

CAO: A Fully Automatic Emoticon Analysis System

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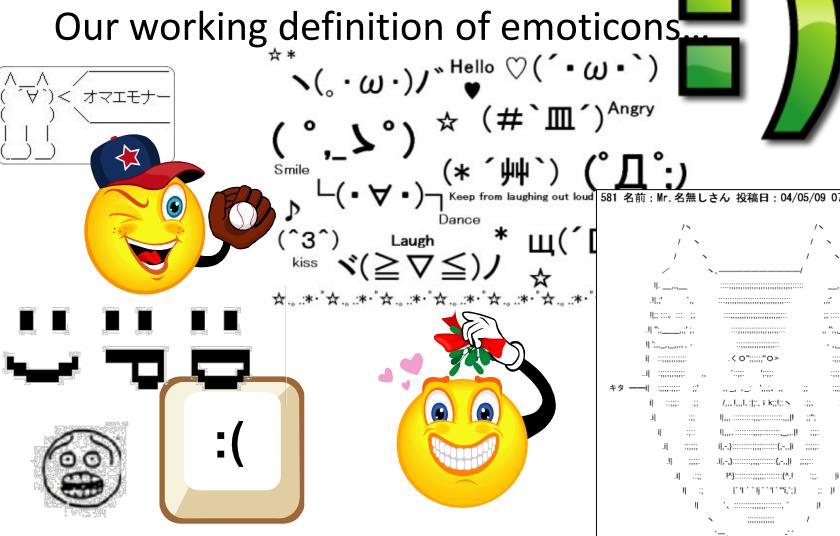
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Presentation Outline

- Emoticons Definition
- Database Construction
- CAO Emoticon Analysis System
- Evaluation of CAO
- Conclusions and Future Work

Our working definition of emoticons...





Emoticons:

 Emoticons are representations of body language in online communication (more-less).

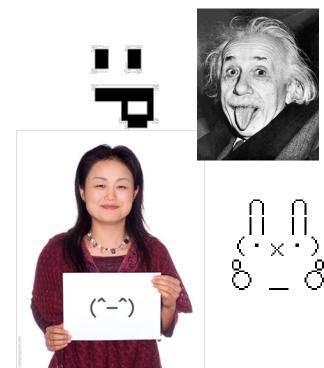








Therefore...





Emoticons:

 Are an important in part of communication [1,2] in online communities (blogs, forums, BBS, e-

mails, chat-rooms, etc.)

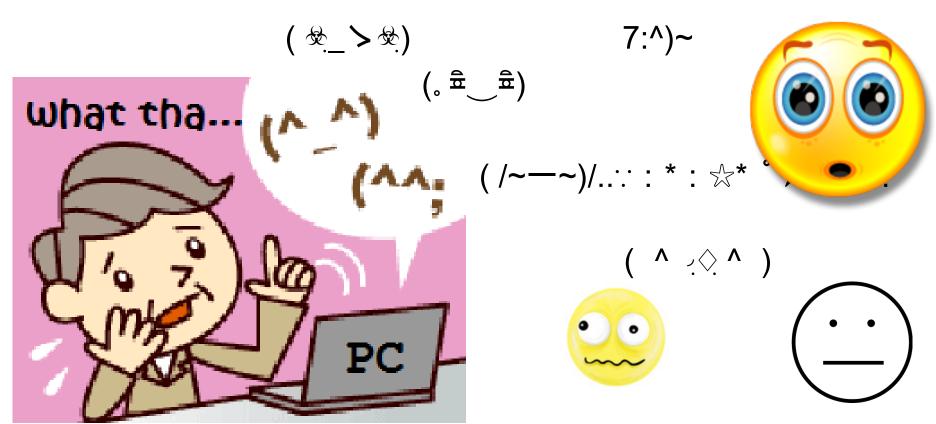


^{1.} Suzuki, N. and Tsuda, K. 2006. Automatic emoticon generation method for web community, WBC2006, pp. 331-334.

^{2.} Derks, D., Bos, A.E.R., von Grumbkow, J. 2007. Emoticons and social interaction on the Internet: the importance of social context, Computers in Human Behavior, 23, pp. 842-849.

Emoticons:

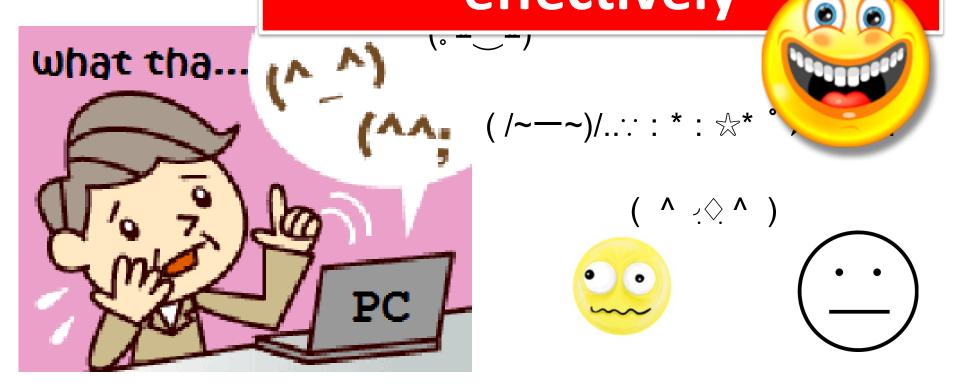
But sometimes are difficult to understand



Emoticons:

But somet

Need to analyze them effectively



- Can be roughly divided into:
 - 1-line Western (text-base or pictures)

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 - 1-line Western (text-base or pictures)











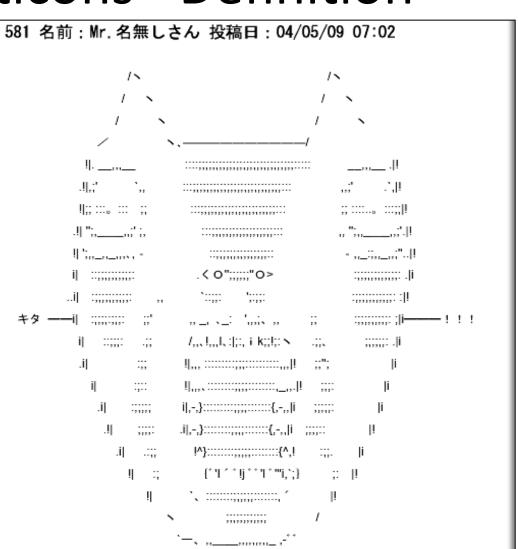
- Can be roughly divided into:
 - 1-line Western (text-base or pictures)
 - 1-line Eastern

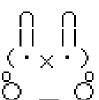
- Can be roughly divided into:
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- Can be roughly divided into:
 - 1-line Western (text-base or pictures)
 - 1-line Eastern
 - Multiline Eastern

- Can be roughly
 - 1-line Wester
 - 1-line Eastern
 - Multiline East







Emoticons:

- Can be roughly divided into:
 - 1-line Western (text-base or pictures)
 - 1-line Eastern
 - Multiline Eastern

We focused on these, because...

- Can be roughly divided into:
 - 1-line Western
- ← There already is some research+ we were a little more ambitious

- 1-line Eastern
- Multiline Eastern ← We are not that crazy

Emoticons:

- Can be roughly divided into:
 - 1-line Western
- ← There already is some research+ we were a little more ambitious

- 1-line Eastern
- Multili Eastern ← We are not that crazy

Only a little research done here

Some examples:

```
\(*^o^*)/
·°·(/Д`;)·°·
(;^_^A
(°. °)
(^ -)y--~
(。·_·。)人(。·_·。)
```

Some examples:

```
\(*^o^*)/
·°·(/Д`;)·°·
(;^ ^A
(°. °)
(^ -)y--~
(。'_'。)人(。'_'。)
```

Suddenly came inspiration!

Some examples:

```
\(*^o^*)/
·°·(/Д`;)·°·
(;^ ^A
(°. °)
(^ -)y--~
(。-_-。)人(。-_-。)
```

Suddenly came inspiration!

Since emoticons are representations of body language...

Some examples:

```
\(*^o^*)/
                        Suddenly came
·°·(/Д`;)·°
            A structural approach
(;^ ^A
              to body language
                                      s are
             could be applicable
(^ -)y--~
                                      s of
                 here as well!
(======Д==
(。'_'。)人(。'_'。)
```

Theory of kinesics:

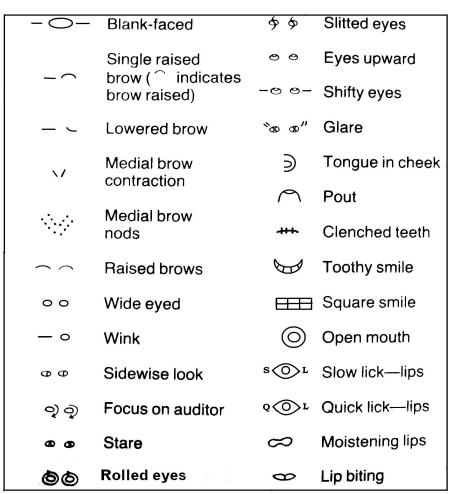
- Non-verbal behavior is used in everyday communication systematically and can be described structurally.
- A minimal part = a kineme, the smallest meaningful set of body movements, e.g. raising eyebrows, etc.

Birdwhistell (1952, 1970)

Birdwhistell, R. L. 1952. Introduction to kinesics: an annotation system for analysis of body motion and gesture, University of Kentucky Press.

Birdwhistell, R. L. 1970. Kinesics and Context, University of Pennsylvania Press, Philadelphia.

- Theory of kinesics:
- Non-verbal behavior is communication system described structurally.
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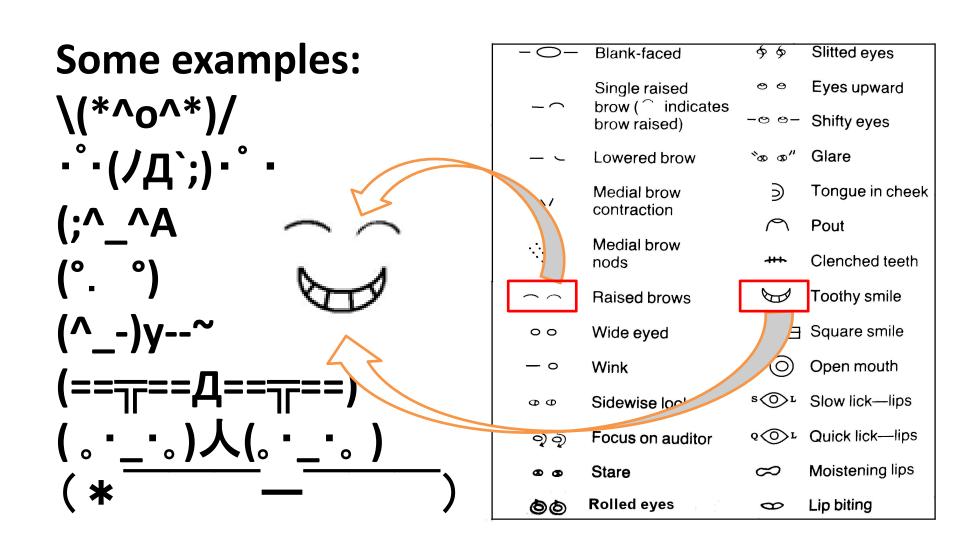
Birdwhistell, R. L. 1952. Introduction to kinesics: a

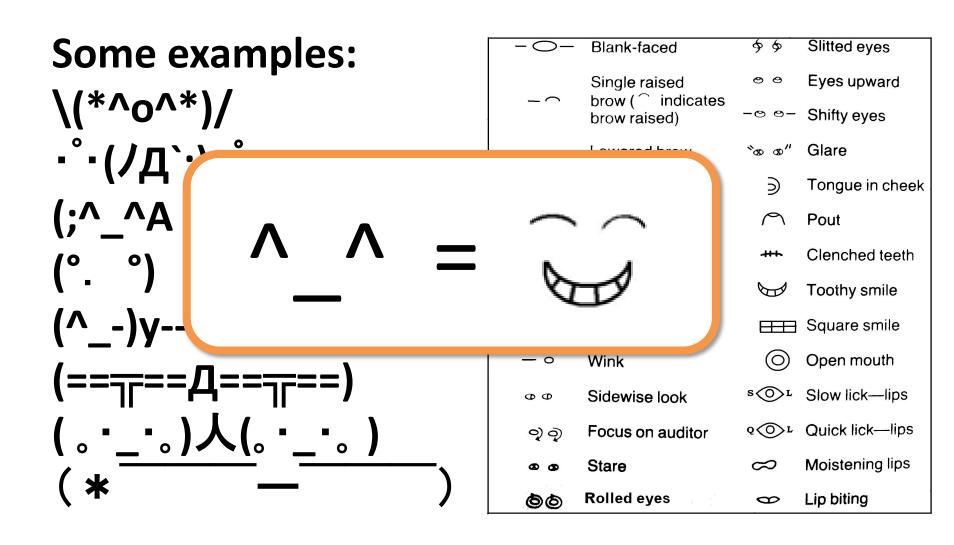
motion and gesture, University of Kentucky Press.

Birdwhistell, R. L. 1970. Kinesics and Context, University of Pennsylvania Press, Philadelphia.

Some examples:

| -0- | Blank-faced | \$ \$ | Slitted eyes | | |
|-----------------------|-------------------------------|-----------------------------------|-----------------|--|--|
| | Single raised | 00 | Eyes upward | | |
| - 0 | brow (indicates brow raised) | -0 0- | Shifty eyes | | |
| _ < | Lowered brow | °∞ ∞″ | Glare | | |
| \/ | Medial brow | € | Tongue in cheek | | |
| | contraction | \sim | Pout | | |
| *:: | Medial brow nods | ** | Clenched teeth | | |
| ~ ~ | Raised brows | | Toothy smile | | |
| 00 | Wide eyed | | Square smile | | |
| - 0 | Wink | \bigcirc | Open mouth | | |
| O O | Sidewise look | $s \bigcirc L$ | Slow lick—lips | | |
| රා රා | Focus on auditor | $\delta \bigcirc\!\!\!\bigcirc r$ | Quick lick—lips | | |
| 6 0 6 0 | Stare | \approx | Moistening lips | | |
| 6 | Rolled eyes | 8 | Lip biting | | |



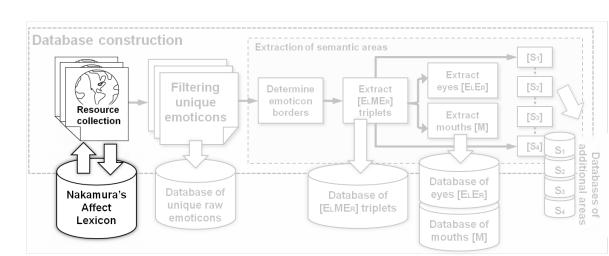


```
\(*^o^*)/
   – Additional area:
   – Bracket:
   – Additional area:
   – Face:
   – Additional area:
   – Bracket:
   – Additional area:
                                        Assumption:
```

Emoticons could be analyzed by dividing them to areas (kinemes)!

Visited 7 online emoticon dictionaries:

- 1. Face-mark Party, 2. Kaomo-jiya,
- 3. Kao-moji-toshokan, 4. Kaomoji-café,
- 5. Kaomoji Paradise, 6. Kaomojisyo and
- 7. Kaomoji Station.



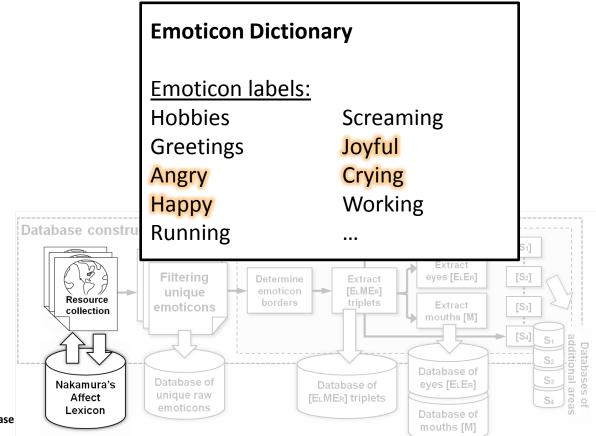
http://www.facemark.jp/facemark.htm, http://kaomojiya.com/, http://www.kaomoji.com/kao/text/,

http://kaomoji-cafe.jp/, http://rsmz.net/kaopara/,

http://matsucon.net/material/dic/,

http://kaosute.net/jisyo/kanjou.shtml

 Used an affect analysis system to select and categorize only emotion-related labels.

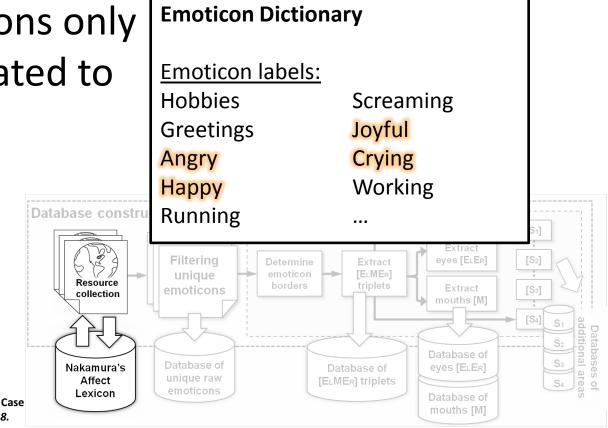


Ptaszynski, M., Dybala, P., Rzepka, R. and Araki, K. 2009. Affecting Corpora: Experiments with Automatic Affect Annota-tion System - A Case Study of the 2channel Forum, In *Proceedings PACLING-09, pp. 223-228*.

 Used an affect analysis system to select and categorize only emotion-related labels.

 Extract emotions only from labels related to emotions

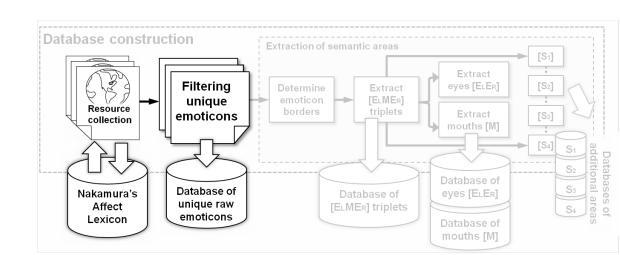
\(*^o^*)/



 Obtained 10,137 unique emoticons classified with emotion types.

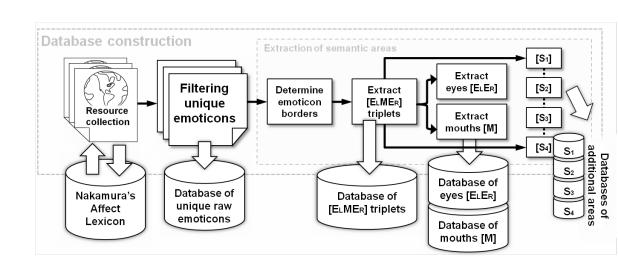
| joy, delight | liking, fondness | anger | surprise, amazement | sadness, gloom | excite- cite- ment | dis- like | shame, shyness | fear | relief | Over- all | Emoticons |
|-----------------|---------------------|-------|------------------------|-------------------|--------------------------|--------------|-------------------|------|--------|--------------|---------------|
| 3128 | 1988 | 1238 | 1227 | 1203 | 1124 | 704 | 526 | 179 | 99 | 11416 | All extracted |
| 1972 | 1972 | 1221 | 1196 | 1169 | 1120 | 698 | 511 | 179 | 99 | 10137 | Unique |
| 63% | 99% | 99% | 97% | 97% | 99% | 99% | 97% | 100% | 100% | 89% | Ratio |





- Automatically divide emoticons into:
 - Eyes [E]: ^ ^
 - Mouths [M]: o
 - Additional areas (inside emoticon) [S]: * *
 - Additional areas (outside emoticon) [S]: \ /





- We have a set of databases!
 - Raw emoticons
 - Triplets (E-M-E)
 - Eyes (E-E)
 - Mouths (M)
 - Additional (S)

| \ | (| * | Λ | 0 | Λ | * |) | / |
|---|---|---|---|---|---|---|---|---|
| 1 | 1 | | | | | | / | / |

| - | joy, delight | liking, fondness | anger | surprise, amazement | sadness, gloom | excite- | dis- like | shame, shyness | fear | relief | Over- | Emoticons |
|---|-----------------|---------------------|-----------|------------------------|-------------------|--------------|--------------|-------------------|------|--------|-------|----------------|
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| _ | 63% | 99% | 99% | 97% | 97% | 99% | 99% | 97% | 100% | 100% | 89% | Ratio |
| | areas | | E_LME_R | S_1 | B_1 | S_2 | E_{L} | E_R | M | S_3 | B_2 | S ₄ |
| | joy, d | elight | 1298 | 1469 | | 653 | 34 | 19 | 336 | 671 | | 2449 |
| | anger | • | 741 | 525 | | 321 | 18 | 38 | 239 | 330 | | 1014 |
| | sadne | SS, | 702 | 350 | | 303 | 29 | 91 | 170 | 358 | | 730 |
| | fear | | 124 | 72 | | 67 | 5. | 2 | 62 | 74 | | 133 |
| | shame shyne | | 315 | 169 | | 121 | 11 | 0 | 85 | 123 | | 343 |
| | liking fondn | * | 1079 | 1092 | | 802 | 30 |)5 | 239 | 805 | | 1633 |
| | dislik | e | 527 | 337 | | 209 | 16 | 51 | 179 | 201 | | 562 |
| | excite | ment | 670 | 700 | | 268 | 24 | 13 | 164 | 324 | | 1049 |
| | relief | | 81 | 50 | | 11 | 3 | 8 | 26 | 27 | | 64 |
| | surpr amaz | ise, ement | 648 | 405 | | 231 | 18 | 33 | 154 | 279 | | 860 |
| | overa | 11 | 6185 | 5169 | _ | 2986 | 192 | 20 | 1654 | 3192 | _ | 8837 |

Already annotated with emotion types!







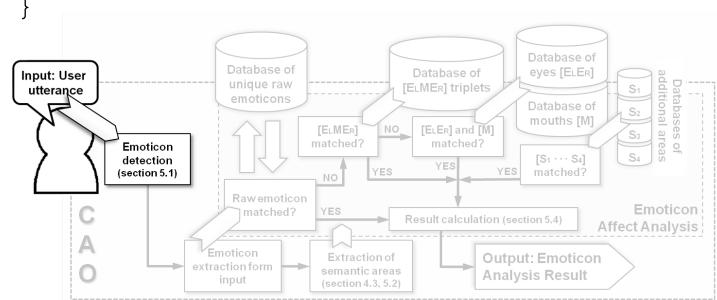


CAO – Emoticon Analysis System

Constructed CAO system for emoticon analysis with these databases.

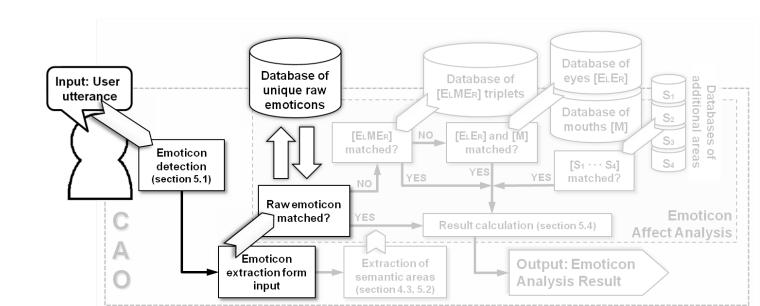
CAO – Emoticon Analysis System

- Emoticon detection in (any) input
 - Use 455 characters most frequently (>10 times)
 appearing in emoticons (x₁,x₂,...x₄₅₅)
 - If (any three x appear in a row) {
 there is an emotion in input
 }



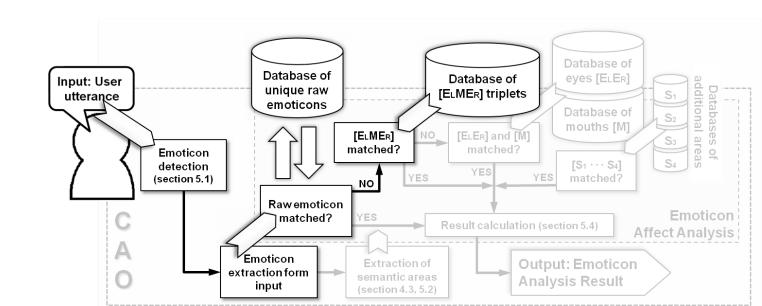
CAO – Emoticon Analysis System

- Emoticon extraction from input (+ affect analysis)
 - Three steps:
 - Looking for a "raw" emoticon (+checking emotion labels)



- Emoticon extraction from input (+ affect analysis)
 - Three steps:
 - Looking for a "raw" emoticon (+checking emotion labels)
 - Looking for a triplet (+checking emotion labels)

If no "raw" emoticon



- Emoticon extraction from input (+ affect analysis)
 - Three steps:
 - Looking for a "raw" emoticon (+checking emotion labels)
 - Looking for a triplet (+checking emotion labels)
 - Checking all combinations of triplets (eyes x mouth*)

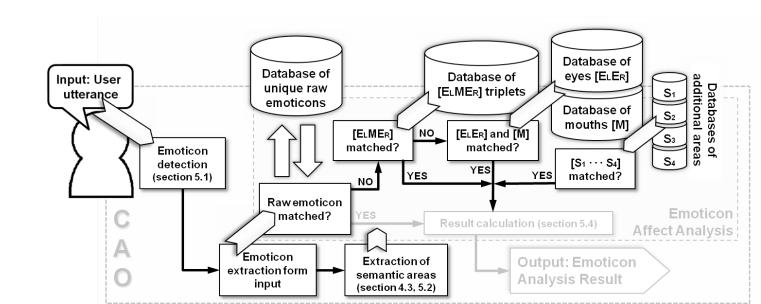
(+checking emotion labels) Database of Database of eyes [ELER] Database of Input: User unique raw [ELMER] triplets utterance emoticons Database of mouths [M] [ELMER] [ELER] and [M] matched? matched? Emoticon detection (section 5.1) ΝО Rawemoticon matched? Result calculation (section 5.4) Affect Analysis Emoticon Output: Emoticon extraction form semantic areas **Analysis Result** input (section 4.3, 5.2)

If no triplet

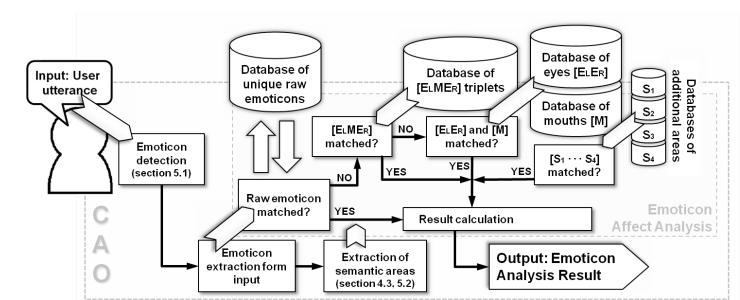
*)Eyes=1,920 Mouths=1,654 All combinations:

ExM=3,175,680

- Emoticon extraction from input
 - Finally:
 - Extract additional areas (+checking emotion labels)



- Emoticon extraction from input
 - Finally:
 - Extract additional areas (+checking emotion labels)
 - Summarize scores (to determine emotion types statistically most probable for this emoticon)

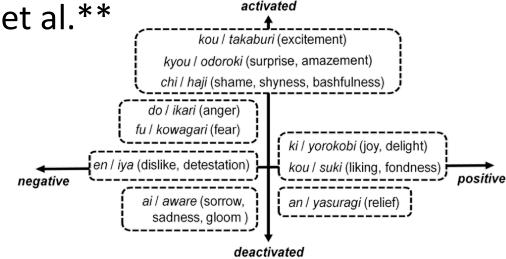


- Test set
 - A large corpus of blogs from: Ameba Blog*
 - 354,288,529 Japanese sentences in
 - 12,938,606 downloaded and parsed web pages
 - written by 60,658 unique bloggers

- Randomly extracted 1000 middle-sized* sentences as the test set
 - 418 of those sentences included emoticons.
 - annotate the sentences with 42 people (10 sentences per 1 person)
 - Question: What emotion was expressed in the sentence?
 - annotate emoticons from the sentences (different samples than in sentences)
 - Question: What emotion could be expressed with this emoticon?
 - Answers (emotion type, random order): a) System's;
 b) Similar**; c) Completely different; d) Other (from the seven remaining);

Test Set Gold standard

- Estimation of:
 - Emotion types (10 types)
 - General emotive features (valence and activation)* adjusted to Japanese like in Ptaszynski et al.**



^{*)} Russell, J. A. 1980. A circumplex model of affect, J. of Personality and Social Psychology, 39(6), pp. 1161-1178.

^{**)} Ptaszynski, M., Dybala, P., Shi, W., Rzepka, R. and Araki, K. 2009. Towards Context Aware Emotional Intelligence in Machines: Computing Contextual Appropriateness of Affective States, In *Proceedings of IJCAI-09*, pp. 1469-1474.

Results

| Detection | | | | | |
|--|-------------|----------|-------------|--|--|
| | | System | | | |
| | | Emoticon | No emoticon | | |
| Users | Emoticon | 394 | 24 | | |
| | No emoticon | 0 | 582 | | |
| No. of agreements=976 (97.6%) Kappa=0.95 | | | | | |

In 24/418 cases there were no 3 usual chars in a row

Results

| Extraction | | | | | | |
|--------------------------------|--------------------------------|--------------------|--|--|--|--|
| \overline{R} | Р | F-score | | | | |
| 94.3% | 100% (| 97.1% | | | | |
| $\left(\frac{394}{418}\right)$ | $\left(\frac{394}{394}\right)$ | $2\frac{P*R}{P+R}$ | | | | |

Errors only for the undetected emoticons

- Results
- Emotion Estimation on Separate Emoticons
 - Emotion types: 93.54%
 - General emotive features: 97.39%

Accuracy in determining probable emotion types a certain emoticon could be used to express

- Results
- Emotion Estimation on Sentences
 - Emotion types: 80.2%
 - General emotive features: 94.63%

Accuracy in determining emotion types expressed in a sentence, only with the use of emoticon*

*) a sentence needs to contain at least one emoticon

- Results
- Emotion Estimation on Sentences
 - Emotion types: 80.2%
 - General emotive features: 94.63%
- The results were
 worse because meaning in
 sentences is conveyed also
 through lexical channel;
 but,
- Results for general features
 were high → People
 sometimes misinterpret
 specific emotion type, but
 rarely valence/activation;

Accuracy in determining emotion types expressed in a sentence, only with the use of emoticon*

*) a sentence needs to contain at least one emoticon

Conclusions

- Presented a prototype system for automatic affect analysis of Eastern type emoticons, CAO.
- Inspired by Theory of Kinesics
- Gathered database of +10,000 emoticons and (almost) automatically expanded it to +3 mln.

Conclusions

- CAO is capable of:
 - Detecting emoticons in any input
 - Extracting emoticons form input
 - Dividing emoticons into semantic areas (eyes, mouths, etc.)
 - Estimating potential emotion types expressed by emoticons.
 - Affect analysis of sentences including emoticons
- CAO got almost ideal results in all tasks.

Future Work

Possible applications:

- Affect analysis/annotation of corpora
- Emotion detecting in online communication
 - Support for Internet messengers, blog services, forums, etc.
- Sentiment analysis (when looking only at valence)
- Detecting irony*

^{*)} Carvalho, P., Sarmento, L., Silva, M. J., and de Oliveira, E. 2009. Clues for detecting irony in user-generated contents: oh..!! it's "so easy";-). In Proceeding of the 1st international CIKM Workshop on Topic-Sentiment Analysis For Mass Opinion (Hong Kong, China, November 06 - 06, 2009)



Thank you for your attention!

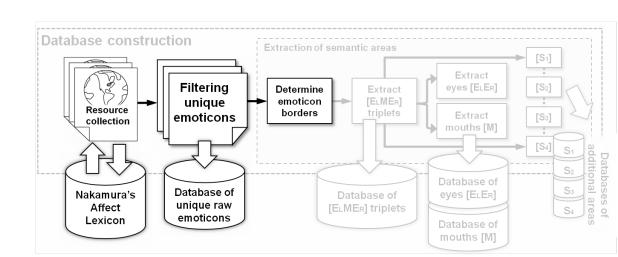
Read more in: "A Fully Automatic Emoticon Analysis System Based on Theory of Kinesics"



Details on Extraction of Emoticon Areas

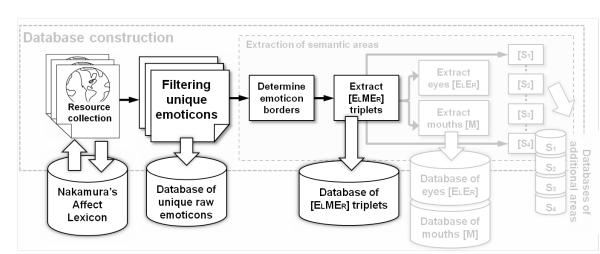
Determined all possible emoticon borders:





- Extract eye-mouth-eye triplets
 - Get rid of what is behind brackets
 (inclusively with brackets)
 - Get rid of additional areas from within emoticons (the only detail done manually)
- Make a database of emoticon triplets

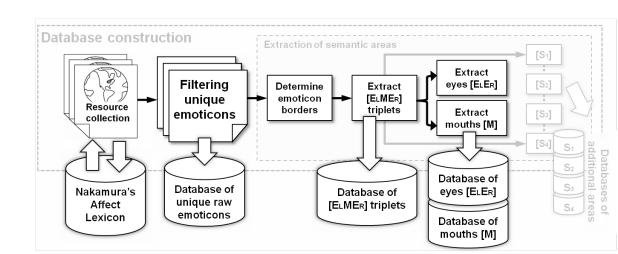




- Extract eyes and mouths
 - If an eye has more than 1 character, both eyes are the same;

```
if (n characters from left and right
match) {n=eye};
ifelse (take n-1,n-2,n-3,...)
```

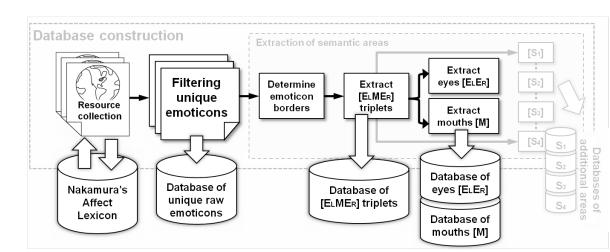
```
(*^{\circ}O^{\wedge}*)/
```



- Extract eyes and mouths
 - If an eye has 1 character, eyes could be the same or different;

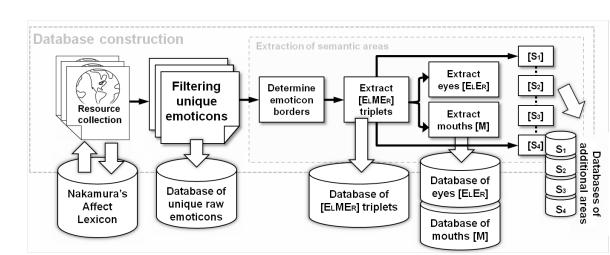
```
else(
    take 1 char. from left and right as eyes;
    mouth is what is left inside;
)
```





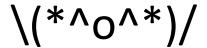
- Extract additional areas
 - Localize and extract additional areas
 - Make database of additional areas



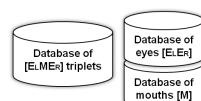


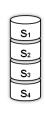
We have a database!

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| areas | | E_LME_R | S_1 | \mathbf{B}_{1} | S_2 | E_L | E_R | M | S_3 | B_2 | S_4 |
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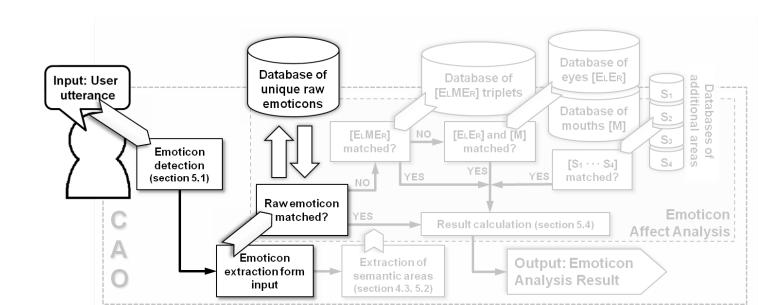




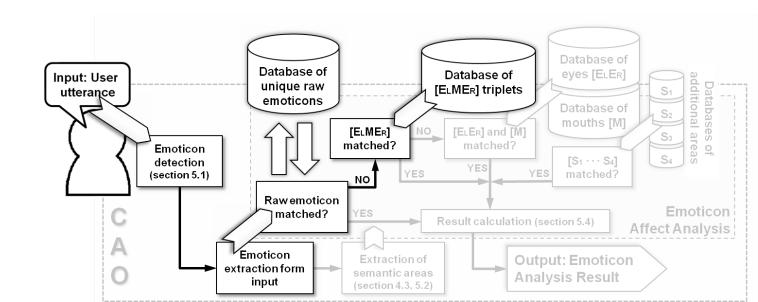
Details on Affect Analysis of Emoticons

In: CAO – Emoticon Analysis System

- Emoticon affect analysis (along with extraction)
 - Emotion list extraction
 - For [1.]: Check emotion types annotated on raw emoticons

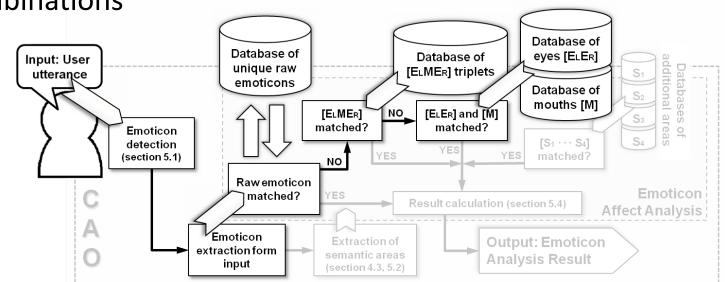


- Emoticon affect analysis
 - Emotion list extraction
 - For [1.]: Check emotion types annotated on raw emoticons
 - For [2.]: Check emotion types annotated on triplets

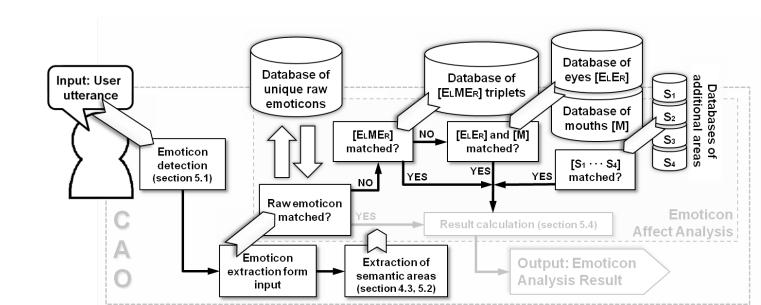


- Emoticon affect analysis
 - Emotion list extraction
 - For [1.]: Check emotion types annotated on raw emoticons
 - For [2.]: Check emotion types annotated on triplets

 For [3.]: Check emotion types annotated on separate ExM combinations

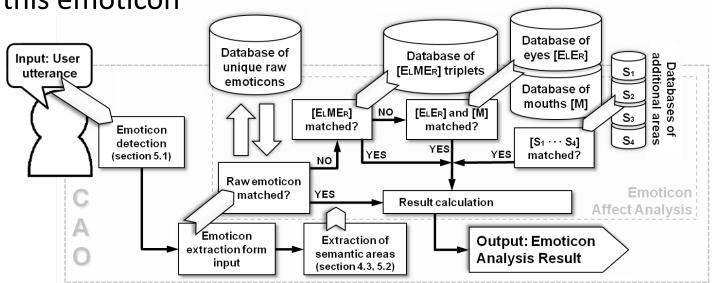


- Emoticon affect analysis
 - Finally
 - Check emotion types annotated on additional areas



- Emoticon affect analysis
 - Finally
 - Check emotion types annotated on additional areas
 - Summarize score

 Give output: list of emotions most probably expressed with this emoticon



Summarizing scores

In: Evaluation of CAO

- Summarizing scores
 - Occurrence
 - Sum of all emotion types found for all elements
 - Frequency
 - Sum for each element divided by number of all elements in each database
 - Unique frequency
 - Sum for each element divided by number of unique elements in each database

Detailed Description

- Training Set*
- Raw emoticon $\frac{1972}{63\%} \frac{1972}{99\%} \frac{1221}{99\%}$ database (Tr.S. gold standard)

| joy, delight | liking, fondness | anger | surprise, amazement | sadness, gloom | excite- cite- ment | dis- like | shame, shyness | fear | relief | Over- all | Emoticons |
|-----------------|---------------------|-------|------------------------|-------------------|--------------------------|--------------|-------------------|------|--------|--------------|---------------|
| 3128 | 1988 | 1238 | 1227 | 1203 | 1124 | 704 | 526 | 179 | 99 | 11416 | All extracted |
| 1972 | 1972 | 1221 | 1196 | 1169 | 1120 | 698 | 511 | 179 | 99 | 10137 | Unique |
| 63% | 99% | 99% | 97% | 97% | 99% | 99% | 97% | 100% | 100% | 89% | Ratio |

- Take emoticon from a database (e.g. from "joy")
- Process
- Check result with gold standard

*) In training set evaluation we matched only triplets and all possible; matching also raw would give all 100%

- Training Set*
- Raw emoticon database (Tr.S. gold
 - Take emoticon from a
 - Process
 - Check result with gold standard
 - Ranking:
 - 1. Occurrence
 - 2. Unique Frequency
 - 3. Frequency

(differences not significant=all equally good)

| | Emotion type | CAO: Occur- rence | Freq- uency | Unique Freq- uency |
|-----|--------------|-------------------------|----------------|--------------------------|
| | anger | 0.811 | 0.771 | 0.767 |
| | dislike | 0.631 | 0.800 | 0.719 |
| a ¯ | excitement | 0.786 | 0.769 | 0.797 |
| | fear | 0.451 | 0.936 | 0.858 |
| | fondness | 0.915 | 0.778 | 0.783 |
| | joy | 0.944 | 0.802 | 0.860 |
| | relief | 0.600 | 0.990 | 0.985 |
| , | shame | 0.706 | 0.922 | 0.910 |
| | sorrow | 0.814 | 0.809 | 0.791 |
| | surprise | 0.862 | 0.866 | 0.874 |
| - | All approx. | 0.852 | 0.804 | 0.818 |

- Test set
 - A large corpus of blogs from: Ameba Blog*
 - 354,288,529 Japanese sentences in
 - 12,938,606 downloaded and parsed web pages
 - written by 60,658 unique bloggers
- Randomly extracted 1000 middle-sized** sentences as the test set
 - 418 of those sentences included emoticons.

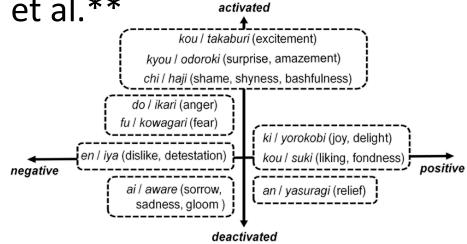
Test Set Gold standard

- annotate the sentences with 42 people (10 sentences per 1 person) Question: What emotion was expressed in the sentence?
- annotate emoticons from the sentences (different samples than in sentences)

Question: What emotion could be expressed with this emoticon?

- *) www.ameblo.co.jp
- **)20-50 characters in Japanese

- Estimation of:
 - Emotion types (specific)
 - General emotive features (valence and activation)* adjusted to Japanese like in Ptaszynski et al.**



^{*)} Russell, J. A. 1980. A circumplex model of affect, J. of Personality and Social Psychology, 39(6), pp. 1161-1178.

^{**)} Ptaszynski, M., Dybala, P., Shi, W., Rzepka, R. and Araki, K. 2009. Towards Context Aware Emotional Intelligence in Machines: Computing Contextual Appropriateness of Affective States, In *Proceedings of IJCAI-09*, pp. 1469-1474.

Results

| Detection | | | | | | | |
|---|-------------|----------|-------------|--|--|--|--|
| System | | | | | | | |
| | | Emoticon | No emoticon | | | | |
| Users | Emoticon | 394 | 24 | | | | |
| | No emoticon | 0 | 582 | | | | |
| No. of agreements=976 (97.6%), Kappa=0.95 | | | | | | | |

Results

| Extraction | | | | | | | | |
|--------------------------------|--------------------------------|--------------------|--|--|--|--|--|--|
| \overline{R} | Р | F-score | | | | | | |
| 94.3% | 100% | 97.1% | | | | | | |
| $\left(\frac{394}{418}\right)$ | $\left(\frac{394}{394}\right)$ | $2\frac{P*R}{P+R}$ | | | | | | |

Results

Emotion Estimation on Separate Emoticons

| CAO | | | | | | | | | |
|------------|----------|----------|----------|--------------------|----------|--|--|--|--|
| Occurrence | | Frequ | uency | Unique Frequency 😃 | | | | | |
| Types | 2D space | Types | 2D space | Types | 2D space | | | | |
| 0.891472 | 0.966778 | 0.934319 | 0.971044 | 0.935364 | 0.973925 | | | | |

Emotion Estimation on Sentences

| CAO | | | | | | | | | |
|----------|----------|----------|----------|------------------|----------|--|--|--|--|
| Occui | rrence | Freq | uency | Unique Frequency | | | | | |
| Types | 2D space | Types | 2D space | Types | 2D space | | | | |
| 0.755171 | 0.908911 | 0.800896 | 0.940582 | 0.802012 | 0.946291 | | | | |

- 1. Unique Frequency
- 2. Frequency
- 3. Occurrence

Comparing CAO to other systems

In: Evaluation of CAO

Comparing CAO to other systems

| $egin{aligned} \mathbf{Research} & 	o \ (\mathbf{approach}) \ \mathbf{Capability} \ \downarrow \end{aligned}$ | Tanaka et al. (2005) (kernel methods) | Yamada et al. (2007) (n-grams) | Kawakami (2008) (database) | CAO (theory of kinesics) |
|---|---|---------------------------------------|----------------------------------|---|
| 1. Detection whether input equals emoticon | × | × | Х | O |
| 2. Detection of emoticon in sentence input | O (included in 3.) | X | Х | 0 |
| 3. Extraction of emoticon from any string of characters | 0 | X | Х | 0 |
| 4. Division into semantic areas | X | × | × | 0 |
| 5. Database coverage | 1,075 | 693 | 31 | 10,137 (expanded automatically to over 3 million) |
| 6. Classification of emotion types | 6 types (BBS-based; | 7 types Subjective (Subjective) | 6 types (Subjective) | 10 types (Language/ Culture Based) |
| 7. Emotion esti- mation of separate emoticons | O (included in 8.) | 0 | 0 | 0 |
| 8. Affect Analysis of sentences with emoticons | О | × | Х | 0 |

- Comparing CAO to other systems
 - In Training set:Comparison with Yamada et al. (2007)

Upgraded with our database and emotion classification

Comparing CAO to other systems

In Training
set:
Comparison
with Yamada
et al. (2007)
(their best <
our worst)

| E 4: | Yamada et al (2007) | | | CAO: | | Unique |
|-----------------|---------------------|-------|-------|--------|-------|--------|
| Emotion type | 1- | 2- | 3- | Occur- | Freq- | Freq- |
| турс | gram | gram | gram | rence | uency | uency |
| anger | 0.702 | 0.815 | 0.877 | 0.811 | 0.771 | 0.767 |
| dislike | 0.661 | 0.809 | 0.919 | 0.631 | 0.800 | 0.719 |
| excitement | 0.700 | 0.789 | 0.846 | 0.786 | 0.769 | 0.797 |
| fear | 0.564 | 0.409 | 0.397 | 0.451 | 0.936 | 0.858 |
| fondness | 0.452 | 0.436 | 0.448 | 0.915 | 0.778 | 0.783 |
| joy | 0.623 | 0.792 | 0.873 | 0.944 | 0.802 | 0.860 |
| relief | 1.000 | 0.999 | 1.000 | 0.600 | 0.990 | 0.985 |
| shame | 0.921 | 0.949 | 0.976 | 0.706 | 0.922 | 0.910 |
| sorrow | 0.720 | 0.861 | 0.920 | 0.814 | 0.809 | 0.791 |
| surprise | 0.805 | 0.904 | 0.940 | 0.862 | 0.866 | 0.874 |
| All approx. | 0.675 | 0.751 | 0.802 | 0.852 | 0.804 | 0.818 |

T. Yamada, S. Tsuchiya, S. Kuroiwa, F. Ren, "Classification of Facemarks Using N-gram", International Conference on NLP and Knowledge Engineering, pp. 322-327, 2007.

- Comparing CAO to other systems
 - In Test set: their best < our worst (or 2nd worst)

| | Emotion Estimation on Separate Emoticons | | | | | | | | | |
|----------|--|----------|----------|----------------------|-----------|----------|------------------|----------|--|--|
| Ya | mada et al. (20 | 007) | CAO | | | | | | | |
| 1 gram | 2 gram | 3 gram | Occur | rence | Frequ | iency | Unique Frequency | | | |
| 1-gram | 2-gram | 3-gram | Types | 2D space | Types | 2D space | Types | 2D space | | |
| 0.721347 | 0.865117 | 0.877049 | 0.891472 | 0.966778 | 0.934319 | 0.971044 | 0.935364 | 0.973925 | | |
| | | | Emotion | Estimation on | Sentences | | | | | |
| Ya | mada et al. (20 | 007) | | | CA | 10 | | | | |
| 1 gram | 2 gram | 3 gram | Occur | rence | Frequ | iency | Unique Frequency | | | |
| 1-gram | 2-gram | 3-gram | Types | 2D space | Types | 2D space | Types | 2D space | | |
| 0.685714 | 0.797659 | 0.714819 | 0.755171 | 0.908911 | 0.800896 | 0.940582 | 0.802012 | 0.946291 | | |
| | | | | | | | | | | |

T. Yamada, S. Tsuchiya, S. Kuroiwa, F. Ren, "Classification of Facemarks Using N-gram", International Conference on NLP and Knowledge Engineering, pp. 322-327, 2007.

Statistical significance of results

Statistical significance or differences in training set evaluation

| Emotion | Yamada (| et al (2007) i | mproved | CAO: | Unique | | Unique | |
|------------|----------|----------------|---------|------------|-----------|-----------|----------|----------|
| type | 1-gram | 2-gram | 3-gram | Occurrence | Frequency | Frequency | Position | Position |
| anger | 0.702 | 0.815 | 0.877 | 0.811 | 0.771 | 0.767 | 0.476 | 0.476 |
| dislike | 0.661 | 0.809 | 0.919 | 0.631 | 0.8 | 0.719 | 0.556 | 0.591 |
| excitement | 0.7 | 0.789 | 0.846 | 0.786 | 0.769 | 0.797 | 0.56 | 0.516 |
| fear | 0.564 | 0.409 | 0.397 | 0.451 | 0.936 | 0.858 | 0.652 | 0.671 |
| fondness | 0.452 | 0.436 | 0.448 | 0.915 | 0.778 | 0.783 | 0.46 | 0.389 |
| joy | 0.623 | 0.792 | 0.873 | 0.944 | 0.802 | 0.86 | 0.522 | 0.421 |
| relief | 1 | 0.999 | 1 | 0.6 | 0.99 | 0.985 | 0.599 | 0.621 |
| shame | 0.921 | 0.949 | 0.976 | 0.706 | 0.922 | 0.91 | 0.538 | 0.566 |
| sorrow | 0.72 | 0.861 | 0.92 | 0.814 | 0.809 | 0.791 | 0.553 | 0.52 |
| surprise | 0.805 | 0.904 | 0.94 | 0.862 | 0.866 | 0.874 | 0.52 | 0.523 |
| All approx | . 0.675 | 0.751 | 0.802 | 0.852 | 0.804 | 0.818 | 0.517 | 0.469 |
| | X | Y | Z | А | В | С | | |

| XvsY | not sign | AvsB | not sign |
|------|----------|------|----------|
| XvsZ | not sign | AvsC | not sign |
| YvsZ | not sign | BvsC | not sign |

In most cases differences were not significant (results equally good).

| | A | В | С |
|---|---|---------|---------|
| X | N | Y (<5%) | Y (<5%) |
| Υ | N | Ν | N |
| Z | N | N | N |

This means that although in training set Occurrence scored higher, Freq and UniqFreq were more probable to achieve better results in test set evaluation.