Objective:

To train machine learning or deep learning model and classify upcoming news on the fly with good accuracy by building end to end machine learning pipeline.

To make containerized application, which is scalable, robust, fault tolerant.

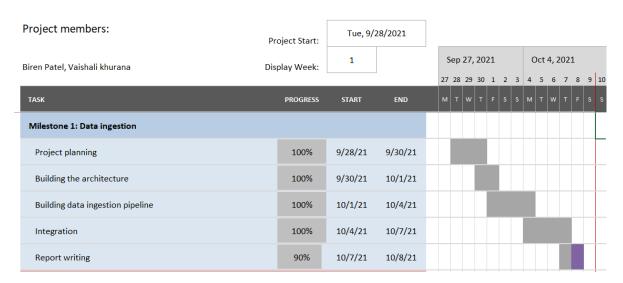
Planning:

We are using agile methodology to build the project. Task level details are mentioned in below Gantt chart.

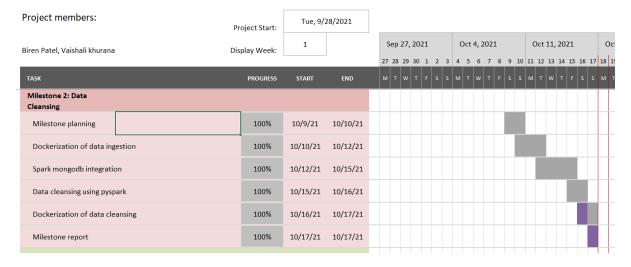
#Sprint: 4

#People: 2

Sprint:1



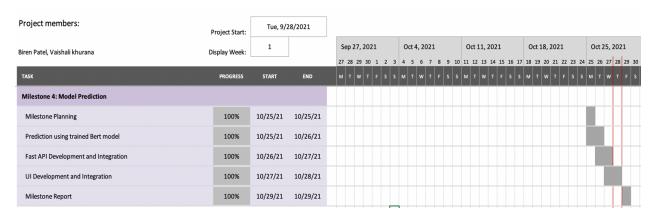
Sprint:2



Sprint:3

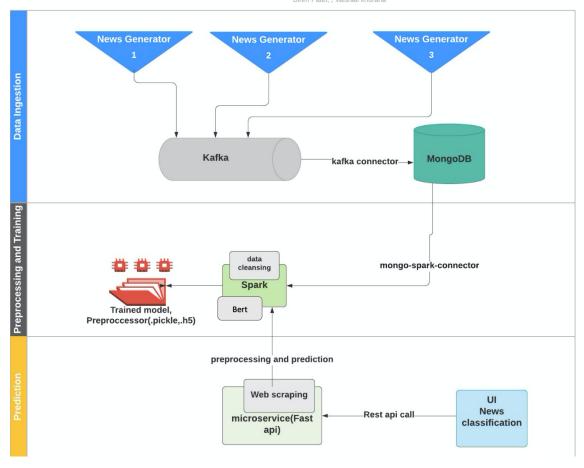
Project members:	Project Start:	Tue, 9/2	28/2021																								
Biren Patel, Vaishali khurana	Display Week:	1		S	ep 2	7, 2	021			00	t 4,	202	21			0	ct 1	1, 2	021			(Oct :	18, 2	021	L	
				27	28	29	30 1	2	3	4	5	6	7 8	9	10	11	12	13	14 1	5 1	6 1	7 18	3 19	20	21	22 2	23 24
TASK	PROGRESS	START	END						s			w ·	T F			м		w		F S	: !	м		w			s s
Milestone 3: Model Training																											П
Milestone Planning	100%	10/18/21	10/18/21																				L				
Training Spark ml model	100%	10/18/21	10/19/21																			ı					П
Training Pre trained Bert(Transformer) model	100%	10/20/21	10/21/21																								
Retraining Model	100%	10/22/21	10/22/21																								
Model saving, loading and deployment	100%	10/22/21	10/22/21																						1		П
Integration with mlflow	90%	10/23/21	10/24/21																								
Milestone report	100%	10/24/21	10/24/21																								

Sprint:4

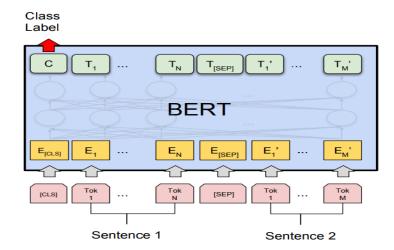


Architecture:

NEWS ARTICLES CLASSIFIER Biren Patel, , Vaishali khurana



Bert model (Transformer) Architecture:



Components and Description:

Component	Description						
Data Ingestion	Multithreaded service that is collecting the data						
	from web using rapid api and custom news						
	generator, passing it to kafka queue. Kafka						
	consumer is used to sync the data between kafka						
	and database. Finally, data is dumped into						
	MongoDb database						
Preprocessor and Trainer	Reading data from mongo db to spark session						
	using mongo-spark-connector.						
	As a part of feature selection we are using						
	category and summary columns.						
	Data cleansing including stop word removal,						
	tokenization, tf-idf vectorization using pyspark.						
	This cleaned data is used for training bert model						
	and achieved training accuracy of 87.8% and test						
	accuracy of 82.9%.						
Predictor and UI	Predicting news category based on the news						
	summary entered by user in UI. Clicking						
	"Predict" button calls backend rest API which						
	firstly processes the data using the trained						
	tokenizer and then predict using the trained Bert						
	model.						

Environment Details:

Docker environment with kafka broker running on 9092 port, zookeeper running on 2181, mongo-db running on 27017, mongo-express running on 8082, spark-master running on 8080, spark-worker running on 8081, predictor service running on 8888 port.

Producer is connecting to kafka and producing news records to kafka queue, consumer is consuming news records from kafka and dumping to mongo db, preprocessor_trainer is cleansing the news data and training machine learning model on spark node using pyspark and predictor service has UI integration with Fast API in backend.

Hardware Details:

- o NVIDIA GPU
- o Memory requirements: Minimum 4GB RAM dedicated for docker

• Tools/libraries used:

- o Docker
- o Pycharm
- o Kafka
- Zookeeper
- o MongoDB
- MongoExpress
- Spark
- Pyspark
- o colab notebook
- o Tensorflow: 2.0.0
- Seaborn
- o mlflow
- o Pretrained Bert model(Transformer): uncased L-8 H-512 A-8
- o Fast API
- o HTML
- CSS
- o Jinja2Templates

• Coding Standards:

o PEP-8 Coding Standards

1. Data Ingestion:

• What goes in as an input:

We have used rapid api and custom news generator as data sources.

```
(env) PS C:\Users\u1105889\P6\Capstone\Capstone\Kafka to Mysql\news_producer> python news_producer.py
sports
['cricket', 'hockey']
sending... {'title': 'There Were 2 Mass Shootings In Texas Last Week, But Only 1 On TV', 'date': '2018-05-26', 'summary': 'She left her husband. He killed their children. Just another day
in America.', 'category': 'CRIME', 'source': 'https://aww.huffingtonpost.com/entry/texas-amanda-painter-mass-shooting.us_5b081ab&e&b882d69caad89')
sending... {'title': 'PRESS RELEASE: NACON: Announcing Cricket 22: A New Era Of Cricket Games Has Arrived!', 'date': '2021-10-07 06:00:00', 'summary': 'Announcing Cricket 22:\n\n\n\n\n New
Era Of Cricket Games Has Arrived!\n\nIncluding The Ashes, Big Bash, The Hundred, Caribbean Premier League, Cricket 22 Is The Biggest Cricket Simulation Ever Made\n\n\nLeguin, October 7: B
ig Ant Studios and Nacon are thrilled to announce that the long-awaited Cricket 22: The Official Game of The Ashes will arrive this November. A true next-generation effort that builds on
the massive success of Cricket 19, Cricket 22 will deliver the most robust, substantial game of cri', 'category': 'sports', 'source': 'yahon.com')
sending... {'title': "Will Smith Joins Diplo And Nicky Jam For The 2018 World Cup's Official Song", 'date': '2018-05-26', 'summary': 'Of course it has a song.', 'category': 'ENTERTAINHENT
', 'source': 'https://www.huffingtonpost.com/entry/will-smith-joins-diplo-and-nicky-jam-for-the-official-2018-world-cup-song.us_5b0972f4eb6fdb2as42021'}s
sending... {'title': 'Cricket 22 Release Date, India Price Announced', 'date': '2021-10-07 07:046012." 'summary': 'Cricket 22 will be the first cricket game on next-gen consoles, the Play's
tation 5 and the Xbox Series S/X. On Thursday, Melbourne-based developer Big Ant Studios announced Cricket 22: Will be available November 25 on PC, PS4, PS5, Xbox One and Xbox Series S/X, a couple of weeks ahead of the start of the Ashes. Cricket 22 for Nintendo Switch a
rrives January 2022. In addition to the Ashes, Cricket, 'category
```

• How the input is being processed:

Multithreaded application that is collecting the data from web using rapid api and custom news generator, passing it to kafka queue. Kafka consumer is dumping the data into mongo db.

```
Kafka Topic | Partitions | Partition Replicas |

confluent_moff_6lksql_processing_log | 1 | 1 |

news | 1 | 1 |

ksql> PRINT news FROM BEGINNING LIMIT 5;

Key format: "\(\text{L}'\text{L}''\text{L}''\text{L}'''} - no data processed

Value format: DSON or KAFKA_STRING

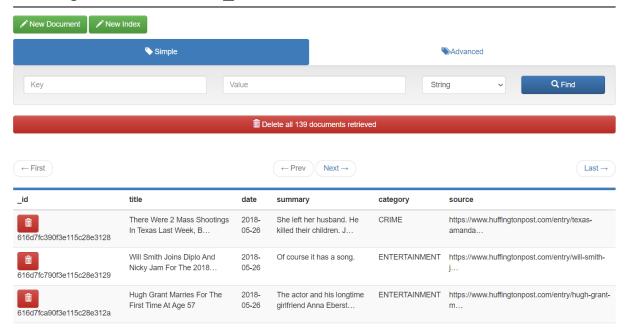
rowtime: 2821/19/10 17:35:37.431 Z, key: <null>, value: {"title": "There Were 2 Mass Shootings In Texas Last Week, But Only 1 On TV", "date": "2818-05-26", "summary": "She left her husban d. He killed their children. Just another day in America.", "category": "CRIME", "source": "https://www.huffingtonpost.com/entry/texas-amanda-painter-mass-shooting_us_5b881ab4e4b8802469ca ad89"), partition: 0

rowtime: 2821/19/10 17:35:39.867 Z, key: <null>, value: {"title": "PRESS RELEASE: NACON: Announcing Cricket 22: A New Era Of Cricket Games Has Arrived!", "date": "2021-10-07 06:00:00", "s ummary": "Announcing Cricket 22: Nninhn\nA New Era Of Cricket Games Has Arrived!\n\nIncluding The Ashes, Big Bash, The Hundred, Caribbean Premier League, Cricket 22 is The Biggest Cricket Simulation Ever Made\n\nLessure. The Cricket Simulation Ever Made\n\nLessure. The Cricket Simulation Ever Made\n\nLessure. The Unit Simulation Ever Made\n\nLessure
```

What comes out as an output:

News records saved in news_collection under news database MongoDb.

Viewing Collection: news_collection

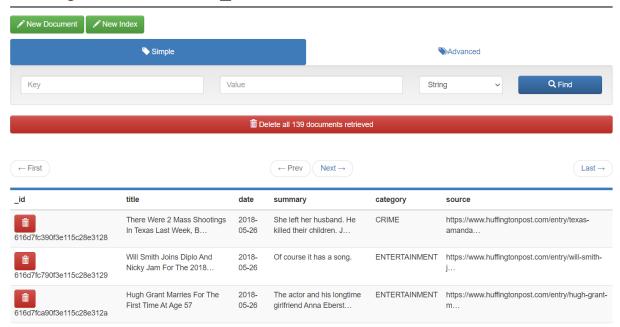


2. Data Cleansing:

• What goes in as an input:

News records saved in news collection under news database MongoDb

Viewing Collection: news_collection



• How the input is being processed:

Records from mongo db collection is processed using pyspark over spark node which is connected through mongo_spark_connector.

As a part of feature selection we are using category and summary columns.

Data cleansing including stop word removal, tokenization, tf-idf vectorization using pyspark

After feature selection and tf-idf vectorization:

```
root
|-- summary: string (nullable = true)
|-- category: string (nullable = true)
|-- tf: vector (nullable = true)
|-- idf: vector (nullable = true)
|-- label: double (nullable = false)
```

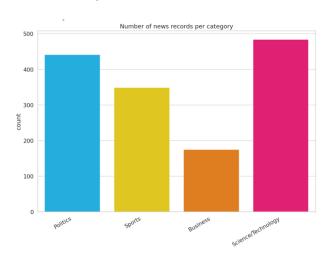
• What comes out as an output:

Processed and cleaned data, saved the preprocessed pipeline

3. Model Training:

• What goes in as an input:

Spark dataframe having cleaned data.



• How the input is being processed:

After the records from mongo db collection is processed and cleaned, these records are further used for training using pyspark random forest model over spark node which is connected through mongo spark connector.

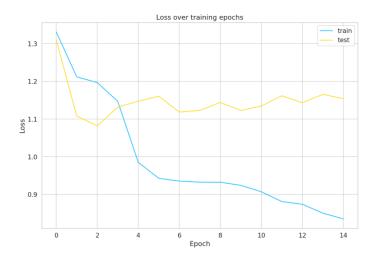
Input data is divided into 80% training and 20% test data.

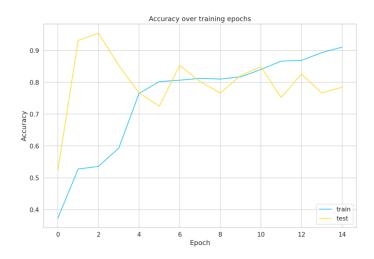
Pretrained Bert model(uncased_L-8_H-512_A-8) is also used for training for 15 epochs with training accuracy of 87.8% and test accuracy of 82.9%, with following parameters:

Layer (type)	Output	Shape	Param #
input_ids (InputLayer)	[(None	, 512)]	0
bert (BertModelLayer)	(None,	512, 512)	41109504
lambda (Lambda)	(None,	512)	0
dropout (Dropout)	(None,	512)	0
dense (Dense)	(None,	512)	262656
dropout_1 (Dropout)	(None,	512)	0
dense_1 (Dense)	(None,	4)	2052
Total params: 41,374,212 Trainable params: 41,374,212 Non-trainable params: 0			

• What comes out as an output:

The trained model and tokenizer are saved in registry.





	precision	recall	f1-score	support
Science/Technology	0.78	0.98	0.87	121
Sports	0.89	0.78	0.83	87
Politics	0.99	0.83	0.90	111
Business	0.53	0.52	0.53	44
accuracy			0.83	363
macro avg	0.80	0.78	0.78	363
weighted avg	0.84	0.83	0.83	363

4. Model Prediction:

• What goes in as an input:

News Summary text entered by user in UI

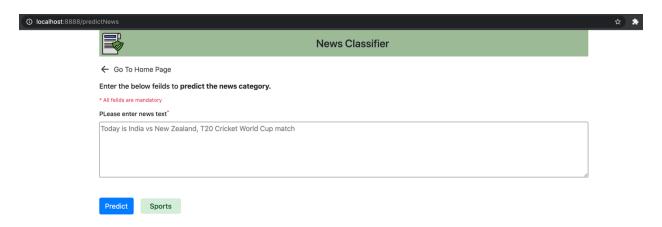


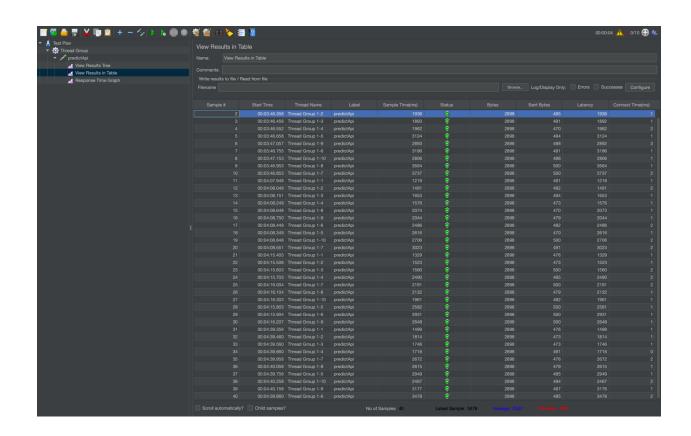
• How the input is being processed:

Clicking "Predict" button calls backend rest API which firstly processes the data using the trained tokenizer and then predict using the trained Bert model and returned response is displayed on UI.

• What comes out as an output:

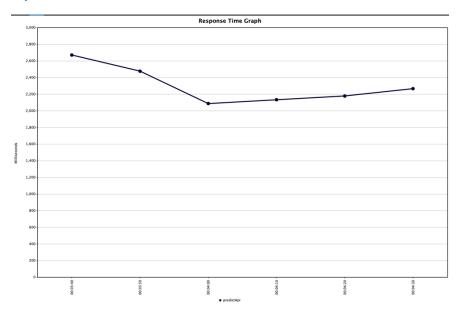
Predicted news category is returned to the UI.





API Stress Testing:

• API Response Time



Summary Report (API Throughput)



Challenges encountered and how we tackled them:

We faced below mentioned challenges:

- Finding better legal data sources
 - o We have used Rapid API and custom dataset.
- Rapid API rate limiting makes the pipeline slow
 - o For training, we have used the dataset dumped into mongodb
- Data Labeling
 - o Annotating correct labels to custom dataset was done manually
- Connecting MongoDb and spark
 - Tried different ways of connecting spark to mongodb then implemented it with mongo-spark-connector
- Authorization issue in MongoDb connection
 - o Configured appropriate parameters to resolve it

- Loading model from HDFS
 - o Configured HDFS in spark node.
- Python-Tensorflow version compatibility
 - o Installed compatible libraries versions
- Slow prediction because of Hardware limitation
 - o Increased docker memory allocation and processor cores
- Loading Bert tokenizer in predictor service
 - Used save and load methods of pytorch
- Responsive UI Design and Integration for mobile support
 - Used Jinja2Templates

Future Scope:

- Further scale optimizations
- Implementing re-training mechanism using feedback feature
- Once we have large volume of labeled data, we will train our own model in place of transfer learning
- Adding multi language support

Github link:

https://github.com/biren162/Capstone