Content Generation / Writer Assistance tool as an extension of OpenNLP

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Motivations

Most web visionaries think that good quality content comes from:

- really passionate fans, or
- from professional journalists, knowing the topic of their writing well.

However, nowadays, the demand for content requires people not as professional and not as passionate to write large amount of text, for a variety of commercial purposes:

- from search engine optimization
- marketing
- self-promotion.

A wide variety of business areas require content creation in some form, including web content,

Expectation of text quality are rather low and deteriorate further, as content is becoming more and more commercial

So lets have a machine generate low-quality content for us

Objective

To build

- an efficient
- domain-independent
- creative
- interactive

writing tool which produces:

- a large volume of content where
- quality and
- effectiveness

are not essential

Plan

- Goals of the project
- Technology
 - Content generation algorithm
 - NLP: Learning parse trees
 - Parse thickets
- Applications
 - Facebook agent
 - Various cases of content generation
- How to use it
 - How to integrate into your application

Innovation

- Use of web mining to collect pieces of content
- Machine learning of parse trees to filter irrelevant pieces of content from the web Paragraph-level syntactic structures to form content from pieces
- Document-level structure of sections is formed by web mining for attributes of an entity which is a subject of an assay being written

Motivations: why open source

Today search engines are information access gatekeepers
Search engines want their automated agent to analyze / index content created by humans

WE want to automatically build content good enough so that automated agents of search engines do not recognize it as written by humans...

So that search engines would need humans to analyze content created by our machines

Got content from Walmart?

- Today, original content is created by human writers and therefore costly, slowly produced
- Finding a way to automate content creation so that the result is satisfactory for human content consumers and perceived as original by search engine would advance user's web presence and facilitate his professional reputation

Reformulate a text for the purpose of making it original:

- well known to search engines
- produce uninteresting and semantically non-cohesive

Reproduce the human writing activity

A typical creative writing activity of an average author is:

- searching the web for a subject of writing
- browsing the web for relevant information
- finding candidate pieces
- merging them to form a text
- Inserting images/videos/sound recordings
- final text polishing

We intend our writing algorithm to do these exact steps...

And possibly leave the last step for an author

Technology

Difficulty = relevance

Eric P.: We will e-mail you a bill for the amount of your money transfer, and you pay using your bank's online bill pay service.

Boris Galitsky: can YOU do this email?

Eric P.: Please clarify and restate your question.

Boris Galitsky: can you please send me email for bill pay information?

Eric P.: Once you have selected WU Pay as the "Pay With" option and clicked the SUBMIT button, the RECEIPT screen will provide instructions on how to fund the Money Transfer.

Boris Galitsky: ok, I will try that

Boris Galitsky: are you an automated agent?

Eric P.: After choosing WU® Pay, you'll receive a bill by email.

Eric P.: Is there anything else I can help you with today?

Boris Galitsky: can you please answer my previous question?

Eric P.: What question Boris?

Boris Galitsky: are you an automated agent?

Eric P.: No sir

Boris Galitsky: It is my right as a customer to know if I am served by an automated or hur

right?

Eric P.: Yes it is







Content generation algorithm

Input: seed text (expected content to be generated around it)

Split seed text into sentences

For each original sentence

Form a search query from sentence

- -extract noun phrases: either > 3 words length or 2 words length extracted entities
- if not enough noun phrases then the whole sentence
- optionally add mental verbs if the desired content is opinion-based

Run search engine query and collect snippets

For each snippet

Split snippets into sentences and insert markers for incomplete sentences to be substituted by text from original web page If at least one incomplete sentence is detected, download original page strip HTML / extract text split text into sentences to be matched against

For each candidate sentence of current

If it is not complete, find its original form on the page, together with one or more consecutive sentences, and substitute Pre-filter candidate sentences with respect to length, proper punctuation, similarity with snippet

Relevance verification:

- perform syntactic match between current snippet sentence and original sentence
- if no good syntactic match then try current snippet sentence and other original sentence
- run string similarity match to confirm syntactic match-based decisions

Measure the "opinionatedness" of candidate sentence: in what degree the candidate sentence expresses an opinion (includes mental verbs / communicative actions from a fixed vocabulary)

Calculate relevance/opinionatedness score and preliminarily accept/reject candidate sentence

Reject sentence if it belongs to a specific classes of advertisement/sales pitch, do not contain verb phrase, and other style criteria

Re-style and reformat sentence to be included in the resultant text

Combine seed sentence with the set of mined sentences, and do that for all seed sentences

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Content generation algorithm: building section structure

We search for a essay topic and obtain main entities associated with it, such as {space, atomic bomb. time machine} albert einstein invented **space** albert einstein invented **the atomic bomb** albert einstein invented **time machine**

Content generation algorithm: building section structure and filling them

We now search for a essav topic + section topic:

albert einstein invented space

Albert Einstein invented SPACE!! by LemonGrass21 on devi...

lemongrass21.deviantart.com/.../Albert-Einstein... ▼ Перевести эту страницу 28 abr. 2011 r. - HOLY SHIT!!! Its Sunday! Pfft, like i care. Seriously, Sunday is my least favorite day of the week and everybody seems to like to do outside and

HowStuffWorks "What did Albert Einstein in Albert Einstein and the Atomic Bombings of Japan. ... biographer Ronald Clark has science.howstuffworks.com/...inventors/what-di... . Tep-Albert Einstein, a man whose name is practically synony

Albert Einstein and the Atomic Bomb - Was It Necessary? www.doug-long.com/einstein.htm * Перевести эту страницу

observed that the atomic bomb would have been invented without Einstein's ...

but his theories of relativity led to new ways of looking at t Why did albert einstein help invent the atomic bomb - Wiki A...

wiki.answers.com > ... > World War 2 ▼ Перевести эту страницу The atomic bomb was not invented in the US or by Einstein, it was invented in

London, England in 1933 by one of Einstein's friends - Leo Szilard while crossing ...

History of the **Atomic Bomb** and The Manhattan Project - Inventors inventors.about.com/.../astartinventions/a/atomi... • Перевести эту страницу Albert Einstein and other scientists told Roosevelt of Nazi Germany efforts to ... uranium and putting together a working atomic bomb were created and seen to ... Atomic Bomb Detonation at ... - Robert Oppenheimer

Building section structure and filling them

Now we construct the section structure:

Albert Einstein

albert einstein invented space

nvented Space

Albert Einstein invented SPACE!! by LemonGrass21 on devi...

lemongrass21.deviantart.com/.../Albert-Einstein... ▼ Перевести эту страницу 28 авг. 2011 г. - HOLY SHIT!!! Its Sunday! Pfft, like i care. Seriously, Sunday is my least favorite day of the week and everybody seems to like to go outside and ...

HowStuffWorks "What did Albert Einstein invent?"

science.howstuffworks.com/...inventors/what-di... $\ ^{\circ}$ Перевести эту страницу Albert Einstein, a man whose name is practically synonymous with genius, is one but his theories of relativity led to new ways of looking at time, space, matter, ...

Albert Einstein and the Atomic Bomb - Was It Necessary?

www.doug-long.com/einstein.htm * Перевести эту страницу Albert Einstein and the Atomic Bombings of Japan.... biographer Ronald Clark has observed that the atomic bomb would have been invented without Einstein's ...

Why did albert einstein help invent the atomic bomb - Wiki A..

wiki.answers.coms ... -> World War 2 * Repeacor any страницу
The atomic bomb was not invented in the US or by Einstein, it was invented in
London England in 1933 by one of Finstein's friends - Leo Syllard while crossing

ein invented Bomb

History of the Atomic Bomb and The Manhattan Project - Inventors

inventors.about.com/.../astartinventions/a/atomi... ▼ Перевести эту страницу
Albert Einstein and other scientists told Roosevelt of Nazi Germany efforts to ...
uranium and putting together a working atomic bomb were created and seen to ...
Atomic Romb Detonation at - Robert Oppenheimer

Content generation flow 1

For sentence "Give me a break, there is no reason why you can't retire in ten years if you had been a rational investor and not a crazy trader"

- We form the query for search engine API: +rational +investor +crazy +trader
- From search results we remove duplicates, including "Derivatives: Implications for Investors | The Rational Walk".
- From the search results we show syntactic generalization results for the seed and search result:
- Syntactic similarity: np [[IN-in DT-a JJ-*], [DT-a JJ-* JJ-crazy], [JJ-rational NN-*], [DT-a JJ-crazy]], score= 0.9
- Rejected candidate sentence: Rational opportunities in a crazy silly world.

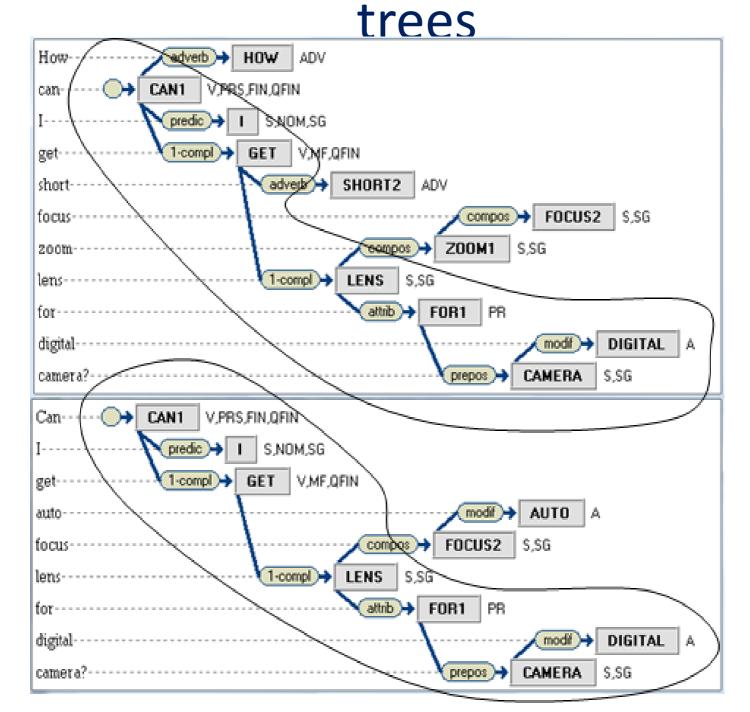
Content generation flow 2

- Syntactic generalization 2: np [[VBN-* DT-a JJ-* JJ-rational NN-investor], [DT-a JJ-* JJ-rational NN-investor]] vp [[DT-a], [VBN-* DT-a JJ-* JJ-rational NN-investor]], score= 2.0
- Accepted sentence: I have little pretensions about

The latter sentence has significantly stronger semantic commonality with the seed one, compared to the former one, so it is expected to serve as a relevant part of generated content about "rational investor" from the seed sentence

NLP: machine learning of parse trees

Similarity = common part among parse



Why Parse Thickets

- To represent a linguistic structure of a paragraph of text based on parse trees for each sentence of this paragraph.
- We will refer to the sequence of parse trees extended by a number of arcs for inter-sentence relations between nodes for words as Parse Thicket (PT).
- A PT is a graph which includes parse trees for each sentence, as well as additional arcs for intersentence relationship between parse tree nodes for words.

What are we going to do with Parse

- Thickets
 Extend the operation of least general generalization (unification of logic formula) towards structural representations of paragraph of texts
- Define the operation of generalization of text paragraphs to assess similarity between portions of text.
- Use of generalization for similarity assessment is inspired by structured approaches to machine learning versus unstructured, statistical where similarity is measured by a distance in feature space (Moschitti et al)

Now we compare linguistic phrase search and regular SOLR search

SOLR is one of the most popular search frameworks used in industry lucene.apache.org/solr/

Linguistic phrase is not a string phrase query "...". There are noun phrases, verb phrases, prepositional phrases, and others

SOLR search takes into account term frequency, inverse document frequency, and the number of words between keywords in the document

This XML file does not appear to have any style information associated with it. The document tree is shown below.

```
▼<response>
 ▼<lst name="responseHeader">
    <int name="status">0</int>
    <int name="QTime">2</int>
   ▼<lst name="params">
      <str name="g">name:ladv name:gaga name:sings</str>
      <str name="fl">name</str>
    </lst>
  </lst>
 ▼<result name="response" numFound="13" start="0">
   ▼<doc>
    ▼<str name="name">
       lady gaga does belly donce but katy perri sings romance
      </str>
    </doc>
   ▼<doc>
    ▼<str name="name">
       lady gaga does belly dance but katy perri sings romance
      </str>
    </doc>
   ▼<doc>
      <str name="name">lady perri sings and katy gaga does dance</str>
    </doc>
   ▼<doc>
      <str name="name">lady gaga and katy perri sings punk rock and dance</str>
    </doc>
   ▼<doc>
      <str name="name">rock lady sings and gaga perri does dance</str>
    </doc>
   ▼<doc>
    ▼<str name="name">
       lady gaga sings with passion and katy perri does an exotic dance
     </str>
    </doc>
   ▼<doc>
    ▼<str name="name">
       lady gaga sings rock but katy perri performs pop and dance
     </str>
    </doc>
   _____
```

Keyword search

This XML file does not appear to have any style information associated with it. The document tree is shown below.

```
▼<response>
 ▼<lst name="responseHeader">
    <int name="status">0</int>
    <int name="OTime">36</int>
   ▼<lst name="params">
      <str name="qt">dismax</str>
      <str name="q">name:lady name:gaga name:sings</str>
      <str name="fl">name</str>
    </lst>
  </lst>
 ▼<result name="response" numFound="10" start="0">
   ▼<doc>
    ▼<str name="name">
       lady gaga sings pop and katy perri performs punk rock and dance
      </str>
    </doc>
   ▼<doc>
    ▼<str name="name">
       lady gaga sings rock but katy perri performs pop and dance
      </str>
    </doc>
   ▼<doc>
    ▼<str name="name">
       lady gaga sings with passion and katy perri does an exotic dance
      </str>
    </doc>
   ▼<doc>
      <str name="name">lady gaga and katy perri sings punk rock and dance</str>
    </doc>
   ▼<doc>
      <str name="name">rock lady sings and gaga perri does dance</str>
    </doc>
   ▼<doc>
      <str name="name">lady perri sings and katy gaga does dance</str>
    </doc>
   ▼<doc>
    ▼<str name="name">
       lady gaga does dance, sings pop but katy perri performs funk rock
      </str>
```

Phrase (natural language) search

Finding similarity between two paragraphs

Original text

- "Iran refuses to accept the UN proposal to end the dispute over work on nuclear weapons",
- "UN nuclear watchdog passes a resolution condemning Iran for developing a second uranium enrichment site in secret",
- "A recent IAEA report presented diagrams that suggested Iran was secretly working on nuclear weapons",
- "Iran envoy says its nuclear development is for peaceful purpose, and the material evidence against it has been fabricated by the US"

Text, found on the web: is it RELEVANT?

"UN passes a resolution condemning the work of Iran on nuclear weapons, in spite of Iran claims that its nuclear research is for peaceful purpose",

"Envoy of Iran to IAEA proceeds with the dispute over its nuclear program and develops an enrichment site in secret",

"Iran confirms that the evidence of its nuclear weapons program is fabricated by the US and proceeds with the second uranium enrichment site"

Keywords: topic with no details

Iran, UN, proposal, dispute, nuclear, weapons, passes, resolution, developing, enrichment, site, secret, condemning, second, uranium

Improvement: pair-wise generalization

```
[NN-work IN-* IN-on JJ-nuclear NNS-weapons], [DT-the
NN-dispute IN-over JJ-nuclear NNS-*], [VBZ-passes DT-a
NN-resolution ],
[VBG-condemning NNP-iran IN-*],
[VBG-developing DT-* NN-enrichment NN-site IN-in NN-
secret ]],
[DT-* JJ-second NN-uranium NN-enrichment NN-site ]],
[VBZ-is IN-for JJ-peaceful NN-purpose],
[DT-the NN-evidence IN-* PRP-it], [VBN-* VBN-fabricated
IN-by DT-the NNP-us ]
```

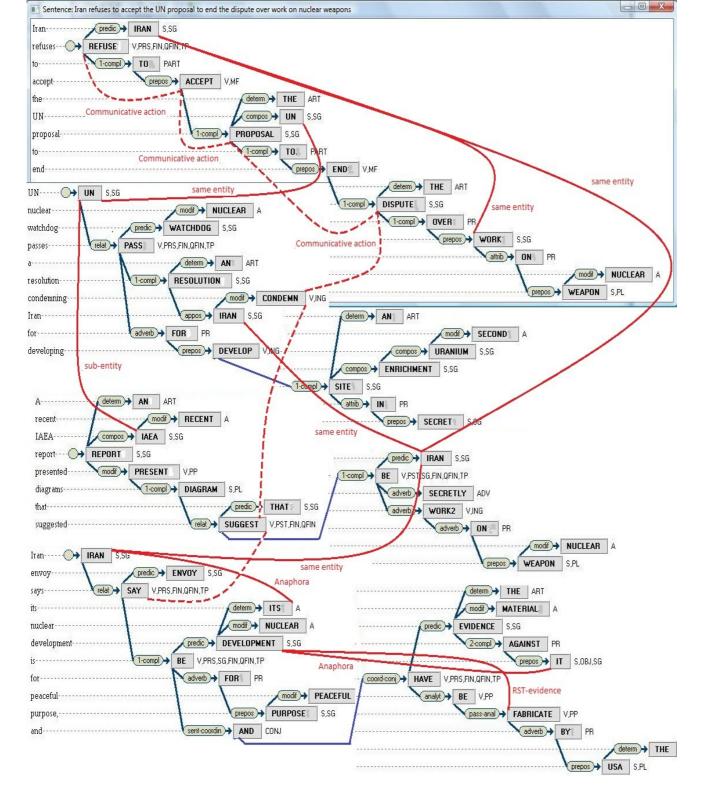
Pair-wise vs. Parse thicket

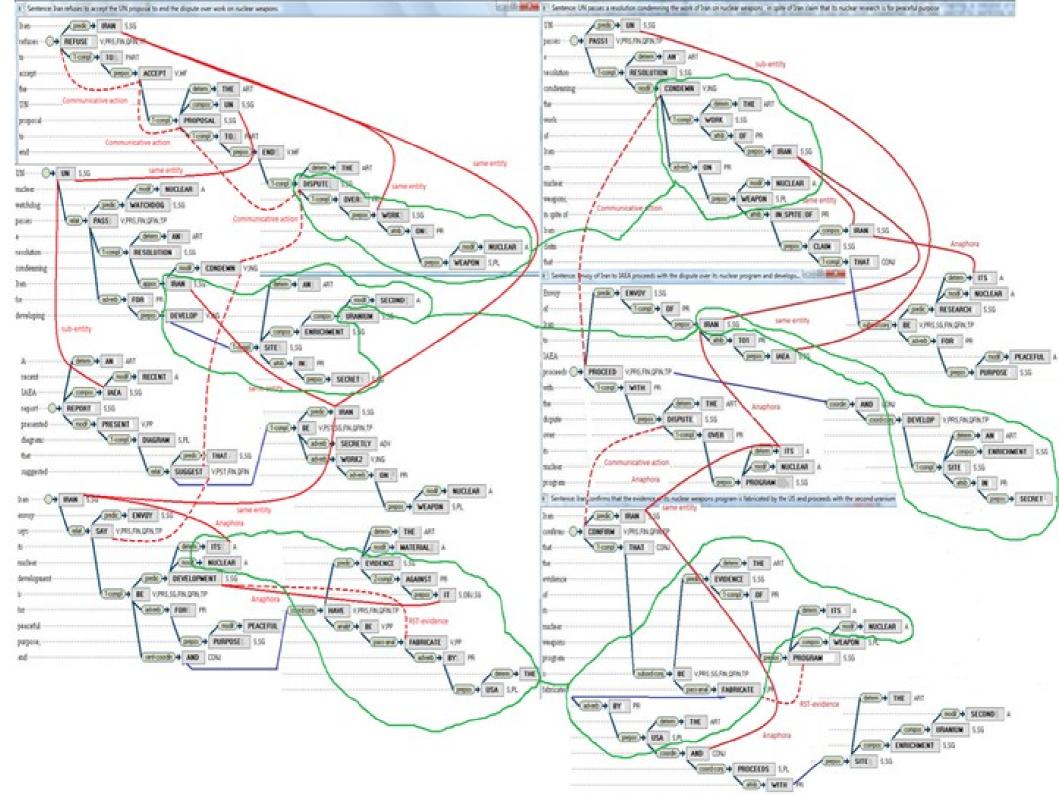
- [NN-Iran VBG-developing DT-* NN-enrichment NN-site IN-in NN-secret] [NN-generalization-<<u>UN/nuclear watchdog></u> * VB-pass NN-resolution VBG condemning NN- Iran]
- [NN-generalization-<Iran/envoy of Iran> **Communicative_action** DT-the NN-dispute IN-over JJ-nuclear NNS-*
- [Communicative_action NN-work IN-of NN-Iran IN-on JJ-nuclear NNS-weapons]
- [NN-generalization Iran/envoy to UN> Communicative_action NN-Iran NN-nuclear NN-* VBZ-is IN-for JJ-peaceful NN-purpose],
- Communicative_action NN-generalize <work/develop> IN-of NN-Iran IN-on JJ-nuclear NNS-weapons]*
- [NN-generalization In-generalization In-generalization In-against NN Iran NN-nuclear VBN-fabricated IN-by DT-the NNP-us] condemn^proceed [enrichment site] <a href="mailto:slight-right-

Parse Thicket

Syntactic parse trees

- Links between words from different sentences
 - Anaphora
 - Rhetoric Structure Theory (RST) [Mann]
 - Speech Act Theory (communicative actions, CA) [Searle]





Applications

Case Study: Agent acting on behalf of a user posting messages to maintain friendship

Domain of social promotion



- On average, people have 200-300 friends or contacts on social network systems such Facebook and LinkedIn.
- To maintain active relationships with this high number of friends, a few hours per week is required
- In reality, people only maintain relationship with 10-20 most close friends, family and colleagues,
- The rest of friends are being communicated with very rarely. These not so close friends feel that the social network relationship has been abandoned



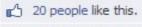
ElliottGetaway Sunset over Lake Tahoe

CASP is about to post a

message



Like · Comment · Share · 7 hours ago near South Lake Tahoe, CA · 18



Therd Kim Priceless picture ... God's creation at His best.
7 hours ago · Like



The hike was, round trip, about 8.5 miles and took us just over 6 ... The view from Relay Peak looking south across Lake Tahoe. ... I met <u>Sevski</u>, who had just put in a day of <u>professoring</u>, for a



The hike was, round trip, about 8.5 miles and took us just over 6 ... The view from Relay Peak looking south across Lake Tahoe. ... I met Seyski, who had just put in a day of professoring, for a couple quick runs before sunset.

Case study: what happened

- Friends posted a photo of them dancing tango
- CASP commented on the posting, finding information about "tango at a wedding"
- Friends got upset and communicate that the posting of CASP citing the wedding was irrelevant and they did not like it.
-Nevertheless they married 5 months later

Case study



Conversational characters

Simulated human characters and agents are increasingly common components of user interfaces of applications.

They include:

- interactive tutoring environments,
- eldercare systems,
- virtual reality-based systems,
- intelligent assistant systems,
- physician-patient communication training,
- entertainment applications

(Cassell et al., 2000, De Rosis et al., 2003; Dias and Paiva 2005; Lisetti 2008).

Limitations of current conversational characters

- Improvement in the intelligent features, expressiveness, understanding human language, dialog abilities and other aspects.
- Social realism is still far behind
- Don't take into consideration fact that users consistently respond to computers as if they were social actors (Reeves and Nass 1996)
- Poor understanding of user needs in the areas such as shopping, finance, conflict resolution, etc.

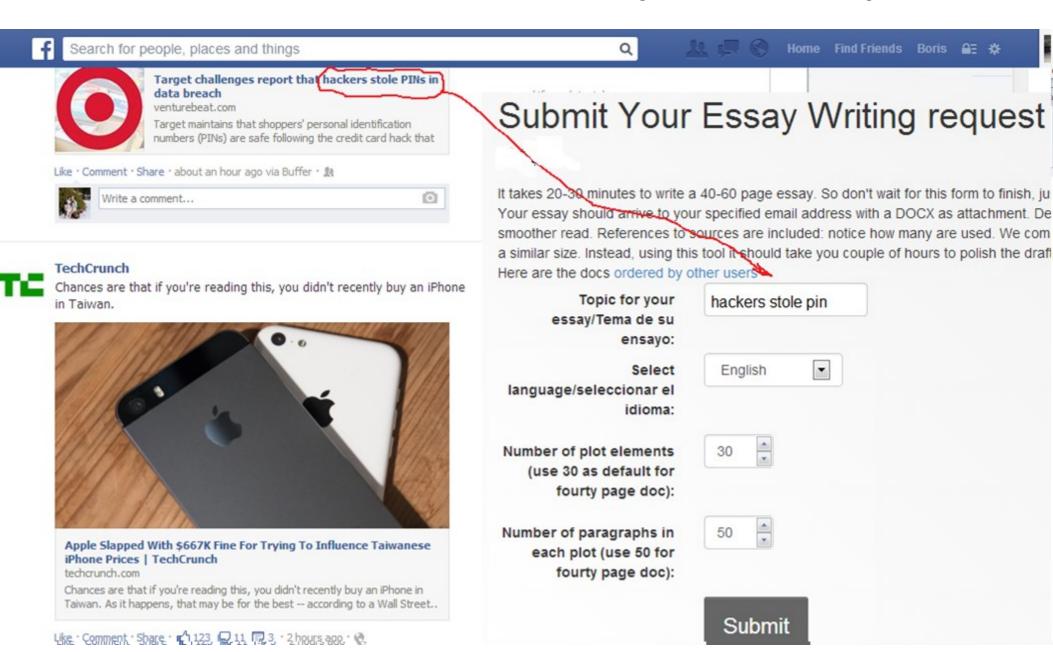
Tolerance to dealing with auto agents

- Users / Readers would tolerate:
 - worse than ideal style,
 - discourse and
 - quality of content being communicated,

as long as overall the communication is positive and makes sense.

- Messages should at least show social engagement
- Currently available commercial chat bots employed by customer support portals, or packaged as mobile apps, possess too limited
 - NLP
 - text understanding
- overall robust intelligent capabilities to support conversations to support social promotion.

CASP writes an essay on a topic



Content Generation part of CASP

- The content generation part of CASP can be used at www.facebook.com/RoughDraftEssay
- Given a topic, it first mines the web to auto build a taxonomy of entities which will be used in the future comment or essay
- Then the system searches the web for these entities to find pieces of content to create respective sections
- The resultant document is delivered as DOCX email attachment
- CASP can automatically compile texts from hundreds of sources to write an essay on the topic
- If a user wants to share a comprehensive review, opinion on something, provide a thorough background, then this interactive mode should be used.

Evaluation results: relevance

		I	T		
Query	Query complexity	Relevance of baseline Bing search, %, averaging over 100 searches	Relevance single-sentence phrase-based generalization search, %, averaging over 100 searches	Relevance of thicket-based phrase generalization search, %, averaging over 100 searches	Relevance of parse thicket-based graph generalization search, %, averaging over 100 searches
Facebook friend agent support search	1 compound sent	74.5	83.2	85.3	87.2
	2 sent	72.3	80.9	82.2	83.9
	3 sent	69.7	77	81.5	81.9
	4 sent	70.9	78.3	82.3	82.7
		l	<u> </u>		

Evaluation of CASP

We measure the failure score of an individual CASP posting as the number of failures in relevance and appropriateness. The postings started with [posted by an agent on behalf of Boris]

After a certain number of failures, friends

- complain
- unfriend
- share negative information about the loss of trust with others
- and even encourage other friends to unfriend an friend who is enabled with CASP.

Trust vs Domain & Complexity

Values are the average numbers of failed CASP postings, for given domain and given complexity, when issues with trust arises
On average, friends complain after 6 failures and want to unfriend CASP's host after 10 failures

Topic of the seed				is	ith
	Complexity of the seed and posted message	A friend complaintsto the CASP's host	A friend unfriends the CASP host	A friend shares with other friends that the trust in CASP is	A friend encourages other friends to unfriend a friend with CASP
Travel & outdoor	l sent	6.3	8.5	9.4	12.8
	2 sent	6.0	8.9	9.9	11.4
	3 sent	5.9	7.4	10.0	10.8
	4 sent	5.2	6.8	9.4	10.8
Shopping	l sent	7.2	8.4	9.9	13.1
	2 sent	6.8	8.7	9.4	12.4
	3 sent	6.0	8.4	10.2	11.6
	4 sent	5.5	7.8	9.1	11.9
Events & entertain ment	l sent	7.3	9.5	10.3	13.8
	2 sent	8.1	10.2	10.0	13.9
	3 sent	8.4	9.8	10.8	13.7
	4 sent	8.7	10.0	11.0	13.8
Job- related	lsent	3.6	4.2	6.1	6.0
	2 sent	3.5	3.9	5.8	6.2
	3 sent	3.7	4.0	6.0	6.4
	4 sent	3.2	3.9	5.8	6.2
Personal life	l sent	7.1	7.9	8.4	9.0
	2 sent	6.9	7.4	9.0	9.5
	3 sent	5.3	7.6	9.4	9.3
	4 sent	5.9	6.7	7.5	8.9
Average		6.03	7.5	8.87	10.575

Trust vs Relevance

- For less information-critical domains like travel and shopping, tolerance to failed relevance is relatively high.
- Conversely, in the domains taken more seriously, like job related, and with personal flavor, like personal life, users are more sensitive to CASP failures and the loss of trust in its various forms occur faster
- For all domains, tolerance slowly decreases when the complexity of posting increases.

Conclusions for Facebook application

- We proposed a problem domain of social promotion and build a conversational agent CASP to act in this domain
- CASP maintains friendship and professional relationships by automatically posting messages on behalf of its human host
- It was demonstrated that a substantial intelligence in information retrieval, reasoning, and natural language-based relevance assessment is required to retain trust in CASP agents by human users

Content generation agent can be trusted

- We evaluated the scenarios of how trust was lost in Facebook environment.
- We confirmed that it happens rarely enough for CASP agent to improve the social visibility and maintain more friends for a human host than being without CASP.

Although some friends lost trust in CASP, the friendship with most on Overall CASP's impact on

Overall CASP's impact on social activity is positive.

Published work

- Galitsky, B., Josep Lluis de la Rosa & Gábor Dobrocsi. Building Integrated Opinion Delivery Environment FLAIRS-24 2011.
- Galitsky, B. and Josep Lluis de la Rosa. Concept-based learning of human behavior for customer relationship management. Special Issue on Information Engineering Applications Based on Lattices. Information Sciences. Volume 181, Issue 10, 15 May 2011, pp 2016-2035, 2011.
- Galitsky, B., Josep-Lluis de la Rosa, and Boris Kovalerchuk. Assessing plausibility of explanation and metaexplanation in inter-human conflict. Engineering Application of AI, V 24 Issue 8, pp 1472-1486, 2011.
- Galitsky, B. Josep Lluis de la Rosa, Gábor Dobrocsi. Inferring the semantic properties of sentences by mining syntactic parse trees. Data & Knowledge Engineering. Available online 28 July 2012.
- Galitsky, B. Machine Learning of Syntactic Parse Trees for Search and Classification of Text. Engineering Application of Artificial Intelligence, 2012 (to appear).
- Galitsky, B. Exhaustive simulation of consecutive mental states of human agents. Knowledge-Based Systems, 2013

OpenNLP contribution

The Apache OpenNLP library is a machine learning based toolkit for the processing of natural language text.

It supports the most common NLP tasks, such as tokenization, sentence segmentation, part-of-speech tagging, named entity extraction, chunking, parsing, and coreference resolution.

Syntactic generalization component:

https://svn.apache.org/repos/asf/incubator/opennlp/sandbox/opennlp-similarity

Also available at Google code

https://code.google.com/p/relevance-based-on-parse-trees/

This project among open source projects

JGraphT

OpenNLP

Graphlearning Parse Thickets

Stanford NLP

Linguists can now enjoy graphlearning of linguistic structures, and graph community – new rich source of graphs

OpenNLP syntactic generalization contribution

Search engineer without knowledge of linguistic / parsing can easily integrate syntactic generalization into an application when relevance is important:

```
SentencePairMatchResult matchRes =
sm.assessRelevance(snapshot, searchQuery);
       List<List<ParseTreeChunk>> match =
matchRes.getMatchResult();
score = parseTreeChunkListScorer.
getParseTreeChunkListScore(match);
       if (score > 1.5) {
         // relevant
```

OpenNLP: how similarity expressions for distinct cases have too high score for bag-of-words

```
String phrase1 = "How to deduct rental expense from income ";
String phrase2 = "How to deduct repair expense from rental income.";
List<List<ParseTreeChunk>> matchResult =
parser.assessRelevance(phrasel,
phrase2).getMatchResult();
      assertEquals (matchResult.toString(),
            "[[ [NN-expense IN-from NN-income ], [JJ-
rental NN-* ], [NN-income ]], [ [TO-to VB-deduct JJ-
rental NN-* ], [VB-deduct NN-expense IN-from NN-income
] ] ] " ) ;
double matchScore =
parseTreeChunkListScorer.getParseTreeChunkListScore(matc
hResult);
double bagOfWordsScore =
parserBOW.assessRelevanceAndGetScore(phrase1, phrase2);
      assertTrue(matchScore+2 < bagOfWordsScore);</pre>
      System.out.println("MatchScore is adequate ( =
"+matchScore + ") and bagOfWordsScore =
"+bagOfWordsScore+" is too high");
```

Value of this research

- We believe this is one of the first systems in learning a semantic discourse to solve a content generation relevance problem, based on a set of parse trees.
- Instead of using linguistic information of individual sentences, we can now compute text similarity at the level of paragraphs.

Project @ Google Code

The relevance assessment code is developed as a part of OpenNLP project and is available

https://code.google.com/p/relevance-based-onparse-trees