally relevant age difference

Approach



- SN chat data (Netlog, 2010-2011)
 - 380k posts
 - 87k users
 - Data point = combined posts of a single user
 - Self-reported age, gender, and location
- Classes: age (binary), gender, age+gender
- 5-fold cross-validation
- SVM with linear kernel
- Features:
 - token n-grams
 - character n-grams

Results



- Gender
 - -~70%
 - Adding different types of features (LIWC, POS patterns, sentiment, etc) boosts f-scores slightly

Results



- Age:
 - Distinguish between users above and below age of consent (16 in Belgium), -16 versus +18 has priority
 - Optimize recall
 - Using cost and confidence parameters in SVMs
 - Up to 95% recall for -16; 92% recall for +18

Ref: Janneke van de Loo , Guy De Pauw, Walter Daelemans, Text-Based Age and Gender Prediction for Online Safety, International Journal of Cyber-Security and Digital Forensics (IJCSDF), 2016, 46-60.

Predator Detection

- Two classifiers
 - LiBSVM
 - Classify at the post level, aggregate at user level
 - Classify at the user level directly
 - Weighted voting of previous
 - Additional constraints
 - E.g. only one pedophile per conversation

Claudia Peersman, Frederik Vaassen, Vincent Van Asch, Walter Daelemans. Conversation Level Constraints on Pedophile Detection in Chat Rooms. CLEF 2012 (PAN), 2012.



Overall test results

- Grooming detection
 - Predator detection
 - 72 % f-score, 89% precision, 60% recall
 - Suspicious posts
 - 30% f-score, 36% precision, 26% recall

More info?



Walter Daelemans:

walter.daelemans@uantwerpen.be



Guy De Pauw: guy.depauw@uantwerpen.be


	0000	
		Q
		+
56		Ы
)Ŏ		ŏ

DISCUSSION

discussion



 Is normalization and automatic detection accurate enough for applications in cybersecurity?

Precision - Recall trade-off

 Should we protect children and young people in social networks against their will?

Protection - privacy trade-off



Thank you! Els.lefever@ugent.be

http://www.amicaproject.be/