# Computational Linguistics Seminar spaCy NLP

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 $The \ banner \ image \ is \ a \ fragment \ of \ Primordial \ Soup \ at \ https://regenaxe.com/2017/01/17/primordial-soup/appendixed-com/2017/01/17/primord$ 

#### Overview

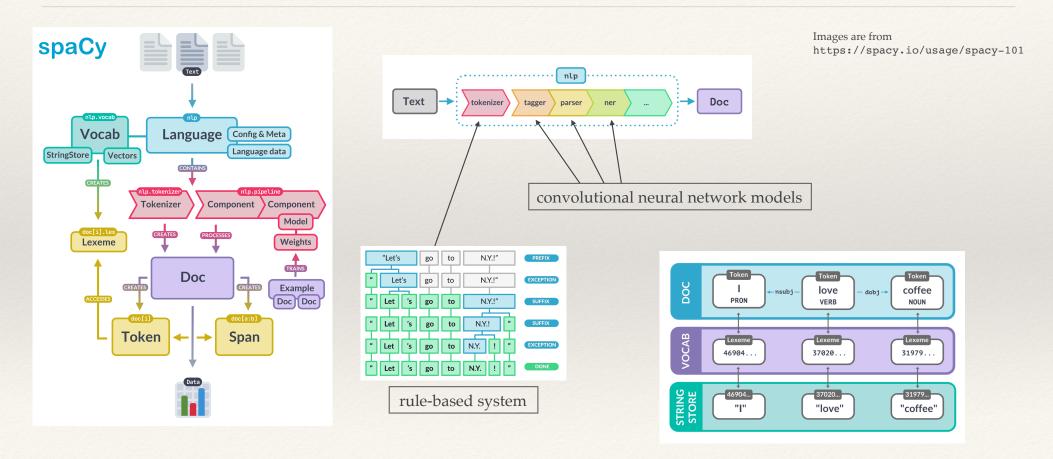
spaCy

- \* Assignment
- \* Some spaCy concepts
  - \* <a href="https://spacy.io/usage/spacy-101">https://spacy.io/usage/spacy-101</a>
- \* Pattern Matching
  - \* token patterns and phrase patterns
- Named Entities
- Vectors and similarities

### spaCy concepts

- \* Architecture
  - \* Text, documents, pipelines and annotations
  - \* The vocabulary: Tokens versus Lexemes
- \* Doc, Span and Token

#### Architecture



### Docs, Spans and Tokens

- \* Documents hold all annotations
- \* Tokens are created by the tokenizer when the document is first created
- Other components add to the Doc or Token object

name	description	creates
tagger	Part-of-speech tagger	Token.tag, Token.pos
parser	Dependency parser	Token.dep, Token.head, Doc.sents, Doc.noun_chunks
ner	Named entity recognizer	Doc.ents, Token.ent_iob, Token.ent_type

### Docs, Spans and Tokens

- \* Spans are slices of documents
- \* When you loop over them...
  - \* ... you get tokens (just like with the document)
- Named entities, sentences and noun chunks:
  - \* are all stored on the Document
  - \* are all instances of Span

```
1  nlp = spacy.load("en_core_web_sm")
2  for nc in nlp("We are eating pizza").noun_chunks:
3     print(type(nc), nc)

<class 'spacy.tokens.span.Span'> We
<class 'spacy.tokens.span.Span'> pizza
```

### Downloading a model

\* \$> python -m spacy download en\_core\_web\_sm

```
[12:24:24] .../site-packages/en_core_web_sm> ls -al
total 32
drwxr-xr-x
            6 marc admin 192 Feb 12 07:24 .
drwxr-xr-x 212 marc admin 6784 Feb 26 07:50 ...
-rw-r--r--
          1 marc admin 236 Feb 12 07:24 __init__.py
          3 marc admin 96 Feb 12 07:24 __pycache_
drwxr-xr-x
drwxr-xr-x 14 marc admin 448 Feb 12 07:24 en_core_web_sm-3.0.0
-rw-r--r-- 1 marc admin 9362 Feb 12 07:24 meta.json
[[12:24:30] .../site-packages/en core web sm> ls -al en core web sm-3.0.0/
total 208
                          448 Feb 12 07:24 .
drwxr-xr-x 14 marc admin
                          192 Feb 12 07:24 ..
drwxr-xr-x 6 marc admin
-rw-r--r 1 marc admin 6253 Feb 12 07:24 accuracy.json
drwxr-xr-x 3 marc admin
                          96 Feb 12 07:24 attribute_ruler
-rw-r--r-- 1 marc admin
                          5257 Feb 12 07:24 config.cfg
drwxr-xr-x 3 marc admin
                          96 Feb 12 07:24 lemmatizer
                          9362 Feb 12 07:24 meta.json
-rw-r--r-- 1 marc admin
drwxr-xr-x 5 marc admin
                          160 Feb 12 07:24 ner
                          160 Feb 12 07:24 parser
drwxr-xr-x 5 marc admin
                          128 Feb 12 07:24 senter
drwxr-xr-x 4 marc admin
drwxr-xr-x 4 marc admin
                          128 Feb 12 07:24 tagger
drwxr-xr-x 4 marc admin
                         128 Feb 12 07:24 tok2vec
-rw-r--r- 1 marc admin 77375 Feb 12 07:24 tokenizer
drwxr-xr-x 6 marc admin
                         192 Feb 12 07:24 vocab
```

## Pattern Matching

```
import spacy
from spacy.matcher import Matcher
from spacy.matcher import PhraseMatcher

nlp = spacy.load("en_core_web_sm")
```

#### Matching on tokens

Here we define patterns by using a dicitonary for each token. The following patterns matches 'iPhone X':

```
[{"TEXT": "iPhone"}, {"TEXT": "X"}]),
```

Instead of accessing the text you can access many of the features on a token, the following matches "2018 FIFA World Cup":

```
[{"IS_DIGIT": True}, {"LOWER": "fifa"}, {"LOWER": "world"}, {"LOWER": "cup"}]
```

You can access parts of speech and lemmas:

```
[{"LEMMA": "love", "POS": "VERB"}, {"POS": "NOUN"}]
```

And use some Kleene operators (possible values are "!", "?", "\*" and "+", where "!" is negation, as in, no match):

```
[{"LEMMA": "buy"}, {"POS": "DET", "OP": "?"}, {"POS": "NOUN"}]
```

#### Named Entities

```
import spacy

nlp = spacy.load("en_core_web_sm")

doc = nlp("Apple is looking at buying U.K. startup for $1 billion")

for ent in doc.ents:
    print(ent.text, ent.start_char, ent.end_char, ent.label_)
```

#### Similarities

```
1 nlp lg = spacy.load("en core web lg")
 2 nlp lg("Fido barks.").vector
array([-1.07563362e-02, 3.33379984e-01, -3.28269988e-01, -5.88053286e-01,
        3.62993330e-01, 2.02390000e-01, 1.34178683e-01, -2.58746654e-01,
       -1.32898003e-01, 3.90106648e-01, -1.44556671e-01, 4.27457958e-01,
       -1.66819990e-01, -1.20200336e-01, -1.77853659e-01, -5.72146773e-02,
       2.89449006e-01, 1.45733312e-01, 9.77416709e-02, -3.44655663e-01,
       -2.71650016e-01, 3.74561340e-01, 2.62319326e-01, 2.68040001e-02,
       -3.74459922e-02, -1.40609995e-01, -3.32466692e-01, -3.50136645e-02,
       1.43150330e-01, -1.68436036e-01, -1.11395337e-01, -4.99066599e-02,
       -6.44636676e-02, 2.66300350e-01, -3.86333466e-03, -8.14373270e-02,
       3.79196644e-01, -1.44066676e-01, -3.57766636e-02, 2.55699337e-01,
        3.49776983e-01, -7.86300004e-02, 1.81003332e-01, -3.06970000e-01,
       -9.42760035e-02, 3.29153299e-01, -1.45003334e-01, -7.31186643e-02,
       -1.73419669e-01, 7.76770040e-02, 6.54873326e-02, -5.40900230e-03,
       1.55579999e-01, -9.49332118e-03, -4.53666635e-02, 1.59826681e-01,
       7.14799985e-02, 3.65343317e-02, -2.74064630e-01, -9.20736715e-02,
       -1.52166588e-02, 3.49733353e-01, 9.33466628e-02, -8.52633193e-02,
       -6.92400038e-02, -9.46443379e-02, 8.48719850e-02, 6.66899979e-02,
       -3.33857328e-01, -1.24433441e-02, -4.43583280e-01, -1.17006667e-01,
       -3.37433331e-02, 1.04824997e-01, -2.76716679e-01, 3.26154679e-01,
       3.21750015e-01, -3.46729994e-01, 1.04659997e-01, 2.76700165e-02,
        4.07203324e-02, -1.29903331e-02, -5.44013321e-01, -4.73100059e-02,
```

#### Mini Presentation

- \* Think of which topic really interests you
  - \* Does not have to be on the schedule, but should be related to the seminar
  - \* Possible topics
    - \* dive into textblob or polyglot
    - \* what is available for continuous integration
    - \* noSQL databases
- \* Prepare to talk about it for 5-10 minutes
- \* Contact me

### Schedule

Date	Торіс	Notes
Feb 5	Introduction	
Feb 12	Software Engineering 101	Some pre-class preparation, no assignment
Feb 19	NLP tools	Some pre-class preparation (installing tools), spaCy assignment
Feb 26	spaCy	Some pre-class preparation (reading <a href="https://spacy.io/usage/spacy-101">https://spacy.io/usage/spacy-101</a> ), assignment due
Mar 12	Web services	Light reading on web services, Flask assignment
Mar 5	Databases	
Mar 19	Packaging and distributing code	Flask assignment due, PyPI assignment
Mar 26	Docker containers and DockerHub	PyPI assignment due, Docker assignment
Apr 2	-	No class (Good Friday)
Apr 9	Machine learning packages & techniques	Some pre-class preparation (installing and testing tools), ML assignment
Apr 16	Testing and continuous integration	
Apr 23	Hadoop and MapReduce, ML assignment due	
Apr 30	Wrap up, reviewing	