Objective:

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

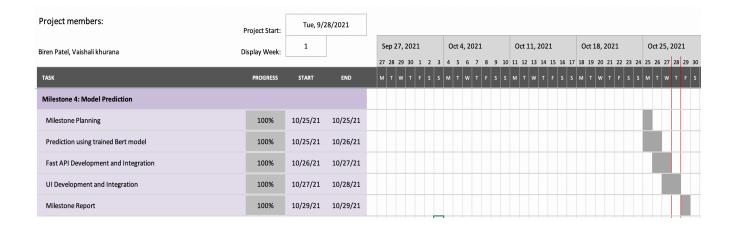
To make containerized application which is scalable, robust, fault tolerant.

Planning:

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

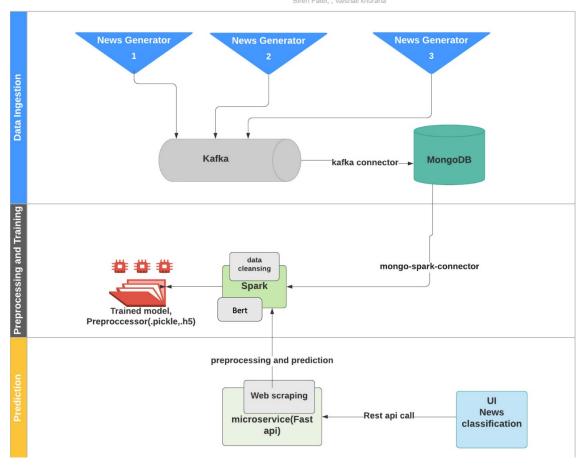
#Sprint: 4

#People: 2

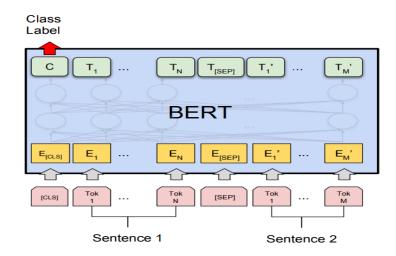


Architecture:

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%.
Prediction	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:

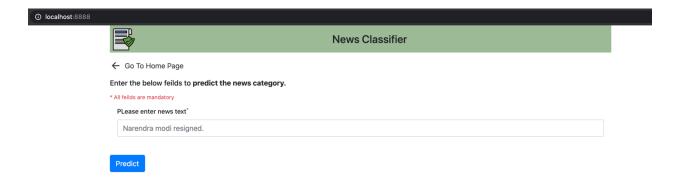
- o NVIDIA GPU
- o Memory requirements: Minimum 4GB RAM dedicated for docker

- Tools/libraries used:
 - Docker
 - o Pycharm
 - o Kafka
 - o Zookeeper
 - o MongoDB
 - MongoExpress
 - o Spark
 - o Pyspark
 - o colab notebook
 - o Tensorflow: 2.0.0
 - o Seaborn
 - o mlflow
 - o Pretrained Bert model(Transformer): uncased_L-8_H-512_A-8
 - o Fast API
 - o HTML
 - o CSS
 - o Jinja2Templates

Model Prediction:

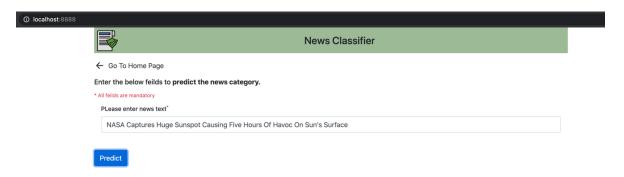
• What goes in as an input:

News Summary text entered by user in UI



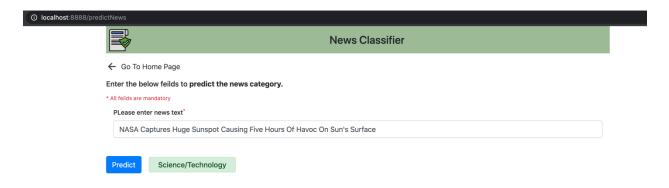
• How the input is being processed:

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.



Challenges encountered and how we tackled them:

We faced below mentioned challenges:

- 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