

Natural Language Watermarking

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Problem and Key Idea

Performing **natural language watermarking**, which uses the structure of the sentence constituents in natural language text to insert a watermark.

Outline

- **Why Natural Language (NL) watermarking?**
- **NL Watermarking vs. Image Watermarking**
- **Natural Language Processing (NLP) for Watermarking**
- **Linguistic Steganography**
- **NL Watermarking**
- **Conclusions**

Why Natural Language Watermarking?

- **Authenticating the source of a document**
- **Proving or denying ownership on a document**
- **Controlling distribution and reuse of intellectual property**
- **Digital libraries, on-line news channels, online stores etc.**
- **Content protection, text auditing, meta-data binding, tamper-proofing, traitor tracing**

Natural Language vs. Image Watermarking

- **Same goal, different methods**
- **The amount of redundancy is very low in text**
- **Evaluation of stealthiness is harder for text**
- **Limitations of human visual system is high**
 - LSB vs. Synonym Substitution

Natural Language vs. Image Watermarking

LSB Embedding

original



watermarked



difference image




Synonym Substitution

“The tyrant will always find a pretext for his tyranny.”

“The despot will always find a pretext for his despotism”.

Natural Language vs. Image Watermarking

- **Sentences have combinatorial syntax and semantics:**
 - complex representations are constructed using structurally simple constituents
 - the semantic content of a sentence is a function of the semantic content of its constituents together with its syntactic structure
- **Transformational Grammar Theory**

“Ned loves Jody”  “Jody is loved by Ned”
- **Not the surface but the underlying structure is altered**
 - a rough analogy to modifying DCT coefficients
- **Achieving higher robustness**

NLP for Watermarking

- **Natural Language Processing (NLP) aims to design algorithms that will analyze, understand, and generate natural language automatically**
- **Electronic Data Resources and Tools**
 - **Corpora**
 - **Dictionaries e.g., WordNet, Verbnet**
 - **Parsers, Generators, Machine Translation and Question Answering Systems**

NLP for Watermarking: Linguistic Transformations

- Synonym Substitution
- Syntactic Transformations
- Semantic Transformations

NLP for Watermarking:

Syntactic Transformations

Transformation	Original sentence	Transformed sentence
Passivization	The slobbering dog kissed the big boy.	The big boy was kissed by the slobbering dog.
Topicalization	I like bagels.	Bagels, I like.
Clefting	He bought a brand new car.	It was a brand new car that he bought.
Preposing	I like big bowls of beans.	Big bowls of beans are what I like.
There-construction	A unicorn is in the garden.	There is a unicorn in the garden.
Fronting	“What!” Alice cried.	“What!” cried Alice.

NLP for Watermarking:

Syntactic Transformations

- **Verb Alternations:**
 - Levin Verb Classes
 - [Spray/Load Alternation]

“Jack sprayed paint on the wall. ⇔

Jack sprayed the wall with paint.”

“The farmer loaded apples into the cart . ⇔

The farmer loaded the cart with apples .”

NLP for Watermarking:

Semantic Transformations

- Based on **co-references**

- **Pruning** : removing repeated information

Yet Iceland has offered a residency visa to ~~ex-chess champion~~ Bobby Fischer in recognition of a 30-year-old match that put the country on the map.

- **Grafting**: adding or repeating information

He, **an American citizen**, is being detained in Japan and is wanted in the US for violating international sanctions against the former Yugoslavia by playing there in 1992.

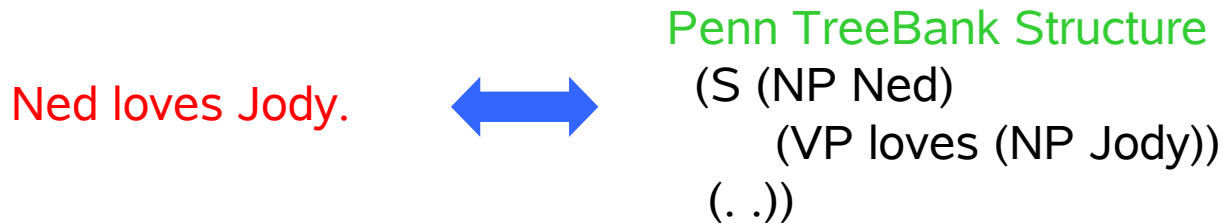
- **Substitution**: replacing information

Ex-chess champion's historic win over Russian Boris Spassky in Reykjavik in 1972 shone the international spotlight on Iceland as never before.

NLP for Watermarking

- **NL Parsing**

- **processing sentences to determine their structure**



- **NL Generation**

- **constructing NL output from non-linguistic information representations according to some communication specifications**

Sentence realization

DSYNTS:

love [class:"verb"]

(I Ned [class:"proper_noun"]

 II Jody [class:"proper_noun"]

)

END:

NLP for Watermarking

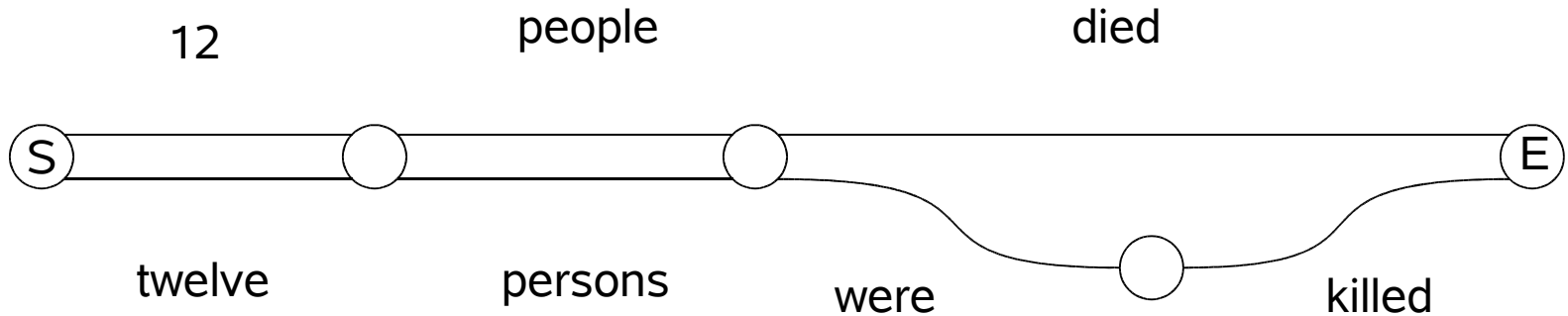
- **Paraphrasing**

- **Parallel corpus**

After the latest Fed rate cut, stocks rose across the board.

Winners strongly outpaced losers after Greenspan cut interest rates again.

- **Finite-State Automata Reduction**



Previous Work in Linguistic Steganography

- **Mimicry Text:** Using PCFGs to Generate Cover Text

(Wayner, 1992)

Rule	Code	Prob
S =>AB	0	0.5
S =>CB	1	0.5
A =>She	00	0.25
A =>He	01	0.25
A =>Susan	10	0.25
A =>Alex	11	0.25
B =>likes D	0	0.5
B =>detests D	10	0.25
B =>wants D	110	0.125
B =>hates D	111	0.125
C =>Everybody B	0	0.5
C =>The lady B	10	0.25
C =>A nice kid B	11	0.25
D =>milk.	00	0.25
D =>apples.	01	0.25
D =>pie.	10	0.25

Position	Prefix	Output
°1011010	1	CB
1°011010	0	Everybody B
10°11010	110	Everybody wants D
10110°10	10	Everybody wants pie.

Previous Work in Linguistic Steganography

- **Synonym Substitution:** Using mixed radix form

(Winstein, 1999)

	0 wonderful			$(101)_2 = 5$
	1 decent		0 city.	
Midshire is a	2 fine	little		
	3 great		1 town.	$\begin{pmatrix} a_1 & a_0 \\ 5 & 2 \end{pmatrix} = 2a_1 + 1a_0 = 5$
	4 nice			

- **NICETEXT:** Using a dictionary table and style template

(Chapman and Davida, 97)

Type	Code	Word
name-male	0	Ned
name-male	1	Tom
name-female	0	Jody
name-female	0	Tracy

Style

name-female name-male name-male

Payload

011

Cover Text

jody tom tom

Previous Work in NL Watermarking

- **Synonym Substitution (Atallah et al., 2000)**
 - If $m_i \bmod k = 1$ and $A(w_i) + r_i \bmod k$ is a quadratic residue modulo p , then w_i is kept same.
 - $m_i \bmod k$ is the current bit of watermark
 - $A(w_i)$ is the ASCII value of word w_i
 - p is a 20 digit prime key
 - k is the number of bits in the watermark
 - r_0, r_1, \dots, r_{k-1} is the sequence of pseudo-random numbers generated using p as key

Previous Work in NL Watermarking

- **Using sentence tree structure** (Atallah et al., 2001, 2002)
 - DCT analogy
 - Selection depends on the tree structure
 - Nodes are **labeled in pre-order** traversal
 - Then a node label j is converted to **1** if $j + H(p)$ is a **quadratic residue modulo p**
 - B_i is generated according to post-order traversal
 - A **rank d_i** is assigned for each sentence s_i using
$$d_i = H(B_i) \oplus H(p)$$
 - Starting from least-ranked sentence s_j the watermark is inserted **s_j 's successor in the text** by altering B_{j+1} using **linguistic transformations**

Previous Work in NL Watermarking

- **With Syntactic Transformations**

(Atallah et. al, 2001)

(S (NP Ned)

(VP loves (NP Jody))

(. .))

- **With Semantic Transformations**

(Atallah et al., 2002)

The EU ministers will tax aviation fuel as a way of curbing the environmental impact of air travel.

```
author-event-1--|--author--unknown
      |--theme--levy-tax-1--|--agent--set-4--|--member-type--geopolitical-entity
      |                               |--cardinality--unknown
      |                               |--members--(set| "EU nations")
      |--theme--kerosene-1
      |--purpose--regulate-1--|--agent--unknown-1
      |                               |--theme--effect-1--|--caused-by--flight
```

Summary

- **Steganography vs. Watermarking**
 - More complex methods
- **Still in its infancy**
- **Gain pace with synergy**
- **Statistical methods for robust and generic solutions**
- **Easier to work on syntactic structure**
 - Available tools are better developed
- **Harder to Evaluate**
 - Different styles, genres, authors and audience

Conclusion

- **Challenging problem !**
- **Collaboration with image watermarking will help**
- **Existing work is on preliminary level**
- **Wide range of application areas**