Feature transformation with Amazon SageMaker processing job and Feature Store

Introduction

In this lab you will start with the raw Women's Clothing Reviews dataset and prepare it to train a BERT-based natural language processing (NLP) model. The model will be used to classify customer reviews into positive (1), neutral (0) and negative (-1) sentiment.

You will convert the original review text into machine-readable features used by BERT. To perform the required feature transformation you will configure an Amazon SageMaker processing job, which will be running a custom Python script.

Table of Contents

- 1. Configure the SageMaker Feature Store
 - 1.1. Configure dataset
 - 1.2. Configure the SageMaker feature store
 - Exercise 1
- 2. Transform the dataset
 - Exercise 2
 - Exercise 3
- 3. Query the Feature Store
 - 3.1. Export training, validation, and test datasets from the Feature Store
 - o Exercise 4
 - 3.2. Export TSV from Feature Store
 - 3.3. Check that the dataset in the Feature Store is balanced by sentiment
 - Exercise 5
 - Exercise 6
 - Exercise 7

```
In [1]: # please ignore warning messages during the installation
!pip install --disable-pip-version-check -q sagemaker==2.35.0
!conda install -q -y pytorch==1.6.0 -c pytorch
!pip install --disable-pip-version-check -q transformers==3.5.1
```

```
and conflicting behaviour with the system package manager. It is recommen
        ded to use a virtual environment instead: https://pip.pypa.io/warnings/ve
        Collecting package metadata (current_repodata.json): ...working... done
        Solving environment: ...working... done
        ## Package Plan ##
          environment location: /opt/conda
          added / updated specs:
            - pytorch==1.6.0
        The following packages will be UPDATED:
                             conda-forge::ca-certificates-2022.6.1~ --> pkgs/main
          ca-certificates
        ::ca-certificates-2022.07.19-h06a4308 0
        The following packages will be SUPERSEDED by a higher-priority channel:
                             conda-forge::conda-4.14.0-py37h89c186~ --> pkgs/main
          conda
        ::conda-4.14.0-py37h06a4308 0
        Preparing transaction: ...working... done
        Verifying transaction: ...working... done
        Executing transaction: ...working... done
        Retrieving notices: ...working... done
        WARNING: Running pip as the 'root' user can result in broken permissions
        and conflicting behaviour with the system package manager. It is recommen
        ded to use a virtual environment instead: https://pip.pypa.io/warnings/ve
In [2]: import boto3
        import sagemaker
        import botocore
        config = botocore.config.Config(user_agent_extra='dlai-pds/c2/w1')
        # low-level service client of the boto3 session
        sm = boto3.client(service name='sagemaker',
                          config=config)
        featurestore runtime = boto3.client(service name='sagemaker-featurestore-
                                            config=config)
        sess = sagemaker.Session(sagemaker client=sm,
                                 sagemaker featurestore runtime client=featuresto
        bucket = sess.default bucket()
        role = sagemaker.get execution role()
        region = sess.boto_region_name
```

WARNING: Running pip as the 'root' user can result in broken permissions

1. Configure the SageMaker Feature Store

1.1. Configure dataset

The raw dataset is in the public S3 bucket. Let's start by specifying the S3 location of it:

```
In [3]: raw_input_data_s3_uri = 's3://dlai-practical-data-science/data/raw/'
    print(raw_input_data_s3_uri)

s3://dlai-practical-data-science/data/raw/
```

List the files in the S3 bucket (in this case it will be just one file):

```
In [4]: !aws s3 ls $raw_input_data_s3_uri
2021-04-30 02:21:06 8457214 womens_clothing_ecommerce_reviews.csv
```

1.2. Configure the SageMaker feature store

As the result of the transformation, in addition to generating files in S3 bucket, you will also save the transformed data in the **Amazon SageMaker Feature Store** to be used by others in your organization, for example.

To configure a Feature Store you need to setup a **Feature Group**. This is the main resource containing all of the metadata related to the data stored in the Feature Store. A Feature Group should contain a list of **Feature Definitions**. A Feature Definition consists of a name and the data type. The Feature Group also contains an online store configuration and an offline store configuration controlling where the data is stored. Enabling the online store allows quick access to the latest value for a record via the GetRecord API. The offline store allows storage of the data in your S3 bucket. You will be using the offline store in this lab.

Let's setup the Feature Group name and the Feature Store offline prefix in S3 bucket (you will use those later in the lab):

```
In [5]: import time
   timestamp = int(time.time())

feature_group_name = 'reviews-feature-group-' + str(timestamp)
   feature_store_offline_prefix = 'reviews-feature-store-' + str(timestamp)

print('Feature group name: {}'.format(feature_group_name))
   print('Feature store offline prefix in S3: {}'.format(feature_store_offli)

Feature group name: reviews-feature-group-1661803330
```

Feature group name: reviews-feature-group-1661803330 Feature store offline prefix in S3: reviews-feature-store-1661803330

Taking two features from the original raw dataset (Review Text and Rating), you will transform it preparing to be used for the model training and then to be saved in the Feature Store. Here you will define the related features to be stored as a list of FeatureDefinition.

```
In [6]: from sagemaker.feature_store.feature_definition import (
            FeatureDefinition,
            FeatureTypeEnum,
        feature definitions= [
            # unique ID of the review
            FeatureDefinition(feature_name='review_id', feature_type=FeatureTypeE
            # ingestion timestamp
            FeatureDefinition(feature_name='date', feature_type=FeatureTypeEnum.S
            # sentiment: -1 (negative), 0 (neutral) or 1 (positive). It will be f
            FeatureDefinition(feature_name='sentiment', feature_type=FeatureTypeE
            # label ID of the target class (sentiment)
            FeatureDefinition(feature_name='label_id', feature_type=FeatureTypeEn
            # reviews encoded with the BERT tokenizer
            FeatureDefinition(feature name='input ids', feature type=FeatureTypeE
            # original Review Text
            FeatureDefinition(feature name='review body', feature type=FeatureTyp
            # train/validation/test label
            FeatureDefinition(feature_name='split_type', feature_type=FeatureType
        ]
```

Exercise 1

Create the feature group using the feature definitions defined above.

Instructions: Use the FeatureGroup function passing the defined above feature group name and the feature definitions.

```
feature_group = FeatureGroup(
    name=..., # Feature Group name
    feature_definitions=..., # a list of Feature Definitions
    sagemaker_session=sess # SageMaker session
)
```

```
In [7]: from sagemaker.feature_store.feature_group import FeatureGroup

feature_group = FeatureGroup(
    ### BEGIN SOLUTION - DO NOT delete this comment for grading purposes
    name=feature_group_name, # Replace None
    feature_definitions=feature_definitions, # Replace None
    ### END SOLUTION - DO NOT delete this comment for grading purposes
    sagemaker_session=sess
)

print(feature_group)
```

FeatureGroup(name='reviews-feature-group-1661803330', sagemaker_session=< sagemaker.session.Session object at 0x7fd212934e10>, feature_definitions= [FeatureDefinition(feature_name='review_id', feature_type=<FeatureTypeEnu m.STRING: 'String'>), FeatureDefinition(feature_name='date', feature_type=<FeatureTypeEnum.STRING: 'String'>), FeatureDefinition(feature_name='sen timent', feature_type=<FeatureTypeEnum.STRING: 'String'>), FeatureDefinit ion(feature_name='label_id', feature_type=<FeatureTypeEnum.STRING: 'String'>), FeatureDefinition(feature_name='input_ids', feature_type=<FeatureTypeEnum.STRING: 'String'>), FeatureDefinition(feature_name='review_body', feature_type=<FeatureTypeEnum.STRING: 'String'>), FeatureDefinition(feature_name='review_body', feature_type=<FeatureTypeEnum.STRING: 'String'>)])

You will use the defined Feature Group later in this lab, the actual creation of the Feature Group will take place in the processing job. Now let's move into the setup of the processing job to transform the dataset.

2. Transform the dataset

You will configure a SageMaker processing job to run a custom Python script to balance and transform the raw data into a format used by BERT model.

Set the transformation parameters including the instance type, instance count, and train/validation/test split percentages. For the purposes of this lab, you will use a relatively small instance type. Please refer to this link for additional instance types that may work for your use case outside of this lab.

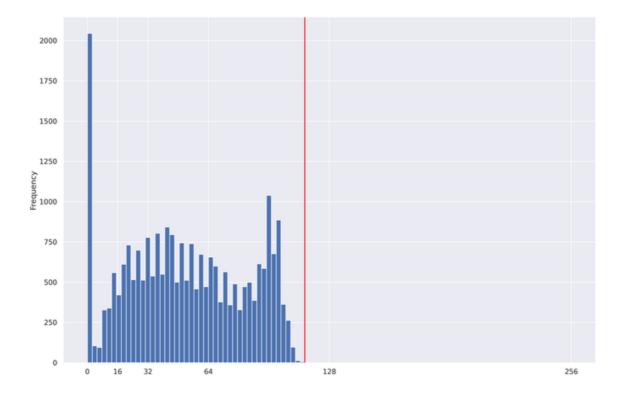
You can also choose whether you want to balance the dataset or not. In this case, you will balance the dataset to avoid class imbalance in the target variable, sentiment.

Another important parameter of the model is the <code>max_seq_length</code>, which specifies the maximum length of the classified reviews for the RoBERTa model. If the sentence is shorter than the maximum length parameter, it will be padded. In another case, when the sentence is longer, it will be truncated from the right side.

Since a smaller <code>max_seq_length</code> leads to faster training and lower resource utilization, you want to find the smallest power-of-2 that captures <code>100%</code> of our reviews. For this dataset, the <code>100th</code> percentile is <code>115</code>. However, it's best to stick with powers-of-2 when using BERT. So let's choose <code>128</code> as this is the smallest power-of-2 greater than <code>115</code>. You will see below how the shorter sentences will be padded to a maximum length.

mean	52.512374
std	31.387048
min	1.000000
10%	10.000000
20%	22.000000
30%	32.000000

```
40%
             41.000000
50%
             51.000000
60%
             61.000000
70%
             73.000000
80%
             88.000000
90%
             97.000000
100%
            115.000000
max
            115.000000
```



```
In [8]: processing_instance_type='ml.c5.xlarge'
    processing_instance_count=1
    train_split_percentage=0.90
    validation_split_percentage=0.05
    test_split_percentage=0.05
    balance_dataset=True
    max_seq_length=128
```

To balance and transform our data, you will use a scikit-learn-based processing job. This is essentially a generic Python processing job with scikit-learn pre-installed. You can specify the version of scikit-learn you wish to use. Also pass the SageMaker execution role, processing instance type and instance count.

```
In [9]: from sagemaker.sklearn.processing import SKLearnProcessor

processor = SKLearnProcessor(
    framework_version='0.23-1',
    role=role,
    instance_type=processing_instance_type,
    instance_count=processing_instance_count,
    env={'AWS_DEFAULT_REGION': region},
    max_runtime_in_seconds=7200
)
```

The processing job will be running the Python code from the file src/prepare_data.py. In the following exercise you will review the contents of the file and familiarize yourself with main parts of it.

Exercise 2

- 1. Open the file src/prepare_data.py. Go through the comments to understand its content.
- 2. Find and review the convert_to_bert_input_ids() function, which contains the RoBERTa tokenizer configuration.
- 3. Complete method encode_plus of the RoBERTa tokenizer . Pass the max_seq_length as a value for the argument max_length . It defines a pad to a maximum length specified.
- 4. Save the file src/prepare_data.py (with the menu command File -> Save Python File).

This cell will take approximately 1-2 minutes to run.

```
In [10]:
       import sys, importlib
       sys.path.append('src/')
       # import the `prepare data.py` module
       import prepare data
       # reload the module if it has been previously loaded
       if 'prepare data' in sys.modules:
           importlib.reload(prepare_data)
       input ids = prepare data.convert to bert input ids("this product is great
       updated_correctly = False
       if len(input_ids) != max_seq_length:
           print('Please check that the function \'convert to bert input ids\' i
           raise Exception('Please check that the function \'convert to bert inp
       else:
           print('###########")
           print('Updated correctly!')
           print('############")
           updated_correctly = True
```

Review the results of tokenization for the given example (|"this product is great!|"):

Launch the processing job with the custom script passing defined above parameters.

```
In [12]: from sagemaker.processing import ProcessingInput, ProcessingOutput
         if (updated_correctly):
             processor.run(code='src/prepare data.py',
                       inputs=[
                             ProcessingInput(source=raw input data s3 uri,
                                            destination='/opt/ml/processing/input
                                            s3_data_distribution_type='ShardedByS
                       ],
                       outputs=[
                             ProcessingOutput(output name='sentiment-train',
                                             source='/opt/ml/processing/output/se
                                             s3 upload mode='EndOfJob'),
                             ProcessingOutput(output_name='sentiment-validation',
                                             source='/opt/ml/processing/output/se
                                             s3 upload mode='EndOfJob'),
                             ProcessingOutput(output name='sentiment-test',
                                             source='/opt/ml/processing/output/se
                                             s3_upload_mode='EndOfJob')
                       ],
                       arguments=['--train-split-percentage', str(train split perc
                                  '--validation-split-percentage', str(validation_
                                  '--test-split-percentage', str(test_split_percen
                                  '--balance-dataset', str(balance_dataset),
                                  '--max-seq-length', str(max_seq_length),
                                  '--feature-store-offline-prefix', str(feature_st
                                  '--feature-group-name', str(feature group name)
                       ],
                       logs=True,
                       wait=False)
         else:
             print('##############")
             print('Please update the code correctly above.')
             print('###############")
```

Job Name: sagemaker-scikit-learn-2022-08-29-20-16-18-304 Inputs: [{'InputName': 'input-1', 'AppManaged': False, 'S3Input': {'S3Ur i': 's3://dlai-practical-data-science/data/raw/', 'LocalPath': '/opt/ml/p rocessing/input/data/', 'S3DataType': 'S3Prefix', 'S3InputMode': 'File', 'S3DataDistributionType': 'ShardedByS3Key', 'S3CompressionType': 'None'}}, {'InputName': 'code', 'AppManaged': False, 'S3Input': {'S3Uri': 's3://s agemaker-us-east-1-504458356961/sagemaker-scikit-learn-2022-08-29-20-16-1 8-304/input/code/prepare_data.py', 'LocalPath': '/opt/ml/processing/input /code', 'S3DataType': 'S3Prefix', 'S3InputMode': 'File', 'S3DataDistribut ionType': 'FullyReplicated', 'S3CompressionType': 'None'}}] Outputs: [{'OutputName': 'sentiment-train', 'AppManaged': False, 'S3Outp ut': {'S3Uri': 's3://sagemaker-us-east-1-504458356961/sagemaker-scikit-le arn-2022-08-29-20-16-18-304/output/sentiment-train', 'LocalPath': '/opt/m l/processing/output/sentiment/train', 'S3UploadMode': 'EndOfJob'}}, {'Out putName': 'sentiment-validation', 'AppManaged': False, 'S3Output': {'S3Ur i': 's3://sagemaker-us-east-1-504458356961/sagemaker-scikit-learn-2022-08 -29-20-16-18-304/output/sentiment-validation', 'LocalPath': '/opt/ml/proc essing/output/sentiment/validation', 'S3UploadMode': 'EndOfJob'}}, {'Outp utName': 'sentiment-test', 'AppManaged': False, 'S3Output': {'S3Uri': 's3 ://sagemaker-us-east-1-504458356961/sagemaker-scikit-learn-2022-08-29-20-16-18-304/output/sentiment-test', 'LocalPath': '/opt/ml/processing/output /sentiment/test', 'S3UploadMode': 'EndOfJob'}}]

You can see the information about the processing jobs using the describe function. The result is in dictionary format. Let's pull the processing job name:

Exercise 3

Pull the processing job status from the processing job description.

Instructions: Print the keys of the processing job description dictionary, choose the one related to the status of the processing job and print the value of it.

Processing job status: InProgress

Review the created processing job in the AWS console.

Instructions:

- open the link
- notice that you are in the section Amazon SageMaker -> Processing jobs
- check the name of the processing job, its status and other available information

```
In [17]: from IPython.core.display import display, HTML
    display(HTML('<b>Review <a target="blank" href="https://console.aws.amazo")</pre>
```

Review processing job

Wait for about 5 minutes to review the CloudWatch Logs. You may open the file src/prepare_data.py again and examine the outputs of the code in the CloudWatch logs.

```
In [18]: from IPython.core.display import display, HTML
    display(HTML('<b>Review <a target="blank" href="https://console.aws.amazo")</pre>
```

Review CloudWatch logs after about 5 minutes

After the completion of the processing job you can also review the output in the S3 bucket.

```
In [19]: from IPython.core.display import display, HTML
    display(HTML('<b>Review <a target="blank" href="https://s3.console.aws.am")</pre>
```

Review \$3 output data after the processing job has completed

Wait for the processing job to complete.

This cell will take approximately 15 minutes to run.

Please wait until ^^ Processing Job ^^ completes above

Inspect the transformed and balanced data in the S3 bucket.

```
In [21]: processing_job_description = running_processor.describe()
         output_config = processing_job_description['ProcessingOutputConfig']
          for output in output_config['Outputs']:
             if output['OutputName'] == 'sentiment-train':
                 processed_train_data_s3_uri = output['S3Output']['S3Uri']
             if output['OutputName'] == 'sentiment-validation':
                  processed_validation_data_s3_uri = output['S3Output']['S3Uri']
              if output['OutputName'] == 'sentiment-test':
                 processed_test_data_s3_uri = output['S3Output']['S3Uri']
         print(processed train data s3 uri)
         print(processed validation data s3 uri)
         print(processed_test_data_s3_uri)
         s3://sagemaker-us-east-1-504458356961/sagemaker-scikit-learn-2022-08-29-2
         0-16-18-304/output/sentiment-train
         s3://sagemaker-us-east-1-504458356961/sagemaker-scikit-learn-2022-08-29-2
         0-16-18-304/output/sentiment-validation
         s3://sagemaker-us-east-1-504458356961/sagemaker-scikit-learn-2022-08-29-2
         0-16-18-304/output/sentiment-test
In [22]:
         !aws s3 ls $processed train data s3 uri/
         2022-08-29 20:30:16
                                4890076 part-algo-1-womens clothing ecommerce revi
         ews.tsv
In [23]:
         !aws s3 ls $processed_validation_data_s3_uri/
         2022-08-29 20:30:17
                                  264236 part-algo-1-womens_clothing_ecommerce_revi
         ews.tsv
In [24]: !aws s3 ls $processed test data s3 uri/
         2022-08-29 20:30:17
                                  273558 part-algo-1-womens clothing ecommerce revi
         ews.tsv
         Copy the data into the folder balanced.
In [25]:
         laws s3 cp $processed train data s3 uri/part-algo-1-womens clothing ecomm
          laws s3 cp $processed_validation_data_s3_uri/part-algo-1-womens_clothing_
          laws s3 cp $processed test_data_s3_uri/part-algo-1-womens_clothing_ecomme
         download: s3://sagemaker-us-east-1-504458356961/sagemaker-scikit-learn-20
         22-08-29-20-16-18-304/output/sentiment-train/part-algo-1-womens clothing
         ecommerce reviews.tsv to balanced/sentiment-train/part-algo-1-womens clot
         hing_ecommerce_reviews.tsv
         download: s3://sagemaker-us-east-1-504458356961/sagemaker-scikit-learn-20
         22-08-29-20-16-18-304/output/sentiment-validation/part-algo-1-womens_clot
         hing_ecommerce_reviews.tsv to balanced/sentiment-validation/part-algo-1-w
         omens clothing ecommerce reviews.tsv
         download: s3://sagemaker-us-east-1-504458356961/sagemaker-scikit-learn-20
         22-08-29-20-16-18-304/output/sentiment-test/part-algo-1-womens_clothing_e
         commerce_reviews.tsv to balanced/sentiment-test/part-algo-1-womens_clothi
         ng ecommerce reviews.tsv
         Review the training, validation and test data outputs:
```

!head -n 5 ./balanced/sentiment-train/part-algo-1-womens_clothing_ecommer

In [26]:

review id sentiment label id input ids review_body date [0, 133, 1521, 1415, 2422, 11962, 804, 1437, 53, 3343 0 1 11, 621, 1437, 5, 1468, 2551, 6162, 4, 888, 1437, 23, 513, 11, 5, 2272, 6 5, 1437, 5, 1468, 16, 430, 11, 5, 29874, 8, 2705, 9042, 131, 114, 24, 56, 70, 57, 5, 29874, 1468, 1437, 24, 74, 33, 57, 203, 35735, 4, 2, 1, 1, 1, The design looked super cute online but in person the material seemed c heap. actually at least in the green one the material is different in t he striped and solid sections; if it had all been the striped material i t would have been much nicer. 2022-08-29T20:24:48Z 1154 [0, 100, 2740, 209, 11, 127, 6097, 1836, 1437, 97 2 3, 1437, 8, 51, 2564, 6128, 6683, 4682, 5, 17507, 4, 13, 162, 1437, 939, 303, 106, 7, 28, 350, 3229, 11, 14, 443, 4, 939, 439, 7, 5, 1400, 8, 1381 , 15, 5, 974, 8, 51, 58, 350, 380, 11, 5, 13977, 4, 939, 109, 5940, 106, 600, 4, 51, 32, 157, 156, 1437, 939, 2638, 5, 10397, 8, 5, 5933, 21, 1969 , 4, 145, 195, 108, 246, 113, 51, 376, 7, 5, 795, 7451, 8, 14, 21, 2051, 19, 162, 4, 492, 106, 10, 860, 1437, 5952, 51, 173, 357, 13, 47, 4, 2, 1, "I ordered these in my typical size 26 and they fit everywhere perfectl y except the butt. for me i found them to be too tight in that area. i w ent to the store and tried on the 27 and they were too big in the waist. i do recommend them though. they are well made i loved the wash and the length was perfect. being 5'3"" they came to the lower ankle and that was fine with me. give them a try hopefully they work better for you." 2022-08-29T20:24:48Z [0, 100, 21, 182, 6649, 7, 1325, 42, 3588, 19, 10 , 36069, 43805, 1437, 1326, 101, 22196, 219, 1104, 18309, 479, 192, 7391, 3493, 4, 5, 3588, 10199, 16, 182, 2579, 8, 939, 222, 860, 24, 15, 8, 24, 16, 10, 182, 2579, 2564, 1437, 9321, 19, 10, 410, 4140, 53, 127, 821, 521 2, 99, 10, 6162, 2129, 1318, 43805, 14, 1326, 101, 24, 21, 793, 8, 4924, 7, 16227, 142, 5, 4204, 1667, 33, 10, 1104, 22196, 219, 6572, 328, 6215, 47, 269, 240, 7, 1649, 110, 12952, 137, 3981, 7, 110, 8259, 916, 4, 12056 I was very shocked to receive this dress 1, 1, 1, 1, 1, 1, 1, 1, 1, 1] with a rusty zipper looks like crusty white rust . see attached pictures . the dress fabric is very nice and i did try it on and it is a very nice lined with a little stretch but my gosh what a cheap poor quality zi pper that looks like it was old and exposed to moisture because the metal parts have a white crusty substance! retailer you really need to check yo ur merchandise before sending to your loyal customers. lately the quality 2022-08-29T20:24:48Z has been declining a 1465 [0, 713, 3588, 16, 182, 11962, 8, 10698, 1341, 15 1 7, 4, 47, 218, 75, 33, 7, 3568, 10, 11689, 19, 24, 1437, 61, 16, 6344, 14 37, 8, 24, 34, 24897, 12189, 4, 5, 13422, 16, 14, 5, 10199, 15, 5, 299, 4 57, 16, 10, 410, 7735, 4, 24, 16, 182, 2016, 8, 13116, 4, 67, 1437, 5, 10 25, 34, 103, 6664, 11918, 8, 5, 1318, 473, 45, 2045, 7, 28, 5, 275, 4, 93 , 1, 1, 1, 1, 1, 1, 1, 1] This dress is very cute and fits quite we 11. you don't have to wear a bra with it which is awesome and it has de lightful pockets. the downside is that the fabric on the top half is a li ttle weird. it is very heavy and stiff. also the inside has some fraying and the quality does not seem to be the best. i'm keeping it though. 2022-08-29T20:24:48Z

review id sentiment label id input ids review_body date [0, 1708, 1437, 24, 95, 630, 75, 2450, 62, 4, 78, 15360 -1 0 1437, 5, 9215, 73, 16526, 606, 23983, 12552, 114, 2468, 2829, 8, 24, 16, 182, 1202, 7, 888, 120, 24, 15, 111, 1169, 8296, 88, 24, 50, 6539, 24, 81 , 110, 471, 4, 8, 1437, 683, 15, 1437, 5, 299, 21, 10, 380, 41783, 8, 89, 21, 350, 203, 1468, 23, 5, 2576, 61, 156, 24, 356, 269, 35156, 4, 9574, 1 437, 939, 524, 2609, 42, 7, 28, 5, 403, 19, 98, 171, 9, 5, 1964, 939, 907 , 31, 6215, 12056, 4, 5, 1468, 16, 269, 11962, 1437, 53, 5, 3588, 965, 75 , 966, 5, 425, 1589, 1351, 4, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1 But it just doesn't measure up. first the slip/lining c omes unattached if pulled slightly and it is very difficult to actually g et it on - either stepping into it or pulling it over your head. and onc e on the top was a big snug and there was too much material at the botto m which made it look really bulky. unfortunately i am finding this to be the case with so many of the items i buy from retailer lately. the materi al is really cute but the dress isn't worth the price / effort. 2022-08-29T20:24:48Z 7838 [0, 713, 5780, 16, 98, 3035, 328, 5, 6399, 16, 95 1 , 10, 410, 7082, 98, 645, 110, 4505, 1836, 8, 24, 40, 2564, 6683, 4, 2, 1 This print is so cool! the shirt is just a little loose s o order your usual size and it will fit perfectly. 2022-08-29T20:24: 48Z 19592 [0, 100, 437, 4716, 1459, 1437, 8, 939, 1467, 215 -19, 42, 14, 24, 74, 28, 10, 380, 23204, 111, 14, 16, 5, 356, 14, 16, 15266 , 15, 5, 1421, 4, 5, 23204, 16, 12272, 219, 1437, 5, 5397, 21, 98, 380, 8 , 35156, 939, 56, 3605, 442, 24, 4477, 235, 4, 24, 1326, 372, 15, 5, 1421 , 259, 1437, 53, 45, 15, 162, 4, 164, 124, 4, 98, 29673, 4567, 4, 5, 3195 , 12339, 16, 1528, 7, 99, 16, 2343, 8, 24, 34, 410, 22, 31670, 113, 50, 1 8459, 420, 5, 10762, 4, 98, 89, 16, 10, 5262, 828, 9, 4617, 4, 1374, 1437 1, 1, 1 "I'm petite and i knew buying this that it would be a bi g sweater - that is the look that is portrayed on the model. the sweater is scratchy the neck was so big and bulky i had trouble making it lay ri ght. it looks great on the model here but not on me. going back. so bumm ed. the color gray is true to what is shown and it has little ""holes"" o r openings across the shoulders. so there is a tiny bit of detail. overal l not a great sweater." 2022-08-29T20:24:48Z 9833 [0, 713, 21, 95, 45, 34203, 13, 162, 5579, 13724, 196, 350, 203, 101, 10, 21592, 3588, 4, 939, 64, 75, 269, 109, 10, 12138, 5357, 13977, 98, 2532, 14, 74, 146, 42, 3588, 55, 16106, 13, 951, 1493, 4 , 939, 21, 6661, 7, 33, 7, 671, 142, 24, 1411, 157, 19, 6255, 1536, 8, 10 1] This was just not flattering for me--looked too much like a mater nity dress. i can't really do a belted waist so perhaps that would make t his dress more versatile for someone else. i was sorry to have to return

2022-08-29T20:24:48Z

because it goes well with sandals and boots!

review id sentiment label id input ids review body date [0, 133, 23204, 16, 182, 3137, 24382, 8, 1415, 20 13528 -1 0 5, 5, 78, 86, 939, 1381, 24, 15, 1437, 959, 5, 1468, 16, 7174, 8, 2829, 1 92, 149, 4, 67, 1437, 71, 65, 3568, 24, 2198, 13596, 66, 8, 156, 5, 251, 21764, 11789, 7, 3568, 396, 14784, 11, 227, 349, 3568, 4, 71, 5, 371, 86, 939, 5328, 24, 1437, 939, 2967, 10, 739, 4683, 11, 5, 4709, 6195, 405, 14 37, 144, 533, 528, 7, 5, 10079, 14784, 4, 939, 74, 45, 5940, 42, 23204, 4 The sweater is ve ry comfy and looked good the first time i tried it on however the materi al is thin and slightly see through. also after one wear it completely s tretched out and made the long sleeves awkward to wear without washing in between each wear. after the third time i wore it i discovered a large h ole in the armpit most likely due to the excessive washing. i would not recommend this sweater. 2022-08-29T20:24:48Z [0, 100, 1467, 939, 770, 42, 299, 5, 1151, 939, 7 1089 1 2 94, 24, 1437, 98, 2500, 127, 2813, 8458, 24, 439, 454, 24, 439, 15, 1392, 328, 24, 10698, 16467, 11, 127, 2340, 1836, 8, 939, 657, 14, 24, 416, 34, 10, 22220, 6013, 299, 4, 939, 120, 33391, 8378, 939, 3568, 24, 12, 206, 1 368, 417, 11, 2242, 354, 429, 28, 127, 92, 213, 12, 560, 328, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1] I knew i wanted this top the moment i saw it so onto my wishlist it went until it went on sale! it fits beautifully in my normal size and i love that it already has a coordinating tank top. i get compliments whenever i wear it- think hd in paris might be my new go-to! 2022-08-29T20:24:48Z [0, 16587, 5, 3793, 10199, 8, 1365, 2564, 9, 42, 3588, 4, 939, 437, 4343, 14, 24, 473, 492, 10, 11602, 9, 65, 18, 1955, 8, 630, 75, 95, 6713, 101, 10, 14072, 4, 939, 303, 5, 2440, 7, 28, 144, 3420 3, 8, 21, 816, 19, 23256, 23, 184, 8, 303, 14, 5, 6012, 6215, 22, 34108, 1409, 113, 9219, 7494, 851, 24, 10, 1086, 92, 356, 4, 2, 1, 1, 1, 1, 1 , 1, 1, 1, 1] "Love the soft fabric and easy fit of this dress. i'm ple ased that it does give a hint of one's figure and doesn't just hang like a sack. i found the blue to be most flattering and was playing with styli ng at home and found that the wider retailer ""tabby"" leather belt gave it a whole new look." 2022-08-29T20:24:48Z [0, 100, 2638, 209, 9304, 8, 3584, 106, 11, 130, -1 8089, 4, 101, 277, 37102, 1437, 624, 10, 367, 688, 1437, 519, 10610, 106, 117, 55, 87, 10, 891, 9, 498, 1437, 5, 10199, 11, 5, 42613, 443, 554, 7, 3568, 7174, 8, 80, 15029, 2226, 6538, 4, 939, 21, 45, 543, 15, 5, 9304, 1 11, 95, 5328, 106, 7, 5, 558, 111, 8, 516, 16380, 106, 4, 98, 6770, 1437, 25, 939, 269, 222, 657, 5, 9304, 4, 746, 3844, 9, 418, 4, 2, 1, 1, 1, 1, I loved these pants and p urchased them in three colors. like another reviewer within a few weeks having worn them no more than a couple of times the fabric in the crotch

area started to wear thin and two pairs developed holes. i was not hard o n the pants - just wore them to the office - and line dried them. so disa

ppointing as i really did love the pants. total waste of money.

2022-08-29T20:24:48Z

3. Query the Feature Store

In addition to transforming the data and saving in S3 bucket, the processing job populates the feature store with the transformed and balanced data. Let's query this data using Amazon Athena.

3.1. Export training, validation, and test datasets from the Feature Store

Here you will do the export only for the training dataset, as an example.

Use athena_query() function to create an Athena query for the defined above Feature Group. Then you can pull the table name of the Amazon Glue Data Catalog table which is auto-generated by Feature Store.

```
In [29]: feature_store_query = feature_group.athena_query()
         feature store table = feature store query.table name
         query_string = """
             SELECT date,
                 review id,
                 sentiment,
                 label_id,
                 input ids,
                 review_body
             FROM "{}"
             WHERE split_type='train'
          """.format(feature_store_table)
         print('Glue Catalog table name: {}'.format(feature store table))
         print('Running query: {}'.format(query_string))
         Glue Catalog table name: reviews-feature-group-1661803330-1661804658
         Running query:
             SELECT date,
                 review id,
                 sentiment,
                 label id,
                 input_ids,
                 review body
             FROM "reviews-feature-group-1661803330-1661804658"
             WHERE split type='train'
             LIMIT 5
```

Configure the S3 location for the query results. This allows us to re-use the query results for future queries if the data has not changed. We can even share this S3 location between team members to improve query performance for common queries on data that does not change often.

```
In [30]: output_s3_uri = 's3://{}/query_results/{}/'.format(bucket, feature_store_print(output_s3_uri)

s3://sagemaker-us-east-1-504458356961/query_results/reviews-feature-store -1661803330/
```

Exercise 4

Query the feature store.

Instructions: Use feature_store_query.run function passing the constructed above query string and the location of the output S3 bucket.

```
feature_store_query.run(
    query_string=..., # query string
    output_location=... # location of the output S3 bucket
)
```

```
In [31]: feature_store_query.run(
    ### BEGIN SOLUTION - DO NOT delete this comment for grading purposes
    query_string=query_string, # Replace None
    output_location=output_s3_uri # Replace None
    ### END SOLUTION - DO NOT delete this comment for grading purposes
)

feature_store_query.wait()
```

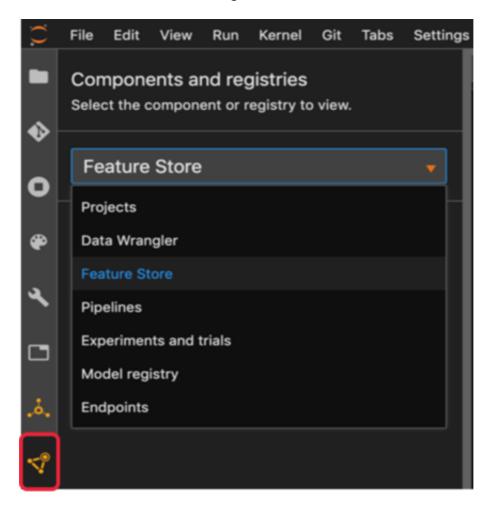
```
In [32]: import pandas as pd
pd.set_option("max_colwidth", 100)

df_feature_store = feature_store_query.as_dataframe()
df_feature_store
```

Out[32]:	
----------	--

	date	review_id	sentiment	label_id	input_ids	review_body
0	2022-08- 29T20:24:48Z	7558	1	2	[0, 41541, 1318, 1437, 1254, 1437, 8, 939, 101, 5, 4600, 19780, 847, 36, 11655, 157, 15, 42, 308	Nice quality details and i like the looser cut (works well on this blouse).
1	2022-08- 29T20:24:48Z	3803	0	1	[0, 35703, 10, 1256, 475, 4791, 548, 6399, 14, 888, 2564, 127, 36, 3340, 45314, 43, 3065, 1368,	Finally a pretty maeve shirt that actually fit my (apparently) giant hulk arms but it's so flo
2	2022-08- 29T20:24:48Z	3316	1	2	[0, 597, 2629, 6683, 328, 2422, 3793, 8, 157, 12, 7078, 328, 657, 5, 155, 73, 306, 21764, 8, 124	Fits perfectly! super soft and well-made! love the 3/4 sleeves and back drop. i can expect i wil
3	2022-08- 29T20:24:48Z	4699	1	2	[0, 100, 657, 42, 23204, 4, 24, 18, 3279, 1437, 3473, 1437, 8, 3793, 4, 2579, 1318, 350, 4, 2, 1	I love this sweater. it's warm comfortable and soft. nice quality too.
4	2022-08- 29T20:24:48Z	16894	0	1	[0, 100, 101, 42, 299, 53, 9574, 939, 40, 28, 3357, 4, 24, 18, 169, 350, 251, 4, 5, 3195, 16, 12	I like this top but unfortunately i will be returning. it's way too long. the color is pretty as

Review the Feature Store in SageMaker Studio



3.2. Export TSV from Feature Store

Save the output as a TSV file:

date review id sentiment label id input ids review body 2022-08-29T20:24:48Z 7558 [0, 41541, 1318, 1437, 12 1 2 54, 1437, 8, 939, 101, 5, 4600, 19780, 847, 36, 11655, 157, 15, 42, 3089, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1] Nice quality details and i like the looser cut (works well on this blouse). 2022-08-29T20:24:48Z [0, 35703, 10, 1256, 475, 3803 0 4791, 548, 6399, 14, 888, 2564, 127, 36, 3340, 45314, 43, 3065, 1368, 162 60, 3701, 7586, 53, 24, 18, 98, 3041, 219, 73, 32123, 219, 939, 56, 7, 67 1, 24, 4, 42, 429, 95, 28, 10, 936, 19, 145, 10, 2514, 6429, 1437, 53, 5, 7400, 10199, 156, 162, 619, 101, 939, 21, 2498, 10, 10178, 8, 156, 162, 3 56, 158, 1178, 2671, 87, 939, 21, 198, 127, 13977, 73, 7903, 1437, 190, 7 7, 9320, 66, 19, 2084, 6149, 1033, 4, 939, 802, 59, 667, 7, 1836, 159, 53 , 938, 75, 686, 114, 14, 74, 190, 6136, 5, 7400, 10199, 936, 8, 202, 989, 929, 11, 5, 3701, 4, 9, 768, 939, 64, 75, 860, 2, 1, 1, 1, 1, 1, 1, 1] Finally a pretty maeve shirt that actually fit my (apparently) giant hulk arms.. but it's so flowy/swingy i had to return it. this might just be a problem with being a larger lady but the excess fabric made me feel like i was wearing a tent and made me look $10 \, \mathrm{x}$ bigger than i was around my wai st/hips even when balanced out with leggings. i thought about trying to size down but wasn't sure if that would even solve the excess fabric prob lem and still leave room in the arms. of course i can't try [0, 597, 2629, 6683, 328, 2022-08-29T20:24:48Z 3316 1 2 2422, 3793, 8, 157, 12, 7078, 328, 657, 5, 155, 73, 306, 21764, 8, 124, 1 874, 4, 939, 64, 1057, 939, 40, 3568, 10, 319, 11, 42, 567, 1136, 73, 314 21, 191, 11, 6056, 4, 2200, 5940, 328, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1 Fits perfectly! super soft and well-made! love the 3/4 sleeves and back d rop. i can expect i will wear a lot in this coming fall/winter season in ca. highly recommend! 2022-08-29T20:24:48Z 4699 1 2 [0, 100, 657, 42, 23204, 4, 24, 18, 3279, 1437, 3473, 1437, 8, 3793, 4, 2579, 1318, 350, 4, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1] I love this sweater. it's warm comfortab le and soft. nice quality too.

Upload TSV to the S3 bucket:

In [35]: !aws s3 cp ./feature_store_export.tsv s3://\$bucket/feature_store/feature_

upload: ./feature_store_export.tsv to s3://sagemaker-us-east-1-5044583569 61/feature_store/feature_export.tsv

Check the file in the S3 bucket:

In [36]: laws s3 ls --recursive s3://\$bucket/feature_store/feature_store_export.ts

2022-08-29 20:33:26

3.3. Check that the dataset in the Feature Store is balanced by sentiment

Now you can setup an Athena query to check that the stored dataset is balanced by the target class sentiment .

Exercise 5

Write an SQL query to count the total number of the reviews per sentiment stored in the Feature Group.

Instructions: Pass the SQL statement of the form

```
FROM table_name
GROUP BY category_column
COUNT(*) AS new_column_name
FROM table_name
```

into the variable query_string_count_by_sentiment . Here you would need to use the column sentiment and give a name count_reviews to the new column with the counts.

```
In [37]: feature_store_query_2 = feature_group.athena_query()

# Replace all None
### BEGIN SOLUTION - DO NOT delete this comment for grading purposes
query_string_count_by_sentiment = """
SELECT sentiment, COUNT(*) AS count_reviews
FROM "{}"
GROUP BY sentiment
""".format(feature_store_table)
### END SOLUTION - DO NOT delete this comment for grading purposes
```

Exercise 6

Query the feature store.

Instructions: Use run function of the Feature Store query, passing the new query string query_string_count_by_sentiment. The output S3 bucket will remain unchanged. You can follow the example above.

Out[38]:

	sentiment	count_reviews
0	0	2051
1	1	2051
2	-1	2051

Exercise 7

Visualize the result of the query in the bar plot, showing the count of the reviews by sentiment value.

Instructions: Pass the resulting data frame df_count_by_sentiment into the barplot function of the seaborn library.

```
sns.barplot(
    data=...,
    x='...',
    y='...',
    color="blue"
)
```

```
In [39]: import seaborn as sns
sns.barplot(
    ### BEGIN SOLUTION - DO NOT delete this comment for grading purposes
    data=df_count_by_sentiment, # Replace None
    x='sentiment', # Replace None
    y='count_reviews', # Replace None
    ### END SOLUTION - DO NOT delete this comment for grading purposes
    color="blue"
```

Out[39]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd1afc93dd0>

Upload the notebook and prepare_data.py file into S3 bucket for grading purposes.

Note: you may need to save the file before the upload.

In [40]: !aws s3 cp ./C2_W1_Assignment.ipynb s3://\$bucket/C2_W1_Assignment_Learner
!aws s3 cp ./src/prepare_data.py s3://\$bucket/src/C2_W1_prepare_data_Lear

upload: ./C2_W1_Assignment.ipynb to s3://sagemaker-us-east-1-504458356961 /C2_W1_Assignment_Learner.ipynb upload: src/prepare_data.py to s3://sagemaker-us-east-1-504458356961/src/C2_W1_prepare_data_Learner.py

Please go to the main lab window and click on Submit button (see the Finish the lab section of the instructions).

In []: