MOOC Kapita Selekta Sistem Cerdas Semester Genap 2021

The World with Language Technologies

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The Speaker

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Education background:



- 2012 2017 Computer Science at Universitas Gadjah Mada, Indonesia
- 2017 2019 Computational Linguistics at Saarland University, Germany

Computer science/computational linguistics at Charles University, Czech Rep.

Research & Professional

- 2018 Research Assistant @ Dept. of Language Science & Technology, Saarland Uni.
- 2019 2020 Research Assistant @ German Research Center for AI (DFKI)
- 2020 now NLP Engineer @ Prosa.ai

Research interest: natural language processing, machine learning **Also likes:** running, traveling, books, and coffee

Gaining linguistic capabilities with technologies

- What are the capabilities?
- What are the technologies?
- Where are we now?
- What are the challenges?
- What can be in the future?

Towards artificial intelligence systems

- Artificial intelligence is an imitation of human intelligence demonstrated by machines to solve some problem
- One of the intelligences we would like to achieve is linguistic intelligence
 - how human use languages (natural languages)
 - both in written and spoken settings.



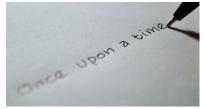
Human intelligence category according to Howard Gardner

What are the capabilities?

Some examples are:

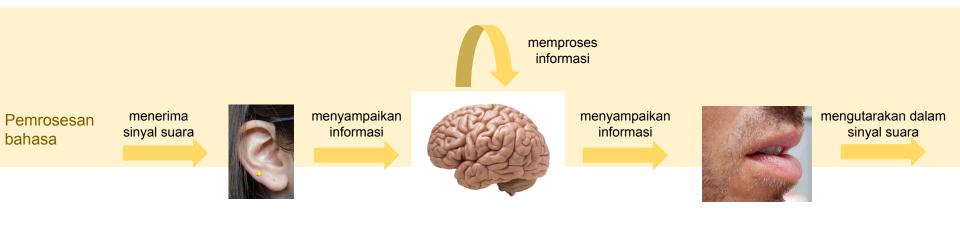
- Producing a speech
- Writing a story
- Responding to a question
- Speaking more than one language
- Giving some recommendation
- Understanding analogy
- Having conversation
- etc





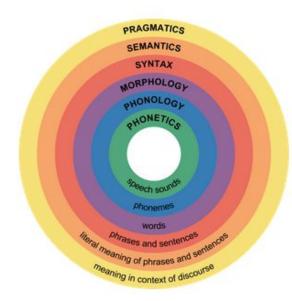


Human communication (spoken)



Language and Linguistics

- Language is the ability to produce and comprehend spoken and written words
- Linguistics is the study of language
- Linguistic level:



Source: <u>lumenlearning.com</u>

Can we make technologies to have such capabilities?

Actually, we have used a lot of language technologies.

- Spam detection



- Keyword suggestion



Q	saya makan nasi	×
Q	saya makan nasi	
Q	saya makan nasi di pagi hari	
Q	saya makan nasi bahasa inggrisnya	
Q	saya makan nasi goreng bahasa inggrisnya	
Q	saya makan nasi bahasa arabnya	
Q	saya makan nasi goreng	

- Machine Translation

🗙 Teks Dokumen		
DETEKSI BAHASA INDONESIA INGGRIS JERMAN	✓ KOREA INDONESIA INGGRIS ✓	
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Kirim masukan

- Dialogue system



Chatbot Prosa.ai

- Auto-complete, auto-correct, and spell-checker

Tersimpan	
Penerima	have 🛞 🛄 🛔
Smart Compose Subjek	I think I has known you before.
Saran penulisan membantu Anda v Good morning, menghemat waktu saat	
membuat draf email. a Let me introduce myse	elf as Cika. I was a fresh graduate student.
y y	tidak 🛞 🛄 i Aku tdak suka makan mei goreng

- Speech recognition (ASR) and Text-to-Speech (TTS)



In order to imitate the language capabilities, the technologies must run some processing on the language as input (both written and spoken).

Hence, the discipline learning how machine process natural languages is called **natural language processing (NLP)**.



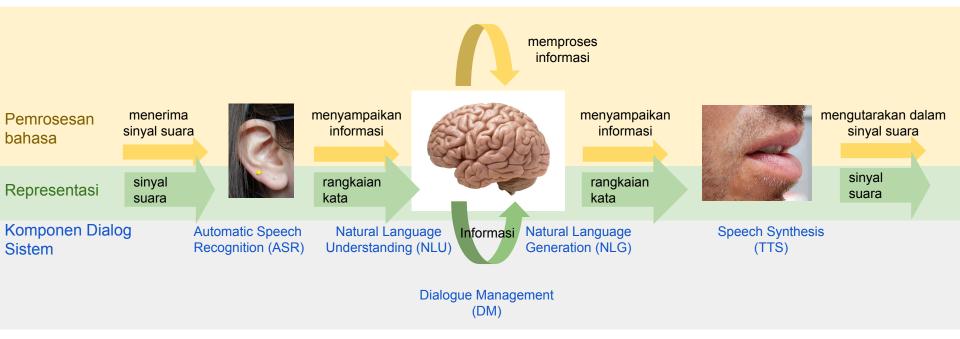
For the sake of simplicity, NLP tasks can be categorized into two types:

- Natural Language Understanding (NLU)
- Natural Language Generation (NLG)

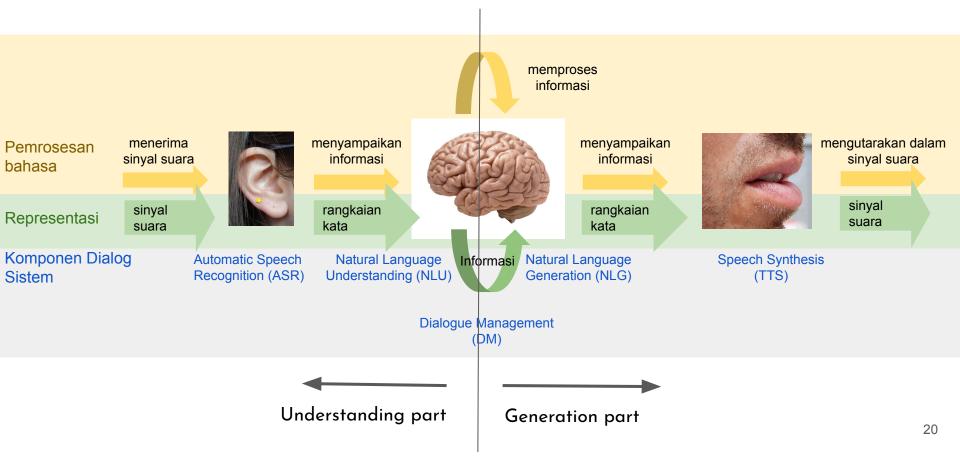
NLP Tasks

- We can be dealing with different **linguistic level**:
 - Phonetics/phonology \rightarrow sound
 - Morphology \rightarrow morphemes
 - Syntax \rightarrow sentence structure
 - Semantics/Pragmatics → meaning
- or **input types**:
 - sound signal
 - text (our main focus)
 - others: picture, table, ontology, etc.

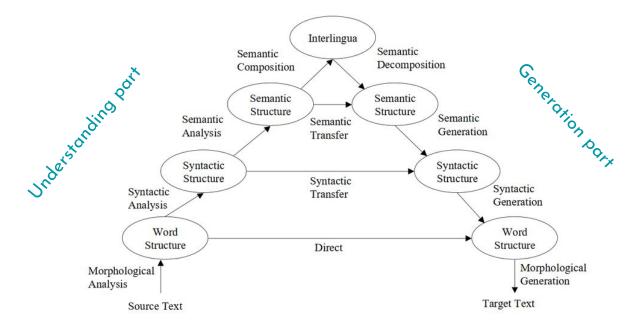
Example 1: Dialogue system



Example 1: Dialogue system



Example 2: Machine Translation



Vauquois triangle for rule-based machine translation systems

Example 3: Text Classification (Understanding)

- To assign some text to some fixed category
- We should have labeled text
- Examples:
 - News Classification: Sport, Health, Finance, Politics, Technology
 - Hate Speech Detection: Yes, No
 - Plagiarism Detection: Yes, No
 - Hoax News Detection: Yes, No
 - Sentiment Classification: Positive, Negative, Neutral

Example 4: Information Extraction (Understanding)

- To extract some structured information from a non/semi-structured information, e.g. texts
- Examples:
 - Template filling
 - Named entity recognition

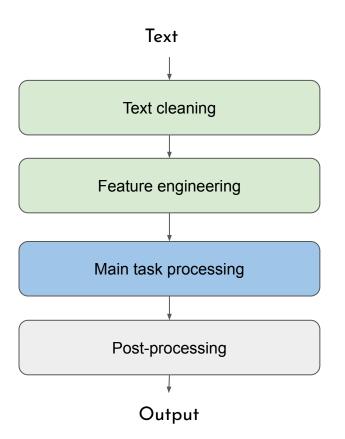
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Event	T		T
	Person	Organisation	Location
Figure 1: An ex	ample of NER a	pplication on an example	text
	S	ource: inspiratron.c	rg

Example 5: Text generation

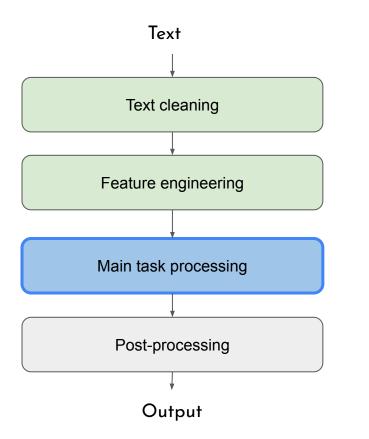
- To generate text given some input:
 - Text seed, table, image, etc..
- Examples:
 - Answer generation in QA system
 - Story/article generation:
 - e.g. <u>http://ai-writer.com</u>, <u>https://notrealnews.net/</u>

How do machines process the languages?

Text processing pipeline



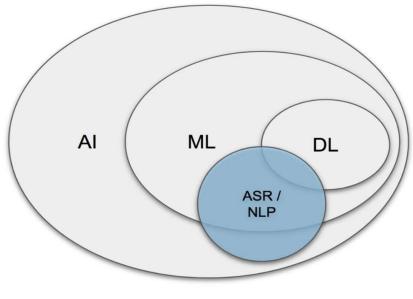
Text processing pipeline: main task processing



Common approaches:

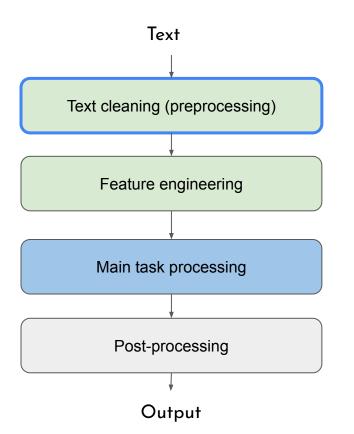
- Rule-based
 - Using handcrafted rules
- Statistical/machine learning (ML)
 - Data-driven
- Neural/deep learning (DL)
 - Also data-driven and various neural network architectures

Text processing pipeline: main task processing



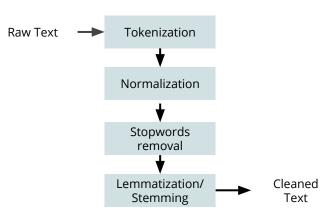
source: sonix.ai

Text processing pipeline: text cleaning

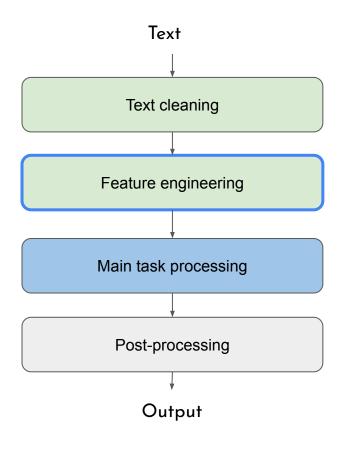


To clean the raw text, so it becomes a kind of text that is useful for our system.

Example pipeline:



Text processing pipeline: feature engineering

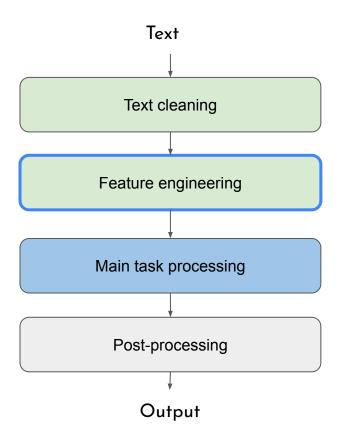


- **Feature** is a property of an object being observed.
- **Objective**: to gain extra information from the existing data that can helps our system solve the problem.
- It requires **understanding of the problem and data** to select appropriate features.

Examples:

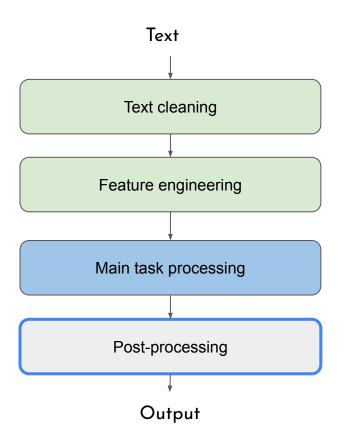
- Part-of-Speech (POS) tags
- $\circ \quad \text{Word position} \quad$
- Word-before, word-after
- N-gram
- Lemmas
- Text length
- o etc

Text processing pipeline: feature engineering



- We need to transform the features into numbers because machines can only read numbers
- Mostly used representation: vectors
- Approaches:
 - Label encoder
 - Bag-of-words
 - Term Frequency-Inverse Document Frequency (TF-IDF)
 - (word) embeddings
 - etc

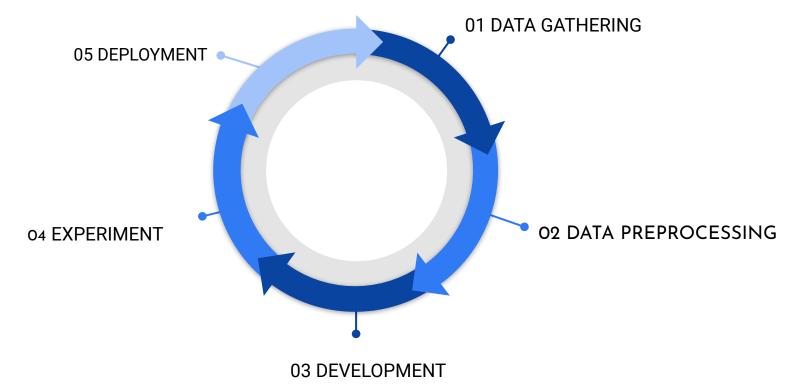
Text processing pipeline: post-processing



The main process may not be "perfect"

Optionally, we can post-process the output of our main approach to get results with better quality

It's not a one-time process



Evaluation

- Research:
 - Performance on test dataset
 - vs baseline system or human participants
- Application:
 - Performance on real users
 - benchmark may vary, depending on the application goal
 - \circ $\,$ e.g. A/B Testing $\,$

At which state are we now?

AI in Sci-fi: what people expect



AI in reality: what people encounter



Capturing "meaning" is not easy

Language Technology Progress

Nice summary for various tasks: <u>https://nlpprogress.com/</u>

- Different **task** may have different performance
- Different language may have different performance
 - English is still the most heavily researched language
- Different **domain** may have different performance
 - E.g. News, sport, legal, health, formal language

Example: Question Answering and Information Retrieval

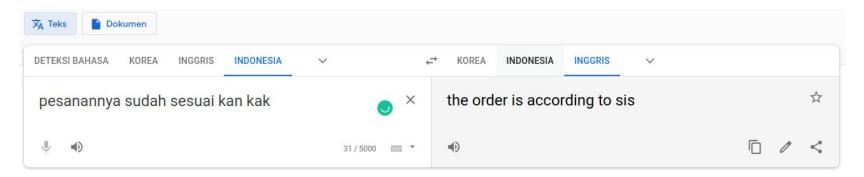
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Q All	🗉 News	🖾 Images	⊘ Maps	▶ Videos	: More	Settings	Tools
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Example: Machine translation

DETEKSI BAHASA KOREA INGGRIS		↔ KOREA INDONESIA INGGRIS ✓		
Tadi aku dimarahi ibu padal angkat jemuran sama masa		My mother scolded me earlier, even though I helped lift the clothesline and cook for my younger siblings		
	90 / 5000	•)	<	

Example: Machine translation





There are many applications

Only for **dialogue systems**:

- Smart Speakers
 - Siri, Alexa, Google Assistant
- Telephone
- Computer games
- Chatbots
- Assistive technologies
 - Therapy, elderly care, psychology consultation
- Built-in car dialogue system
- Research systems

Can we develop a new one?

There are still rooms for progress

Challenges in building language technologies

Ambiguity (at all level)

• Lexical (word)

Kakak datang untuk memberi **tahu**.

• Syntactic

Saya membaca **buku sejarah musik baru**.

• Semantic

Budi berlibur dengan istrinya, begitu juga Arif.

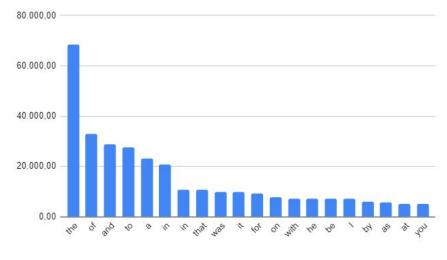
- Pragmatic
 - Jam berapa sekarang?

Bisa memiliki dua makna:

- menanyakan jam (arti sebenarnya)
- menyindir/marah karena orang lain terlambat

Language is creative and infinite

- There are so many words
 - Zipf law: The frequency of a word occurrence in inversely proportional to its rank



Top-20 words in English (from a corpus)

Language is creative and infinite

- New words inclusion
 - e.g. in KBBI: daring, warganet, luring, pramusiwi
- Slang words
 - e.g. mager, woles, baper, bucin
- Code-switching
 - Seriously, harganya mahal banget
 - Nanti **nek misale** Bapak udah nyampe, kamu kabari aku ya
- One may produce a sentence that no one has ever produced
 - Factors: length, choose of words, etc

Data-driven methods NEED data

- Quality vs quantity
 - Should achieve both
 - Internet is a good resource but there is so much noise
- This is why systems using low-resource language struggles

Ethical use of language technologies

- Data collection
- Data privacy in delivery
- The result of language generation:
 - Hate speech, false news, racism, harassment
 - Plagiarism
 - Killing creative industry?

Language Technologies in the future

The research is still going on ...

- Inclusion of other languages (i.e. not just English)
- More application in different domains:
 - The use in healthcare is getting more attention
 - Open-domain understanding system is still unsolved
- Multimodality: combining multiple "intelligence"
 - Text + sound + images

