



Annotating Affective Information on 5.5 Billion Word Corpus of Japanese Blogs

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Presentation Outline

- Introduction
- Emotion Blog Corpora
- Affective Information Annotation
- Evaluation of Affective Annotations
- Statistics of Affective Annotations
- Conclusions and Future Work

- Recently there has been much research done on Sentiment Analysis and Affect Analysis
 - Sentiment Analysis (SA): Positive/Negative
 - Affect Analysis (AA): Emotion types

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 - "John loves his red shiny Porshe boxster."



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 - Sentiment Analysis (SA): Positive/Negative
 - Affect Analysis (AA): Emotion types

"John loves his red shiny Porshe boxster."

SA: attitude=Positive, opinionobject=Porshe boxster (red)AA: emotion=love, emotionobject=Porshe boxster (red)



• Blogs have been recognized as a good source of information on attitudes and emotions.

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- [1] Quan, C. and Ren, F. 2010. "A blog emotion corpus for emotional expression analysis in Chinese", *Computer Speech & Language*, Vol. 24, Issue 4, pp. 726-749.
- [2] Hashimoto, C., Kurohashi, S., Kawahara, D., Shinzato, K. and Nagata, M. 2011. "Construction of a Blog Corpus with Syntactic, Anaphoric, and Sentiment Annotations" [in Japanese], *Journal of Natural Language Processing*, Vol. 18, No. 2, pp. 175-201.
- [3] Aman, S. and Szpakowicz, S. 2007. "Identifying Expressions of Emotion in Text", LNAI 4629, pp. 196-205.
- [4] 阿部修也, 江口萌, 隅田飛鳥, 大崎梓, 乾健太郎. みんなの経験: ブログから抽出したイベントおよびセンチメントのDB化. 言語処理学会 第15回年次大会発表論文集, pp.296-299, March 2009.
- [5] Fujimura, S., Fujimura, K. Okuda, H., ``Blogosonomy: Autotagging Any Text Using Bloggers' Knowledge", IEEE/WIC/ACM International Conference on Web Intelligence, pp. 205 212, 2007,
- [6] G. Leshed and J. Kaye, "Understanding how bloggers feel: recognizing affect in blog posts", CHI'06 extended abstracts on Human factors in computing systems, pp. 1019-1024, 2006.
- [7] J. Li and F. Ren, "Emotion Recognition from Blog Articles", IEEE NLP-KE 2008, pp. 355-362, Beijing, Oct. 2008.
- [8] R. Mihalcea and H. Liu. "A corpus-based approach to finding happiness", In Proceedings of Computational approaches for analysis of weblogs, AAAI Spring Symposium, 2006.
- [9] Dipankar Das, Sivaji Bandyopadhyay, ``Labeling Emotion in Bengali Blog Corpus ? A Fine Grained Tagging at Sentence Level", Proceedings of the 8th Workshop on Asian Language Resources, pages 47-55, 2010.

• It would be good to have a big blog corpus annotated with emotions/attitudes.

• Let's compare Existing Emotion Corpora

Let's compare Existing Emotion Corpora

[10] Changqin Quan, Fuji Ren, A blog emotion corpus for emotionalexpression analysis in Chinese, Computer Speech & Language, Volume 24, Issue 4, October **2010**, Pages 726-749.

[11] Janyce Wiebe, Theresa Wilson, and Claire Cardie (**2005**). Annotatingexpressions of opinions and emotions in language. Language Resources and Evaluation, volume 39, issue 2-3, pp.165-210.

[12] Chikara Hashimoto, Sadao Kurohashi, Daisuke Kawahara, Keiji Shinzato and Masaaki Nagata,

"Construction of a Blog Corpus with Syntactic, Anaphoric, and Sentiment Annotations" [in Japanese], Journal of Natural Language Processing, Vol 18, No. 2, pp. 175-201, **2011**.

[13] J. Minato, D. B. Bracewell, F. Ren and S. Kuroiwa, "StatisticalAnalysis of a Japanese Emotion Corpus for Natural LanguageProcessing", LNCS 4114, **2006**.

[14] Aman, S. and Szpakowicz, S. (**2007**). Identifying Expressions of Emotion in Text. In Proceedings of the 10th InternationalConference on Text, Speech, and Dialogue (TSD-2007), Plzen,Czech Republic, Lecture Notes in Computer Science (LNCS),Springer-Verlag.

[15] Dipankar Das, Sivaji Bandyopadhyay, "Labeling Emotion inBengali Blog Corpus - A Fine Grained Tagging at SentenceLevel", Proceedings of the 8th Workshop on Asian LanguageResources, pages 47-55, 2010.
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[17] Gilad Mishne, Experiments with Mood Classification in BlogPosts. In: Style2005 the 1st Workshop on Stylistic Analysis ofText for Information Access, at SIGIR 2005, August **2005**.

• Let's compare Existing Emotion Corpora

Number of sentences | number of documents

[10]	12,724	500
[11]	10,657	535
[12]	4,186	249
[13]	1,191	(separate sentences)
[14]	5,205	173
[15]	12,149	123
[16]	4,773	(separate sentences)
[17]	?	815,494

Let's compare Existing Emotion Corpora

Number of sentences | number of documents

500	12,724	[10]
6 out of	10,657	[11]
8 are	4,186	[12]
based on htences)	1,191	[13]
blogs	5,205	[14]
	12,149	[15]
(separate sentences)	4,773	[16]
815,494	?	[17]

• Let's compare Existing Emotion Corpora

Number of sentences | number of documents

[10]	12,724	5	00
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[13]	1,191	based on	hte8care
[14]	5,205	blogs	Japanese
[15]	12,149		>
[16]	4,773	(separa	te sentences)
[17]	?	83	15,494

- Let's compare Existing Emotion Corpora
 - Emotion blog corpora are rather small
 - Japanese emotion blog corpora are the smallest

 Let's compare Existing Emotion Corpora Amount of annotations

• Let's compare Existing Emotion Corpora

Amount of		annotated a	syntactic				
annotations	emotion classes	emotive expressions	emotive/ non-emot.	valence/ activation	emotion intensity	emotion objects	annota- tions
annotations	[10] 8	0	0	O/×	0	0	T,POS;
	[11] 0	\bigcirc	0	\bigcirc/\times	0	\bigcirc	T,POS;
•	[12] 0	0	×	\bigcirc/\times	×	0	T,POS,L, DP,NER;
•	[13] 8	0	0	\times / \times	×	×	POS;
	[14] 6	0	0	\times/\times	0	×	×
	[15] 6	0	×	\times / \times	0	×	×
•	[16] 9	0	×	\times/\times	×	×	×
	[17] 123	×	×	\times / \times	×	×	×

• Let's compare Existing Emotion Corpora

notion e						syntactic	
lasses ex	emotive pressions	emotive/ non-emot.	valence/ activation	emotion intensity	emotion objects	annota- tions	
0	0	0	O/×	0	0	T,POS;	
0	0	\bigcirc	\bigcirc/\times	\bigcirc	\bigcirc	T,POS;	
0	0	×	O/×	×	0	T,POS,L, DP,NER;	
8	0	0	\times/\times	×	×	POS;	
6	0	0	\times/\times	0	×	×	
6	0	×	\times/\times	0	×	×	
9	0	X		S	ome are	focused on	
123	×	×	\times/\times	eı	emotion annotations		
	asses ex o 0 0 0 8 6 6 9 123 .	asses expressions a O 0 O 0 O 0 O 0 O 0 O 0 O 0 O 0 O 0 O 0 O 0 O 0 O 0 O 0 O 8 O 6 O 9 O 123 ×	assesexpressionsnon-emot. \circ \bigcirc \bigcirc 0 \bigcirc \bigcirc 0 \bigcirc \land 0 \bigcirc \land 0 \bigcirc \land 6 \bigcirc \bigcirc 6 \bigcirc \land 6 \bigcirc \land 9 \bigcirc \times 123 \times \times	assesexpressions non-emot.activation \circ \bigcirc \bigcirc \bigcirc/\times 0 \bigcirc \bigcirc/\times 0 \bigcirc \land/\times 0 \bigcirc \times/\times 8 \bigcirc \bigcirc/\times 6 \bigcirc \times/\times 6 \bigcirc \times/\times 9 \bigcirc \times 123 \times \times/\times	asses expressions non-emot. activation intensity \circ \circ \circ \circ \circ /\times \circ	asses expressions non-emot. activation intensity objects \circ \circ \circ \circ \circ \circ \circ \circ \circ \circ	

- We need an emotion corpus that is:
- 1. Large
- 2. Annotated with syntax/morphology
- 3. Annotated with emotions

- YACIS corpus
- 1. 354 mil. Sentences in 13 mil. Documents.
- 2. Tokenized, lemmatized, annotated with POS, dependency structure, named entities.

- YACIS is:
- 1. Large 🗹
- 2. Annotated with syntax/morphology 🗹
- 3. Annotated with emotions

- Tools
 - ML-Ask: Affect Analysis system for text-based utterances
 - CAO: Emoticon Analysis system

- ML-Ask
 - Distinguish emotive sentences from non-emotive
 - Annotate emotive expressions in emotive sentences
 - Generalize emotive expressions into:
 - Emotion classes
 - Valence/activation

 Michal Ptaszynski, Pawel Dybala, Rafal Rzepka and Kenji Araki, "Affecting Corpora:Experiments with Automatic Affect Annotation System - A Case Study of the 2channel Forum -" The Conference of the Pacific Association for Computational Linguistics (PACLING-09), September 1-4, 2009, Hokkaido University, Sapporo, Japan, pp. 223-228.
 Michal Ptaszynski, Pawel Dybala, Wenhan Shi, Rafal Rzepka and Kenji Araki, "A System for Affect Analysis of Utterances in Japanese Supported with Web Mining", Journal of Japan Society for Fuzzy Theory and Intelligent Informatics, Special Issue on Kansei Retrieval, Vol. 21, No. 2 (April), pp. 30-49 (194-213), 2009.

• ML-Ask

High Accuracy

- Distinguish emotive sentences from non-emotive
- Annotate emotive expressions in emotive sentences
 High Precision, Low Recall
- Generalize emotive expressions into:
 - Emotion classes
 - Valence/activation

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- CAO
 - Detect emoticons in sentence
 - Extract/analyze emoticons
 - Annotate sentences containing emoticons with
 - Emotion classes
 - Valence/activation

[1] Michal Ptaszynski, Jacek Maciejewski, Pawel Dybala, Rafal Rzepka and Kenji Araki, "CAO: A Fully Automatic Emoticon Analysis System Based on Theory of Kinesics", In IEEE Transactions on Affective Computing, vol. 1, no. 1, pp. 46-59, Jan.-June 2010

[2] Michal Ptaszynski, Jacek Maciejewski, Pawel Dybala, Rafal Rzepka and Kenji Araki, "CAO: A Fully Automatic Emoticon Analysis System", In Proceedings of The Twenty-Fourth AAAI Conference on Artificial Intelligence (AAAI-10), pp. 1026-1032, July 11 - 15, 2010, Atlanta, Georgia, USA

[3] Michal Ptaszynski, Pawel Dybala, Rafal Rzepka and Kenji Araki, "Towards Fully Automatic Emoticon Analysis System (^o^)", In Proceedings of The Sixteenth Annual Meeting of The Association for Natural Language Processing (NLP-2010), pp. 583-586, 2010.

• CAO

High Accuracy

High Accurac

- Detect emoticons in sentence
- Extract/analyze emoticons
- Annotate sentences containing emoticons with
 - Emotion classes
 - Valence/activation *) condition: sentence must contain emoticon

[1] Michal Ptaszynski, Jacek Maciejewski, Pawel Dybala, Rafal Rzepka and Kenji Araki, "CAO: A Fully Automatic Emoticon Analysis System Based on Theory of Kinesics", In IEEE Transactions on Affective Computing, vol. 1, no. 1, pp. 46-59, Jan.-June 2010

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• System output

Input sentence: なぜかレディーががを見ると恐怖感じる(;´艸`)

ML-Ask output: なぜかレディーガガを見ると恐怖感じる(;´艸`) sentence: emotive emotemes: EMOTICON:(;´艸`) emotions:(1),FEAR:恐怖 2D: NEGATIVE, ACTIVE CAO output: Extracted emoticon: (; ´艸`) **Emoticon segmentation:** S1 | BL | S2 | ELMER | S3 | BR | S4 N/A | (| ; | ´艸` | N/A |) | N/A **Emotion score** Fear (0.02708333) Surprize (0.01973684) Dislike (0.0105364) Excitement (0.01018174) Anger (0.00703125) Sorrow (0.004665203) Shame (0.004424779) Joy (0.002962932) Fondness (0.001851166) Relief (0)

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Both systems have been evaluated separately, but not together (CAO supports ML-Ask) CAO output: Extracted emoticon: (;´艸`) **Emoticon segmentation:** S1 | BL| S2 | ELMER | S3 | BR | S4 N/A | (| ; | ´艸` | N/A |) | N/A **Emotion score** Fear (0.02708333) Surprize (0.01973684) Dislike (0.0105364) Excitement (0.01018174) Anger (0.00703125) Sorrow (0.004665203) Shame (0.004424779) Joy (0.002962932) Fondness (0.001851166) Relief (0)

Evaluation of Affective Annotations

- Data set
 - Applied earlier in evaluation of CAO in [1]
 - 1000 random sentences from YACIS
 - 418 emotive, 582 non-emotive
 - One part:

42 laypeople annotated emotion classes on sentences

[1] Michal Ptaszynski, Jacek Maciejewski, Pawel Dybala, Rafal Rzepka and Kenji Araki, "CAO: A Fully Automatic Emoticon Analysis System Based on Theory of Kinesics", In IEEE Transactions on Affective Computing, vol. 1, no. 1, pp. 46-59, Jan.-June 2010

Evaluation of Affective Annotations

• Results

	emotive/ non-emotive	emotion classes	2D (valence and activation)
ML-Ask	98.8%	73.4%	88.6%
CAO	97.6%	80.2%	94.6%
ML-Ask+CAO	100.0%	89.9%	97.5%

Evaluation of Affective Annotations

• Results	On blogs n	nany	ML-Ask was	better
	sentences c	ontain	on blogs th	an on
	emotico	ns	original dat	aset *
	em	otive/	emotion	2D (valence
	10n-	emotive	classes	and activation)
ML-Ask	98	8.8%	73.4%	88.6%
CAO	97	7.6%	80.2%	94.6%
ML-Ask+CAO	10	0.0%	89.9%	97.5%
Bot	h systems		<u>On Y</u>	ACIS blog
comb	bined were		corpu	us the two
alwa	ays better		systen	ns together
			have g	ood results.

* Proof that dataset influences results.

• Emotive / Non emotive

of emotive sentences
of non-emotive sentence
ratio (emotive/non-emotive)

233,591,502 120,408,023 1.94

 2 times more emotive sentences than non-emotive

• Emotion classes

- Emotion classes
- Three emotion classes were dominant
 - Joy 喜 (+), dislike 厭 (-), fondness 好 (+)
 - Japanese generally express more positive emotions on blogs

emotion class	# of sentences	%	emotion class	# of sentences	%
joy dislike fondness fear relief	16,728,452 10,806,765 9,861,466 3,308,288 3,104,774	31% 20% 19% 6% 6%	excitement surprize gloom anger shame	2,833,388 2,398,535 2,144,492 1,140,865 952,188	5% 5% 4% 2% 2%

in YACIS corpus

- Emotion classes
- Three emotion classes were dominant

– Joy 喜 (+), dislike 厭 (-), fondness 好 (+)

 Japanese generally express more positive emotions on blogs

* In previous research, on 2channel two dominating emotions were: dislike (-) and excitement (+/-)

44,492

40,865

-2.188

4%

2%

2%

Michal Ptaszynski, Pawel Dybala, Rafal Rzepka and Kenji Araki, "Affecting Corpora:Experiments with Automatic Affect Annotation System - A Case Study of the 2channel Forum -" The Conference of the Pacific Association for Computational Linguistics (PACLING-09), September 1-4, 2009, Hokkaido University, Sapporo, Japan, pp. 223-228.

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• Emotion classes

In	lexicon:			No significant correlation
emotion	nunber of	emotion	nunber	between number of words in lexicon and frequency of
class	expressions	class	express	
dislike	532	fondness	197 (emotion class! *
excitement	269	fear	147	
sadness	232	surprise	129	* ρ=0.38
joy	224	relief	106	
anger	199	shame	65	_° ~in YACIS corpus

Its good because it	emotion class	# of sentences	%	emotion class	# of sentences	%
lexicon does not influence the results.	joy dislike fondness fear relief	16,728,452 10,806,765 9,861,466 3,308,288 3,104,774	31% 20% 19% 6% 6%	excitement surprize gloom anger shame	2,833,388 2,398,535 2,144,492 1,140,865 952,188	5% 5% 4% 2% 2%

• Comparison with other corpora

- YACIS and KNB
 - KNB:
 - 249 pages, 67,000 words
 - No emotion types, but valence/attitude and opinion-related annotations
 - Compared:
 POSITIVE SENTENCES vs. NEGATIVE SENTENCES in YACIS and KNB

KNB: Chikara Hashimoto, Sadao Kurohashi, Daisuke Kawahara, Keiji Shinzato and Masaaki Nagata, "Construction of a Blog Corpus with Syntactic, Anaphoric, and Sentiment Annotations" [in Japanese], Journal of Natural Language Processing, Vol 18, No. 2, pp. 175-201, 2011.

- YACIS and KNB
- positive negative ratid KNB* emotional 317 208 1.52attitude 489 289 1.69opinion Similar ratio was 449 264 1.70 merit acceptation 125 3.05 41 observed for the or rejection 43 63 0.68 event 1,423 865 1.65 sum two blog corpora: YACIS** 1.74only 22,381,992 12,837,728 large (YACIS) and (ML-Ask) 23,753,762 only+ 13,605,514 1.75 mostly 31,071,945 17,496,901 1.78 (ML-Askonly small (KNB). simple) only+ 32,752,589 18,442,602 1.78mostly p<.05, ** p<.01
- **KNB:** Chikara Hashimoto, Sadao Kurohashi, Daisuke Kawahara, Keiji Shinzato and Masaaki Nagata, "Construction of a Blog Corpus with Syntactic, Anaphoric, and Sentiment Annotations" [in Japanese], Journal of Natural Language Processing, Vol 18, No. 2, pp. 175-201, 2011.

• YACIS and KNB

positive negative ratid KNB* emotional 317 208 1.52attitude 489 289 1.69 opinion 449 264 1.70merit **Japanese generally** acceptation 125 3.05 41 or rejection 43 63 0.68 event express more 1,423 865 1.65 sum positive emotions 1.74 only 22,381,992 12,837,728 23,753,762 only+ 13,605,514 1.75 mostly on blogs. 31,071,945 17,496,901 only 1.78only+ 32,752,589 18,442,602 1.78(confirmation) mostly p<.05, ** p**<.0**1

KNB: Chikara Hashimoto, Sadao Kurohashi, Daisuke Kawahara, Keiji Shinzato and Masaaki Nagata, "Construction of a Blog Corpus with Syntactic, Anaphoric, and Sentiment Annotations" [in Japanese], Journal of Natural Language Processing, Vol 18, No. 2, pp. 175-201, 2011.

- YACIS and Minato et al.
 - Corpus by Minato et al. (2006) [1]
 - Examples (sentences) gathered from Emotion Dictionary [2] and analyzed.
 - Annotations (in general):
 - Emotion classes on sentences
 - POS

 [1] Minato, J., Bracewell, D. B., Ren, F. and Kuroiwa, S. 2006. "Statistical Analysis of a Japanese Emotion Corpus for Natural Language Processing", *LNCS* 4114, pp. 924-928.
 [2] Hiojima, A short distingery of feelings and emotions in English and Japanese. Tokyodo Shuppap, 100

- YACIS and Minato et al.
 - Compared: DISTRIBUTION OF EMOTION CLASSES
 - Between: YACIS, Minato et al. and Nakamura's Emotion Dictionary[3] *

*) dictionary used as affect lexicon in ML-Ask

 Minato, J., Bracewell, D. B., Ren, F. and Kuroiwa, S. 2006. "Statistical Analysis of a Japanese Emotion Corpus for Natural Language Processing", *LNCS* 4114, pp. 924-928.
 I. Hiejima. *A short dictionary of feelings and emotions in English and Japanese*, Tokyodo Shuppan, 1995.
 Nakamura, A. 1993. *Kanjo hyogen jiten*, Tokyodo Publishing

• YACIS vs. Minato et al. vs. Nakamura

	Minato et al.	YACIS	Nakamura
dislike	355	14,184,697	532
joy	295	22,100,500	224
fondness	205	13,817,116	197
sorrow	205	2,881,166	232
anger	160	1,564,059	199
fear	145	4,496,250	147
surprise	25	3,108,017	129
	Minato et al. and Nakamura	Minato et al. and YACIS	YACIS and Nakamura
Spearman's ρ	0.88	0.63	0.25

[1] Minato, J., Bracewell, D. B., Ren, F. and Kuroiwa, S. 2006. "Statistical Analysis of a Japanese Emotion Corpus for Natural Language Processing", *LNCS* 4114, pp. 924-928.

• YACIS vs. Minato et al. vs. Nakamura

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		dislike	355	14,184,697	532
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1	No significant	fondness	205	13,817,116	197
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	correlation	anger	160	1,564,059	199
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	hotwoon	surprise	25	3,108,017	129
			Minato et al.	Minato et al.	YACIS and
	YACIS and		and Nakamura	and YACIS	Nakamura
		Spearman's ρ	0.88	0.63	0.25
	inakamura *		on of previous	calculation	

* confirmation of previous calculation

good because lexicon does not influence the results

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• YACIS vs. Minato et al. vs. Nakamura

		Minato et al.	YACIS	Nakamura			
2 Medium	dislike	355	14,184,697	532			
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between YACIS	fear	145	4,496,250	147			
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and Minato		Minato et al.	Minato et al.	ACIS and			
		and Nakamura	and YACIS	Jakamura			
(could be some	Spearman's ρ	0.88	0.63	0.25			
	· ·						
cimilarities but nothing 1000/ cure)							

similarities, but nothing 100% sure)

[1] Minato, J., Bracewell, D. B., Ren, F. and Kuroiwa, S. 2006. "Statistical Analysis of a Japanese Emotion Corpus for Natural Language Processing", *LNCS* 4114, pp. 924-928.

• YACIS vs. Minato et al. vs. Nakamura

		Minato et al.	YACIS	Nakamura
3 STRONG	dislike	355	14,184,697	532
5.5110116	joy	295	22,100,500	224
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between ivilnato	fear	145	4,496,250	147
	surprise	25	3,108,017	129
and Nakamura		Minato et al.	Minato et al.	YACIS and
		and Nakamura	and YACIS	Nakamura
(both are	Spearman's ρ	0.88	0.63	0.25
1			11	

dictionaries, but differ in: time, media, collecting

person's background, approach, assumptions.....)

[1] Minato, J., Bracewell, D. B., Ren, F. and Kuroiwa, S. 2006. "Statistical Analysis of a Japanese Emotion Corpus for Natural Language Processing", *LNCS* 4114, pp. 924-928.

- What does it all mean?
 - There could be a tendency in Japanese language in general to produce expressions of some emotions
 (dislike 厭・嫌, excitement 昂, sadness 哀・悲, joy 喜,)
 than the other
 (surprise 驚, fear 怖, anger 怒)
 - The number of expressions in language is not equivalent to the frequency of usage
 - Some expressions are used more frequently (relief, fear) than the others (anger sadness)

Conclusions

- Presented a study on Emotion Corpora with a focus on Emotion Blog Corpora
- Used 2 tools for Affective Information Annotation (ML-Ask and CAO) to Annotate YACIS

The tools showed good performance on YACIS

• Presented Statistics of Affective Annotations

Conclusions

- Presented Statistics of Affective Annotations
 - Japanese express on blogs more positive emotions than negative
 - Japanese language seems to have a tendency to generate more expressions in Japanese for some emotion types than the others
 - However, number of expressions and usage frequency are not related
 - Emotion class with many expressions is sometimes expressed rarely (sadness, anger),
 - Emotion class with few expressions is sometimes expressed more often (relief, fear)

Future Work

- Online interface!
- More detailed evaluation
- N-gram version for download without limitations
- Applications
 - Affect Analysis
 - Sentiment Analysis
 - Emotion-aware dialog agent development





Thank you for your attention!

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• Let's compare Existing Emotion Corpora

corpus	scale	language	annotated affective information sy					syntactic	
name	(in senten- ces / docs)	_	emotion classes (standard)	emotive expressions	emotive/ non-emot.	valence/ activation	emotion intensity	emotion objects	annota- tions
Ren- CECps1.0 [10]	12,724 /500	Chinese	8 (Yahoo! news an- notation standard)	0	0	O/×	0	0	T,POS;
MPQA [11]	10,657 /535	English	none (no standard)	\bigcirc	\bigcirc	O/×	\bigcirc	\bigcirc	T,POS;
KNB [12]	4186 /249	Japanese	none (no standard)	0	×	O/×	×	\bigcirc	T,POS,L, DP,NER;
Minato [13] et al.	1,191 separate sentences	Japanese	8 (chosen subjectively)	0	0	\times/\times	×	×	POS;
Aman&Szpa- kowicz [14]	5205 /173	English	6 (face recognition)	\bigcirc	0	\times / \times	0	×	×
Das&Bandyo- padhayay [15]	12,149 /123	Bengali	6 (face recognition)	\bigcirc	×	\times / \times	\bigcirc	×	×
Wakamono Kotoba [16]	4773 separate sentences	Japanese	9 (6 from face reco- gnition plus 3 adde subjectively)	ed 🔿	×	\times/\times	×	×	×
Mishne [17]	815,494 blog posts	_s English	132 (LiveJournal annotation standard)	×	×	\times/\times	×	×	×