Fine-Tuning BERT project

Project: Test strategies to fine-tuning bert model for classification tasks (domain adaptation).

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```
In [41]:
         # Datasets
         import os
         import opendatasets as od
         # Data visualisation
         import altair as alt
         alt.data transformers.enable('default', max rows=None)
         # Text Mining
         from nltk.tokenize import word tokenize
         import numpy as np
         import pandas as pd
         import torch
         from sklearn.model selection import train test split
         from transformers import BertTokenizerFast, BertForSequenceClassification
         from transformers import Trainer, TrainingArguments
         from sklearn.metrics import accuracy score, precision score, f1 score
         from sklearn import preprocessing
         from sklearn.manifold import TSNE
         import plotly.express as px
         from sentence transformers import SentenceTransformer
```

Load data - Kaggle Emotions in Text

i can go from feeling so hopeless to so damned...

2 im grabbing a minute to post i feel greedy wrong

Data Understanding

Sample

```
In [5]: data.shape
Out[5]: (21459, 2)

In [6]: data.head(5)

Out[6]: Text Emotion

O i didnt feel humiliated sadness
```

sadness

anger

```
4
                                        i am feeling grouchy
                                                                anger
           data.columns = list(map(lambda i: i.lower(), data.columns))
In [7]:
           data.head()
In [8]:
Out[8]:
                                                       text
                                                             emotion
          0
                                       i didnt feel humiliated
                                                              sadness
              i can go from feeling so hopeless to so damned...
                                                              sadness
             im grabbing a minute to post i feel greedy wrong
                                                                anger
                 i am ever feeling nostalgic about the fireplac...
                                                                 love
          4
                                        i am feeling grouchy
                                                                anger
```

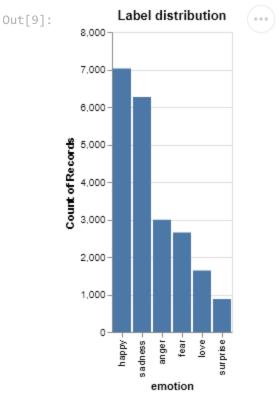
love

Labels distribution

3

i am ever feeling nostalgic about the fireplac...

```
In [9]: hist_labels = alt.Chart(data, title="Label distribution").mark_bar().encode(x = alt.X('encode))
hist_labels
```



Tokens distribution - General

```
2
                  feel
          3 humiliated
          4
          df tokens.shape
In [12]:
          (409965, 1)
Out[12]:
          hist tokens = (alt.Chart(df tokens, title="General Token Distribution")
In [13]:
                           .transform aggregate(
                               count='count()',
                               groupby=['token'])
                           .transform window(
                               rank='rank(count)',
                               sort=[alt.SortField('count', order='descending')])
                           .transform filter(
                               alt.datum.rank < 25)</pre>
                           .mark bar().encode(
                               y='count:Q',
                               x = alt.X('token', sort='-y'),
                               tooltip=['count:Q']))
          hist tokens
                                        General Token Distribution
Out[13]:
            35.000
            30.000 -
            25.000 -
            20,000
            15,000
            10,000-
            5.000-
```

Distribution of tokens per label (with stopwords and with/without duplicates)

token

feeling-

È

of-

```
In [14]: df_tokens_label = data.copy()
    df_tokens_label['text'] = (df_tokens_label['text'].apply(lambda sentence: word_tokenize(
    df_tokens_label = df_tokens_label.explode(['text']).reset_index(drop=True)
    df_tokens_label.columns = ['token', 'label']
In [15]: df_tokens_label.groupby('label').count().sort_values('token', ascending=False)
```

ġ

اً. قٍ

nave-

0

1

i

didnt

```
label
            happy
                   136690
           sadness
                   115719
            anger
                    57049
                    49794
              fear
                    33833
             love
          surprise
                    16880
          df tokens label.drop duplicates().groupby('label').count().sort values('token', ascendin
In [16]:
Out[16]:
                   token
             label
                   10631
            happy
                    9220
           sadness
                    6457
            anger
              fear
                    6044
                    4416
             love
          surprise
                    3102
```

Data Preparation

Out[15]:

In this test we dont will pre-process anything because BERT has default pre-process steps. Before this steps, we will make the MLM pre-processing steps to train the model.

Modeling - Sequence Model

```
In [17]: model_name = 'bert-base-uncased'
    max_length = 512
In [18]: tokenizer = BertTokenizerFast.from_pretrained(model_name, do_lower_case=True, use_cache=
```

Mapping token-index and index-token

```
In [19]: token2idx = tokenizer.get_vocab()
    idx2token = {value:key for key, value in token2idx.items()}

In [20]: test_tensor = tokenizer.encode("i'm working in information retrieval course", return_ten
    list(map(lambda i: idx2token[i], test_tensor[0].tolist()))

Out[20]: ['[CLS]',
    'i',
    "'",
    'working',
    'in',
    'information',
    'retrieval',
```

```
'course',
'[SEP]']
```

Train and validation data split

```
In [21]: data_batch = data.copy()
    data_batch.columns = ['text', 'labels']

In [22]: documents = data_batch[['text']]
    labels = data_batch[['labels']]

# Encoding categorical labels
    le = preprocessing.LabelEncoder()
    label_encoder = le.fit(labels['labels'].tolist())
    labels = label_encoder.transform(labels['labels'])

# Data split 80/20
    train_data, valid_data, train_labels, valid_labels = train_test_split(documents, labels,
```

Tokenize data with truncation and padding

Truncate -> more tokens than max_lenght \ Padding -> less tokens than max_lenght (padding with special_token for padding set in model)

```
In [23]: train_encodings = tokenizer(train_data['text'].tolist(), truncation=True, padding=True,
    valid_encodings = tokenizer(valid_data['text'].tolist(), truncation=True, padding=True,
```

Dataset class definition

```
In [25]: train_dataset = EmotionDataset(train_encodings, train_labels)
  valid_dataset = EmotionDataset(valid_encodings, valid_labels)
```

Load model

```
In [26]: model = BertForSequenceClassification.from_pretrained(model_name, num_labels=len(label_e

Some weights of the model checkpoint at bert-base-uncased were not used when initializin
g BertForSequenceClassification: ['cls.predictions.transform.LayerNorm.bias', 'cls.predi
ctions.bias', 'cls.predictions.decoder.weight', 'cls.predictions.transform.dense.bias',
'cls.seq_relationship.bias', 'cls.predictions.transform.LayerNorm.weight', 'cls.seq_rela
tionship.weight', 'cls.predictions.transform.dense.weight']

- This IS expected if you are initializing BertForSequenceClassification from the checkp
oint of a model trained on another task or with another architecture (e.g. initializing
a BertForSequenceClassification model from a BertForPreTraining model).

- This IS NOT expected if you are initializing BertForSequenceClassification from the ch
```

eckpoint of a model that you expect to be exactly identical (initializing a BertForSeque

```
nceClassification model from a BertForSequenceClassification model). Some weights of BertForSequenceClassification were not initialized from the model checkp oint at bert-base-uncased and are newly initialized: ['classifier.bias', 'classifier.weight']
You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.
```

Training

```
In [27]: def compute metrics(pred):
             labels = pred.label ids
             preds = pred.predictions.argmax(-1)
              # calculate accuracy using sklearn's function
              acc = accuracy score(labels, preds)
              f1 = f1 score(labels, preds, average='macro')
              precision = precision score(labels, preds, average='macro')
              return {
                'accuracy': acc,
                'f1 score': f1,
                'precision': precision
          training args = TrainingArguments(
              output dir='./models/checkpoint', # output directory
              num train epochs=3,
                                       # total number of training epochs
             per device train batch size=16, # batch size per device during training
              per_device_eval_batch_size=20, # batch size for evaluation
              warmup steps=500,
                                                  # number of warmup steps for learning rate schedule
              warmup_steps=500,  # number of warmup steps for learning rate schedule
weight_decay=0.01,  # strength of weight decay
logging_dir='./logs',  # directory for storing logs
load_best_model_at_end=True,  # load the best model when finished training (defau)
              # but you can specify `metric_for_best_model` argument to change to accuracy or othe
                                              # log & save weights each logging steps
              logging steps=500,
              evaluation strategy="steps",
                                                # evaluate each `logging steps`
              save steps=1500,
              learning rate=5e-5,
              seed=42,
          trainer = Trainer(
            model=model,
                                                     # the instantiated Transformers model to be tra
              args=training args,
                                                     # training arguments, defined above
              train dataset=train_dataset,
                                                    # training dataset
              eval dataset=valid dataset,
                                                     # evaluation dataset
              compute metrics=compute metrics, # the callback that computes metrics of interes
```

In [28]: trainer.train()

```
C:\Users\Matheus\anaconda3\lib\site-packages\transformers\optimization.py:306: FutureWar
ning: This implementation of AdamW is deprecated and will be removed in a future versio
n. Use the PyTorch implementation torch.optim.AdamW instead, or set `no_deprecation_warn
ing=True` to disable this warning
  warnings.warn(

***** Running training *****
  Num examples = 17167
  Num Epochs = 3
  Instantaneous batch size per device = 16
  Total train batch size (w. parallel, distributed & accumulation) = 16
  Gradient Accumulation steps = 1
  Total optimization steps = 3219
```

[3219/3219 12:36, Epoch 3/3]

```
1500
                0.186800
                            2000
                0.151200
                            0.924921
        2500
                0.114900
                            3000
                0.087800
                            ***** Running Evaluation *****
         Num examples = 4292
          Batch size = 20
        ***** Running Evaluation *****
         Num examples = 4292
         Batch size = 20
        **** Running Evaluation ****
         Num examples = 4292
         Batch size = 20
        Saving model checkpoint to ./models/checkpoint\checkpoint-1500
        Configuration saved in ./models/checkpoint\checkpoint-1500\config.json
        Model weights saved in ./models/checkpoint\checkpoint-1500\pytorch model.bin
        ***** Running Evaluation *****
          Num examples = 4292
          Batch size = 20
        **** Running Evaluation ****
         Num examples = 4292
         Batch size = 20
        ***** Running Evaluation *****
         Num examples = 4292
          Batch size = 20
        Saving model checkpoint to ./models/checkpoint\checkpoint-3000
        Configuration saved in ./models/checkpoint\checkpoint-3000\config.json
        Model weights saved in ./models/checkpoint\checkpoint-3000\pytorch model.bin
        Training completed. Do not forget to share your model on huggingface.co/models =)
        Loading best model from ./models/checkpoint\checkpoint-3000 (score: 0.1528589129447937).
        TrainOutput(global step=3219, training loss=0.29593553451248933, metrics={'train runtim
        e': 757.9626, 'train samples per second': 67.947, 'train steps per second': 4.247, 'tota
        1 flos': 2302606080827532.0, 'train loss': 0.29593553451248933, 'epoch': 3.0})
In [29]: trainer.evaluate()
        ***** Running Evaluation *****
         Num examples = 4292
          Batch size = 20
                                             [215/215 00:13]
Out[29]: {'eval_loss': 0.1528589129447937,
         'eval accuracy': 0.9394221808014911,
         'eval f1 score': 0.9080865946249088,
         'eval precision': 0.9199482490527001,
         'eval runtime': 13.6386,
         'eval_samples_per_second': 314.695,
         'eval steps per second': 15.764,
         'epoch': 3.0}
```

Save Model

Out[28]:

500

1000

1.007400

0.318700

```
date time = now.strftime("%m-%d-%Y %H-%M-%S")
        model path = "models/emotion-bert-base-uncased %s"%date time
        model.save pretrained(model path)
         tokenizer.save pretrained (model path)
        Configuration saved in models/emotion-bert-base-uncased 11-02-2022 19-45-07\config.json
        Model weights saved in models/emotion-bert-base-uncased 11-02-2022 19-45-07\pytorch mode
        tokenizer config file saved in models/emotion-bert-base-uncased 11-02-2022 19-45-07\toke
        nizer config.json
        Special tokens file saved in models/emotion-bert-base-uncased 11-02-2022 19-45-07\specia
        1 tokens map.json
         ('models/emotion-bert-base-uncased 11-02-2022 19-45-07\\tokenizer config.json',
Out[35]:
          'models/emotion-bert-base-uncased 11-02-2022 19-45-07\\special tokens map.json',
          'models/emotion-bert-base-uncased 11-02-2022 19-45-07\\vocab.txt',
          'models/emotion-bert-base-uncased 11-02-2022 19-45-07\\added tokens.json',
          'models/emotion-bert-base-uncased 11-02-2022 19-45-07\\tokenizer.json')
```

Playground

Getting embeddings from last hidden state of model

Get last hidden state of classification model

```
In []: my_text = "i'm so happy today."
    _input = tokenizer(my_text, padding='max_length', truncation=True, max_length=max_length
    _output = model(**_input, output_hidden_states=True)

In []: # 9 tokens total tokens -> including CLS and SEP.
    tokenizer.tokenize(my_text)

In []: # Last Hidden State Values
    lhs = _output.hidden_states[-1]
    lhs.shape

In []: # Get padding indexes
    attention_mask = _input['attention_mask']

In []: # unsqueeze mask
    mask = attention_mask.unsqueeze(-1).expand(_output.hidden_states[0].shape).float()

In []: # Avoid masked values in embedding by zero this values with lhs*mask
    masked_embeddings = lhs * mask
```

```
In [ ]: masked_embeddings.shape
In []: # Pooling - SUM
        summed = torch.sum(masked embeddings, 1)
        summed.shape
In [ ]: summed.shape
In [ ]: # Pooling - Count
        cnts = torch.clamp(mask.sum(1), min=1e-8)
        cnts.shape
In [ ]: cnts[0][0].item()
In [ ]: # Pooling - Mean
        mean pooled = summed/cnts
        mean pooled.shape
In [ ]: mean_pooled
In [ ]: from sentence_transformers import SentenceTransformer
        model = SentenceTransformer('models/emotion-bert-base-uncased 11-02-2022 18-44-55')
In [ ]: | model.encode(my text)
In [ ]: mean_pooled_cpu = mean pooled.cpu()
In [ ]: from sklearn.metrics.pairwise import cosine similarity
In [ ]: anger_1 = data_batch[data_batch.labels == 'anger']['text'].iloc[0]
        anger 2 = data batch[data batch.labels == 'anger']['text'].iloc[1]
        happy 1 = data batch[data batch.labels == 'happy']['text'].iloc[0]
        happy 2 = data batch[data batch.labels == 'happy']['text'].iloc[1]
        anger 1
In [ ]: |
        anger 2
In [ ]:
        happy 1
In [ ]:
        happy 2
In [ ]:
        cosine similarity([model.encode(happy 1)], [model.encode(anger 2)], dense output=True)
In [ ]:
```

Visualisation t-SNE Bert

```
In [42]: model = SentenceTransformer(model_name)

data_vis = data_batch.head(5000).copy()

X = model.encode(data_vis['text'].tolist())
X_embedded = TSNE(n_components=2).fit_transform(X)

data_vis[['x', 'y']] = X_embedded

No sentence-transformers model found with name C:\Users\Matheus/.cache\torch\sentence_tr
```

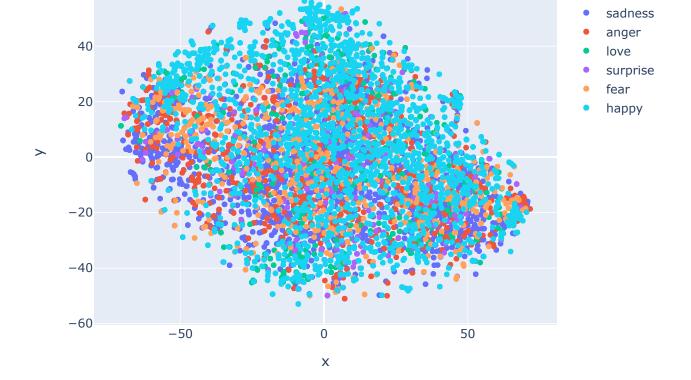
```
ansformers\bert-base-uncased. Creating a new one with MEAN pooling.
loading configuration file C:\Users\Matheus/.cache\torch\sentence transformers\bert-base
-uncased\config.json
Model config BertConfig {
  " name or path": "C:\\Users\\Matheus/.cache\\torch\\sentence transformers\\bert-base-u
ncased",
  "architectures": [
    "BertForMaskedLM"
  "attention probs dropout prob": 0.1,
  "classifier dropout": null,
  "gradient checkpointing": false,
  "hidden act": "gelu",
  "hidden dropout prob": 0.1,
  "hidden size": 768,
  "initializer range": 0.02,
 "intermediate size": 3072,
  "layer norm eps": 1e-12,
  "max position embeddings": 512,
  "model type": "bert",
  "num attention heads": 12,
  "num hidden layers": 12,
  "pad token id": 0,
  "position embedding type": "absolute",
 "transformers version": "4.22.0",
  "type vocab size": 2,
  "use cache": true,
  "vocab size": 30522
loading weights file C:\Users\Matheus/.cache\torch\sentence transformers\bert-base-uncas
ed\pytorch model.bin
Some weights of the model checkpoint at C:\Users\Matheus/.cache\torch\sentence transform
ers\bert-base-uncased were not used when initializing BertModel: ['cls.predictions.trans
form.LayerNorm.bias', 'cls.predictions.bias', 'cls.predictions.decoder.weight', 'cls.pre
dictions.transform.dense.bias', 'cls.seq relationship.bias', 'cls.predictions.transform.
LayerNorm.weight', 'cls.seq relationship.weight', 'cls.predictions.transform.dense.weigh
t']
- This IS expected if you are initializing BertModel from the checkpoint of a model trai
ned on another task or with another architecture (e.g. initializing a BertForSequenceCla
ssification model from a BertForPreTraining model).
- This IS NOT expected if you are initializing BertModel from the checkpoint of a model
that you expect to be exactly identical (initializing a BertForSequenceClassification mo
del from a BertForSequenceClassification model).
All the weights of BertModel were initialized from the model checkpoint at C:\Users\Math
eus/.cache\torch\sentence transformers\bert-base-uncased.
If your task is similar to the task the model of the checkpoint was trained on, you can
already use BertModel for predictions without further training.
loading configuration file C:\Users\Matheus/.cache\torch\sentence transformers\bert-base
-uncased\config.json
Model config BertConfig {
  " name or path": "C:\\Users\\Matheus/.cache\\torch\\sentence transformers\\bert-base-u
ncased",
  "architectures": [
   "BertForMaskedLM"
  "attention probs dropout prob": 0.1,
  "classifier dropout": null,
  "gradient checkpointing": false,
  "hidden act": "gelu",
  "hidden dropout prob": 0.1,
  "hidden size": 768,
  "initializer range": 0.02,
  "intermediate size": 3072,
  "layer norm eps": 1e-12,
  "max position embeddings": 512,
```

```
"num attention heads": 12,
           "num hidden layers": 12,
           "pad token id": 0,
          "position embedding type": "absolute",
          "transformers version": "4.22.0",
          "type vocab size": 2,
          "use cache": true,
           "vocab size": 30522
        loading file vocab.txt
        loading file tokenizer.json
        loading file added tokens.json
        loading file special tokens map.json
        loading file tokenizer config.json
        loading configuration file C:\Users\Matheus/.cache\torch\sentence transformers\bert-base
        -uncased\config.json
        Model config BertConfig {
          " name or path": "C:\\Users\\Matheus/.cache\\torch\\sentence transformers\\bert-base-u
        ncased",
           "architectures": [
             "BertForMaskedLM"
          "attention probs dropout prob": 0.1,
          "classifier dropout": null,
          "gradient checkpointing": false,
          "hidden act": "gelu",
           "hidden dropout prob": 0.1,
           "hidden size": 768,
          "initializer range": 0.02,
          "intermediate size": 3072,
          "layer norm eps": 1e-12,
           "max position embeddings": 512,
          "model type": "bert",
           "num attention heads": 12,
           "num hidden_layers": 12,
          "pad token id": 0,
          "position embedding type": "absolute",
          "transformers version": "4.22.0",
           "type vocab size": 2,
          "use cache": true,
          "vocab size": 30522
        C:\Users\Matheus\anaconda3\lib\site-packages\sklearn\manifold\ t sne.py:800: FutureWarni
        ng: The default initialization in TSNE will change from 'random' to 'pca' in 1.2.
          warnings.warn(
        C:\Users\Matheus\anaconda3\lib\site-packages\sklearn\manifold\ t sne.py:810: FutureWarni
        ng: The default learning rate in TSNE will change from 200.0 to 'auto' in 1.2.
          warnings.warn(
In [43]: fig = px.scatter(
            data vis, x='x', y='y',
             color='labels', labels={'color': 'label'},
            hover data=['text'], title = 'Emotion Visualisation - Before Training')
         fig.show()
```

Emotion Visualisation - Before Training

"model_type": "bert",

60 labels



```
In [46]: model = SentenceTransformer('models/emotion-bert-base-uncased_11-02-2022_19-45-07')

data_vis = data_batch.head(5000).copy()

X = model.encode(data_vis['text'].tolist())

X_embedded = TSNE(n_components=2).fit_transform(X)

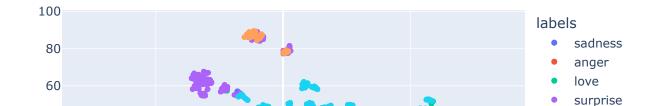
data_vis[['x', 'y']] = X_embedded
```

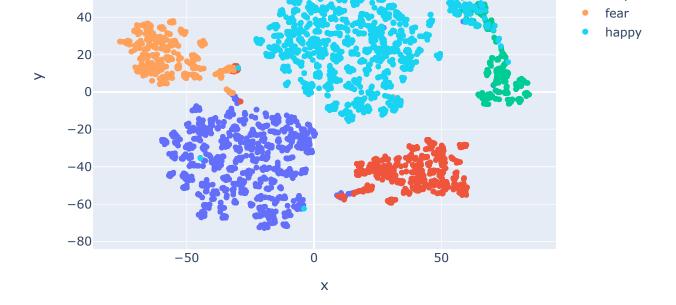
```
No sentence-transformers model found with name models/emotion-bert-base-uncased 11-02-20
22 19-45-07. Creating a new one with MEAN pooling.
loading configuration file models/emotion-bert-base-uncased 11-02-2022 19-45-07\config.j
son
Model config BertConfig {
 " name or path": "models/emotion-bert-base-uncased 11-02-2022 19-45-07",
  "architectures": [
    "BertForSequenceClassification"
 ],
 "attention probs dropout prob": 0.1,
 "classifier dropout": null,
 "gradient checkpointing": false,
 "hidden act": "gelu",
  "hidden dropout prob": 0.1,
  "hidden size": 768,
  "id2label": {
    "0": "LABEL 0",
    "1": "LABEL 1",
    "2": "LABEL 2",
    "3": "LABEL 3",
    "4": "LABEL 4",
    "5": "LABEL 5"
  "initializer range": 0.02,
  "intermediate size": 3072,
  "label2id": {
    "LABEL 0": 0,
    "LABEL 1": 1,
    "LABEL 2": 2,
    "LABEL 3": 3,
```

```
"LABEL 4": 4,
    "LABEL 5": 5
  },
  "layer norm eps": 1e-12,
  "max position embeddings": 512,
  "model type": "bert",
  "num attention heads": 12,
  "num hidden layers": 12,
  "pad token id": 0,
  "position embedding type": "absolute",
  "problem type": "single label classification",
  "torch dtype": "float32",
 "transformers version": "4.22.0",
  "type vocab size": 2,
  "use cache": true,
  "vocab size": 30522
loading weights file models/emotion-bert-base-uncased 11-02-2022 19-45-07\pytorch model.
bin
Some weights of the model checkpoint at models/emotion-bert-base-uncased 11-02-2022 19-4
5-07 were not used when initializing BertModel: ['classifier.bias', 'classifier.weight']
- This IS expected if you are initializing BertModel from the checkpoint of a model trai
ned on another task or with another architecture (e.g. initializing a BertForSequenceCla
ssification model from a BertForPreTraining model).
- This IS NOT expected if you are initializing BertModel from the checkpoint of a model
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All the weights of BertModel were initialized from the model checkpoint at models/emotio
n-bert-base-uncased 11-02-2022 19-45-07.
If your task is similar to the task the model of the checkpoint was trained on, you can
already use BertModel for predictions without further training.
loading file vocab.txt
loading file tokenizer.json
loading file added tokens.json
loading file special tokens map.json
loading file tokenizer config.json
C:\Users\Matheus\anaconda3\lib\site-packages\sklearn\manifold\ t sne.py:800: FutureWarni
ng:
The default initialization in TSNE will change from 'random' to 'pca' in 1.2.
C:\Users\Matheus\anaconda3\lib\site-packages\sklearn\manifold\ t sne.py:810: FutureWarni
ng:
The default learning rate in TSNE will change from 200.0 to 'auto' in 1.2.
   data vis, x='x', y='y',
    color='labels', labels={'color': 'label'},
```

```
In [47]: fig = px.scatter(
            hover data=['text'], title = 'Emotion Visualisation - After Training')
         fig.show()
```

Emotion Visualisation - After Training





Results PNG



alt text