



JAGIELLONIAN UNIVERSITY
IN KRAKOW

4th International Conference
on Affective and Pleasurable Design



Towards Joking, Humor Sense Equipped and Emotion Aware Conversational Systems

Pawel DYBALA*, Motoki Yatsu**, Michal Ptaszynski***,
Rafal Rzepka**, Kenji ARAKI**

* Jagiellonian University, Institute of Middle and Far Eastern Studies, Poland

** Hokkaido University, Graduate School of Information Science and Technology, Japan

*** Department of Computer Science, Kitami Institute of Technology, Japan



1. Martin, R. A. and Lefcourt, H. M.: "Humor and life stress. Antidote to adversity". New York, Springer Verlag, 1996

2. N.A. Yovetich, J.A. Dale, M.A. Hudak. Benefits of humor in reduction of threat-induced anxiety. *Psychological Reports*, 66(1), 51-58 (1990)

3. Richard A. Dienstbier, "The impact of humor on energy, tension, task choices, and attributions: Exploring

- We tend to use humor when discussing difficult subjects - **makes our conversation easier** ⁴



- If it is so good in human-human interaction...



- ...should also work well in human-computer interaction

- If we are going to construct naturally talking machines, we need to include humor in our research
- Important challenge for AI, NLP and HCI:

**to create
a humor-equipped talking
machine**

A humor-equipped system should be able to:

- detect user's humorous behaviors
- react to them adequately
- recognize users' emotions
- on this basis - decide whether or not use humor
- generate humorous contents
- recognize users' reactions to humorous contents
- construct users' humor sense models to individualize its performance.

- See paper for the summary
- Much has been done
- Still to be done:
 - system able to react to users' particular emotional states with humor
 - humor-equipped chatterbot with individualization module (able to adapt to users' humor sense)
 - system able to both recognize and generate humor

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GOAL: construct a system able to:

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Restrictions:

- Japanese
- Chatterbot, text based
- Humor generation: Japanese puns (*dajare*)
- Humor detection: Japanese puns (*dajare*)
- Emotions: affect-from-text recognition
- Individualisation: emotion-based

Puns.



GILSAN
SUNGLASS
DARTS
5 20.00



**DARTH
GATOR**

**GATORLAND
ORLANDO**

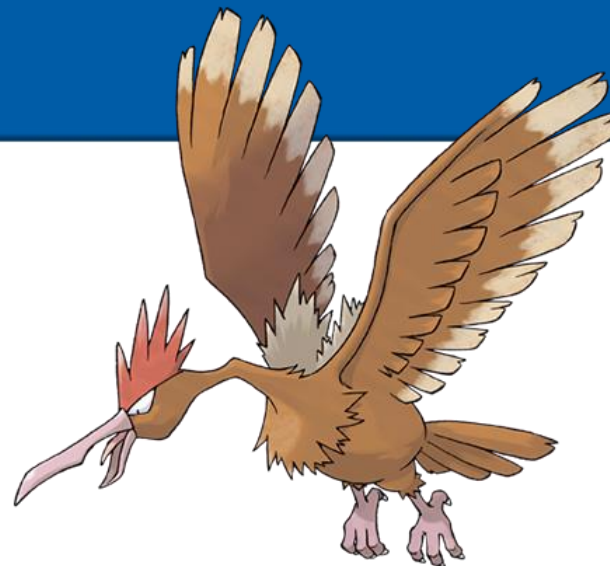


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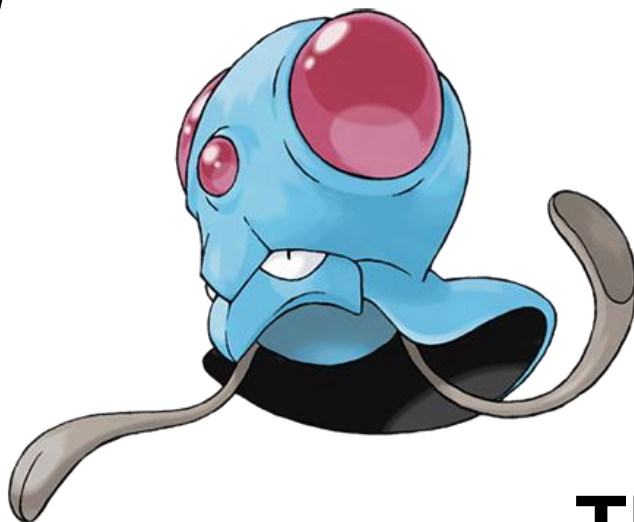




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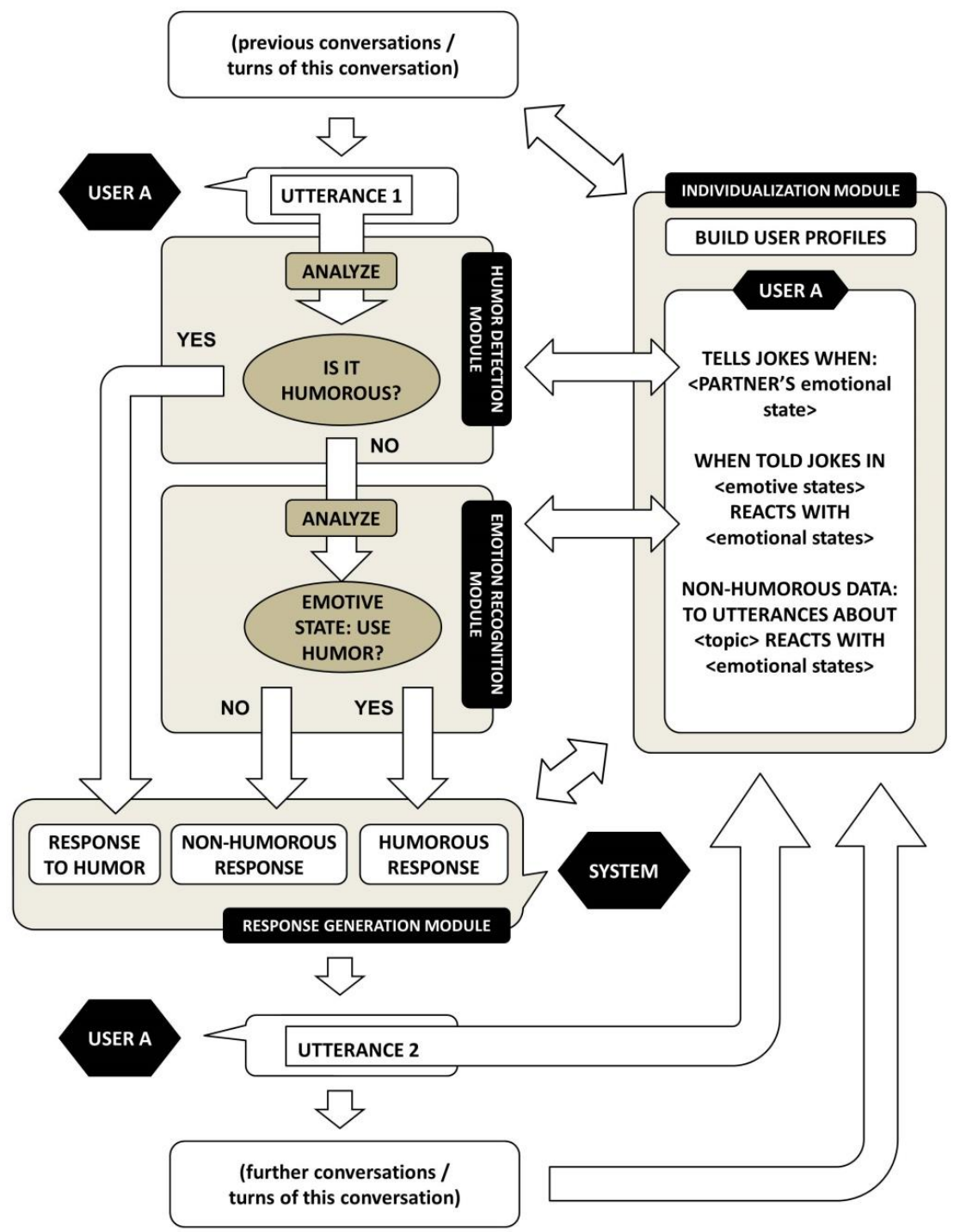
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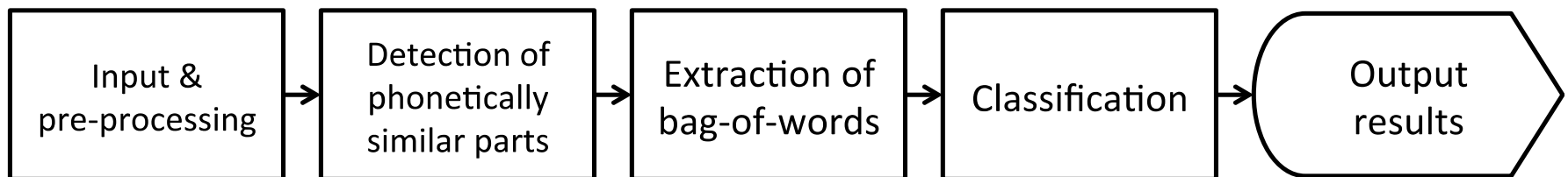
TENTACOOOL

Modules:

- Humor detection module
- Emotion recognition module
- Response generation module
 - Humorous response generator
 - Non-humorous response generator
 - Reaction to humor generator
- Individualization module



- Pun/Non-pun classifier using SVM
- Positive examples were obtained from websites collecting Japanese puns
- Negative examples were sentences with limitation of sentence length
- SVM features
 - Bag of words
 - Exact-match and similarity



20-fold cross validation of the classifier
Size of both the examples: 23,984 sentences,
divided into 22,784 versus 1,200 sentences
for training and tested data with the same
numbers of positive and negative examples

Set	BOW features	Exact-match	Similarity	Euphonic sounds	Precision	Recall	F-measure
F ₁	-	○	○	○	0.908	0.861	0.884
F ₂	○	-	-	-	0.873	0.824	0.848
F ₃	○	○	○	○	0.915	0.633	0.749

Summary:

- This pun detection algorithm can be used in our system
- Still much to be done
- Currently we are working on its full implementation, which will be evaluated experimentally in the nearest future
- Also: funniness assessment

ML-Ask system - a text-based affect analyzer for Japanese^{5, 6}

- Keyword-based
- Uses a lexicon of emotive elements and emotive expressions⁵
- Uses a web mining technique to extract emotive associations from the Internet
- Detects specified emotions from emoticons⁷

5. Ptaszynski, M., Masui, F., Dybala, P., Rzepka, R., Araki, K.: Open Source Affect Analysis System with Extensions. In: Proceedings of iHAI 2013, Sapporo, Japan (2013).

6. Ptaszynski, M., Dybala, P., Rzepka, R., Araki, K.: Affecting Corpora: Experiments with Automatic Affect Annotation System - A Case Study of the 2channel Forum. In Proceedings of PACLING-09, pp. 223--228. Hokkaido University, Sapporo, Japan (2009).

7. Ptaszynski, M., Maciejewski, J., Dybala, P., Rzepka, R., Araki, K.: CAO: A Fully Automatic Emoticon Analysis System Based on Theory of Kinesics. In: IEEE Transactions on Affective Computing, vol. 1, no. 1, 46--59 (2010).

Steps of analysis:

- 1. Specify whether an utterance is emotive** - i.e. if it contains emotive elements (like “whoa!” or “Oh!”)
- 2. Recognize the particular emotion types in utterances described as emotive** – by detecting if it contains emotive expressions, i.e. words or phrases that define specific emotional states (like “happy” or “sad”).
- 3. Use a web mining technique to extract emotion associations** from the Internet (emotive expressions that co-occur most frequently with the input). “I lost my job” -> “sad”, “angry”

Sentence: *Kono hon saa, sugee kowakatta yo.
Maji kowasugi!*

(That book, ya know, 'twas a total killer.
It was just too scary.)

Emotive elements: *saa* (emphasis), *sugee* (totally),
yo (emphasis), *maji* (really),
-sugi (too much), exclamation mark

Emotive value: 6 (above zero -> specify types of
emotions)

Emotive expressions: *kowai* (frightening)

Emotions found: **fear**

Valence: **negative**

Sentence: *Kyou wa atatakai desu ne.*
(It's warm today, isn't it?)

Emotive elements: *-ne* (-isn't it)

Emotive value: 1

(above zero -> specify types of emotions)

Emotive expressions: none

(-> use web mining procedure)

Emotions found on the Web: **joy**

Valence: **positive**

Tested experimentally

	emotive/ non-emotive	emotion classes
ML-Ask	98.8%	73.4%
CAO	97.6%	80.2%
ML-Ask+CAO	100.0%	89.9%

CAO – emoticon recognition module

3 submodules:

- **Non-humorous response generator** (chatbot) for Japanese (currently: 2 Web-based systems) ^{8,9}
- **Response to user's humor generator** – currently 5 different phrases that can be used in reaction to users' humorous utterances
- **Humorous response generator** – generates Japanese puns towards inputted sentence

8. Higuchi, Shinsuke, Rafal Rzepka, and Kenji Araki. "A casual conversation system using modality and word associations retrieved from the web." Proceedings of the Conference on Empirical Methods in Natural Language Processing. Association for Computational Linguistics, 2008.

9. Dybala, P.: Humor to Facilitate HCI. Germany: Lambert Academic Publishing (2011).

PUNDA Japanese pun (*dajare*) generator ^{9, 10}

Futon ga futtonda

Futon flew away

- Internet-based
- Uses templates to generate humorous responses towards users utterances

9. Dybala, P.: Humor to Facilitate HCI. Germany: Lambert Academic Publishing (2011).

10. Dybala, P., Ptaszynski, M., Maciejewski, J., Takahashi, M., Rzepka, R., Araki, K.: Multiagent system for joke generation: Humor and emotions combined in human-agent conversation. Journal of Ambient Intelligence and Smart Environments (Thematic Issue on Computational Modeling of Human-Oriented Knowledge within Ambient Intelligence) 2 (2010), 31--48 (2010).

Input: Sentence (utterance)

USER: *Rabu kome kirai desu yo!* (I don't like love comedies!)

**Extract a base
phrase
rabu (love)**

**Transform using
generation
patterns**

**Phonetic pun
candidates list
(*arabu, irabu,
rabua, rabui...*)**

**Use sentence
patterns to
generate
humorous
utterances**

**Check the co-occurrence with the base
phrase in the Internet
„*rabu*” „*erabu*” (chose): 2 940 000**

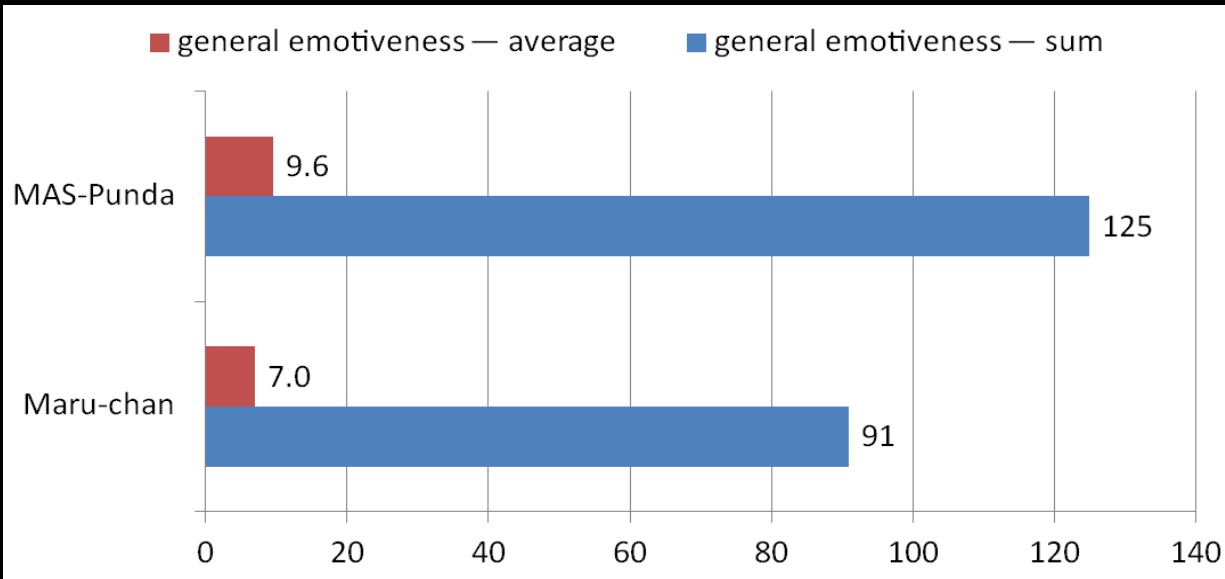
**Output: Humorous response
SYSTEM: *Rabu to ieba, erabu n da yo ne.*
(Speaking of love, you can choose that)**

- The system generates pun candidates with 75% accuracy (compared to human level)
- In our previous research implemented into a conversational system.
- Evaluation experiments showed that it was evaluated as better, more human-like and triggering more positive emotions than a similar system without humor.^{9, 10, 11}
- Still not human level – working on improvements

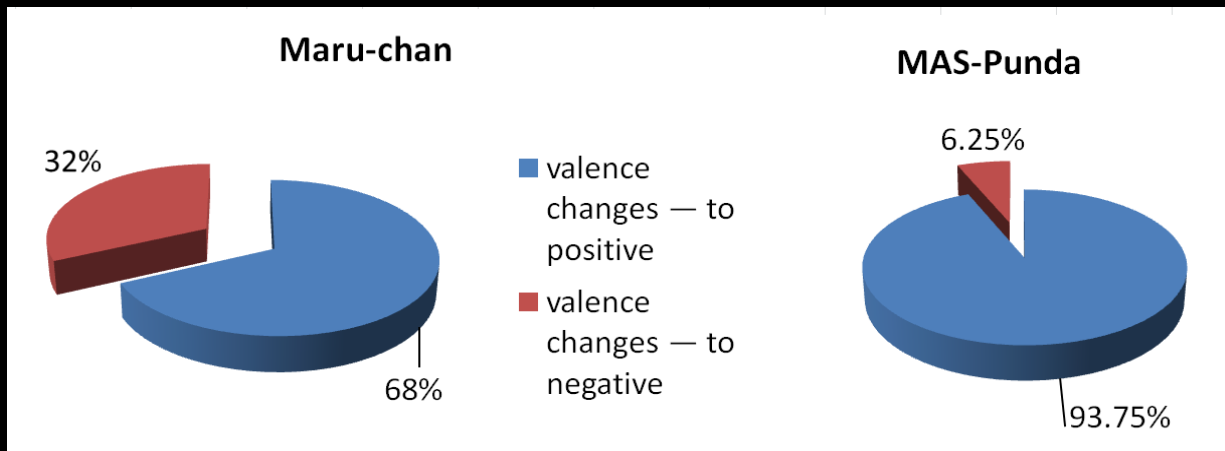
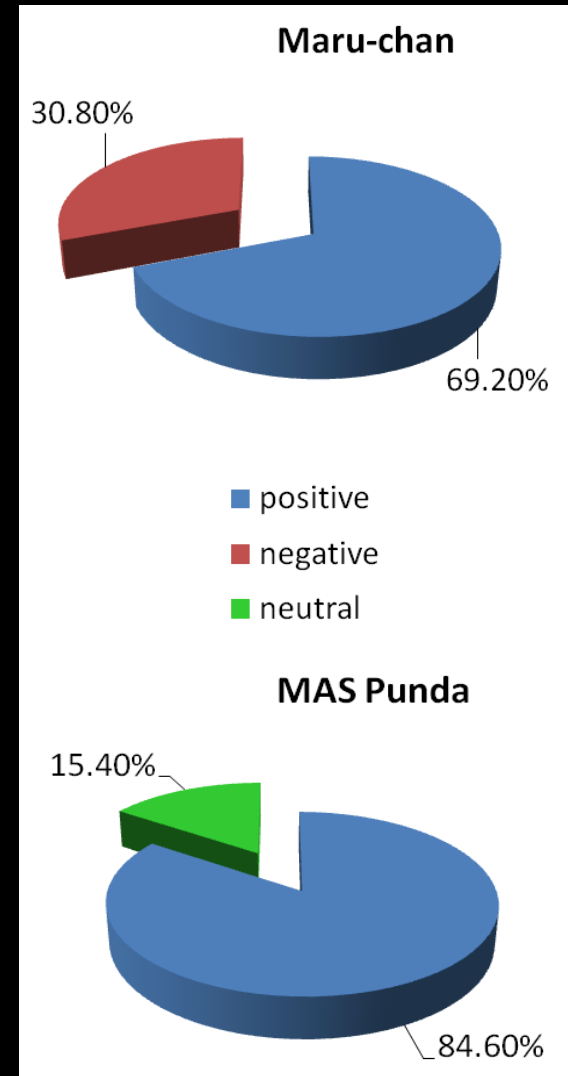
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11. Dybala, P., Ptaszynski, M., Rzepka, R., Araki, K.: Extending the Chain: Humor and Emotions in Human Computer Interaction. *International Journal of Computational Linguistics Research*, Vol. 1, Issue 3, 116--125 (2010).

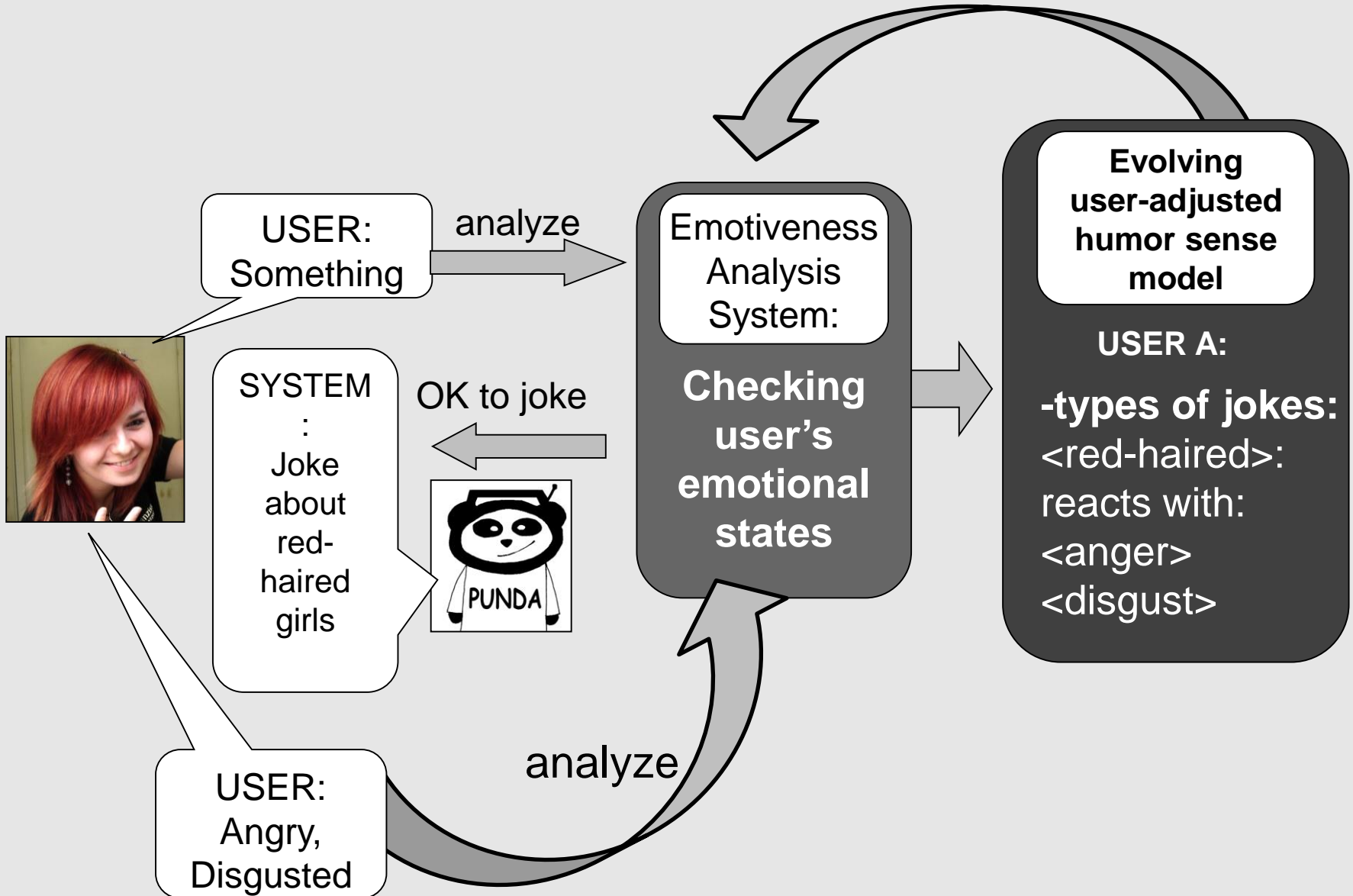


Maru-chan – chatterbot without humor
 MAS-Punda – chatterbot with humor



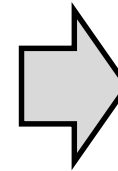
- Uses the ML-Ask affect analyzer to gather data about users' emotive states during conversations.
- On this basis the module builds each user's individual model of sense of humor.

Individualized interaction



User A

- emotion changes when told jokes in emotive state <sadness>:
 - to positive: 7/10
 - to neutral: 2/10
 - to negative: 1/10



OK to tell
jokes
when
<sad>

Emotiveness analysis based decision taking method tested experimentally:

- Implemented into a joking chatterbot
- if user's emotive state negative or neutral - use humor to make him/her feel better.
- Results: enhanced user's moods, evaluated as friendly, human-like and using humor in appropriate moments

- Summary: uses emotiveness analysis based data to individualize system's performance
- More interactions = more complex models
- Problem: at initial stage performance will not be individualized – may be discouraging
- Proposed solution: initial settings based on statistical data (e.g. which age/social groups prefer which types of jokes) – currently gathering
- Individualization module can be used also for features other than humor

- Goal: Construction of a joking, humor sense equipped and emotion aware conversational system
- Currently: Under development, working on modules separately
- In the nearest future: join them into one system, perform evaluation experiments



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Thank you for your attention
