# biomarker\_nlp

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**CHAPTER** 

ONE

# **BIOMARKER NLP**

# 1.1 biomarker extraction module

biomarker\_extraction.disease\_content(dailyMedURL, disease, header=False)

Extract subsection for a particular disease from a drug's DailyMed 'INDICATIONS AND USAGE' section.

Parse the url link using lxml library. Locate the subsection that disucsses the disease in 'INDICATIONS AND USAGE' section from the html's tree structure. If header is set to False, only the text content will be extracted. If header is set to True, the whole subsection including the subheading and its text content will be extracted.

### **Parameters**

- **dailyMedURL** (*str*) An URL link to a drug's DailyMed information page and quoted ("") as a string.
- disease (str) The name of a disease whose text information will be extract from 'INDI-CATIONS AND USAGE' section.
- **header** (*bool*, *optional*) Extract the subheading, by default False. If True, the subheading will be extracted. If False, only the text content will be extracted.

**Returns** Return the subsection text with or without the subheading. If no associated subsection of the disease is found, return None.

Return type None or str

### **Examples**

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

Example (without subheading)

Example (with subheading)

```
>>> url = "https://dailymed.nlm.nih.gov/dailymed/drugInfo.cfm?setid=939b5d1f-9fb2-
_4499-80ef-0607aa6b114e"
>>> disease = "Cervical Cancer"
>>> biomarker_extraction.disease_content(dailyMedURL = url, disease = disease,__
_header = True)
'1.5 Persistent, Recurrent, or Metastatic Cervical Cancer\nAvastin, in__
_combination with paclitaxel and cisplatin or paclitaxel and topotecan, is__
_indicated for the treatment of patients with persistent, recurrent, or metastatic__
_cervical cancer.'
```

### biomarker\_extraction.drug\_brand\_label(dailyMedURL)

Extract drug label at the drug dailyMed label information page.

Parse the url link using lxml library. Locate and extract the drug label from the html's tree structure.

**Parameters dailyMedURL** (str) – An URL link to a drug DailyMed information page and quoted ("") as a string.

Returns Return drug label. If no drug label is found, return a zero length string ("").

Return type str

### **Examples**

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

### Example

```
>>> url = "https://dailymed.nlm.nih.gov/dailymed/drugInfo.cfm?setid=43a4d7f8-48ae-

-4a63-9108-2fa8e3ea9d9c&audience=consumer"

>>> biomarker_extraction.drug_brand_label(dailyMedURL = url)

'SUTENT- sunitinib malate capsule'
```

### See also:

ndc\_code

biomarker\_extraction.gene\_protein\_chemical(text, gene=1, protein=1, chemical=1)

Extract gene, protein, and drug labels from a string.

The function uses three pre-trained NER models from scispacy. Please see: https://allenai.github.io/scispacy/. We use en\_ner\_craft\_md model to recognize genes. Entities labeled with "GGP" in this model are categorized as genes. We use en\_ner\_jnlpba\_md model to recognize proteins. Entities labeled with "PROTEIN" in this model are categorized as proteins. We use en\_ner\_bionlp13cg\_md model to recognize drugs. Entities labeled with "SIMPLE\_CHEMICAL" in this model are categorized as drugs.

### **Parameters**

- text (str) A single string.
- **gene** (*int*, *optional*) Extract genes, by default 1. 0: do not extract genes. 1: extract genes.
- **protein** (*int*, *optional*) Extract proteins, by default 1. 0: do not extract proteins. 1: extract proteins.

• **chemical** (*int*, *optional*) – Extract simple chemicals, by default 1. 0: do not extract simple chemicals. 1: extract simple chemicals.

**Returns** Return a dictionary in which "gene", "proteins", or/and "chemical" are the keys, lists of genes, proteins, or/and chemical are the values.

Return type dic

### **Examples**

Install the necessary packages and pri-trained models

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

Example (Recognize entities)

```
>>> txt = "Patients with EGFR or ALK genomic tumor aberrations should have disease,
→progression on FDA-approved therapy for NSCLC harboring these aberrations prior
→to receiving TECENTRIQ."
>>> biomarker_extraction.gene_protein_chemical(text = txt, gene= 1, protein = 1,
\rightarrowchemical = 1)
{'gene': ['EGFR', 'ALK genomic'], 'protein': ['EGFR', 'TECENTRIQ'], 'chemical': []}
>>> genProChe = biomarker_extraction.gene_protein_chemical(text = txt, gene= 1,_
\rightarrowprotein = 1, chemical = 1)
# get genes
>>> genProChe.get("gene")
['EGFR', 'ALK genomic']
# get proteins
>>> genProChe.get("protein")
['EGFR', 'TECENTRIQ']
# Only detect genes
>>> biomarker_extraction.gene_protein_chemical(text = txt, gene= 1, protein = 0, _
\rightarrowchemical = \emptyset)
{'gene': ['EGFR', 'ALK genomic']}
```

### biomarker\_extraction.is\_accelerated\_approval(text)

Detect if the drug is accelerated approval.

Use keyword matching to detect if the keyword "accelerated approval" appears in the sentence.

**Parameters** text (str) – A string.

**Returns** Return True if "accelerated approval" is mentioned in the string, False otherwise.

Return type bool

### **Examples**

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

### Example

### biomarker\_extraction.is\_accelerated\_approval\_rate(text)

Detect if the drug is accelerated approval based on response rate.

Use keyword matching to detect if the keywords "accelerated approval" and "response rate" appear in the sentence.

**Parameters text** (str) – A string.

**Returns** Return True if "accelerated approval" and "response rate" are mentioned in the string, False otherwise.

Return type bool

### **Examples**

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

### Example

### biomarker\_extraction.is\_firstline(text, medicine, disease)

Detect if first-line treatment is mentioned with a medicine in a sentence.

Use keyword matching to detect if the keywords "first-line treatment" or "first-or second-line treatment", medicine name, and disease name all appear in the sentence.

### **Parameters**

- **text** (*str*) A single sentence.
- **medicine** (*str*) A medicine's name.

**Returns** Return True if the medicine and first-line treatment are mentioned in the sentence, False otherwise.

Return type bool

### **Examples**

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

### Example

### biomarker\_extraction.is\_metastatic(text, disease)

Detect if the metastatic disease is mentioned.

Use keyword matching to detect if the combination of keywords "metastatic" or "unresectable" with disease's name appears in the sentence.

### **Parameters**

- text (str) A string or sentence
- **disease** (*str*) A disease's name

**Returns** Return True if "metastatic" or "unresectable" and the disease are mentioned together, False otherwise.

Return type bool

### **Examples**

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

### Example

```
>>> txt = "TECENTRIQ, in combination with bevacizumab, is indicated for the treatment of patients with unresectable or metastatic hepatocellular carcinoma (HCC) who have not received prior systemic therapy."
>>> disease = "hepatocellular carcinoma"
>>> biomarker_extraction.is_metastatic(text = txt, disease = disease)
True
```

### biomarker\_extraction.ndc\_code(dailyMedURL)

Extract NDC code(s) from the drug dailyMed label information page.

Parse the url link using lxml library. Locate and extract the NDC codes from the html's tree structure. It returns all the codes found in a string. The codes are seperated by comma.

**Parameters dailyMedURL** (str) – An URL link to a drug DailyMed information page and quoted ("") as a string.

**Returns** Return NDC codes(s) in a string. If more than one codes are found, seperate by comma.

### Return type str

### **Examples**

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

### Example

```
>>> url = "https://dailymed.nlm.nih.gov/dailymed/drugInfo.cfm?setid=43a4d7f8-48ae-

$\times 4a63-9108-2fa8e3ea9d9c&audience=consumer"
>>> biomarker_extraction.ndc_code(dailyMedURL = url)

'0069-0550-38, 0069-0770-38, 0069-0830-38, 0069-0980-38'
```

### biomarker\_extraction.section\_content(dailyMedURL, section)

Extract a whole section text content from the drug's DailyMed information page excluding the section heading.

Parse the url link using lxml library. Locate the section using the section's heading from the html's tree structure. Extract the text content of the section excluding the heading.

### **Parameters**

- **dailyMedURL** (*str*) An URL link to a drug DailyMed information page and quoted ("") as a string.
- **section** (str) The header of the section.

**Returns** Return section content. If no such a section is found, return None.

**Return type** None or str

### **Examples**

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

### Example

```
>>> url = "https://dailymed.nlm.nih.gov/dailymed/drugInfo.cfm?setid=43a4d7f8-48ae-
→4a63-9108-2fa8e3ea9d9c&audience=consumer"
>>> sectionHeader = "INDICATIONS AND USAGE"
>>> biomarker_extraction.section_content(dailyMedURL = url, section = sectionHeader)
'1.1 Gastrointestinal Stromal Tumor
SUTENT is indicated for the treatment of adult patients with gastrointestinal.
→stromal tumor (GIST) after disease progression on or intolerance to imatinib
→mesvlate.
     Advanced Renal Cell Carcinoma
1.2
SUTENT is indicated for the treatment of adult patients with advanced renal cell.
→carcinoma (RCC).
1.3
      Adjuvant Treatment of Renal Cell Carcinoma
SUTENT is indicated for the adjuvant treatment of adult patients at high risk of.
→recurrent RCC following nephrectomy.
      Advanced Pancreatic Neuroendocrine Tumors
SUTENT is indicated for the treatment of progressive, well-differentiated_
→ pancreatic neuroendocrine tumors (pNET) in adult patients with unres(eontablesen next page)
→locally advanced or metastatic disease.'
```

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### See also:

disease\_content

### biomarker\_extraction.sent\_subtree(text)

Extract the subtree of the patterns 'in conbination with' and 'used with' based on dependency parsing.

The function uses pattern match to recognize two patterns ('in conbination with' and 'used with') from a sentence. Once such a pattern is recognized, the sentence is parsed as dependency tree by scispacy's nlp\_bionlp13cg mdoel which is based on Stanford Dependency Converter. The "combination" or "used" is used as headword to extract its subtree.

**Parameters** text(str) – A single sentence.

**Returns** Return a list of subtree clauses in string.

Return type list

### **Examples**

Install the necessary packages and pri-trained models

Import the module

```
>>> from biomarker_nlp import biomarker_extraction
```

Example (extract the subtree)

### Another example

# 1.2 negation\_cue\_scope module

negation\_cue\_scope.negation\_detect(text, modelCue)

Dectect if a sentence contains any negation cues.

This function predicts if a sentence contains any negation words by using a pre-trianed negation dectection model that was pre-trained through Aditya and Suraj's (2020) NegBERT transfer learning program. Please see the reference link: https://github.com/adityak6798/Transformers-For-Negation-and-Speculation. The model was trained using 'bioscope\_abstracts' and 'bioscope\_full\_papers' corpora. It was upload to a cloud repository and was freely-available.

### **Parameters**

- text (str) a single sentence.
- modelCue (torch model) pre-trained negation cue detection model

**Returns** True if any negation cues are detected. False if no negation cues are detected.

Return type bool

### See also:

negation\_scope

### **Notes**

NVIDIA GPU will be required. Make sure your machine brings NVIDIA GPU or set GPU as Hardware accelerator if using Colab notebook.

### **Examples**

Install the necessary packages, and make sure turn GPU Hardware accelerator on.

```
>>> If using Colab Notebook, add ! before pip.
$ pip install biomarker_nlp
$ pip install transformers
$ pip install knockknock==0.1.7
$ pip install sentencepiece
```

Load the necessary packages and pre-trained model

### Examples (predict negation)

```
>>> txt = "TECENTRIQ is not indicated for use in combination with paclitaxel for the treatment of adult patients with unresectable locally advanced or metastatic TNBC."
>>> negation_cue_scope.negation_detect(text = txt, modelCue = modelCue)
True
>>> txt = "KEYTRUDA is not recommended for treatment of patients with PMBCL who require urgent cytoreductive therapy."
```

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```
>>> negation_cue_scope.negation_detect(text = txt, modelCue = modelCue)

True
>>> txt = "KEYTRUDA is indicated for the treatment of adult patients with relapsed_
or refractory classical Hodgkin lymphoma (cHL)."
>>> negation_cue_scope.negation_detect(text = txt, modelCue = modelCue)

False
```

negation\_cue\_scope.negation\_scope(text, modelCue, modelScope)

Extract the scope of negation in a sentence.

This function predicts the negation' cues and their scope in a sentence by using two pre-trianed negation models that were pre-trained through Aditya and Suraj's (2020) NegBERT transfer learning program. Please see the reference link: https://github.com/adityak6798/Transformers-For-Negation-and-Speculation. The models were trained using 'bioscope\_abstracts' and 'bioscope\_full\_papers' corpora. One of them is used to performed negation cue detection and negation scope resolution. Another one is used to performed negation scope resolution. They were upload to a cloud repository and was freely-available. This function predicts negation cues first through the negation cue detection model. If any negation cues were predected, it predectes the scope of each negation cue, if there are more than one, through the negation scope resolution model.

### **Parameters**

- text (str) a single sentence.
- modelCue (torch model) pre-trained negation cue prediction model
- modelScopre (torch model) pre-trianed negation scope prediction model

**Returns** a list of negated clauses. Some sentences would contain more than one negation cues, in this case, all of their negated clauses will be extracted. If no negation is found, return an empty list.

Return type list

### See also:

negation\_detect

### **Notes**

The negation cue will not be extracted. NVIDIA GPU will be required. Make sure your machine brings NVIDIA GPU or set GPU as Hardware accelerator if using Colab notebook.

### **Examples**

Install the necessary packages, and make sure turn GPU Hardware accelerator on.

```
>>> If using Colab Notebook, add ! before pip.
$ pip install biomarker_nlp
$ pip install transformers
$ pip install knockknock==0.1.7
$ pip install sentencepiece
```

Load the necessary packages and pre-trained model

### Examples (predict negation scope)

```
>>> txt = "TECENTRIQ is not indicated for use in combination with paclitaxel for the treatment of adult patients with unresectable locally advanced or metastatic through the treatment of adult patients with unresectable locally advanced or metastatic through the treatment of adult patients with unresectable locally advanced or metastatic tributes."

>>> negation_cue_scope.negation_scope(text = txt, modelCue = modelCue, modelScope = txt = "KEYTRUDA is not recommended for treatment of patients with PMBCL who trequire urgent cytoreductive therapy."

>>> negation_cue_scope.negation_scope(text = txt, modelCue = modelCue, modelScope = modelScope)

['KEYTRUDA is', 'recommended for treatment of patients with PMBCL who']
```

# 1.3 negation\_negbert module

This is part of Aditya Khandelwal & Suraj Sawant's (2020) NegBERT program, it is necessary to run the negation cue detection (modelCue) and negation scope resolution (modelScope) models. For more information about NegBERT program, please see: https://github.com/adityak6798/Transformers-For-Negation-and-Speculation.

# **CHAPTER**

# TWO

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