

Natural Language Processing (Automatic Questions Tagging System)

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Contents

1 Introduction

- Project Overview
- Project Goals

2 Data Definition

Dataset

3 Data Preprocessing

- Preprocessing on the Tags Dataset
- Preprocessing on the Questions Dataset

4 Feature Extraction

5 Data Splitting

6 Machine Learning Algorithm (Classification)

- 6.1 LinearSVC
 - 6.1.1 LinearSVC Performance
 - Grid Search
 - Model Accuracy
 - F1 Score
- 6.2 SGDClassifier
 - 6.2.1 SGDClassifier Performance
 - Model Accuracy
 - F1 Score
- 6.3 Logistic Regression

- 6.3.1 Logistic Regression Performance
 - Model Accuracy
 - F1 Score
- 6.4 Decision Tree Classifier
 - 6.4.1 Decision Tree Classifier Performance
 - Model Accuracy
 - F1 Score

7 Visualization

- Total Accuracy of all Models
- 8 Prediction
- 9 Conclusion

Introduction

1.1 Project Overview

The Automatic Question Tags System is a project that aims to assist users in generating question tags for their sentences with the help of Natural Language Processing (NLP) techniques. The system will be designed to analyze the grammatical structure of a given sentence and generate an appropriate question tag that fits the context and meaning of the sentence. The project will involve the use of machine learning algorithms to identify the relevant features in the sentences that can be used to generate the question tags. The system will also be trained on a large dataset of annotated sentences to improve its accuracy and effectiveness.

1.2 Project Goals

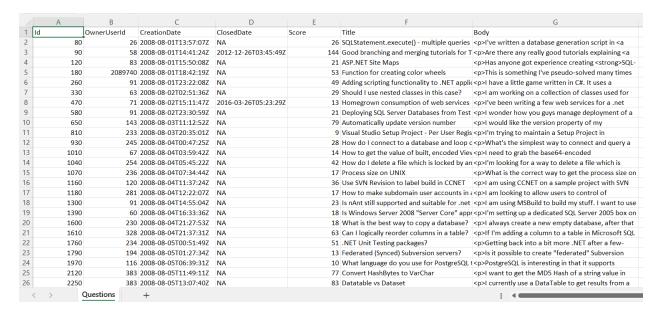
The primary goal of the Automatic Question Tags System project is to develop a tool that can automatically generate question tags for a given sentence. The system will be designed to accurately identify the relevant features in the sentence that can be used to generate the question tag, such as the verb tense, subject, and context. The system will also be trained on a large dataset of annotated sentences to improve its accuracy and effectiveness.

Data Definition

2.1 Dataset

 For our Questions dataset, we got a dataset containing 1264216 records of questions with 6 attributes for each of question, which will be explained in the next section.

1. Dataset Sample:



2. Dataset Info

```
RangeIndex: 1264216 entries,
Data columns (total 7 columns):
     Column
                   Non-Null Count
                                       Dtype
     Id
                                       int64
                    1264216 non-null
     OwnerUserId
                    1249762 non-null
                                       float64
     CreationDate
     ClosedDate
                    1264216 non-null
     Score
     Title
                   1264216 non-null
     Body
dtypes: float64(1), int64(2), object(4)
memory usage: 67.5+ MB
None
```

 For our Tags dataset, we got a dataset containing 3750994 records of tags with one attribute for each tag, which will be explained in the next section.

1. Dataset Sample:

	А	В			
1	Id	Tag			
2	80	flex			
3	80	actionscript-3			
4	80	air			
5	90	svn			
6	90	tortoisesvn			
7	90	branch			
8	90	branching-and-merging			
9	120	sql			
10	120	asp.net			
11	120	sitemap			
12	180	algorithm			
13	180	language-agnostic			
14	180	colors			
15	180	color-space			
16	260	c#			
17	260	.net			
18	260	scripting			
19	260	compiler-construction			
20	330	C++			
21	330	оор			
22	330	class			
23	330	nested-class			
24	470	.net			
25	470	web-services			
26	580	sql-server			
〈 〉 Tags +					

2. Dataset Info

```
RangeIndex: 3750994 entries, 0 to 3750993

Data columns (total 2 columns):

# Column Dtype
--- ---- 0 Id int64

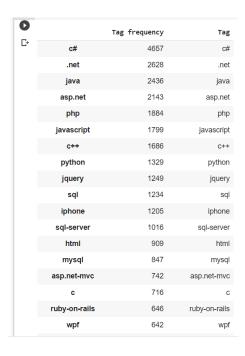
1 Tag object
dtypes: int64(1), object(1)
memory usage: 57.2+ MB

None
```

Data Preprocessing

3.1 Preprocessing of Tags:

the 'tags output' dataframe is converted to a string, and the frequency of each tag is calculated and stored in the 'tags_fre' dataframe. The top 50 most frequent tags are stored in the 'most_tags' dataframe.



The 'tags_output' dataframe is then merged with the 'tags_fre' dataframe, and only those tags with a frequency of at least 271 are retained. The resulting dataframe is stored in 'merge_tags'.

₽		Id	Tag	Tag frequency
	0	80	flex	329
	3	90	svn	291
	7	120	sql	1234
	8	120	asp.net	2143
	10	180	algorithm	300
	99985	1750010	C#	4657
9	99989	1750040	php	1884
9	99990	1750040	mysql	847
	99993	1750070	C#	4657
	99997	1750170	visual-studio	548
40449 rows × 3 columns				

'grouped_tags' is then created by grouping the 'merge_tags' dataframe by 'Id' and concatenating the tags for each 'Id' into a single string. Finally, 'questions_tags' is created by merging the 'questions' dataframe with the 'grouped_tags' dataframe on the 'Id' column.



3.2 Preprocessing of Questions:

The preprocessing of questions (Body column and Title) involves several techniques, including removing HTML tags, tokenization, stop word removal, lemmatization, and cleaning punctuation. We first removed HTML tags from the Body column only using the BeautifulSoup class from the bs4 library. We then tokenized the text using the ToktokTokenizer class from the nltk library. We also removed stop

words, punctuation, and lemmatized the text using the WordNetLemmatizer class from the nltk library, and this is results:

```
sqlstatement.execute( multiple query one state...
0
1
              good branching merge tutorial tortoisesvn
2
                                        asp.net site map
                             function create color wheel
3
              add script functionality .net application
Name: clean title, dtype: object
    write database generation script sql want exec...
    really good tutorial explain branch merge apac...
     anyone get experience create sql-based asp.net...
2
     something pseudo-solved many time never quite ...
3
    little game write c#. us database back-end. tr...
Name: clean body, dtype: object
```

4

Feature Extraction

The feature extraction process in the Automatic Question Tags System involves using the TfidfVectorizer class from the sklearn library to convert the preprocessed text data into numerical vector representations. The vectorizer computes the Term Frequency - Inverse Document Frequency (TF-IDF) scores for each token in the text, which are then used as features. The system also limits the maximum number of features to 1,000 and uses n-grams of up to 3 words to capture the context of the text, The TfidfVectorizer class returns a sparse matrix of

shape (n_samples, n_features), where n_samples is the number of questions and n_features is the number of unique tokens in the text. Each element of the matrix represents the tf-idf weight of a token in a particular question. The resulting matrix is a numerical representation of the text data that can be used as input to a machine learning algorithm.

5

Data Splitting

We use "train test split", The process of dividing the data into these sets is called a train-test split. The train-test split is typically done randomly, with a certain percentage of the data (80%) allocated to the training set and the remaining percentage (20%) allocated to the test set and apply preprocessing to train and test data.

6

Machine Learning Algorithm

Several machine learning algorithms are used to predict the tags for new questions, including SGDClassifier, LogisticRegression, LinearSVC, and DecisionTreeClassifier. The OneVsRestClassifier class from the sklearn library is used to train the multi-label classification model.

6.1 LinearSVC:

```
C=[1,2,3,4,5,6,7,8,9,10]
for i in C:
    svm_model = LinearSVC(C=i)
    clf = OneVsRestClassifier(svm_model)
    clf.fit(x_train, y_train)
    y_pred = clf.predict(x_test)
    print_score(y_pred, clf)
```

6.1.1 LinearSVC Performance

• Grid search: The provided code implements a grid search for the best C parameter in a linear SVM model using the scikit-learn library. The SVM model is trained using the LinearSVC class and the OneVsRestClassifier wrapper for multi-class classification. The C parameter controls the trade-off between maximizing the margin and minimizing the classification error of the SVM model, and is searched over a range of values using a for loop. The performance of the model is evaluated using a custom print_score function that computes and prints various performance metrics for each value of C. The grid search approach provides a systematic and data-driven way to select the best value of C that maximizes the performance of the model on the test set. And this is output:

o Accuracy of LinearSCV: 0.27997809819310093

F1Score of LinearSCV: 0.5156300879192445

6.2 SGDClassifier:

The SGDClassifier is a linear model that trains on a subset of the data at each iteration using stochastic gradient descent. It is well-suited for large-scale and sparse datasets, and can handle a variety of loss functions and penalties. The SGDClassifier is fast and efficient, but may require careful tuning of the learning rate and regularization parameters to achieve good performance.

6.2.1 SGDClassifier Performance

o Accuracy of SGDClassifier: 0.2463953276145282

F1Score of SGDClassifier: 0.45944729498604736

6.3 Logistic Regression:

Logistic Regression is a linear model that estimates the probability of a binary or multi-class outcome using a logistic or softmax function. It is a simple and interpretable model that can handle both continuous and categorical features. Logistic Regression is widely used for binary classification and can be extended to multi-class classification using the one-vs-rest or multinomial approach. Logistic Regression is a well-established model with a strong theoretical foundation and can achieve high performance on a variety of datasets.

6.3.1 Logistic Regression Performance

accuracy of Logistic Regression: 0.22577112611790473

o the F1 Score of Logistic Regression: 0.43906842390323164

6.4 Decision Tree Classifier:

The DecisionTreeClassifier is a tree-based model that constructs a decision tree to recursively partition the data into different classes based on the values of the features. It is a simple and interpretable model that can handle both continuous and categorical features, and can capture non-linear and interaction effects. The DecisionTreeClassifier can be prone to overfitting and may require careful tuning of the hyperparameters, such as the maximum depth, minimum number of samples per leaf, and minimum number of samples per split, to balance between simplicity and complexity and achieve good performance on the test set.

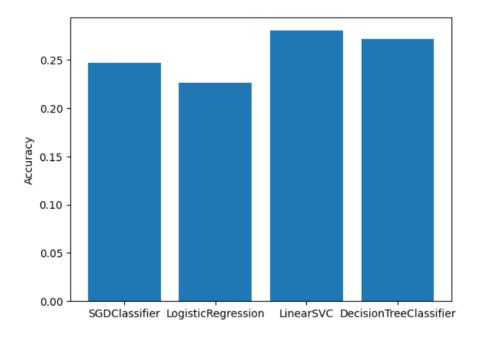
6.4.1 Decision Tree Classifier Performance

- o accuracy of DecisionTreeClassifier: 0.27194743566344226
- the F1 Score of DecisionTreeClassifier: 0.5215875782302147

7

Visualization

Create a bar plot of the accuracy scores for four different classification models evaluated on a given dataset. The plot visualizes the performance of each model and provides a quick and easy way to compare the accuracy scores of the models. The plot can be customized and extended to include additional information and provide more insights into the performance of the models. The visualization can help in selecting the best model for a given task and assessing the performance of the model on the test set.



8

Prediction

This code shows a simple and effective way to predict the tags of a given text input using machine learning and text preprocessing techniques. The code takes a text input 'how to write ml code in python and c#' and preprocesses it by removing stop words and punctuations, and applying lemmatization. The code then converts the preprocessed text into a feature vector using a TF-IDF vectorizer object, and applies a trained classifier object to predict the label of the text input.

```
x = 'how to write ml code in python and c#'
x=[token_remove_stop_words_punc_lemma(x)]
print(x)
xt = tidf_X1.transform(x)
s = clf.predict(xt)
```

This is the output:

```
['write ml code python c#']

m.inverse_transform(s)

[('c#', 'python')]
```

9

Conclusion

The Automatic Question Tags System project is an innovative application of NLP techniques that aims to assist users in generating question tags for their sentences. The system will use machine learning algorithms to identify the relevant features in the sentences and improve its accuracy over time. The project has many potential applications in language learning, content creation, and social media, and has the potential to improve communication and language skills for users across a variety of domains.