

Textklassifikation in Python

Einführung

Im vorigen Kapitel hatten wir die Formel zur Berechnung der Wahrscheinlichkeit hergeleitet, dass ein Document d in zu einer Klasse bzw. Kategorie c gehört, bezeichnet als $P(c|d)$.

Wir haben die Standardformel für $P(c|d)$, so wie sie in verschiedenen wissenschaftlichen Abhandlungen verwendet wird¹, in eine numerisch stabile Form gewandelt.

Für unsere Implementierung in Python benutzen wir einen auf "Naive Bayes" basierenden Klassifikator. Eine formale Einführung in das Konzept eines Naive-Bayes-Klassifikators befindet sich im vorigen Kapitel "Einführung in die Textklassifikation".

Python ist ideal für die Textklassifikation wegen seiner Stringklasse und deren mächtigen Methoden. Weiterhin werden mit dem Modul `re` leistungsstarke Tools zur Verfügung gestellt, die weit über den Rahmen anderer Programmiersprachen hinausgehen.

Die vorliegende Python-Implementierung eines Naive-Bayes-Klassifikators ist allerdings nicht auf Effizienz optimiert.

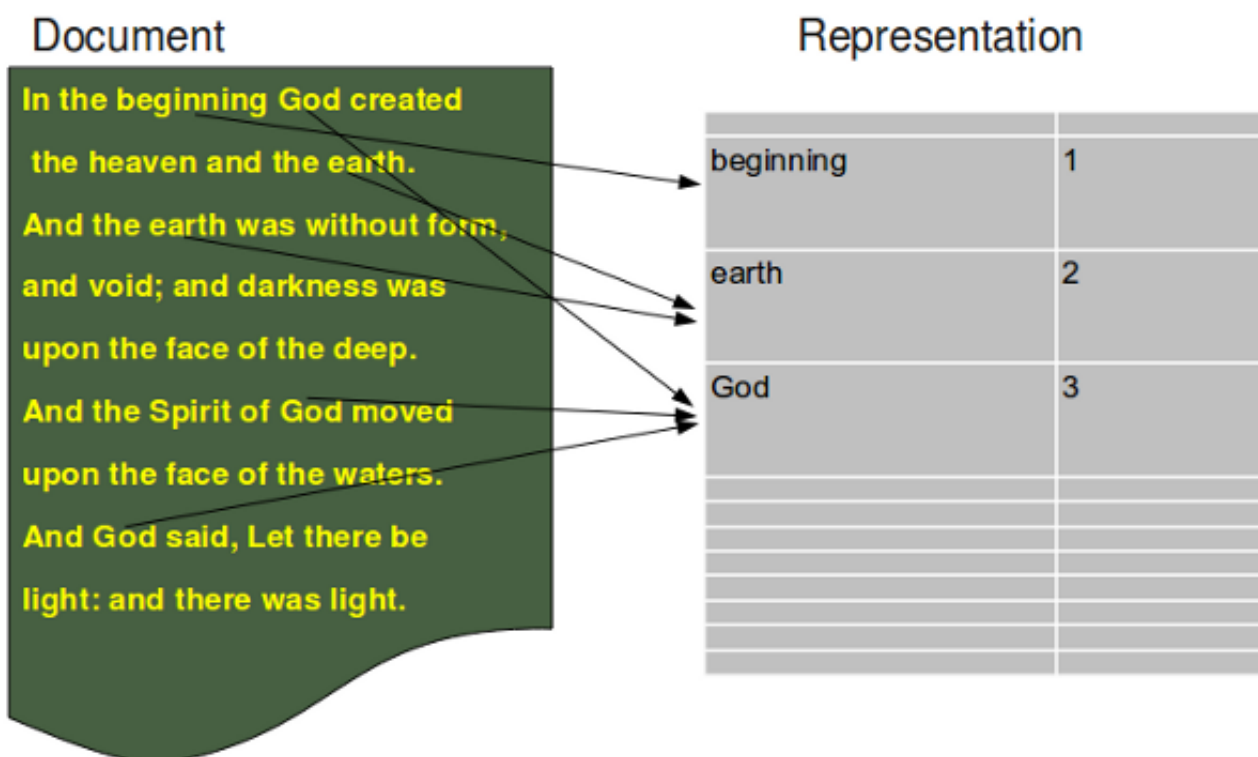


Python-Implementierung

Warnung: Die vorliegende Implementierung ist Python 3 geschrieben und nicht Python 2.x kompatibel!

Dokumenten-Repräsentation

Die Dokumente werden in unserer Implementierung nach dem sogenannten "bag of words"-Modell implementiert, was wir im folgenden Diagramm illustrieren:



Benötigte Module

Unsere Implementierung benötigt das Modul `re` für reguläre Ausdrücke und die Betriebssystemschnittstelle im `os`-Modul:

```
import re, os
```

BagOfWords-Klasse

```
class BagOfWords(object):
    """ Implementing a bag of words, words corresponding with their frequency
    of usages in a "document"
    for usage by the Document class, DocumentClass class and the Pool
    class."""

    def __init__(self):
        self.__number_of_words = 0
        self.__bag_of_words = {}

    def __add__(self, other):
        """ Overloading of the "+" operator to join two BagOfWords """
        erg = BagOfWords()
        sum = erg.__bag_of_words
        for key in self.__bag_of_words:
            sum[key] = self.__bag_of_words[key]
            if key in other.__bag_of_words:
                sum[key] += other.__bag_of_words[key]
        for key in other.__bag_of_words:
            if key not in sum:
                sum[key] = other.__bag_of_words[key]
        return erg

    def add_word(self, word):
        """ A word is added in the dictionary __bag_of_words"""
        self.__number_of_words += 1
        if word in self.__bag_of_words:
            self.__bag_of_words[word] += 1
        else:
            self.__bag_of_words[word] = 1

    def len(self):
        """ Returning the number of different words of an object """
        return len(self.__bag_of_words)

    def Words(self):
        """ Returning a list of the words contained in the object """
        return self.__bag_of_words.keys()

    def BagOfWords(self):
        """ Returning the dictionary, containing the words (keys) with their
        frequency (values)"""
        return self.__bag_of_words
```

```
def WordFreq(self, word):
    """ Returning the frequency of a word """
    if word in self.__bag_of_words:
        return self.__bag_of_words[word]
    else:
        return 0
```

Dokumenten-Klasse

```
class Document(object):
    """ Used both for learning (training) documents and for testing documents.
    The optional parameter learn
    has to be set to True, if a classifier should be trained. If it is a
    test document learn has to be set to False. """
    _vocabulary = BagOfWords()

    def __init__(self, vocabulary):
        self.__name = ""
        self.__document_class = None
        self._words_and_freq = BagOfWords()
        Document._vocabulary = vocabulary

    def read_document(self, filename, learn=False):
        """ A document is read. It is assumed, that the document is either
        encoded in utf-8 or in iso-8859... (latin-1).
        The words of the document are stored in a Bag of Words, i.e.
        self._words_and_freq = BagOfWords() """
        try:
            text = open(filename, "r", encoding='utf-8').read()
        except UnicodeDecodeError:
            text = open(filename, "r", encoding='latin-1').read()
        text = text.lower()
        words = re.split("[^\wäöüÄÖÜß]*", text)

        self._number_of_words = 0
        for word in words:
            self._words_and_freq.add_word(word)
            if learn:
                Document._vocabulary.add_word(word)

    def __add__(self, other):
        """ Overloading the "+" operator. Adding two documents consists in
        adding the BagOfWords of the Documents """
        res = Document(Document._vocabulary)
        res._words_and_freq = self._words_and_freq + other._words_and_freq
        return res

    def vocabulary_length(self):
        """ Returning the length of the vocabulary """
        return len(Document._vocabulary)

    def WordsAndFreq(self):
        """ Returning the dictionary, containing the words (keys) with their
        frequency (values) as contained
```

```

        in the BagOfWords attribute of the document"""
        return self._words_and_freq.BagOfWords()

    def Words(self):
        """ Returning the words of the Document object """
        d = self._words_and_freq.BagOfWords()
        return d.keys()

    def WordFreq(self, word):
        """ Returning the number of times the word "word" appeared in the
        document """
        bow = self._words_and_freq.BagOfWords()
        if word in bow:
            return bow[word]
        else:
            return 0

    def __and__(self, other):
        """ Intersection of two documents. A list of words occuring in both
        documents is returned """
        intersection = []
        words1 = self.Words()
        for word in other.Words():
            if word in words1:
                intersection += [word]
        return intersection

```

Die Klasse DocumentClass

Die Klasse DocumentClass ist die Klasse für unser Dokumentenklassen. Sie erbt von der Klasse Document.

```

class DocumentClass(Document):
    def __init__(self, vocabulary):
        Document.__init__(self, vocabulary)
        self._number_of_docs = 0

    def Probability(self, word):
        """ returns the probability of the word "word" given the class "self"
        """
        voc_len = Document._vocabulary.len()
        SumN = 0
        for i in range(voc_len):
            SumN = DocumentClass._vocabulary.WordFreq(word)
        N = self._words_and_freq.WordFreq(word)
        erg = 1 + N
        erg /= voc_len + SumN
        return erg

    def __add__(self, other):
        """ Overloading the "+" operator. Adding two DocumentClass objects
        consists in adding the
        BagOfWords of the DocumentClass objectss """
        res = DocumentClass(self._vocabulary)
        res._words_and_freq = self._words_and_freq + other._words_and_freq

```

```

        return res

    def SetNumberOfDocs(self, number):
        self._number_of_docs = number

    def NumberOfDocuments(self):
        return self._number_of_docs

```

Die Pool-Klasse

Die Klasse pool ist die Klasse, in der die Dokumentenklassen gelernt werden, und die die Klassifikationsmethoden zur Verfügung stellt:

```

class Pool(object):
    def __init__(self):
        self.__document_classes = {}
        self.__vocabulary = BagOfWords()

    def sum_words_in_class(self, dclass):
        """ The number of times all different words of a dclass appear in a
class """
        sum = 0
        for word in self.__vocabulary.Words():
            WaF = self.__document_classes[dclass].WordsAndFreq()
            if word in WaF:
                sum += WaF[word]
        return sum

    def learn(self, directory, dclass_name):
        """ directory is a path, where the files of the class with the name
dclass_name can be found """
        x = DocumentClass(self.__vocabulary)
        dir = os.listdir(directory)
        for file in dir:
            d = Document(self.__vocabulary)
            print(directory + "/" + file)
            d.read_document(directory + "/" + file, learn = True)
            x = x + d
        self.__document_classes[dclass_name] = x
        x.SetNumberOfDocs(len(dir))

    def Probability(self, doc, dclass = ""):
        """Calculates the probability for a class dclass given a document
doc"""
        if dclass:
            sum_dclass = self.sum_words_in_class(dclass)
            prob = 0

            d = Document(self.__vocabulary)
            d.read_document(doc)

            for j in self.__document_classes:
                sum_j = self.sum_words_in_class(j)
                prod = 1
                for i in d.Words():

```

```

        wf_dclass = 1 +
self.__document_classes[dclass].WordFreq(i)
        wf = 1 + self.__document_classes[j].WordFreq(i)
        r = wf * sum_dclass / (wf_dclass * sum_j)
        prod *= r
        prob += prod * self.__document_classes[j].NumberOfDocuments()
/ self.__document_classes[dclass].NumberOfDocuments()
        if prob != 0:
            return 1 / prob
        else:
            return -1
    else:
        prob_list = []
        for dclass in self.__document_classes:
            prob = self.Probability(doc, dclass)
            prob_list.append([dclass, prob])
        prob_list.sort(key = lambda x: x[1], reverse = True)
        return prob_list

def DocumentIntersectionWithClasses(self, doc_name):
    res = [doc_name]
    for dc in self.__document_classes:
        d = Document(self.__vocabulary)
        d.read_document(doc_name, learn=False)
        o = self.__document_classes[dc] & d
        intersection_ratio = len(o) / len(d.Words())
        res += (dc, intersection_ratio)
    return res

```

Benutzung des Klassifikators

Damit Sie einen Klassifikator lernen und testen können, stellen wir einen ein "**Lern und Testset zum Download**" zur Verfügung. Dabei handelt es sich um sechs Klassen mit englischen Witzen der Rubriken "clinton", "lawyer", "math", "medical", "music", "sex".

```

from NaiveBayes import Pool
import os

DClasses = ["clinton", "lawyer", "math", "medical", "music", "sex"]

base = "../learn/"
p = Pool()
for i in DClasses:
    p.learn(base + i, i)

base = "../test/"
for i in DClasses:
    dir = os.listdir(base + i)
    for file in dir:
        res = p.Probability(base + i + "/" + file)
        print(i + ": " + file + ": " + str(res))

```