

Handwritten Scripts Classifier

USING GOOGLE CLOUD AUTOML AND APIS

Names	SRN
Shubham S	PES1UG20CS420
Shashank S	PES1UG20CS394

Objective

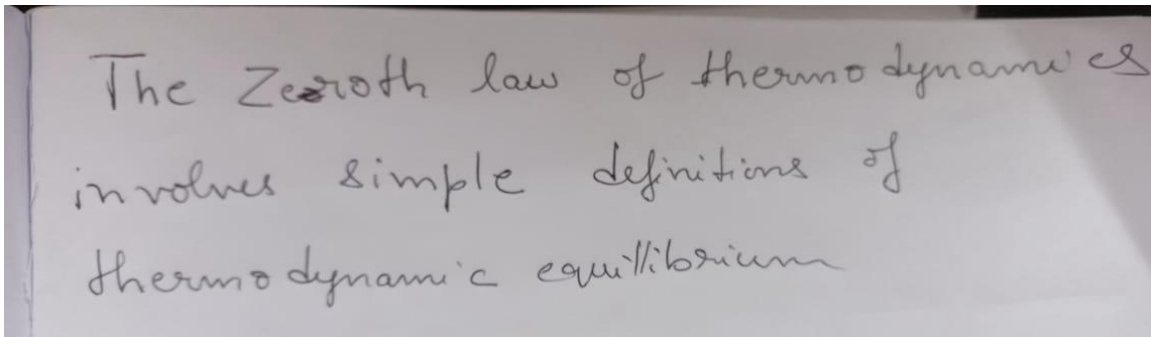
The goal of our application is to take in handwritten text images written by students and to give out the text and convert this into speech in any language the evaluator prefers. Natural Language processing is also done on this text to identify entities and other features.

A custom dataset which has sentences classified into 3 labels (Physics , Chemistry , Biology) is taken and trained using AutoML . The text identified from the image is given to this model which finally classifies it to the appropriate subject.

Pretrained APIs used

- Cloud Vision
To identify the Text present in the Image.
- Text to speech
To convert the text we obtained from the image to an audio file.
- Translate
To translate the text we obtained to any language chosen.
- Natural language
To analyze the text read.

Sample Input



Code

```
import os, io
import pandas as pd

from google.cloud import texttospeech_v1
from google.cloud import translate_v2
from google.cloud import language_v1
```

```

from google.cloud import vision_v1

os.environ['GOOGLE_APPLICATION_CREDENTIALS'] = 'google_credentials.json'
client = vision_v1.ImageAnnotatorClient()
client1=texttospeech_v1.TextToSpeechClient()
client2=translate_v2.Client()

FOLDER_PATH = 'C:\Handwritten to text\Handwritten\Pics'
IMAGE_FILE = '7th.jpeg'
FILE_PATH = os.path.join(FOLDER_PATH, IMAGE_FILE)

with io.open(FILE_PATH, 'rb') as image_file:
    content = image_file.read()

image = vision_v1.types.Image(content=content)
response = client.document_text_detection(image=image)

docText = response.full_text_annotation.text
print(docText)
pages = response.full_text_annotation.pages

#Confidence level of each letter detected (commented for convinient
output)
"""
for page in pages:
    for block in page.blocks:
        print('block confidence:', block.confidence)

        for paragraph in block.paragraphs:
            print('paragraph confidence:', paragraph.confidence)

            for word in paragraph.words:
                word_text = ''.join([symbol.text for symbol in
word.symbols])

                print('Word text: {0} (confidence: {1}').format(word_text,
word.confidence))

                for symbol in word.symbols:
                    print('\tSymbol: {0} (confidence:
{1}').format(symbol.text, symbol.confidence))
"""
text = docText

```

```

synthesis_input = texttospeech_v1.SynthesisInput(ssml=text)

voice1 = texttospeech_v1.VoiceSelectionParams(
    language_code = 'en-in',
    ssml_gender= texttospeech_v1.SsmlVoiceGender.FEMALE )

print(client1.list_voices)
audio_config= texttospeech_v1.AudioConfig(
    audio_encoding = texttospeech_v1.AudioEncoding.MP3
)

response1 = client1.synthesize_speech (
    input=synthesis_input,
    voice=voice1,
    audio_config=audio_config
)

with open('audio.mp3','wb',) as output:
    output.write(response1.audio_content)

target = input(" Enter the language to which you wish to translate : ")

output1 = client2.translate(text,target_language = target)

print(output1)

synthesis_input1 =
texttospeech_v1.SynthesisInput(ssml=output1['translatedText'])

voice2 = texttospeech_v1.VoiceSelectionParams(
    language_code = target,
    ssml_gender= texttospeech_v1.SsmlVoiceGender.FEMALE )

print(client1.list_voices)
audio_config= texttospeech_v1.AudioConfig(
    audio_encoding = texttospeech_v1.AudioEncoding.MP3
)

response2 = client1.synthesize_speech (
    input=synthesis_input1,

```

```

        voice=voice2,
        audio_config=audio_config
    )

with open('audio1.mp3','wb',) as output10:
    output10.write(response2.audio_content)

def sample_analyze_entities(docText):
    client3 = language_v1.LanguageServiceClient()
    type_ = language_v1.Document.Type.PLAIN_TEXT
    language = "en"
    document = {"content": docText, "type_": type_, "language": language}
    encoding_type = language_v1.EncodingType.UTF8
    response = client3.analyze_entities(request = {'document': document,
'encoding_type': encoding_type})
    for entity in response.entities:
        print(u"Representative name for the entity:
{}".format(entity.name))
        print(u"Entity type:
{}".format(language_v1.Entity.Type(entity.type_).name))
        print(u"Saliency score: {}".format(entity.saliency))
        for metadata_name, metadata_value in entity.metadata.items():
            print(u"{}: {}".format(metadata_name, metadata_value))
        for mention in entity.mentions:
            print(u"Mention text: {}".format(mention.text.content))
            print(u"Mention type:
{}".format(language_v1.EntityMention.Type(mention.type_).name) )
            print(u"Language of the text: {}".format(response.language))

sample_analyze_entities(docText)

```

Identified Text

```
Windows PowerShell
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Install the latest PowerShell for new features and improvements! https://aka.ms/PSWindows

PS C:\Handwritten to text\Handwritten> & 'C:\Users\Shashank S\AppData\Local\Programs\Python\Python310\python.exe' 'c:\Users\Shashank S\.vscode\extensions\ms-python.python-2022.10.1\pythonFiles\lib\pyth
hon\debugpy\adapter\..\..\debugpy\launcher' '53809' '--' 'c:\Handwritten to text\Handwritten\Handwritten text to plain text and speech[code].py'
The Zeroth law of thermodynamics
involves simple definitions of
thermodynamic equilibrium
<bound method TextToSpeechClient.list_voices of <google.cloud.texttospeech_v1.services.text_to_speech.client.TextToSpeechClient object at 0x0000022540D00310>>
Enter the language to which you wish to translate : █
```

Converting Text to different language (both text and to speech) with NLP analysis .

```
Windows PowerShell
Copyright (C) Microsoft Corporation. All rights reserved.

Install the latest PowerShell for new features and improvements! https://aka.ms/PSWindows

PS C:\Handwritten to text\Handwritten> & 'C:\Users\Shashank S\AppData\Local\Programs\Python\Python310\python.exe' 'c:\Users\Shashank S\.vscode\extensions\ms-python.python-2022.10.1\pythonFiles\lib\pyt
hon\debugpy\adapter\..\..\debugpy\launcher' '53809' '--' 'c:\Handwritten to text\Handwritten\Handwritten text to plain text and speech[code].py'
The Zeroth law of thermodynamics
involves simple definitions of
thermodynamic equilibrium
<bound method TextToSpeechClient.list_voices of <google.cloud.texttospeech_v1.services.text_to_speech.client.TextToSpeechClient object at 0x0000022540D00310>>
Enter the language to which you wish to translate : kn
{'translatedText': 'ಉಷ್ಣತೆಯು ಒಂದೇ ಆಗಿರುವ ಎರಡು ವಸ್ತುಗಳ ನಡುವೆ ಉಷ್ಣವರ್ಧನೆಯು ಸುಲಭವಾಗಿರುತ್ತದೆ', 'detectedSourceLanguage': 'en', 'input': 'The Zeroth law of thermodynamics\ninvolves simple definitions
of\nthermodynamic equilibrium'}
<bound method TextToSpeechClient.list_voices of <google.cloud.texttospeech_v1.services.text_to_speech.client.TextToSpeechClient object at 0x0000022540D00310>>
Representative name for the entity: law
Entity type: OTHER
Saliency score: 0.5302996635437012
Mention text: law
Mention type: COMMON
Representative name for the entity: Zeroth
Entity type: PERSON
Saliency score: 0.24963222444057465
Mention text: Zeroth
Mention type: PROPER
Representative name for the entity: thermodynamics
Entity type: OTHER
Saliency score: 0.11976754665374756
Mention text: thermodynamics
Mention type: COMMON
Representative name for the entity: equilibrium
Entity type: OTHER
Saliency score: 0.06128700450062752
Mention text: equilibrium
Mention type: COMMON
Representative name for the entity: definitions
Entity type: OTHER
Saliency score: 0.039013542234897614
Mention text: definitions
Mention type: COMMON
Language of the text: en
PS C:\Handwritten to text\Handwritten> █
```

AutoML Dataset

←

PCB_data

PCM_data_tcn

?

IMPORT

BROWSE

ANALYSE

All7,947

Labelled7,947

Unlabelled0

FilterFilter labels+

Biology3,048

Chemistry2,867

Physics2,032

ADD NEW LABEL

FilterFilter items

Text

Labels

▼

...it didn't really take that long. \n\nTube tech was established in 1904 by John ...

Physics

▼

In a way more efficient version I exchange 100litres of water to move 15700Litr...

Physics

▼

This is a homework problem we were given. I know that normally you could pot...

Chemis

▼

Excellent example to illustrate the difference!!

Biology

▼

Those are all good ideas, thank you!

Biology

▼

Yeah, if people thought this through, they'd be deploying masks in many more pl...

Biology

▼

Yes, there are different types of reactions. The example you provided is an Acid-...

Chemis

▼

Read the section about the intron again, see if that helps.

Biology

▼

Man, just go to the doctor to be sure, don't take risks

Chemis

▼

Mountain Squirrel.

Biology

Items per page:10▼

1 – 10 of many

<

>

Training jobs and models

✓PCM_

✓PCM_data (Version 1)
Model type: Text classification

TRAIN NEW MODEL

Labeling tasks

If your data still needs to be labeled, create a labeling task to have others label it for you

CREATE LABELLING TASK

Training

Training

+ CREATE

REFRESH

LEARN

TRAINING PIPELINES

CUSTOM JOBS

HYPERPARAMETER TUNING JOBS

Training pipelines are the primary model training workflow in Vertex AI. You can use training pipelines to create an AutoML-trained model or a custom-trained model. For custom-trained models, training pipelines orchestrate custom training jobs and hyperparameter tuning with additional steps like adding a data set or uploading the model to Vertex AI for prediction serving. [Learn more](#)

Region

us-central1 (Iowa)

?

Filter

Enter a property name

?

|||

Name	ID	Status	Job type	Model type	Created	Elapsed time	Labels
PCM_data	6419078900206272512	Finished	Training pipeline	Text classification (Single-label)	21 Jul 2022, 22:19:10	8 hr 14 min	—

PAGE 6

Model

PCM_data

Version 1

VIEW DATA SET

LEARN

EVALUATE

DEPLOY AND TEST

BATCH PREDICT

VERSION DETAILS

Filter

Filter labels

All labels

0

Biology

0.89065

Physics

0.84203

Chemistry

0.81729

Confidence threshold

0.5

All labels

Average precision

0.854

Precision

81.6%

Recall

65.2%

Created

22 Jul 2022, 06:33:16

Total items

7,947

Training items

6,357

Validation items

795

Test items

795

To evaluate your model, set the confidence threshold to see how precision and recall are affected. The best confidence threshold depends on your use case. Read some [example scenarios](#) to learn how evaluation metrics can be used.

Precision-recall curve

Precision-recall by threshold

Confusion matrix

☐ Item counts

This table shows how often the model classified each label correctly (in blue), and which labels were most often confused for that label (in grey).

True label	Predicted label		
	Physics	Biology	Chemistry
Physics	74%	9%	17%
Biology	6%	78%	16%
Chemistry	10%	16%	74%

Classifying The text

Deploy your model

Endpoints are machine learning models made available for online prediction requests. Endpoints are useful for timely predictions from many users (for example, in response to an application request). You can also request batch predictions if you don't need immediate results.

DEPLOY TO ENDPOINT

Name	ID	Status	Models	Region	Monitoring	Most recent monitoring job	Most recent alerts	Last updated	API	Notification
PCM_endpoint	8565802510793572352	Active	1	us-central1	Disabled	—	—	22 Jul 2022, 09:38:24	Sample request	

Test your model

PREVIEW

The Zeroth law of thermodynamics involves simple definitions of thermodynamic equilibrium

PREDICT

Filter

Filter labels

Biology

0.041

Chemistry

0.164

Physics

0.796