Processing pipelines

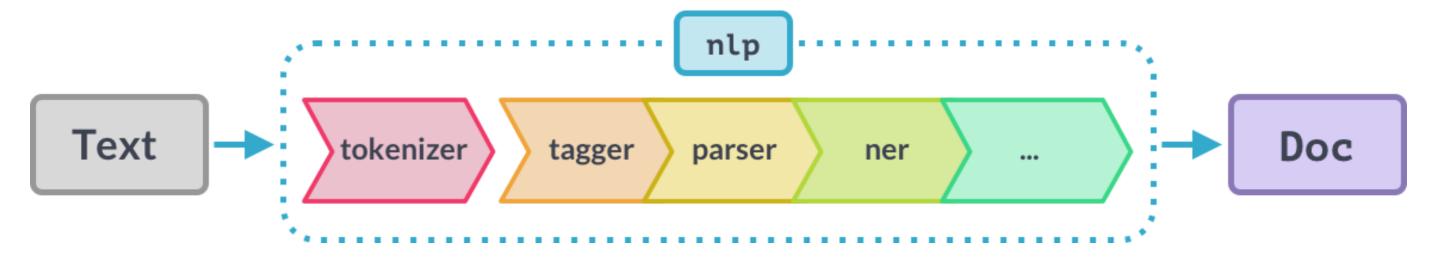
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What happens when you call nlp?

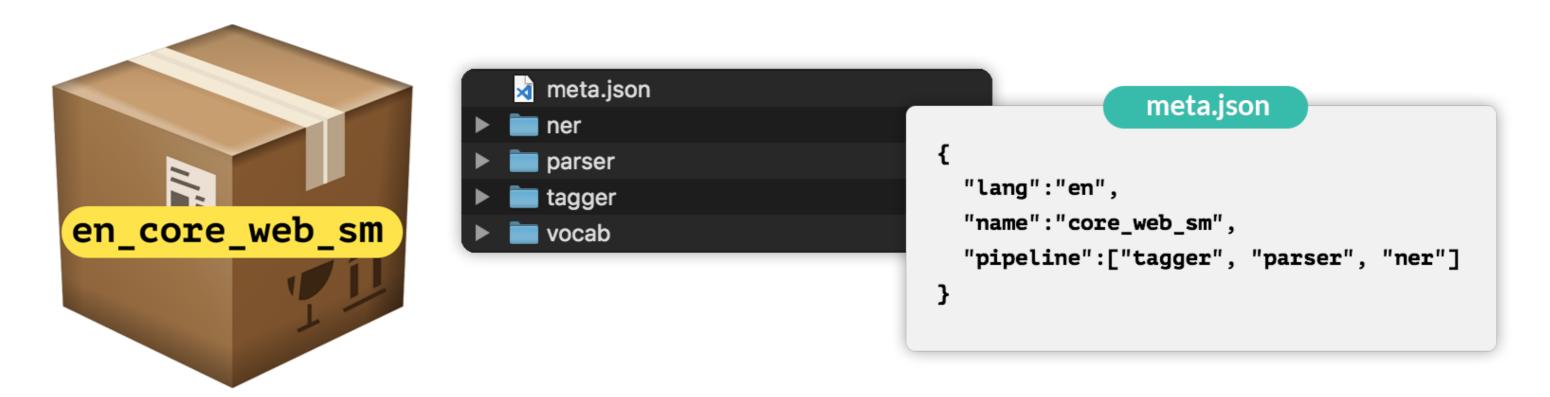


```
doc = nlp("This is a sentence.")
```

Built-in pipeline components

Name	Description	Creates
tagger	Part-of-speech tagger	Token.tag
parser	Dependency parser	Token.dep , Token.head , Doc.sents , Doc.noun_chunks
ner	Named entity recognizer	Doc.ents, Token.ent_iob, Token.ent_type
textcat	Text classifier	Doc.cats

Under the hood



- Pipeline defined in model's meta.json in order
- Built-in components need binary data to make predictions

Pipeline attributes

nlp.pipe_names : list of pipeline component names

```
print(nlp.pipe_names)
['tagger', 'parser', 'ner']
 nlp.pipeline: list of (name, component) tuples
print(nlp.pipeline)
[('tagger', <spacy.pipeline.Tagger>),
 ('parser', <spacy.pipeline.DependencyParser>),
 ('ner', <spacy.pipeline.EntityRecognizer>)]
```

Let's practice!

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Custom pipeline components

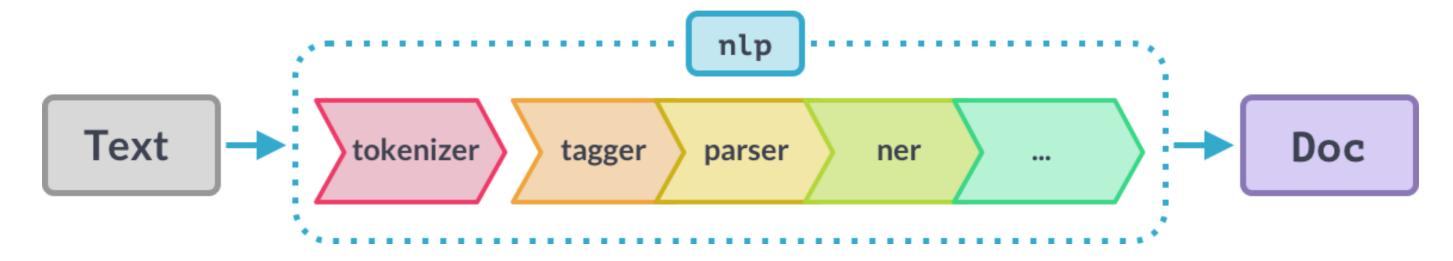
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Why custom components?



- Make a function execute automatically when you call nlp
- Add your own metadata to documents and tokens
- Updating built-in attributes like doc.ents

Anatomy of a component (1)

- Function that takes a doc, modifies it and returns it
- Can be added using the nlp.add_pipe method

```
def custom_component(doc):
    # Do something to the doc here
    return doc
nlp.add_pipe(custom_component)
```

Anatomy of a component (2)

```
def custom_component(doc):
    # Do something to the doc here
    return doc

nlp.add_pipe(custom_component)
```

Argument	Description	Example
last	If True, add last	<pre>nlp.add_pipe(component, last=True)</pre>
first	If True, add first	<pre>nlp.add_pipe(component, first=True)</pre>
before	Add before component	<pre>nlp.add_pipe(component, before='ner')</pre>
after	Add after component	<pre>nlp.add_pipe(component, after='tagger')</pre>



Example: a simple component (1)

```
# Create the nlp object
nlp = spacy.load('en_core_web_sm')
# Define a custom component
def custom_component(doc):
   # Print the doc's length
    print('Doc length:' len(doc))
   # Return the doc object
    return doc
# Add the component first in the pipeline
nlp.add_pipe(custom_component, first=True)
# Print the pipeline component names
print('Pipeline:', nlp.pipe_names)
```

```
Pipeline: ['custom_component', 'tagger', 'parser', 'ner']
```

Example: a simple component (2)

```
# Create the nlp object
nlp = spacy.load('en_core_web_sm')
# Define a custom component
def custom_component(doc):
    # Print the doc's length
    print('Doc length:' len(doc))
    # Return the doc object
    return doc
# Add the component first in the pipeline
nlp.add_pipe(custom_component, first=True)
# Process a text
doc = nlp("Hello world!")
```

```
Doc length: 3
```



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Extension attributes

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Setting custom attributes

- Add custom metadata to documents, tokens and spans
- Accessible via the ._ property

```
doc._.title = 'My document'
token._.is_color = True
span._.has_color = False
```

registered on the global Doc, Token or Span using the set_extension method

```
# Import global classes
from spacy.tokens import Doc, Token, Span
# Set extensions on the Doc, Token and Span
Doc.set_extension('title', default=None)
Token.set_extension('is_color', default=False)
Span.set_extension('has_color', default=False)
```

Extension attribute types

- 1. Attribute extensions
- 2. Property extensions
- 3. Method extensions

Attribute extensions

Set a default value that can be overwritten

```
from spacy.tokens import Token

# Set extension on the Token with default value
Token.set_extension('is_color', default=False)
doc = nlp("The sky is blue.")

# Overwrite extension attribute value
doc[3]._.is_color = True
```

Property extensions (1)

- Define a getter and an optional setter function
- Getter only called when you retrieve the attribute value

```
# Define getter function

def get_is_color(token):
    colors = ['red', 'yellow', 'blue']
    return token.text in colors

# Set extension on the Token with getter

Token.set_extension('is_color', getter=get_is_color)

doc = nlp("The sky is blue.")

print(doc[3]._.is_color, '-', doc[3].text)
```

```
blue - True
```

Property extensions (2)

Span extensions should almost always use a getter

```
from spacy.tokens import Span

# Define getter function
def get_has_color(span):
    colors = ['red', 'yellow', 'blue']
    return any(token.text in colors for token in span)

# Set extension on the Span with getter
Span.set_extension('has_color', getter=get_has_color)
doc = nlp("The sky is blue.")
print(doc[1:4]._.has_color, '-', doc[1:4].text)
print(doc[0:2]._.has_color, '-', doc[0:2].text)
```

```
True – sky is blue
False – The sky
```

Method extensions

- Assign a function that becomes available as an object method
- Lets you pass arguments to the extension function

```
# Define method with arguments
def has_token(doc, token_text):
    in_doc = token_text in [token.text for token in doc]
# Set extension on the Doc with method
Doc.set_extension('has_token', method=has_token)
doc = nlp("The sky is blue.")
print(doc._.has_token('blue'), '- blue')
print(doc._.has_token('cloud'), '- cloud')
```

```
True - blue
False - cloud
```

Let's practice!

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Scaling and performance

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Processing large volumes of text

- Use nlp.pipe method
- Processes texts as a stream, yields Doc objects
- Much faster than calling nlp on each text

BAD:

```
docs = [nlp(text) for text in LOTS_OF_TEXTS]
```

GOOD:

```
docs = list(nlp.pipe(LOTS_OF_TEXTS))
```

Passing in context (1)

- Setting as_tuples=True on nlp.pipe lets you pass in (text, context) tuples
- Yields (doc, context) tuples
- Useful for associating metadata with the doc

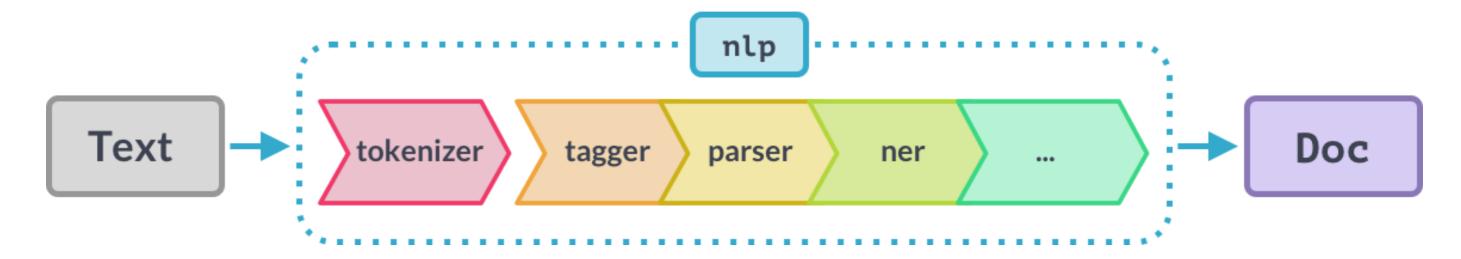
```
data = [
    ('This is a text', {'id': 1, 'page_number': 15}),
    ('And another text', {'id': 2, 'page_number': 16}),
]
for doc, context in nlp.pipe(data, as_tuples=True):
    print(doc.text, context['page_number'])
```

```
This is a text 15
And another text 16
```

Passing in context (2)

```
from spacy.tokens import Doc
Doc.set_extension('id', default=None)
Doc.set_extension('page_number', default=None)
data = [
    ('This is a text', {'id': 1, 'page_number': 15}),
    ('And another text', {'id': 2, 'page_number': 16}),
for doc, context in nlp.pipe(data, as_tuples=True):
    doc._.id = context['id']
    doc._.page_number = context['page_number']
```

Using only the tokenizer



don't run the whole pipeline!

Using only the tokenizer (2)

• Use nlp.make_doc to turn a text in to a Doc object

BAD:

```
doc = nlp("Hello world")
```

GOOD:

```
doc = nlp.make_doc("Hello world!")
```

Disabling pipeline components

• Use nlp.disable_pipes to temporarily disable one or more pipes

```
# Disable tagger and parser
with nlp.disable_pipes('tagger', 'parser'):
    # Process the text and print the entities
    doc = nlp(text)
    print(doc.ents)
```

- restores them after the with block
- only runs the remaining components

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