Deep learning for natural language processing Introduction to natural language processing

Benoit Favre <benoit.favre@univ-amu.fr>

Aix-Marseille Université, LIF/CNRS

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What is Natural Language Processing?

What is Natural Language Processing (NLP)?

- Allow computer to communicate with humans using everyday language
- Teach computers to reproduce human behavior regarding language manipulation
- Linked to the study of human language through computers (Computational Linguistics)

Why is it difficult?

- People do not follow rules strictly when they talk or write: "r u ready?"
- Language is ambiguous: "time flies like an arrow"
- Input can be noisy: speech recognition in the subway

NLP is everywhere

- Spell checker / grammar correction (Word)
- Information retrieval / search (Google)
- Machine translation (Google)
- Information extraction (Ask.com)
- Question answering (Jeopardy)
- Automatic summarization (Google news)
- Call routing (Telcos)
- Sentiment analysis (Amazon)
- Spam filtering (Email)
- Writing recognition (Cheque processing)
- Voice dictation (Dragon, Nuance)
- Speech synthesis (In-car GPS)
- Dialog systems (Siri/OK Google/Alexa...)

Domains reltated to NLP

- Artificial intelligence
- Formal language theory
- Machine learning
- Linguistics
- Psycholinguistics
- Cognitive Sciences
- Philosophy of language

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Communication channel

From the point of view of the source (the speaker)

- Intent: the message we want to communicate
- @ Generation: the message in linguistic form
- **③** Production: the muscular action which leads to sound production

From a receiver point of view (listener)

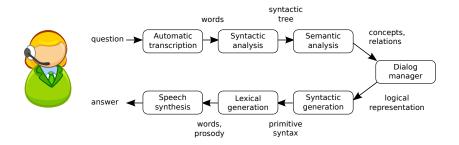
- Perception: how the sound is transmitted to neurons
- Analysis: interpretation of the linguistic message (syntactic, semantic...)
- Integration: believe or not the information, reply...

Processing levels

"John loves Mary"

- Lexical : segment character stream in words, identify linguistic units John/firstname-male loves/verb-love Mary/firstname-female
- Syntax : identify grammatical structures (S (NP (NNP John)) (VP (VBZ loves) (NP (NNP Mary))) (. .))
- Semantic : represent meaning love(person(*John*), person(*Mary*))
- Pragmatic : what is the function of that sentence in context? Is it reciprocal ? Since when ? What does it entail ? know(John, Mary)

Modular approach



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Language ambiguity

- Phonetic
 - I don't know! I don't no!
- Graphical
- Phonetic and graphical
 - I live by the bank (river bank or financial institution)
- Etymology
 - I met an Indian (from India or native American)
 - I love American wine (from USA or from the Americas)
- Syntactic
 - He looks at the man with a telescope
 - He gave her cat food
- Referential
 - She is gone. Who?
- Notational conventions
 - Birth date: 08/01/05

(wikipedia)

Basic NLP tasks

- Syntax
 - Word / sentence segmentation
 - Morphological analysis
 - Part-of-speech tagging
 - Syntactic chunking
 - Syntactic parsing
- Semantic
 - Word sense disambiguation
 - Semantic role labeling
 - Logical form creation
- Pragmatic
 - Coreference resolution
 - Discourse parsing

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Word segmentation

Character sequence \rightarrow word sequence (tokenization)

- Split according to delimiters [:,.!?']
- What about compounds? Multiword expressions?
- URLs (http://www.google.com), variable names (theMaximumInTheTable)
- In Chinese, no spaces between words:
 - ▶ 男孩喜歡冰淇淋。→ 男孩 (the boy) 喜歡 (likes) 冰淇淋 (ice cream)。

Morphological analysis

Split words in relevant factors

- Gender and number
 - flower, flower+s, floppy, flopp+ies
- Verb tense
 - parse, pars+ing, pars+ed
- Prefixes, roots and suffixes
 - geo+caching
 - re+do, un+do, over+do
 - ► pre+fix, suf+fix
 - geo+local+ization
- Agglutinative languages
 - pronouns are glued to the verb (Arabic, spanish...)
- Rich morphology
 - Turkish, Finish
- $\bullet \rightarrow$ Lemmatization task: find canonical word form

Part-of-speech tagging

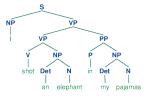
• Syntactic categories

Noun	Adverb	Discourse marker
Proper name	Determiner	Foreign words
Verb	Preposition	Punctuation
Adjective	Conjunctions	Pronouns

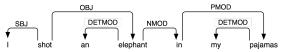
- Each word can have multiple categories
- Example : time flies like an arrow
 - flies: verb or noun?
 - like: preposition or verb?

Syntactic analysis

• Constituency parsing



Dependency parsing



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Word sense disambiguation (WSD)

What is the sense of each word in its context?

- red: color? wine? communist?
- fly: what birds do? insect?
- **bank**: river? financial institution?
- book: made of paper? make a reservation?

Word meaning highly depends on domain

- apple: fruit? company?
- to pitch: a ball? a product? a note?

(日) (四) (日) (日) (日)

Semantic parsing

Syntax is ambiguous

- The man **opens** the door
- The door opens
- The key **opens** the door

Semantic roles

- Who performed the action? the agent
- Who receives the action? the patient
- Who helps making the action? the instrument
- When, where, why?

John	sold	his car	to his brother	this morning
agent	predicate	instrument	patient	time

Reference resolution

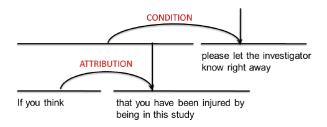
- Link all references to the same entity
 - "Alexander Graham Bell (March 3, 1847 –August 2, 1922)[4] was a Scottish-born[N 3] scientist, inventor, engineer, and innovator who is credited with patenting the first practical telephone." (Wikipedia)

Ambiguity

- Pronouns (it, she, he, we, you, who, whose, both...)
- Noun phrases (the young man, the former president, the company...)
- Proper names ("Victoria": South-African city, Canadian region, Queen, model...)

Discourse analysis

Relationship between sentences of a text, argument structure.



"Fully Automated Generation of Question-Answer Pairs for Scripted Virtual Instruction", Kuyten et al, 2012

Relation type (Rhetorical Structure Theory)

- Background
- Elaboration
- Preparation
- Contrast
- Objective

- Cause
- Circumstances
- Interpretation
- Justification
- Reformulation

Create a logical form

- Predicate representation
 - Can be used to infer new
- John loves Mary but it is not reciprocal.

 $\exists x, y, name(x, "John") \land name(y, "Mary") \land loves(x, y) \land not(loves(y, x))$

• John sold his car this morning to his brother.

 $\exists x, y, z, name(x, "John") \land brother(x, y) \land car(z) \\ \land owns(x, z) \land sell(x, y, z) \land time("morning")$

History of natural language processing

- 1950: Theory (test de Turing, grammaires de Chomsky)
 - Automatic translation during the cold war
- 1960: Toy systems
 - SHRDLU "place the red box next to the blue circle", ELIZA "the therapis"
- 1970:
 - Prolog (logic-base language for NLP), Dictionaries of semantic frames
- 1980: Dictation, Development of grammars
- 1990
 - Transition "introspection" \rightarrow "corpus"
 - Evaluation campaigns
 - Neural networks are "forgotten"
- 2000
 - Machine learning
 - Applications: speech recognition, machine translation
- 2010...
 - Deep learning

Notion of corpus

- Language in the wild
 - Email
 - Forums
 - Chats
 - Speech recordings
 - Video

• Manual Annotation of all elements we want to predict

- $\blacktriangleright \ {\sf Text} \to {\sf topic}$
- ▶ Sentence → parse tree
- Review \rightarrow sentiment

Methodology

Corpus-based natural language processing

- Define a task
- Write an annotation guide
- Collect raw data
- Ask people to annotate that data
- Oreate a system to perform the task
- O Evaluate the output of the system

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NLP Systems

- Input
 - Raw text, audio...
 - Sentences, contextualized words
 - Output of another system
- Output
 - n classes (ex: topics)
 - Structure (ex: syntactic parse)
 - Novel text (ex: translation, summary)
 - Commands for a system (ex: chatbots)
- Process
 - output = f(input)
 - Deterministic vs random (evaluations need to be repeatable)
 - Parametrisable: output = f(input, parameters)

What is the deep learning promise?

- "Classic" NLP system development
 - Requires a lot of feature engineering
 - Is affected by cascading errors
 - Hard to account for unlabeled data
 - Limited architectures and overly complex (ex: speech recognition...)
 - The curse of annotated data (you need linguists)
- Deep learning
 - Feature extraction is learned within the model
 - End-to-end training
 - Much more flexibility in model architectures
 - Can use tons of data
 - A step towards AI?

• Is this the end of linguistic expertise?