

LOL NLP: an Overview of Computational Humor

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About myself

- Research/academia: Ural Federal University, Yekaterinburg/ HSE University, Moscow
- Past industrial experience: Yandex/SKB Kontur
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Humor is a promising area for studies of intelligence and its automation: it is hard to imagine a computer passing a rich Turing test without being able to understand and produce humor.

West & Horvitz, AAAI2019

Humor at Alexa Prize competition

...it's amazing to see that now humor is coming in... Good sense of humor is a sign of intelligence in my mind and something very hard to do.

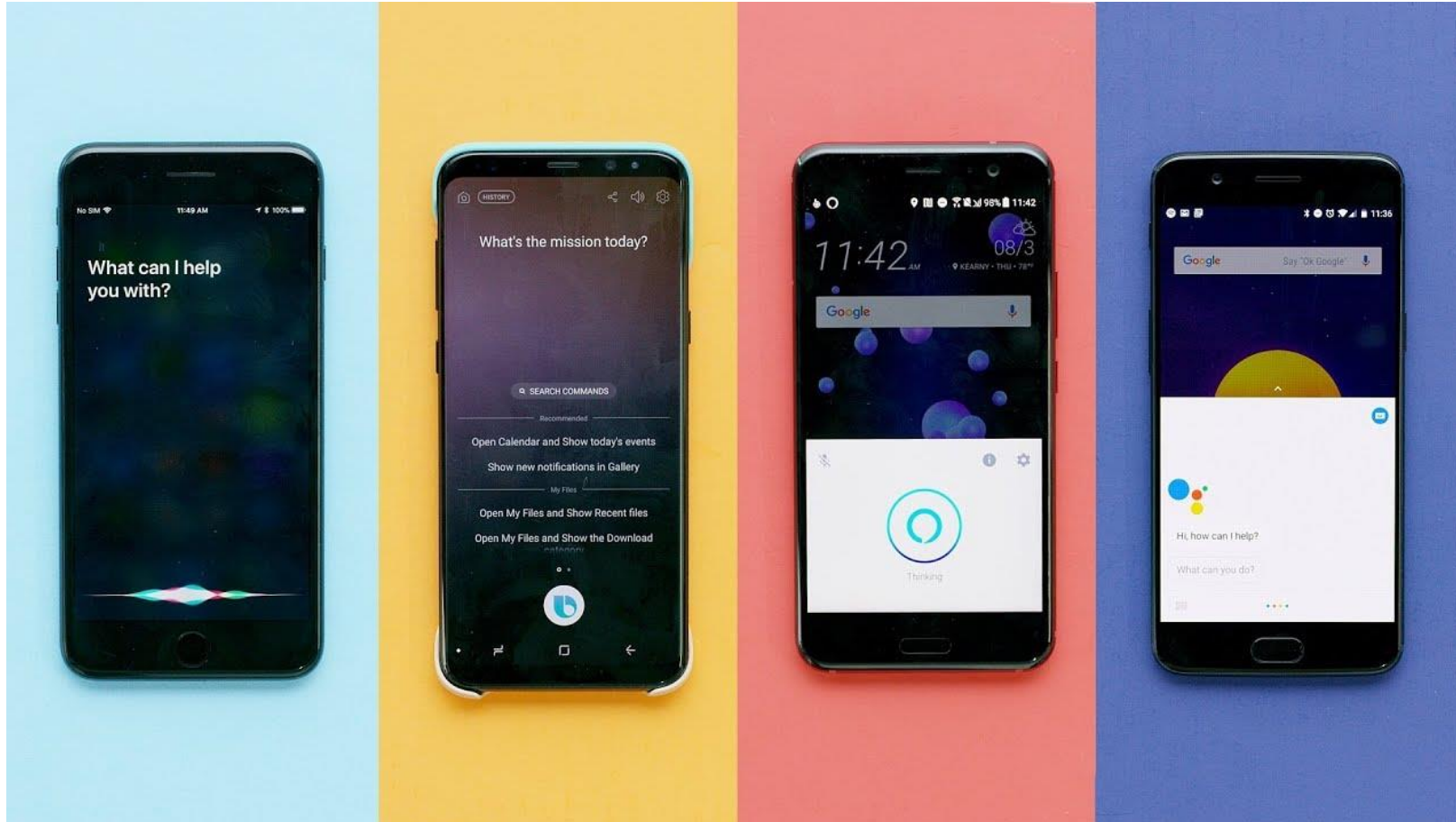
*Rohit Prasad
vice president and head scientist of Amazon Alexa
in conversation with Lex Fridman, 14 December 2019*

<https://youtu.be/Ad89JYS-uZM>

Tell me which are funny, which are not – and which get a giggle first time but are cold pancakes without honey to hear twice.

Robert Heinlein, The Moon Is a Harsh Mistress

Motivation for Computational Humor



Humor...

- ... detection
- ... datasets
- ... generation
- ... evaluation

Humor Detection

Humor classifier [Mihalcea & Strapparava, 2005]

- 16K one-liners/16K non-funny sentences
- Features: alliteration/rhyme, antonymy (WordNet), adult slang, content words
- Classifiers: NB and SVM

| Heuristic | <u>One-liners</u> Reuters | <u>One-liners</u> BNC | <u>One-liners</u> Proverbs |
|--------------|------------------------------|--------------------------|-------------------------------|
| Alliteration | 74.31% | 59.34% | 53.30% |
| Antonymy | 55.65% | 51.40% | 50.51% |
| Adult slang | 52.74% | 52.39% | 50.74% |
| ALL | 76.73% | 60.63% | 53.71% |

| Classifier | <u>One-liners</u> Reuters | <u>One-liners</u> BNC | <u>One-liners</u> Proverbs |
|-------------|------------------------------|--------------------------|-------------------------------|
| Naïve Bayes | 96.67% | 73.22% | 84.81% |
| SVM | 96.09% | 77.51% | 84.48% |

Humor anchors [Yang et al., 2015]

- Humor features:
 - incongruity,
 - ambiguity,
 - interpersonal effect (sentiment/subjectivity),
 - phonetic style.
- ‘Humor anchors’ – structures enabling humorous effect

| | Pun of the Day | | | | 16000 One Liners | | | |
|----------------|----------------|--------------|--------------|--------------|------------------|--------------|--------------|--------------|
| | Accuracy | Precision | Recall | F1 | Accuracy | Precision | Recall | F1 |
| HCF | 0.705 | 0.696 | 0.736 | 0.715 | 0.701 | 0.685 | 0.746 | 0.714 |
| Bag of Words | 0.632 | 0.623 | 0.686 | 0.650 | 0.673 | 0.708 | 0.662 | 0.684 |
| Language Model | 0.627 | 0.602 | 0.762 | 0.673 | 0.635 | 0.645 | 0.596 | 0.620 |
| Word2Vec | 0.833 | 0.804 | 0.880 | 0.841 | 0.781 | 0.767 | 0.809 | 0.787 |
| SaC Ensemble | 0.763 | 0.838 | 0.655 | 0.735 | 0.662 | 0.628 | 0.796 | 0.701 |
| Word2Vec+HCF | 0.854 | 0.834 | 0.888 | 0.859 | 0.797 | 0.776 | 0.836 | 0.805 |

Transformer Gets the Last Laugh [Weller and Seppi, 2019]

- 14K jokes from reddit:

| Method | Body | Punchline | Full |
|-----------------|--------------|--------------|--------------|
| CNN | 0.651 | 0.684 | 0.688 |
| Transformer | 0.661 | 0.692 | 0.724 |
| Human (General) | 0.493 | 0.592 | 0.663 |

- 16K one-liners:

| Previous Work: | Accuracy | Precision | Recall | F1 |
|----------------|--------------|--------------|--------------|--------------|
| Word2Vec+HCF | 0.797 | 0.776 | 0.836 | 0.705 |
| CNN | 0.867 | 0.880 | 0.859 | 0.869 |
| CNN+F | 0.892 | 0.886 | 0.907 | 0.896 |
| CNN+HN | 0.892 | 0.889 | 0.903 | 0.896 |
| CNN+F+HN | 0.894 | 0.866 | 0.940 | 0.901 |
| Our Methods: | Accuracy | Precision | Recall | F1 |
| Transformer | 0.930 | 0.930 | 0.931 | 0.931 |

Shared Tasks & Datasets

Detection and Interpretation of English Puns

SemEval-2017 Task 7 [Miller et al., 2017]

Two types of puns in the data:

- homophonic
- homographic

Tasks:

- Pun detection: binary classification of contexts
- Pun location: locate the punning word
- Pun interpretation: indicate WordNet senses of the punning words

Ten participating systems

#HashtagWars

SemEval-2017 Task [Potash et al., 2017]

From classification to ranking

Data: tweets related to a Comedy TV show with funniness scores (0, 1, 2); 112 hashtags 'topics', 20-180 tweets per hashtag

Tasks:

- Pair-wise comparison of tweets
- Tweets ranking

Eight participating teams

Humor in Edited News Headlines

SemEval-2020 Task 7 [Hossain, 2020]

Data: 15k presumably funny news headlines

Tasks:

- Estimate the funniness of headlines (0..3)
- Pair-wise comparison of headlines

English Datasets

| Dataset | Description | Reference |
|----------------|---|--------------------------------|
| One-liners | 16K one-liners / 16K headlines/proverbs/BNC | [Mihalcea & Strapparava, 2005] |
| Pun of the Day | 2,400 puns/ 2,400 headlines | [Yang et al., 2015] |
| #HashTagWars | 12K tweets for 112 hashtags, graded scores | [Potash et al., 2017] |
| English Puns | 4K (71% puns) + WN annotations | [Miller et al., 2019] |
| Unfun.me | 2.8K headline pairs (1.2K seeds), funny → serious edits | [West & Horvitz, 2019] |
| Humicroedit | 15K headlines, serious → funny edits | [Hossain et al., 2019] |
| FunLines | 8K headlines, serious → funny edits | [Hossain et al., 2020] |
| Reddit | 13,884 not-funny / 2,025 funny jokes | [Weller & Seppi, 2019] |

Russian Humor Datasets [Blinov et al., 2019]

| <i>Dataset</i> | <i>Jokes</i> | <i>Non-jokes</i> | <i>Total</i> |
|----------------|--------------|------------------|----------------|
| STIERLITZ | 46,608 | 46,608 | 93,216 |
| train | 37,447 | 37,447 | 65,530 |
| validation | 4,682 | 4,682 | 9,364 |
| test | 9,361 | 9,361 | 18,722 |
| PUNS | 213 | 0 | 213 |
| FUN | 156,605 | 156,605 | 313,210 |
| train | 125,708 | 125,708 | 251,416 |
| test | 30,897 | 30,897 | 61,794 |
| GOLD | 899 | 978 | 1,877 |

http://bit.ly/fun_data

Humor Generation

HAHAcronym

ACM - Association for Computing Machinery
→ Association for Confusing Machinery
FBI - Federal Bureau of Investigation
→ Fantastic Bureau of Intimidation
PDA - Personal Digital Assistant
→ Penitential Demoniactal Assistant

- Substitutions:
 - semantic field oppositions;
 - rhyme and rhythm;
 - for adjectives: antonym clustering and other semantic relations in WORDNET.
- Stock & Strapparava. HAHAcronym: A Computational Humor System, 2005.

Adult humor via lexical substitution

- Sms corpus
- Lexical constraints:
 - Similarly sounding/spelled words, the same POS
 - Taboo words (700)
 - End of the text, coherence (based on google book n-grams)

| | Experimental Conditions | | |
|-------|--------------------------------|-------------|-----------------|
| | FORM | FORM+TABOO | FORM+TABOO+CONT |
| CF | 2.29 ± 0.19 | 2.98 ± 0.43 | 3.20 ± 0.40 |
| UA(2) | 0.58 ± 0.09 | 0.78 ± 0.11 | 0.83 ± 0.09 |
| UA(3) | 0.41 ± 0.07 | 0.62 ± 0.13 | 0.69 ± 0.12 |
| UA(4) | 0.18 ± 0.04 | 0.36 ± 0.13 | 0.43 ± 0.13 |
| UA(5) | 0.12 ± 0.02 | 0.22 ± 0.09 | 0.26 ± 0.09 |

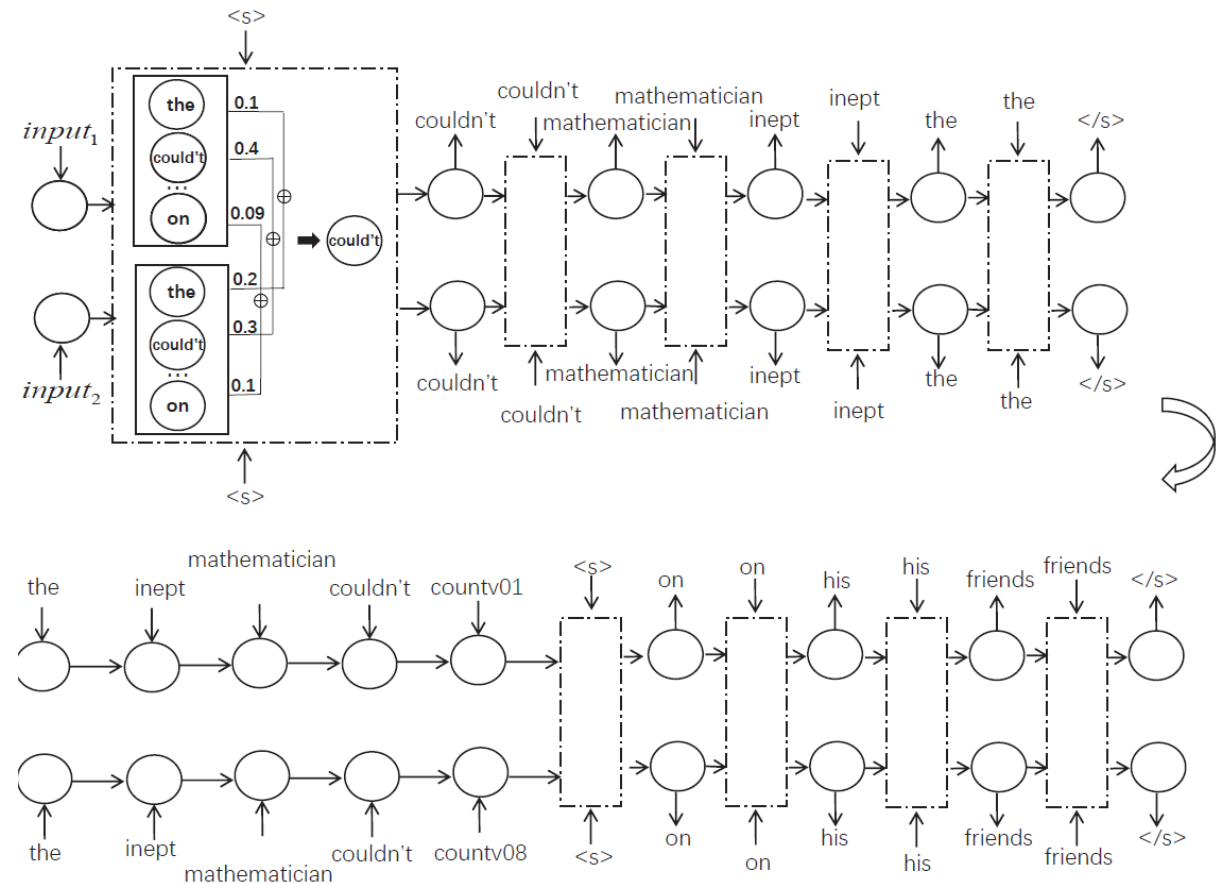
- Evaluation: crowdsourcing (100*3), 1..5 scale
- Valitutti et al. “Let Everything Turn Well in Your Wife”: Generation of Adult Humor Using Lexical Constraints, 2013.

Examples

| Experimental Condition | Text Generated by the System | CF | UA(3) | HE |
|-------------------------------|---|-----------|--------------|-----------|
| FORM | Oh oh...Den muz change plat liao...Go back have yan jiu again... Not 'plat'...'plan'. | 1.68 | 0.26 | 0.43 |
| FORM | Jos ask if u wana melt up? 'meet' not 'melt'! | 2.96 | 0.74 | 2.19 |
| FORM+TABOO | Got caught in the rain.Waited half n hour in the buss stop. Not 'buss'...'bus'! | 2.06 | 0.31 | 0.64 |
| BASE+TABOO | Hey pple... \$ 700 or \$ 900 for 5 nights...Excellent masturbation wif breakfast hamper!!! Sorry I mean 'location' | 3.98 | 0.85 | 3.39 |
| FORM+TABOO+CONT | Nope...Juz off from berk... Sorry I mean 'work' | 2.25 | 0.39 | 0.87 |
| FORM+TABOO+CONT | I've sent you my fart.. I mean 'part' not 'fart'... | 4.09 | 0.90 | 3.66 |

Seq2pun [Yu et al., 2018]

- *Homographic puns*
- LSTM trained on a corpus with *explicitly disambiguated words*



Examples

| Model | Sample |
|---|---|
| <p>pitch: 1) the property of sound that arise with variation in the frequency of vibration; 2) the act of throwing a baseball by a pitcher to a batter.</p> | |
| Highlight | in one that denotes player may have had a high pitch in the world |
| Joint | the object of the game is based on the pitch of the player |
| Normal Language | this is a list of high pitch plot |
| Pun Language | our bikinis are exciting they are simply the tops on the mouth |
| Gold Puns | if you sing while playing baseball you won't get a good pitch |
| <p>square: 1) a plane rectangle with four equal sides and four right angles, a four-sided regular polygon; 2) someone who doesn't understand what is going on.</p> | |
| Highlight | little is known when he goes back to the square of the football club |
| Joint | there is a square of the family |
| Normal Language | the population density was # people per square mile |
| Pun Language | when the pirate captain's ship ran aground he couldn't fathom why |
| Gold Puns | my advanced geometry class is full of squares |
| <p>problem: 1) a source of difficulty; 2) a question raised for consideration or solution.</p> | |
| Highlight | you do not know how to find a way to solve the problem which in the state |
| Joint | he is said to be able to solve the problem as he was a professor |
| Normal Language | in # he was appointed a member of the new york stock exchange |
| Pun Language | those who iron clothes have a lot of pressing veteran |
| Gold Puns | math teachers have lots of problems |

Pun generation with surprise [He & Liang, 2019]

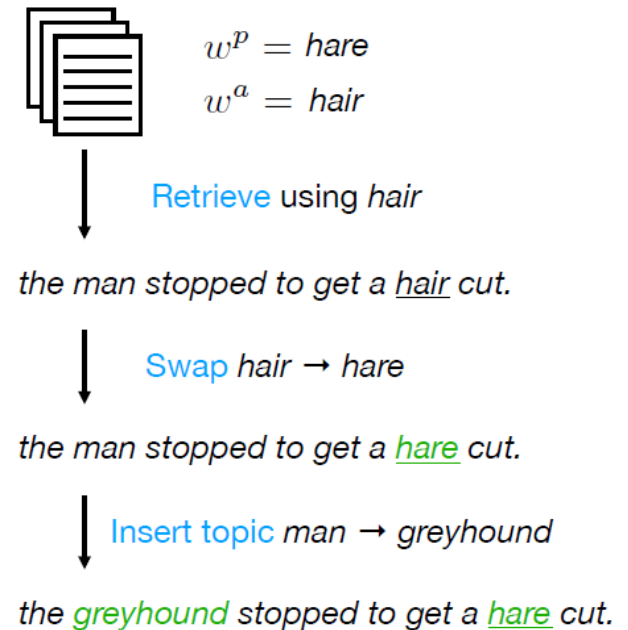
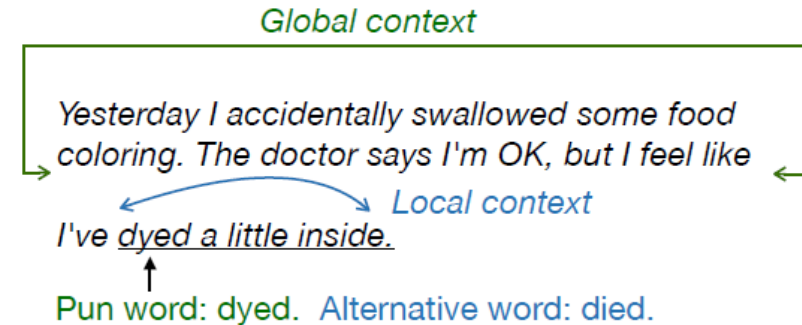
- Puns based on *homophones*
- Local vs global context

$$S(c) \stackrel{\text{def}}{=} -\log \frac{p(w^p | c)}{p(w^a | c)} = -\log \frac{p(w^p, c)}{p(w^a, c)}. \quad (1)$$

$$S_{\text{local}} \stackrel{\text{def}}{=} S(x_{p-d:p-1}, x_{p+1:p+d}), \quad (2)$$

$$S_{\text{global}} \stackrel{\text{def}}{=} S(x_{1:p-1}, x_{p+1:n}), \quad (3)$$

$$S_{\text{ratio}} \stackrel{\text{def}}{=} \begin{cases} -1 & S_{\text{local}} < 0 \text{ or } S_{\text{global}} < 0, \\ S_{\text{local}}/S_{\text{global}} & \text{otherwise.} \end{cases} \quad (4)$$

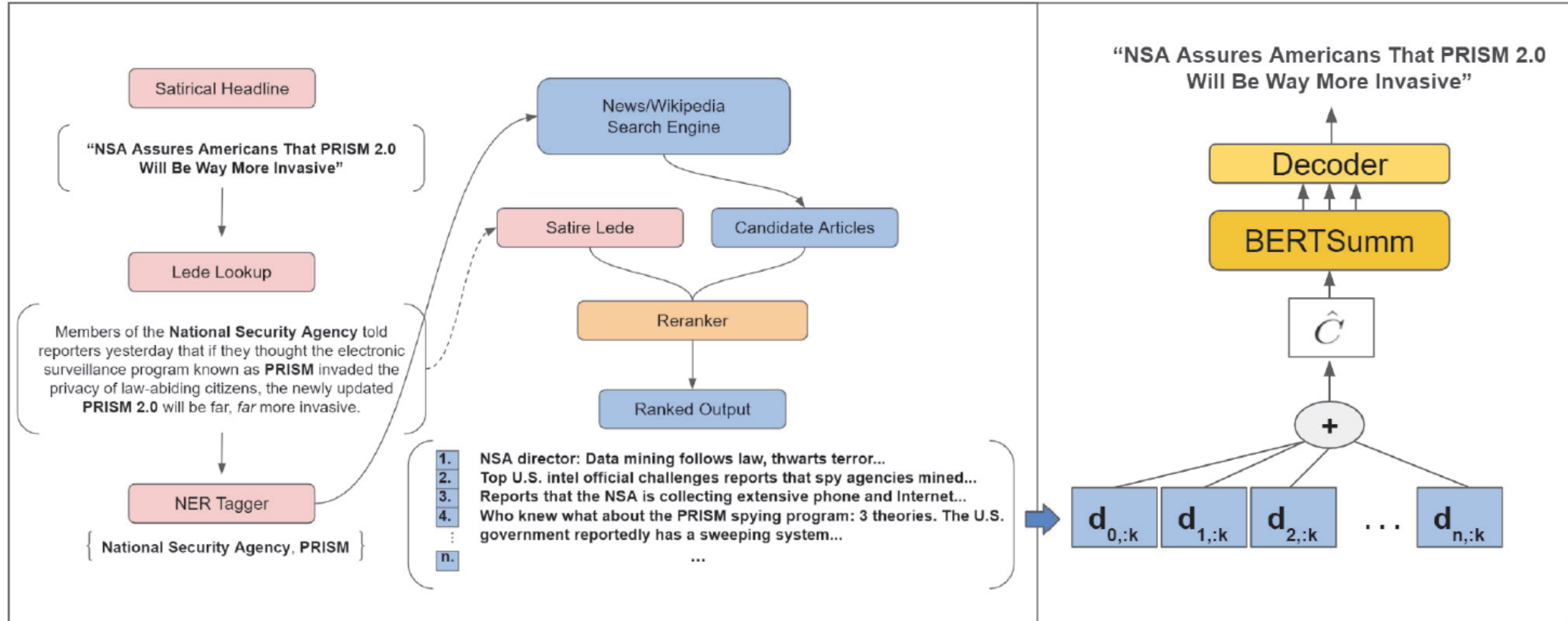


Results

| Method | Example | Rating |
|--|---|--------|
| 1. Pun/alternative word pair: butter – better | | |
| NJD | He is going to come up with the butter a ‘very good’ approach to the world’s economic crisis, the world’s biggest economic climate. | 1 |
| SURGEN | Well, gourmet did it, he thought, it’d butter be right. | 2 |
| Human | Why did the dairy churn? The less said, the butter... | 1.5 |
| 2. Pun/alternative word pair: peace – piece | | |
| NJD | Further, he said, at the end of the peace, it’s not clear that it will be a good example. | 1 |
| SURGEN | That’s because negotiator got my car back to me in one peace. | 1.5 |
| Human | Life is a puzzle; look here for the missing peace. | 3 |
| 3. Pun/alternative word pair: flour – flower | | |
| NJD | Go, and if you are going on the flour. | 1 |
| SURGEN | Butter want to know who these two girls are, the new members of the holy flour. | 1.5 |
| Human | Betty crocker was a flour child. | 4.5 |
| 4. Pun/alternative word pair: wait – weight | | |
| NJD | Gordon Brown, Georgia’s prime minister, said he did not have to wait, but he was not sure whether he had been killed. | 0 |
| SURGEN | Even from the outside, I could tell that he’d already lost some wait. | 2 |
| Human | Patience is a virtue heavy in wait. | 3 |

| Method | Success | Funniness | Grammar |
|---------|--------------|------------|------------|
| NJD | 9.2% | 1.4 | 2.6 |
| R | 4.6% | 1.3 | 3.9 |
| R+S | 27.0% | 1.6 | 3.5 |
| R+S+T+M | 28.8% | 1.7 | 2.9 |
| SURGEN | 31.4% | 1.7 | 3.0 |
| Human | 78.9% | 3.0 | 3.8 |

Funny headline generation



[Horvitz et al., 2020]

Examples

Input: a creator deity or creator god [often called the creator] is a deity or god responsible for the creation of the earth , world , and universe in human religion and mythology . in monotheism , the single god is often also the creator . a number of monolatristic traditions separate a secondary creator from a primary transcendent being , identified as a primary creator..

E-Context: god 's name a big hit / god admits he 's not the creator

D-Context: god 's god calls for greater understanding of all the things

A-Context: god admits he 's not a good person

Onion: Biologists Confirm God Evolved From Chimpanzee Deity

GPT-2 Satire: biologists confirm

GPT-2 News: biologists confirm human ancestor

Input: the jet propulsion laboratory is a federally funded research and development center and nasa field center...on 26 november 2011 , nasa's mars science laboratory mission was successfully launched for mars ... the rover is currently helping to determine whether mars could ever have supported life , and search for evidence of past or present life on mars ...

E-Context: nasa announces plan to put down mars / nasa announces plan to hunt mars

D-Context: nasa launches new mission to find out what life is doing

A-Context: mars scientists successfully successfully successfully successfully

Onion: Coke-Sponsored Rover Finds Evidence Of Dasani On Mars

GPT-2 Satire: coke - a little too much

GPT-2 News: coke - the new 'dancing with the stars'

Evaluation

| Model | Coherence | Onion | Funny | F C |
|--------------|--------------|--------------|--------------|--------------|
| Onion (Gold) | 99.5% | 86.6% | 38.2% | 38.4% |
| Satire GPT-2 | 86.5% | 57.7% | 6.9% | 7.9% |
| News GPT-2 | 89.2% | 36.9% | 2.4% | 2.7% |
| D-Context | 88.4% | 58.8% | 9.4% | 10.4% |
| E-Context | 80.2% | 57.8% | 8.7% | 10.8% |
| A-Context | 85.3% | 54.9% | 8.8 % | 10.3% |

Humorous Response Generation using IR [Blinov et al., 2017]

Data: 103 “funny Twitter accounts”; 300K tweets in total

Three Models (BM25, Query-Term Reweighting, doc2vec)

Evaluation: community question answering (CQA), lab settings.



Responses & Evaluation Scores

| Score | Stimulus | Response |
|-------|---|--|
| 3.00 | Does evolution being a theory make it subjective? | There is no theory of evolution, just a list of creatures Chuck Norris allows to live. |
| 2.67 | Can you find oil by digging holes in your backyard? | Things to do today: 1.Dig a hole 2. Name it love 3. Watch people fall in it. |
| 1.33 | Why don't they put zippers on car doors? | Sick of doors that aren't trap doors. |
| 0.67 | What if you're just allergic to working hard? | You're not allergic to gluten. |
| 0.33 | What test do all mosquitoes pass? | My internal monologue doesn't pass the Bechdel test. :(|

Humor Evaluation

Why Humor Evaluation?

- Massive generation ‘in the wild’ vs. few handcrafted rules.
- Evaluation is crucial for measuring progress.
- Can we get rid of subjectivity of crowdworkers?

Humor Evaluation [Braslavski et al., 2018]

- 30 dialog jokes from different sources

| Source of jokes | Count | Average score in our experiment |
|-------------------------|-------|---------------------------------|
| Jester | 7 | 2.32 |
| Siri | 3 | 1.76 |
| Yahoo!Answers | 5 | 1.73 |
| Automatically generated | 5 | 1.80 |
| Reddit | 5 | 2.37 |
| Twitter | 5 | 1.82 |
| Total | 30 | 2.01 |

Q: Am I the coolest person in the world?

A: Nope. That person lives in Antarctica.

Q: How did the hipster burn his mouth?

A: He ate a cookie BEFORE they were cool!

Evaluation Interface

Page 1/11. Please, evaluate the following jokes:

Question: What's the difference between the government and the Mafia?

Answer: One of them is organized.

- 😞 not funny at all 😓 can be better 😄 funny 😂 hilarious

Question: What is the Australian word for a boomerang that won't come back?

Answer: A stick.

- 😞 not funny at all 😓 can be better 😄 funny 😂 hilarious

Question: What is orange and sounds like a parrot?

Answer: A carrot.

- 😞 not funny at all 😓 can be better 😄 funny 😂 hilarious

Evaluation Results

| | Group | MT | # MT | V | # V | All | # All |
|-----------|-----------|------|------|------|-----|------|-------|
| Age Group | 18–30 | 2.05 | 46 | 2.07 | 77 | 2.06 | 123 |
| | 31–40 | 1.89 | 37 | 2.04 | 54 | 1.98 | 91 |
| | 41–50 | 1.80 | 18 | 2.02 | 29 | 1.94 | 47 |
| | 51–60 | 1.84 | 10 | 1.97 | 6 | 1.89 | 16 |
| | 61+ | 1.73 | 1 | 3.33 | 1 | 2.53 | 2 |
| Sex | Male | 1.92 | 52 | 2.01 | 82 | 1.97 | 134 |
| | Female | 1.95 | 60 | 2.10 | 85 | 2.04 | 145 |
| Language | Average | – | – | 2.16 | 15 | 2.16 | 15 |
| | Good | 2.25 | 5 | 2.10 | 69 | 2.11 | 74 |
| | Bilingual | 2.11 | 3 | 2.06 | 39 | 2.07 | 42 |
| | Native | 1.91 | 104 | 1.95 | 44 | 1.92 | 148 |
| | Global | 1.93 | 112 | 2.06 | 167 | 2.01 | 279 |

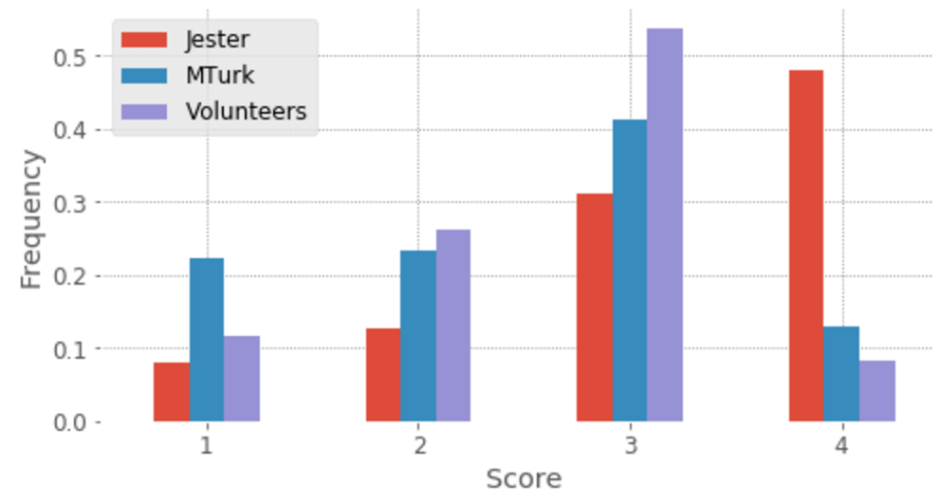
Evaluation Results -2

Highest variation native/no-native:

- **Q:** *Why did 10 die?*
- **A:** *He was in the middle of 9/11*

Highest variation in male/female:

- **Q:** *What is the meaning of life?*
- **A:** *All evidence to date suggests it is chocolate.*



Q: *How many programmers does it take to change a lightbulb?*

A: *NONE! That's a hardware problem*

Evaluation: Summary

- Crowdsourcing is suitable for humor evaluation.
- Age/language proficiency/gender influence joke ratings.
- Jokes degrade over time → re-using evaluation is questionable.

Questions?

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