AI-based Web Application development and deployment



Estimated Time: 60 minutes

Overview

In this project, we make use of the embedded Watson AI libraries, to create an application that would perform sentiment analysis on a provided text. We then deploy the said application over the web using Flask framework.

Project guidelines

For the completion of this project, you'll have to complete the following 8 tasks, based on the knowledge you have gained through the course.

Tasks and objectives:

- · Task 1: Clone the project repository
- Task 2: Create a sentiment analysis application using Watson NLP library
- Task 3: Format the output of the application
- Task 4: Package the application
- . Task 5: Run Unit tests on your application
- · Task 6: Deploy as web application using Flask
- Task 7: Incorporate Error handling
- · Task 8: Run static code analysis

Let's get started !

About Embeddable Watson AI libraries

In this project, you'll be using embeddable libraries to create an AI powered Python application.

Embeddable Watson Al libraries include the NLP library, the text-to-speech library and the speech-to-text library. These libraries can be embedded and distributed as part of your application. For your convenience, these libraries have been pro-installed on Skills Network Labs Cloud IDE for use in this project.

The NLP library includes functions for sentiment analysis, emotion detection, text classification, language detection, etc. among others. The speech-to-text library contains functions that perform the transcription service and generates written text from spoken audio. The text-to-speech library generates natural sounding audio from written text. All available functions, in each of these libraries, calls pretrained AI models that are all available on the Cloud IDE servers, available to all users for free.

These libraries may also be accessed through your personal systems. The guidelines for the same are available on the Watson AI library page.

Task 1: Clone the project repository

The Github repository of the project is available on the URL mentioned below.

- 1. 1
- $1. \ \texttt{https://github.com/ibm-developer-skills-network/zzrjt-practice-project-emb-ai.git}$

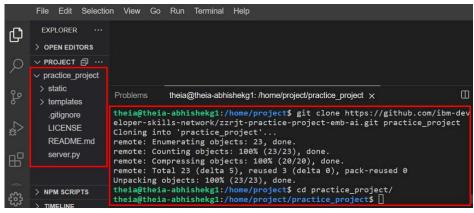
Copied! Executed!

Clone this GitHub repo, using the Cloud IDE terminal to your project to a folder named practice_project. Once the cloning is complete, use the terminal to change the current directory practice_project.

- ► Click here for hint
- ▼ Click here for solution
 - 1. 1
- 1. git clone https://github.com/ibm-developer-skills-network/zzrjt-practice-project-emb-ai.git practice_project
- cd practice_project

Conjed

Upon completion, the project tab should have the folder structure as shown in the image.



Task 2: Create a sentiment analysis application using Watson NLP library

NLP sentiment analysis is the practice of using computers to recognize sentiment or emotion expressed in a text. Through NLP, sentiment analysis categorizes words as positive, negative or neutral.

Sentiment analysis is often performed on textual data to help businesses monitor brand and product sentiment in customer feedback, and understanding customer needs. It helps attain the attitude and mood of the wider public which can then help gather insightful information about the context.

For creating the sentiment analysis application, we'll be making use of the Watson Embedded AI Libraries. Since the functions of these libraries are already deployed on the Cloud IDE server, there is no need of importing these libraries to our code. Instead, we need to send a POST request to the relevant model with the required text and the model will send the appropriate response.

A sample code for such an application could be

```
1. 1
2. 2
3. 3
4. 4
5. 5
6. 6
7. 7
8. 8

1. import requests
2. 3. def <function_name>(<input_args>):
4. url = 'relevant_url>'
5. headers = {kheader_dictionary>}
6. myobj = {kinput_dictionary_to_the_function>}
7. response = requests.post(url, json = myobj, headers=header)
8. return response.text

Copied!
```

Note: The response of the Watson NLP functions is in the form of an object. For accessing the details of the response, we can use text attribute of the object by calling response.text and make the function return the response as simple text.

For this project, you'll be using the BERT based Sentiment Analysis function of the Watson NLP Library. For accessing this funciton, the URL, the headers and the input json format is as follows.

```
1. 1
2. 2
3. 3
1. URL: 'https://sn-watson-sentiment-bert.labs.skills.network/v1/watson.runtime.nlp.v1/NlpService/SentimentPredict'
2. Headers: ("grpc-metadata-mm-model-id": "sentiment_aggregated-bert-workflow_lang_multi_stock")
3. Input json: { "raw_document": { "text": text_to_analyse } }
```

Here, text to analyze is being used as a variable that holds the actual written text which is to be analyzed.

In this task, you need to create a file named sentiment_analysis.py in practice_project folder. In this file, write the function for running sentiment analysis using the Watson NLP BERT Seniment Analysis function, as discussed above. Let us call this function setiment_analyzer. Assume that text to be analysed is passed to the function as an argument and is stored in the variable text_to_analyse.

▼ Click here for the solution sentiment_analysis.py

1. 1 2. 2 3. 3 4. 4 5. 5 6. 6

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```
1. import requests
3. def sentiment_analyzer(text_to_analyse):
      url = https://sn-watson-sentiment-bert.labs.skills.network/v1/watson.runtime.nlp.v1/NlpService/SentimentPredict
      myobj = { "raw_document": { "text": text_to_analyse } }
       header = {"grpc-metadata-mm-model-id": "sentiment_aggregated-bert-workflow_lang_multi_stock"}
      response = requests.post(url, json = myobj, headers=header)
      return response.text
```

Copied!

This application can now be called using Python shell. To test the application, open a Python shell in the terminal with the following command. Make sure that the current directory is practice_project.

- 1. python3.11

Copied! Executed!

In the python shell, import the function sentiment analyzer.

- ► Click here for hint
- ▼ Click here for solution
- 1. from sentiment_analysis import sentiment_analyzer

Copied!

After successful import, test your application with the text "I love this new technology." The result expected is as shown in the image below. To exit the python shell, press Ctrl+Z or type exit().

```
>>> sentiment_analyzer("I love this new technology")
'{"documentSentiment":{"score":0.996954, "label":"SENT_POSITIVE", "mixed":false, "sentimentMentions":[{"span":{"begin":0, "end":26, "text":"I love this new technology"}, "sentimentprob":{"positive":0.9941229, "neutral":0.0028645627, "negative":0.0030124863}}]],
"targetedSentiments":{"targetedSentiments":{}, "producerId":{"name":"Aggregated Sentiment Workflow", "version":"0.0.1"}}, "producer
Id":{"name":"Aggregated Sentiment Workflow", "version":"0.0.1"}}'
```

This completes the Task 2. Note that in the output, the information relevant to us is only the label and the score. In the following task, you will extract this information from this output.

Task 3: Format the output of the application

The output of the application created is in the form of a dictionary, but has been formatted as a text. To access relevant pieces of information from this output, we need to first convert this text into a dictionary. Since dictionaries are the default formatting system for JSON files, we make use of the in-built Python library json.

First, in a Python shell, import the json library.

- 1. 1
- 1. import json

Let's see how this works.

Copied!

Next, run the sentiment analyzer function for the text "I love this new technology", just like in Task 2, and store the output in a variable called response.

- 2. 2
- 1. from sentiment_analysis import sentiment_analyzer
- response = sentiment_analyzer("I love this new technology")

Copied!

Now, pass the response variable as an argument to the json.loads function and save the output in formatted response. Print formatted response to see the difference in the formatting.

- 1. 1 2. 2
- formatted_response = json.loads(response)
- print(formatted_response)

Copied!

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The expected output of the above mentioned steps is shown in the image below

```
theia@theia-abhishekg1:/home/project/practice_project$ python3.11
Python 3.11.2 (main, Feb 8 2023, 14:49:25) [GCC 11.3.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>> import json
>>> from sentiment_analysis import sentiment_analyzer
>>> response = sentiment_analyzer("I love this new technology")
>>> formatted_response = json.loads(response)
>>> formatted_response = json.loads(response)
('documentSentiment': { 'score': 0.996954, 'label': 'SENT_POSITIVE', 'mixed': False, 'sentimentMentions': [{ 'span': { 'begin': 0, 'en d': 26, 'text': 'I love this new technology'}, 'sentimentprob': { 'positive': 0.9941229, 'neutral': 0.0028645627, 'negative': 0.0938
124863}}]), 'targetedSentiments': { 'targetedSentiments': {}, 'producerId': { 'name': 'Aggregated Sentiment Workflow', 'version': '0.0.1'}}
>>> []
```

Note that the absence of single quotes on either side on the response indicates that this is no longer a text, but is a dictionary instead. To access the correct information from this dictionary, we need to access the keys appropriately. Since this is a nested dictionary structure, i.e. a dictionary of dictionaries, the following statements need to be used to get the label and the score outputs from this response.

```
1. 1
2. 2
1. label = formatted_response['documentSentiment']['label']
2. score = formatted_response['documentSentiment']['score']
Copied[
```

Check the contents of label and score to verify the output.

```
>>> label= formatted_response['documentSentiment']['label']
>>> score= formatted_response['documentSentiment']['score']
>>> label
'SENT_POSITIVE'
>>> score
0.996954
```

Now, for Task 3, incorporate the above mentioned technique and make changes to the sentiment_analysis.py file. The expected output from calling the seniment_analyzer function should now be a dictionary with 2 keys, label and score, each having the appropriate value extracted from the response of the Watson NLP function. Verify your changes by testing the modified function in a python shell.

▼ Click here for the solution

```
1. 1
2. 2
3. 3
3
4. 4
5. 5
6. 6
6. 7
7. 7
8. 8
8
9. 9
10. 10
11. 11
12. 12
1. import requests
2. import json
3.
4. def sentiment_analyzer(text_to_analyse):
5. url = 'https://sn.watson-sentiment-bert.labs.skills.network/v1/watson.runtime.nlp.v1/NlpService/SentimentPredict'
6. myobj = { "ram_document": { "text": text to_analyse } }
7. header = { "grpc-metadata-mm-model-id": "sentiment_aggregated-bert-workflow_lang_multi_stock"}
8. response = requests.post(url, json = myobj, headers=header)
9. formatted_response[ olds(response.text)
10. label = formatted_response[ documentSentiment][ label']
11. score = formatted_response[ documentSentiment'][ 'score']
12. return { 'label': label, 'score': score}
```

Copied!

At completion, the expected output of the function is shown in the image below.

```
theia@theia-abhishekg1:/home/project/practice_project$ python3.11

Python 3.11.2 (main, Feb 8 2023, 14:49:25) [GCC 11.3.0] on linux

Type "help", "copyright", "credits" or "license" for more information.

>>> from sentiment_analysis import sentiment_analyzer

>>> sentiment_analyzer("I love this new technology")

{'label': 'SENT_POSITIVE', 'score': 0.996954}
```

To exit the python shell, type exit() or press Ctr1+Z.

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Task 4: Package the application

In this task, you have to package the final application you created in tasks 2 and 3.

Let's keep the name of the package as SentimentAnalysis. The steps involved in packaging are:

- 1. Create a folder in the working directory, with the name as the package name.
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You may use a terminal command or the Cloud IDE console to create the required folder

▼ Click here for solution

- 1. 1
- 1. mkdir SentimentAnalysis

Copied!

- 2. Put (or move) the application code, also called module, in the package folder.
- ▼ Click here for hint

You may use a terminal command or the Cloud IDE console to move the 'sentiment analysis.py' file to the 'SentimentAnalysis' folder

▼ Click here for solution

1. 1

1. mv ./sentiment_analysis.py ./SentimentAnalysis

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- 3. Create the __init__.py file, referencing the module.
- ▼ Click here for hint

Import the module/function from the current folder in the init file

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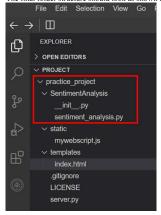
Insert this following line to __init__.py

1. 1

from . import sentiment_analysis

Copied!

The final folder structure should look as shown in the image below.



SentimentAnalysis is now a valid package and can be imported into any file in this project.

To test this, run a python shell in the terminal and try importing the sentiment_analyzer function from the package.

▼ Click here for the hint

The syntax for this import is

1. 1

1. from package_name.module_name import function_name

Copied!

- ▼ Click here for the solution
- 1. 1

1. from SentimentAnalysis.sentiment_analysis import sentiment_analyzer

```
Copied!
```

No error message received after import statement would indicate that the package is now ready for usage. Test the function by running the following statement in the shell.

```
    1. 1
    sentiment_analyzer("This is fun.")
```

Copied!

The output received would look as shown below.

```
>>> sentiment_analyzer("This is fun.")
{'label': 'SENT_POSITIVE', 'score': 0.997183}
```

To exit the python shell, type exit() or press Ctrl+Z.

Task 5: Run Unit tests on your application

Since now we have a functional aplication, it is required that we run unit tests on some test cases to check the validity of its outputs.

For running unit tests, we need to create a new file that calls the required application function from the package and tests its for a known text and output pair.

For this, complete the following steps.

- 1. Create a new file in practice_project folder, called test_sentiment_analysis.py.
- 2. In this file, import the sentiment_analyzer function from the SentimentAnalysis package. Also import the unittest library.
- ▼ Click here for the solution
 - 1. 1
- 1. from SentimentAnalysis.sentiment_analysis import sentiment_analyzer
- import unittest

Copied!

- 3. Create the unit test class. Let's call it TestSentimentAnalyzer. Define test_sentiment_analyzer as the function to run the unit tests.
- ▼ Click here for the solution
- 1. 1
- class TestSentimentAnalyzer(unittest.TestCase):
 def test_sentiment_analyzer(self):

Copied!

4. Define 3 unit tests in the said function and check for the validity of the following statement - label pairs.

```
"I love working with Python": "SENT_POSITIVE"
"I hate working with Python": "SENT_NEGATIVE"
"I am neutral on Python": "SENT_NEUTRAL"
```

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Use 'assertEqual' function to compare the 'label' of the output with the label expected.

▼ Click here for solution

```
1. 1
2. 2
3. 3
4. 4
5. 5
5. 5
6. 6
7. 7
8. 8

1. class TestSentimentAnalyzer(unittest.TestCase):
2. def test_sentiment analyzer(self):
3. result_1 = sentiment_analyzer('I love working with Python')
4. self.assertEqual(result_1['label'], 'SENT_POSITIVE')
5. result_2 = sentiment_analyzer('I hate working with Python')
6. self.assertEqual(result_2['label'], 'SENT_NEGATIVE')
7. result_3 = sentiment_analyzer('I am neutral on Python')
8. self.assertEqual(result_3['label'], 'SENT_NEGATIVE')
```

5. Call the unit tests.

▼ Click here for solution

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Add the following line at the end of the file.

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1. 1

unittest.main()

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Now that the file is ready, execute the file to perform unit tests. Upon successful execution, the output of this file should be as shown in the image below.

Task 6: Deploy as web application using Flask

Now that the application is ready, it is time to deploy it for usage over a web interface. To ease the process of deployment, you have been provided with 3 files which are going to be used for this task.

1. index.html in templates folder.

This file has the code for the web interface that has been designed for this lab. This is being provided for you in completion and is to be used as is. You are not required to make any changes to this file.

The interface is as shown in the image.

NLP - Sentiment Analysis using BERT

Please enter the text to be analyzed

Run Sentiment Analysis

Result of Sentiment Analysis

Clicking the Run Sentiment Analysis button, on the html interface, calls this javascript file which executes a GET request and takes the text provided by the user as input. This text, saved in a variable named textToAnalyze is then passed on to the server file to be sent to the application. This file is also being provided to you in completion and is expected to be used as is. You are not required to make any changes to this file.

server.py in the practice_project folder.

This task revolves around the completion of this file. You can complete this file by completing the following 5 steps.

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^{2.} mywebscript.js in static folder.

a. Import the relevant libraries and functions

In this file, you'll need the Flask library along with its render_template function (for deploying the HTML file) and request function (to initiate the GET request from the web page).

You also would need to import the sentiment_analyzer function from the SentimentAnalysis package.

Add the relevant lines of code, importing the said functions, in server.py

▼ Click here for solution

```
1. 1
2. 2
```

1. from flask import Flask, render template, request

from SentimentAnalysis.sentiment_analysis import sentiment_analyzer

Copied!

b. Initiate the Flask app by the name Sentiment Analyzer

Put the knowledge gained in Module 2 of this course and add the statement to server.py, that initiates the application and names it Sentiment Analyzer.

▼ Click here for solution

```
1. 1
```

app = Flask("Sentiment Analyzer")

Copied!

c. Define the function sent_analyzer

The purpose of this function is two fold. First, the function should send a GET request to the HTML interface to receive the input text. Note that the GET request should reference textToAnalyze variable as defined in the mywebscript.js file. Store the incoming text to a variable text_to_analyze. Now, as the second function, call your sentiment_analyzer application with text_to_analyze as the argument.

Also, format the returning output of the function in a formal text. For e.g. The given text has been identified as POSITIVE with a score of 0.99765.

▼ Click here for hint

- 1. Use request.args.get to initiate the GET request.
- 2. The label, received as "SENT_CLASS" (where class can be POSITIVE, NEGATIVE or NEUTRAL), will have to be split on '_' to access the class name individually.

▼ Click here for solution

The function should read like this.

```
1. 1
2. 2
3. 3
4. 4
5. 5
6. 6
7. 7
1. @app.route("/sentimentAnalyzer")
2. def sent_analyzer():
3. text_to_analyze = request.args.get('textToAnalyze')
4. response = sentiment_analyzer(text_to_analyze)
5. label = response['abel']
6. score = response['score']
7. return "The given text has been identified as {} with a score of {}.".format(label.split('')[1], score)
```

Copied!

Note: The function uses the Flask decorator @app.route("/sentimentAnalyzer") as referenced in the mywebscript.js file.

d. Render the HTML template using render_index_page

This function should simply run the render_template function on the HTML template, index.html.

▼ Click here for solution

```
1. 1
2. 2
3. 3
1. @app.route("/")
2. def render_index_page():
3. return render template('index.html')
```

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e. *Run the application on localhost:5000 *

Finally, upon file execution, run the application on host: 0.0.0 (or localhost) on port number 5000.

▼ Click here for solution

1. 1

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To deploy the application, exectue the file server.py from the terminal.

1. 1
1. python3.11 server.py
Copied! Executed!

The output would look like this.

```
theia@theia-abhishekg1:/home/project/practice_project$ python3.11 server.py

* Serving Flask app 'Sentiment Analyzer'

* Debug mode: off

WARNING: This is a development server. Do not use it in a production deployment. Use a production WSGI server instead

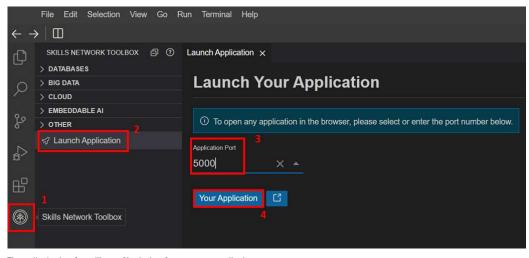
* Running on all addresses (0.0.0.0)

* Running on http://127.0.0.1:5000

* Running on http://172.22.138.184:5000

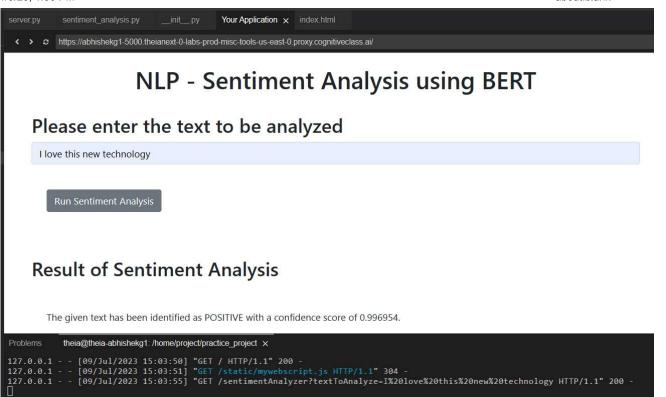
Press CTRL+C to quit
```

The app is now running on localhost:5000. To access the application, go to the Skills Network Toolbox tab and click on Launch Application. Enter the Appplication port as 5000, and click on Your Application.



The application interface will open. Use the interface to test your application.

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To stop the application, press Ctrl+C.

Task 7: Incorporate Error handling

To incorporate error handling, we need to identify the different forms of error codes that may be received in response to the GET query initiated by the sent analyzer function in server.py.

This is already a part of the Watson NLP Library functions and can be observed on the terminal console where the code is running.

Consider the image shown below.

NLP - Sentiment Analysis using BERT

Please enter the text to be analyzed I love this new technology Run Sentiment Analysis

Result of Sentiment Analysis

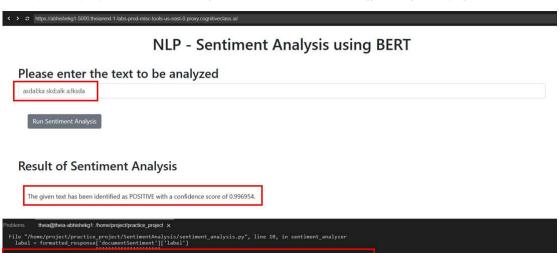
The given text has been identified as POSITIVE with a confidence score of 0.996954.

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The codes indicate that the initial GET request was successful (200), the request was then successfully transferred to the Watson Library (304) and then the GET request to generate the response was also conducted successfully.

In the case of invalid entries, the system responds with 500 error code, indicating that there is something wrong at the server end.

Invalid entry could be anything that the model is not able to interpret. However, in the situation of this error, this application output doesn't get updated.



Note that the output on the interface is the same as before, the text being analyzed is a random text and the Watson AI libraries are throwing a 500 error confirming that the model has not been able to process the request.

To fix this bug in our application, we need to study the response received from the Watson AI library function, when the server generates 500 error. To test this, we need to retrace the steps taken in Task 2, and test the Watson AI library with an invalid string input.

Open a python shell in the terminal and run the following commands to check the required output.

7.0.9.1 - - [10/)u1/2023 15:46:39] "GET /sentimentAnalyzer?textToAnalyze=asdal;ka%20skd;alk%20a;lk%20a HTTP/1.1" 500

```
1. 1
2. 2
3. 3
4. 4
5. 5
6. 6
7. 7
8. 8
9. 9
10. 10
1. import requests
2. url = "https://sn-watson-sentiment-bert.labs.skills.network/v1/watson.runtime.nlp.v1/NlpService/SentimentPredict"
3. headers = {"grpc-metadata-mm-model-id": "sentiment_aggregated-bert-workflow_lang_multi_stock"}
4. myobj = { "raw_document": { "text": "as987da-652d aweadsa" } }
5. response = requests.post(url, json = myobj, headers=headers)
6. print(response.status_code)
7.
8. myobj = { "raw_document": { "text": "Testing this application for error handling" } }
9. response = requests.post(url, json = myobj, headers=headers)
10. print(response.status_code)
```

The console response looks as shown in the image below. The red boxes indicate the invalid text and its status code received, and the yellow boxes indicate the valid text and its status code received.

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```
theia@theia-abhishekgl:/home/project/practice_project$ python3.11
Python 3.11.2 (main, Feb 8 2023, 14:49:25) [GCC 11.3.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>> import requests
>>> url = "https://sn-watson-sentiment-bert.labs.skills.network/v1/watson.runtime.nlp.v1/NlpService/Sent imentPredict"
>>> headers = {"grpc-metadata-mm-model-id": "sentiment_aggregated-bert-workflow_lang_multi_stock"}
>>> myobj = { "raw_document": { "text": "as987da-6s2d aweadsa" } }
>>> response = requests.post(url, json = myobj, headers=headers)
>>> print(response.status_code)
500
>>>
>>> myobj = { "raw_document": { "text": "Testing this application for error handling" } }
>>> response = requests.post(url, json = myobj, headers=headers)
>>> print(response.status_code)
>>> print(response.status_code)
```

This enables you to modify the application in such a fashion, that we can send different outputs for different status codes.

In the first part of this task, you have to modify the sentiment_analyzer() function to return the both label and score as None in case of invalid text entry.

▼ Click here for hint

Make an if-else conditional statement in the sentiment analyzer function to add the necessary functionality.

▼ Click here for the solution sentiment analysis.py

```
5. 5
6. 6
 10. 10
 11. 11
12. 12
 13. 13
  1. import requests
  3. def sentiment_analyzer(text_to_analyse):
           url = 'https://sn-watson-sentiment-bert.labs.skills.network/v1/watson.runtime.nlp.v1/NlpService/SentimentPredict'
myobj = { "raw_document" : { "text": text_to_analyse } }
header = { "grpc-metadata-mm-model-id": "sentiment_aggregated-bert-workflow_lang_multi_stock"}
            response = requests.post(url, json = myobj, headers=header)
            formatted response = json.loads(response.text)
            if response.status_code == 200:
                  label = formatted_response['documentSentiment']['label']
score = formatted_response['documentSentiment']['score']
 10.
 11.
            elif response.status_code == 500:
 12.
                  label = None
 13.
 15.
            return {'label': label, 'score': score}
Copied!
```

Now, in server.py, the response to be sent to the console should also be different for the valid and invalid input types.

For invalid input, let the console print Invalid input ! Try again.

▼ Click here for hint

Make an if-else conditional statement in sent analyzer function of server.py to check whether "label" is None or not.

▼ Click here for the solution

```
1. 1
2. 2
3. 3
4. 4
4
5. 5
6. 6
6. 7
7. 7
8. 8
9. 9
9
1. def sent_analyzer():
2.    text_to_analyze = request.args.get('textToAnalyze')
3.    response = sentiment_analyzer(text_to_analyze')
4.    label = response['label']
5.    score = response['label']
6.    if label is None:
7.        retrum "Invalid input! Try again."
8.    else:
9.        return "The given text has been identified as {} with a score of {}.".format(label.split('_')[1], score)
```

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Copied!

Now, your application is capable of responding appropriately to any form of inputs

https://abhishekg1-8080.theianext-1-labs-prod-misc-tools-us-east-0.proxy.cognitiveclass.ai/

NLP - Sentiment Analysis using BERT

Please enter the text to be analyzed

Kayqa huk prueba kay ruwanapa ruwayninta qhawanapaq.

Run Sentiment Analysis

Result of Sentiment Analysis

Invalid input! Try again.

Task 8: Run static code analysis

Finally, in Task 8, we check the quality of your coding skills as per the PEP8 guidelines by running static code analysis.

Normally, this is done at the time of packaging and unit testing the application. However, we have kept this step at the of this project since the codes are updated in all tasks before this. Once your files for this project are now ready, let us test them for adherence to the PEP8 guidelines.

The first step in this process is to install the PyLint library using the terminal.

▼ Click here for the solution

1. 1

1. python3.11 -m pip install pylint

Copied!

Next, use pylint to run static code analyse server.py.

▼ Click here for the solution

On terminal bash execute the following command.

1. 1

pylint server.py

Copied!

If all aspects of PEP8 guide have been incorporated in your code, then the score generated should be 10/10. In case it isn't follow the instructions given by the modify to correct the code appropriately.

```
theia@theia-abhishekg1:/home/project/practice_project$ pylint server.py

Your code has been rated at 10.00/10 (previous run: 10.00/10, +0.00)
```

This concludes the practice project.

(Optional) Additional Exercises

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Interested learners can the try the following exercises on your own, for enhanced understanding of the concepts learnt through this project. No solution is being provided for these exercises. However, feel free to discuss and share your solutions with your peers in the course discussion forums.

- 1. Run static code analysis on sentiment_analysis.py. Try to achieve a 10/10 score. Hint: Docstrings
- 2. Test the capacity of your application in handling sentences of languages other than English, for e.g. French, German, etc. See if the application responds with an invalid text error.
- 3. Currently, if the application is run WITHOUT supplying an input, i.e. leaving the text blank, the model still throws the same error of invalid text. Try including a special case, where a blank input receives a different error message.

Conclusion

Congratulations on completing this project.

With the completion of this project, you have:

- 1. Created an AI based sentiment analysis application using Watson NLP embedded libraries.
- 2. Formatted the output received from the Watson NLP library function to extract relevant information from it.
- 3. Packaged the application and made it importable to any python code for usage.
- 4. Ran unit tests on the application and checked the validity of its outputs for different inputs.
- 5. Deployed the application using Flask framework.
- 6. Incorporated error handling capability in the application, such that a response code of 500 receives an appropriate response from the application.
- 7. Ran static code analysis on the code files to confirm their adherence to the PEP8 guidelines.

Author(s)

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Changelog

Date	Version	Changed by	Change Description
2023-08-29	1.3	Ritika Joshi	updated the instructions
2023-07-11	1.2	Abhishek Gagneja	Added new functionalities
2023-07-10	1.1	Abhishek Gagneja	Changes in instructions and images
2023-06-30	1.0	Abhishek Gagneja	Initial version created

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