Paraphrase Identification

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Github Repo Link -

https://github.com/manasvi-26/Paraphrase-Identification

Introduction

The major project allotted to our team required us to build a paraphrasing system that both detects paraphrased sentence-pairs and also generates paraphrased sentences when an input sentence is fed into the system. As we move to the second phase of the project, the primary aim was to test out various paraphrase identification tasks that have been happening in the research community. The intention was to complete building the paraphrase identification task after selecting a Transformer model that could help with the same.

In this document, we discuss the internal discussions we have had, the work that we have done since the last submission, progress tracking against the timeline we had submitted as part of phase 1, and key decisions made along with the reasoning behind the same.

The main task for this phase was Paraphrase Identification

Identifying the Dataset

Effort was put from our side to understand the datasets primarily used in paraphrase identification tasks. The following datasets were under consideration initially:

- Quora Question Pair (QQP)
- PAWS
- Microsoft Research Paraphrase Corpus (MRPC)

According to the PAWS paper -

- The best accuracy for the QQP dataset was 95.3%
- The best accuracy for the PAWS_QQP dataset was 89.9%
- The best accuracy for the PAWS_WiKi dataset was 93.8%

We chose the PAWS dataset for the first round of understanding since PAWS is a much larger dataset.

Identifying a Model to Train

The paraphrase identification task is basically a binary classification task where we determine whether or not a given pair of sentences in the dataset are paraphrases of each other or not. We came across research works that have tried solving the paraphrase identification task using primarily the following transformer models or their modifications:

- BERT
- T5

XLNet

Our best model was chosen to be **BERT - specifically the bert-base-uncased model**

Experimentation

Attempt 1:

The first attempt was to use a pre-trained T5 model from the Huggingface library, encode the sentences in the sentence pair and check for the cosine similarity. This gave very poor results and the accuracy was as low as 57%.

Method:

- · We load the PAWS dataset
- We get a pre-trained **T5 model** and it's tokenizer from HuggingFace transformers library
- We put both sentences in each data point through the encoder in the model and got the sentence vectors from the last hidden state
- We calculated sklearn's cosine similarity on this and used a threshold (of about 0.95 to 0.98) to label it as a paraphrase

Findings

- · Accuracy was 57%
- · No Fine tuning was done on this
- This focused more on sentence semantic similarity and hence the PAWS dataset was not fully utilized explaining the lower accuracy

Attempt 2

After the first attempts, we decided to incorporate **Fine-Tuning** for our models. This was done using a framework called **Fast AI**, which basically facilitates Transfer Learning on pre trained models.

Transfer learning is a machine learning method where a model developed for a task is reused as the starting point for a model on a second task. It basically means fine tuning and re training an already pre trained generic model for specific tasks to obtain much better results. The models and datasets used here were:

- PAWS and MRPC with BERT-base-uncased (93.1% accuracy with PAWS and 82.84% accuracy with MRPC)
- PAWS with XLNET (90% accuracy)

The Datasets were PAWS and MRPC

PAWS



The label column determines whether sentence1 and sentence2 are paraphrases of each other

This dataset had 49401 Training data points, and 2000 Testing data points

MRPC

| | Quality | #1 ID | #2 ID | #1 String | #2 String |
|---|---------|---------|---------|---|---|
| 0 | 1 | 702876 | 702977 | Amrozi accused his brother , whom he called " | Referring to him as only " the witness " , Amr |
| 1 | 0 | 2108705 | 2108831 | Yucaipa owned Dominick 's before selling the c | Yucaipa bought Dominick 's in 1995 for \$ 693 m |
| 2 | 1 | 1330381 | 1330521 | They had published an advertisement on the Int | On June 10 , the ship 's owners had published \dots |
| 3 | 0 | 3344667 | 3344648 | Around 0335 GMT , Tab shares were up 19 cents \dots | Tab shares jumped 20 cents , or 4.6 $\%$, to set |
| 4 | 1 | 1236820 | 1236712 | The stock rose \$ 2.11 , or about 11 percent , | PG & E Corp. shares jumped \$ 1.63 or 8 percent |
| | | | | | |

The Quality column determines whether #1 String and #2 String are paraphrases or not

This dataset had 3668 training points and 1725 testing points

Attempt 2a - Bert-base-uncased with PAWS and MRPC:

PAWS Data

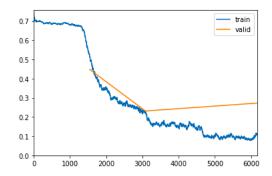
Method:

- Loading the PAWS dataset and pre processing
- Initializing the Learner Fast ai object
- Preparing a Databunch for the learner
- Training for 4 epochs
 - We trained the model on a Colab GPU.
- · Saving the model

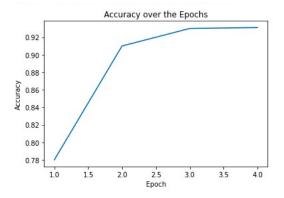
Findings

- Initial accuracy without any fine tuning was 53.6%
- · The accuracy values at each epoch

| epoch | train_loss | valid_loss | accuracy |
|-------|------------|------------|----------|
| 0 | 0.522695 | 0.447943 | 0.788000 |
| 1 | 0.230631 | 0.231232 | 0.918500 |
| 2 | 0.150596 | 0.251808 | 0.931000 |
| 3 | 0.112169 | 0.272477 | 0.931000 |



Decrease of Training loss and Validation loss over training iterations



- Final accuracy value was 93.1%. This was more than the paper about models trained on QQP and PAWS which was 87%
- Final F1 score 92.65%
- · Some sample predictions:

```
SENTENCE 1 -> The exception was between late 2005 and 2009 when he played in Sweden with Carlstad United BK , Serbia with FK Borac Čača SENTENCE 2 -> The exception was between late 2005 and 2009 , when he played in Sweden with Carlstad United BK , Serbia with FK Borac Ča PREDICTED = 1 ACTUAL = 1

SENTENCE 1 -> The Tabaci River is a tributary of the River Leurda in Romania .

SENTENCE 2 -> The Leurda River is a tributary of the River Tabaci in Romania .

PREDICTED = 0 ACTUAL = 0

SENTENCE 1 -> He played with the A-level Kane County Cougars in 1993 and the AA Portland Sea Dogs .

SENTENCE 2 -> He played in 1993 with the A - Level Portland Sea Dogs and the AA Kane County Cougars .

PREDICTED = 0 ACTUAL = 0

SENTENCE 1 -> Winarsky is a member of the IEEE , Phi Beta Kappa , the ACM and Sigma Xi .

SENTENCE 2 -> Winarsky is a member of ACM , the IEEE , the Phi Beta Kappa and the Sigma Xi .

SENTENCE 1 -> In 1938 he became the government anthropologist of the anglo-Egyptian Sudan and led fieldwork with the Nuba .

SENTENCE 2 -> In 1938 he became the Government Anthropologist of the Egyptian-Anglo Sudan and conducted fieldwork with the Nuba .

PREDICTED = 0 ACTUAL = 0
```

MRPC Data

Method

· Loading the MRPC dataset and pre processing

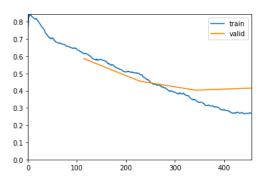
- Initializing the Learner Fast ai object
- · Preparing a Databunch for the learner
- · Training for 4 epochs
 - We trained the model on a Colab GPU.

Findings

- Initial accuracy without any fine tuning was 33.85%
- · Accuracy at each epoch:

epoch train_loss valid_loss accuracy

| 0 | 0.615572 | 0.585727 | 0.704348 |
|---|----------|----------|----------|
| | | | |
| 1 | 0.498354 | 0.455152 | 0.805797 |
| 2 | 0.346862 | 0.402653 | 0.822029 |
| 3 | 0.268923 | 0.414614 | 0.828406 |



- Decrease of Training loss and Validation loss over training iterations
- Final Accuracy score = 82.84%
- Final F1 score = 87.2%

Attempt 2b - XLNet with PAWS:

In this attempt, the model architecture used was XLNet.

Method:

- A pre-trained XLNet model imported from huggingface library
- · preprocessing the PAWS dataset Created a custom Tokenizer and numericalizer
- · Preparing the databunch
- Fast Ai learner: Adam Optimizer imported from huggingface was used.
- Training: The model was trained for a total of 3 epochs

Findings:

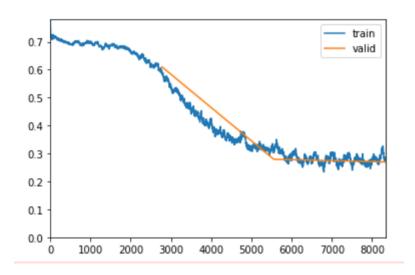
Final accuracy: 90%

Below is the accuracy value for each epoch -

| epoch | train_loss | valid_loss | accuracy | error_rate | time |
|-------|------------|------------|----------|------------|-------|
| 0 | 0.586868 | 0.610249 | 0.669028 | 0.330972 | 25:40 |
| 1 | 0.307644 | 0.279444 | 0.891093 | 0.108907 | 26:37 |
| 2 | 0.272397 | 0.269873 | 0.900000 | 0.100000 | 26:12 |

Below is the training loss and validation loss -

Number of Training Iterations Vs Loss



Some Sample Paraphrase results -

```
SENTENCE 1 -> In the reviews below will be the highest rating for the show in red ,
              and the lowest evaluation for the show will be in blue episode
SENTENCE 2 -> In the reviews below will be the lowest rating for the show in red ,
              and the highest rating for the show will be in blue sequence .
PREDICTED = 0 ACTUAL = 0
SENTENCE 1 -> From the merger of the Four Rivers Council and the Audubon Council , the Shawnee Trails Council was born .
SENTENCE 2 -> Shawnee Trails Council was formed from the merger of the Four Rivers Council and the Audubon Council .
PREDICTED = 1 ACTUAL = 1
SENTENCE 1 -> The family moved to Camp Hill in 1972 , where he attended Trinity High School in Harrisburg , Pennsylvania .
SENTENCE 2 \rightarrow 111972, the family moved to Camp Hill, where he visited the Trinity High School in Harrisburg, Pennsylvania. PREDICTED = 1 + 111912
SENTENCE 1 -> Components of elastic potential systems store mechanical energy if
              they are deformed when forces are applied to the system
SENTENCE 2 -> Components of elastic potential systems store mechanical energy if
              they are deformed to the system when applied to forces .
PREDICTED = 1 ACTUAL = 1
SENTENCE 1 -> Aamir Khan has agreed to play immediately after reading Mehra 's script in `` Rang De Basanti '' .
\textbf{SENTENCE 2 -> Mehra agreed to act in `` Rang De Basanti '' immediately after reading Aamir Khan 's script .\\
PREDICTED = 0 ACTUAL = 0
```

So our final scores were from our best saved model which was: bert-base-cased fine tuned on PAWS dataset

Accuracy = 93.1%

F1 score = 92.65%

These scores are comparable to that of the PAWS paper mentioned above. It is more than the score they got on PAWS_QQP dataset (89%).

Agreement with Proposed Timeline

Our proposed timeline expected a report on the paraphrase generation task as well. However, after much deliberation with everyone including the TA, it was decided to keep information to a minimum on the generation task. The paraphrase generation task was slightly explored until the point where we decided to focus on the task at hand.

The first attempt at the paraphrase identification task was done with the help of the T5 model that was trained on the PAWS dataset.

Plans for Next Phase

Our current classification F1 and accuracy scores are close to the baseline. In the next phase, the primary aim would be to generate paraphrased sentences. Based on the completion of the paraphrased sentence generation task, attempts will be made to 'beat' the deadline.

Learnings from Attempt 1:

The major learning was the fact that we can not very likely proceed with the idea of "off the shelf" models. And hence, we would have to look at means of training the model, irrespective of what we select finally.