

# Natural-language understanding in robotics

*(for domestic service robots, actually)*

Mauricio Matamoros

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# On Language

## Definition

Saussure [1] formally defines the term  $\text{language}_1$  (*fr. langue*, *de. „Spreache“*) as the “work of a collective intelligence”, in contrast with  $\text{language}_2$  (*fr. parole*, *de. „Wort, Redeweise“*) which denotes the particular utterances of an individual.

# On Language

## Philology

Philology studies the historical growth, change, and adaptation of languages, specially focusing in *meaning*. It pays special attention to written literature.

## Linguistics

Linguistics studies the phonetics and structure of languages, as well as their changes over time and the relationships among them.

# On Language

## Linguistics

Linguistics studies the phonetics and structure of languages, as well as their changes over time and the relationships among them.

- **Phonology and Phonetics** study the sounds of a language.
- **Morphology** studies the structure of words and its transforms.
- **Syntax** studies the grammatical aspects of a language, i.e. the structure of sentences.
- **Semantics** studies the meaning of words and sentences.
- **Pragmatics** studies the meaning of a sentence within a given context.

# Formal and natural languages

According with Tarski [2] these are some of the differences between natural and formal languages:

## Natural Languages

- Not finished
- Constantly changing
- Universal:  
*anything that can be  
though, can be spoken of*

## Formal Languages

- ①  $\exists$ , a structural definition  $\forall$  symbol
- ②  $\exists$ , a structural definition  $\forall$  axioms or primitive statements
- ③ Inference rules allow transform sentences into axioms
- ④ All statements are unambiguous

## Controlled Natural Languages

Natural languages are often untractable. Therefore all applications make use of a *Controlled Natural Language* subset that constrains the lexicon and syntax to whatever is relevant within the application domain.



# Natural Language Processing

## Natural Language Processing

Natural Language Processing (NLP) is the branch of Artificial Intelligence concerned about the interactions between computers and human (natural) languages, focusing on the creation of computer programs to process and analyze natural language.

## Natural Language Understanding

Natural Language Understanding (NLU) is a sub-branch of NLP focused on enabling computers to process and understand human languages.

# Natural Language Processing: Applications

- Spelling correction
- Grammar checking
- Question answering
- Dialog Systems
- Language detection
- Machine Translation
- Text classification
- Summarization
- Sentiment analysis
- Opinion mining
- Conversational Agents
  - Chatbots
  - Customer service
  - Client satisfaction survey

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## Robotics

# Approaches

## Classic

- Analytical
- Inspired in linguistics
- Requires programmer expertise

## Stochastic

- Probabilistic
- Machine-learning inspired
- No expertise required

# Approaches

## Classic

### Pros

- No corpora required
- Computationally cheap
- Easier to ground

### Cons

- Hard to develop
- Memory expensive
- Human-error prone

## Stochastic

### Pros

- Easy to develop
- Relatively fast
- Robust

### Cons

- Requires [very] large corpora
- Computationally expensive
- Extending requires retraining

E L I Z A

# E L I Z A

---

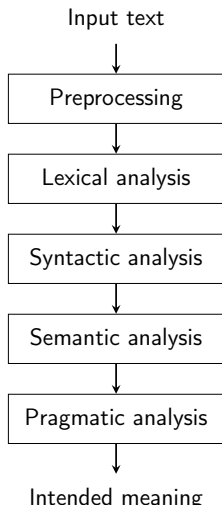
```
1 tar -xzf Chatbot-Eliza-1.08.tar.gz
2 cd Chatbot-Eliza-1.04
3 perl Makefile.PL
4 make
5 sudo make install
6 cd examples
7 ./single
```

---

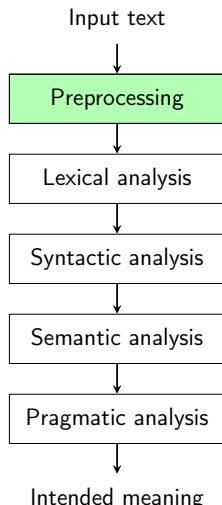
E L I Z A



# Summary of analytic NLP methods



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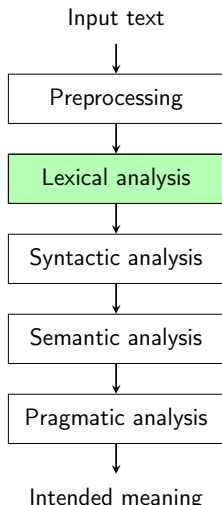


## Preprocessing\*

Cleans and split the whole text into paragraphs, paragraphs into sentences, and sentences into words (tokenization).

It is specially important in non-space delimited languages.

# Summary of analytic NLP methods



## Lexical analysis (Lexer)\*

Separates known and unknown words, and finds each word's canonical form (root).

- Stemmer
- Lemmatizer

### Example

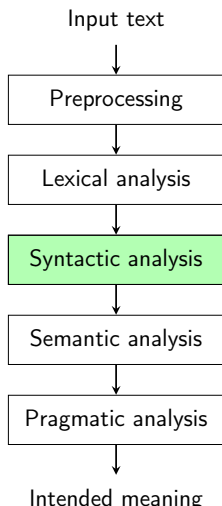
Hand me the apple juice

Gib mir den Apfelsaft

Hand · me (I) · the · apple-juice

gib (geben) · mir (ich) · den (der) · apfelsaft

# Summary of analytic NLP methods



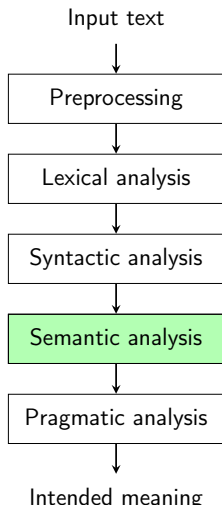
## Syntactic analysis (Parser)

Analyzes the relationship between the words in a sentence, and identifies their roles (POS-tagging and dependency graphs).

Also rejects unsupported inputs.

- Context free grammars (up to  $O(n^3)$ )
  - Generalized Phrase-Structure Grammar
  - Head-driven Phrase-Structure Grammar
  - Lexical-Functional Grammar
- Mildly Context-Sensitive Grammars ( $NP$ )
  - Tree-Adjoining Grammar
  - Combinatory Categorical Grammar
- Immediate Dominance, Linear Precedence, etc

# Summary of analytic NLP methods



## Semantic analysis (Semantic parser)

Determines the meaning of a sentence and provides a structural [logic?] representation which can be grounded, i.e. mapped to representations of objects in the real world.

- First (and  $n - th$ ) order logic
- Lambda calculus
- Semantic networks
- Semantic frames
- Description logics
- Semantic augmentation of CFG
- Natural Semantic Metalanguages
- Pustejovsky's Generative Lexicon
- Ontologies

# Exercise: Lexical Analyzers

## Lemmatizer

```
1 from nltk.stem import WordNetLemmatizer
2 nlp = WordNetLemmatizer()
3
4 print nlp.lemmatize('getting', 'v')
5 print nlp.lemmatize('feet', 'n')
6 print nlp.lemmatize('deadly', 'r')
7 print nlp.lemmatize('xyzing', '')
```

## Stemmer

```
1 from nltk.stem import SnowballStemmer
2 nlp = SnowballStemmer('english')
3
4 print nlp.stem('getting')
5 print nlp.stem('feet')
6 print nlp.stem('deadly')
7 print nlp.stem('xyzing')
```

# Summary of statistical NLP methods

|                            | Preproc. | Lex | Syn | Sem |
|----------------------------|----------|-----|-----|-----|
| Entropy Maximization       | 😊        | 😐   | 😊   | 😐   |
| Hidden Markov Models       | 😐        | 😐   | 😊   | 😞   |
| Markov Decision Processes  | —        | —   | 😐   | 😊   |
| Neural Networks            | 😞        | 😞   | 😐   | 😞   |
| Probabilistic CFG          | —        | —   | 😊   | 😊   |
| Statistical Decision Trees | —        | —   | 😊   | 😐   |
| Support Vector Machines    | 😊        | —   | 😊   | —   |

Other approaches (Not suitable for robotics)

- Clustering
- K-means
- Linear Regression
- Markov Random Fields
- Mixture Models
- Perceptron and Kernel methods

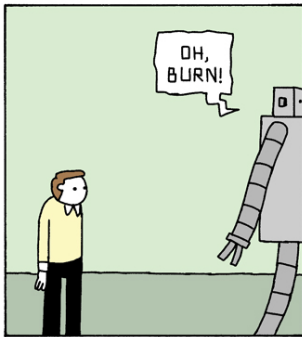
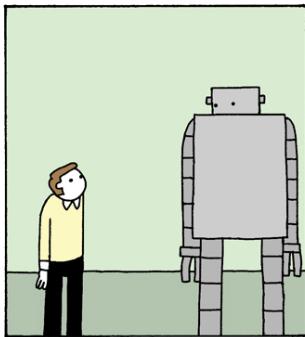
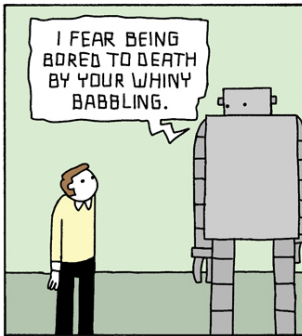
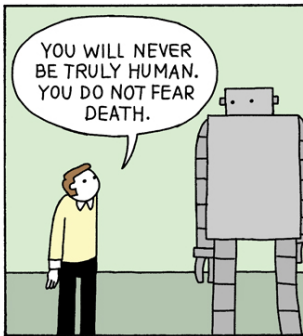
# Summary of statistical NLP methods

Take the best of both worlds to meet application requirements

## In robotics

- No preprocessing (input clean from ASR)
- No lexical analysis (naive replacement)
- POS-Tagger (Stanford, if any)
- *Ad-hoc* semantics (mostly first order logic, if any)





# How can we communicate with a computer?

- Coding

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- Interface HW like smartphones (e.g. Telegram)

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- Gestures and sign language

# How can we communicate with a computer?

- Coding
- Interface HW like smartphones (e.g. Telegram)
- Gestures and sign language
- Talk to it!

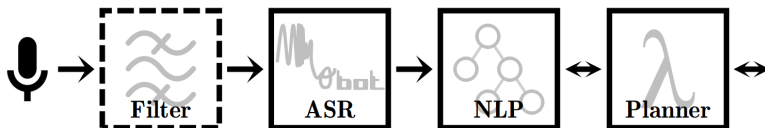
# How can we communicate with a computer?

## Observation

Always keep in mind

People do NOT want to read a manual

# Golden pipeline



# Golden pipeline



Cardioid for best results.



Normally absent (HARK is among the most used).



Mostly grammar-based solutions, but cloud services are used when possible.



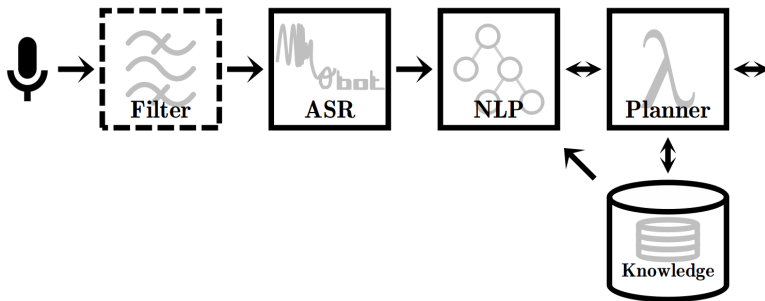
Varies widely. Stanford CoreNLP and LU4R are becoming popular. Mostly keyword spotting.



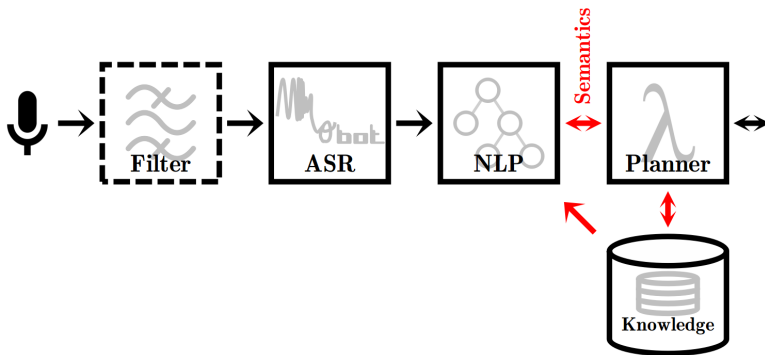
Varies widely.  $\lambda$ -calculus, KnowRob, and Watson are becoming popular. Mostly state machines.



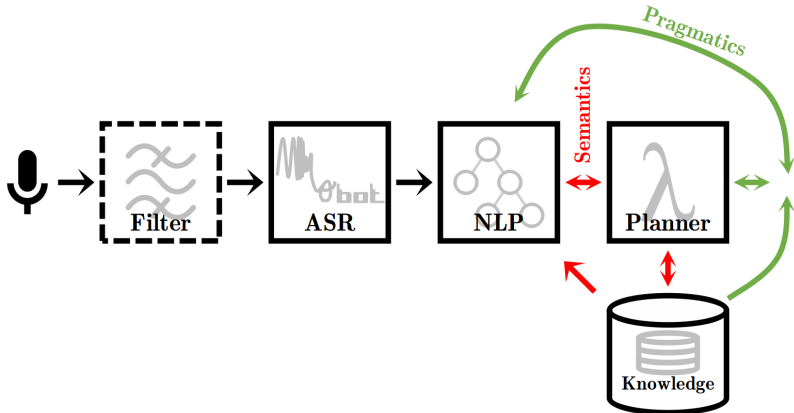
# Golden pipeline



# Golden pipeline



# Golden pipeline



# Challenges

## Example: Ambiguity

*John saw the girl with a telescope*

# Challenges

## Example: Ambiguity

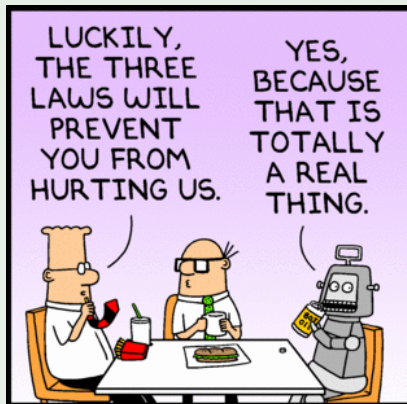
*John saw the girl with a telescope*



|      |  |  |
|------|--|--|
| John | John used a telescope to cut the girl (in halves)    | John used a telescope to see the girl      |
| Girl | John used a saw to cut the girl that had a telescope | John had seen the girl who had a telescope |

# Challenges

## Example: Sarcasm and Irony



# Challenges

- Action Grounding
- Contextualized, time-sensitive dialogs
- Environmental awareness and reasoning

Robot actions alter the physical world

- Recognize what's doable and what isn't
- Consider world's physics
- Consider abstractions

# Challenges

- Action Grounding
- Contextualized, time-sensitive dialogs
- Environmental awareness and reasoning

It shouldn't take longer to ask than to act

- People won't explain twice
- People won't explain the obvious

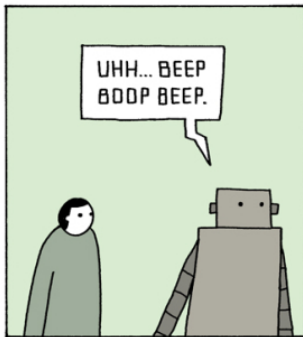
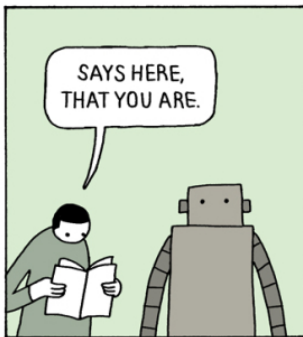
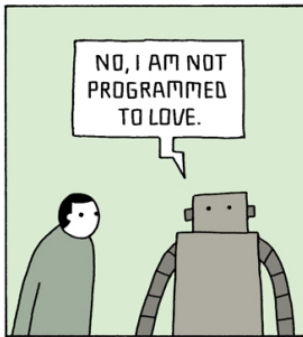
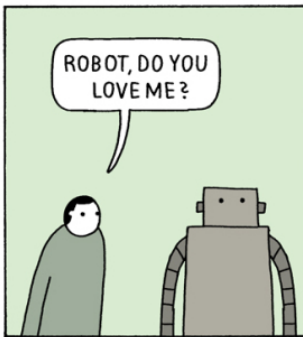


# Challenges

- Action Grounding
- Contextualized, time-sensitive dialogs
- Environmental awareness and reasoning

Previous deeds and surroundings alter the interaction

- ...*just like the other day*
- Most information is omitted when speaking



# Types of sentences

- Declarative
- Interrogative
- Imperative

# Types of sentences

- **Declarative**  
Facts. Store new knowledge.
- Interrogative
- Imperative

## Example

I lost my keys

# Types of sentences

- Declarative
- **Interrogative**  
Questions. Retrieve knowledge.
- Imperative

## Example

Have you seen my keys?

# Types of sentences

- Declarative
- Interrogative
- **Imperative**  
Commands. Trigger an action.

## Example

Find my keys, NOW!

# Keyword spotting

## Strategy

Find relevant words and discard the rest

*Please, **stop** it!*                      (*stop*)  $\rightarrow$  stop()

*Could you help me to  
**find** my **keys**?*                      (*find, keys*)  $\rightarrow$  find(keys)

*I'd like to have **cheese**  
**fingers** and a **coke***                      (*cheese, fingers, coke*)  $\rightarrow$   
deliver(cheese-fingers, coke)

# Keyword spotting and Pattern matching

## Strategy

Use keyword-based patterns to discern similar information

It is common to rely on the (augmented) CFG used by the ASR

## Example

*... and place the napkin over the cup*

→ place/VB · (the/DT · napkin/NN)/NP · (over/IN)/PP · (the/DT · cup/NN)/NP

→  $S : VB_{\text{man}} NP_{\text{obj}} PP_{\text{stack}} NP_{\text{obj}}$

→ place(napkin, glass)



# Frame-based dialog

## Strategy

Fill the gaps (slots) in a dialog frame

Each slot in the frame has an associated question, and a set of keywords or patterns

- ① Keyword-spot the main verb  $V$  (or find the most likely one)
- ② Use  $V$  to select the adequate frame  $F$  (or the most likely one)
- ③ Use Keyword-spotting/pattern-matching to fill  $F$
- ④ If there are gaps
  - ① Select a gap and make the associated question
  - ② Return to 3
- ⑤ Confirm information
- ⑥ Run

# Moving the turtle using NL

```
import re
from robot import Robot

def parse(s):
    keywords = ['advance', 'step', 'turn', 'stop', 'spin']
    parts = re.split(r'\s+', s.lower())
    for word in parts:
        if word in keywords:
            return 'robot.{}()'.format(word)
    return None
#end def

def main():
    robot = Robot()
    while(True):
        s = raw_input('? ')
        action = parse(s)
        if not action is None:
            eval(action)
#end def
```

# Moving the turtle using NL

```
import re
from robot import Robot

def parse(s):
    parts = re.split(r'\s+', s.lower())
    for word in parts:
        switcher = {
            'walk'      : "step",
            'run'       : "advance",
            ...
            'stop'      : "stop",
            'spin'      : "spin",
        }
        func = switcher.get(word, None)
        if not func is None:
            return 'robot.{}()'.format(func)
    return None
#end def
```

# Moving the turtle using NL

## Command

*Please start spinning*

# Moving the turtle using NL

```
import nltk
from robot import Robot

def parse(s):
    parts = nltk.word_tokenize(s.lower())
    stemmer = nltk.stem.SnowballStemmer('english')
    for word in parts:
        switcher = {
            'walk'      : "step",
            'run'       : "advance",
            ...
            'stop'      : "stop",
            'spin'      : "spin",
        }
        func = switcher.get(stemmer.stem(word), None)
        if not func is None:
            return 'robot.{}()'.format(func)
    return None
#end def
```

# Moving the turtle using NL

Command

*Turn left*

# Moving the turtle using NL

```
def parse(s):  
    ...  
    switcher = {  
        'turn'      : "turn({})".format(parts),  
        'walk'      : "robot.advance(0.5)",  
        'run'       : "robot.advance(1.5)",  
        'advance'   : "robot.advance()",  
        'stop'      : "robot.stop()",  
        'spin'      : "robot.spin()",  
    }  
    func = switcher.get(stemmer.stem(word), None)  
    ...  
#end def  
  
def turn(parts):  
    if 'left' in parts:    robot.rotate(90)  
    elif 'right' in parts: robot.rotate(-90)  
    else:                 robot.turn()  
#end def
```

# Moving the turtle using NL

## Now with a grammar (turning only)

```
1  % start S
2  S[F=?f, A='') -> VB_STOP[F=?f]
3  S[F=?f, A=?a] -> VB_TURN[F=?f] DIR[A=?a]
4  S[F=?f, A=?a] -> VB_TURN[F=?f] PREP DET DIR[A=?a]
5  VB_TURN[F='rotate'] -> 'rotate' | 'turn' | 'go' | 'drive' | 'head'
6  VB_STOP[F='stop'] -> 'stop' | 'halt' | 'break' | 'enough'
7  PREP -> 'to' | 'towards'
8  DET -> 'the'
9  DIR[A='90'] -> 'left'
10 DIR[A='-90'] -> 'right'
11 DIR[A='180'] -> 'back' | 'around'
```



# Moving the turtle using NL

---

```
def parse(s):
    try:
        fgrammar = nltk.grammar.FeatureGrammar.fromstring(
            grammar)
        fparser = nltk.parse.FeatureChartParser(fgrammar)
        trees = list(fparser.parse(nltk.word_tokenize(s.lower())
            ))
        if len(trees) < 1:
            return None
        tree = trees[0]
        return 'robot.{}({})'.format( trees[0].label()[ 'F' ],
            trees[0].label()[ 'A' ])
    except Exception as ex:
        print ex
        return None
#end def
```

---

# Coupling an ASR engine (Pocketsphinx)

---

```
from pocketsphinx import LiveSpeech

speech = LiveSpeech()
for phrase in speech:
    recognized = str(reco)
    # Do NLP
#end for
```

---

## Coupling an ASR engine (Pocketsphinx)

---

```
from pocketsphinx import LiveSpeech, get_model_path

speech = LiveSpeech(
    verbose          = False,
    sampling_rate    = 16000,
    buffer_size      = 2048,
    no_search        = False,
    full_utt         = False,
    hmm              = os.path.join('model/en-us/en-us'),
    lm               = 'model/en-us/en-us.lm.bin',
    dic              = 'data/en-us/robot.dict'
)
for phrase in speech:
    recognized = str(reco)
    # Do NLP
#end for
```

---

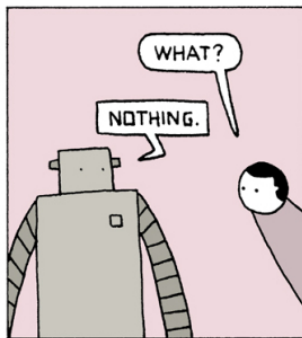
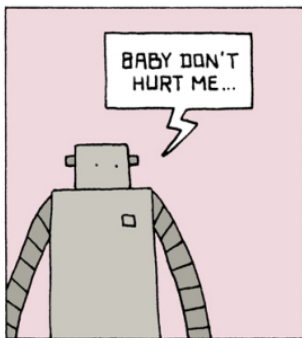
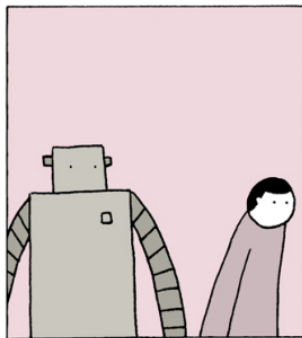
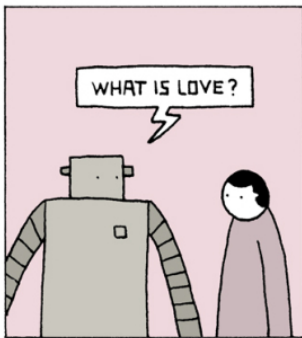
# Coupling an ASR engine (Google)

---

```
import speech_recognition as sr

reco = sr.Recognizer()
with sr.Microphone() as source:
    reco.adjust_for_ambient_noise(source)
while True:
    try:
        audio = reco.listen(source)
        recognized = reco.recognize_google(audio, language='en-US')
        # Do NLP
    except:
        continue
#end while
```

---



# Hausarbeit

## Exercise 1

Based on Example #5, code a ROS node that can execute exactly three instructions in a row, such as:

*Walk five steps, turn left, and run.*

## Exercise 2

Modify the ROS node so it can execute commands made of up to three instructions (i.e. between one and three).

## Suggested Bibliography

- 1 Nitin Indurkha and Fred J. Damerau. *Handbook of natural language processing, volume 2*. CRC Press, 2010.
- 2 Dan Jurafsky and James H Martin. *Speech and language processing: An introduction to natural language processing, computational linguistics, and speech recognition*. Prentice Hall, Pearson Education International, 2009.

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- ① James Allen. *Natural language understanding*. Pearson, 1995.
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- ⑤ George A. Miller. *Wordnet: a lexical database for english*. Communications of the ACM, 38(11):39–41, 1995.
- ⑥ Ruslan Mitkov. *The Oxford handbook of computational linguistics*. Oxford University Press, 2005.



# References

- [1] Ferdinand de Saussure. Cours de linguistique générale. Payot, Lausanne-Paris, 1916.
- [2] Alfred Tarski. The concept of truth in formalized languages. Logic, semantics, metamathematics, 2:152–278, 1956.