Group Number	Group Number 26	
Member Names (BITS ID)	Sunil Mittal (2021SC04968)	
Member Names (BITS ID)	Vikram Panwar (2021SC04958)	
Member Names (BITS ID)	Indira Saha (2021sc04956)	
Domain	Sentiment Analysis in Social Media	

	PAPER 1	PAPER 2	PAPER 3
Title of the paper	Transformer Based Multi-Grained Attention Network for Aspect-Based Sentiment Analysis	TSA-CNN-AOA: Twitter sentiment analysis using CNN optimized via arithmetic optimization algorithm	Transformer-based deep learning models for the sentiment analysis of social media data
Authors	JIAHUI SUN PING HAN ZHENG CHENG ENMING WU WENQING WANG	Serpil Aslan Soner Kızıloluk Eser Sert	Sayyida Tabinda Kokab Sohail Asghar Shehneela Naz
Year of publication	2020	2012	2022
Network used	Transformer model	Convolutional neural network	Dilated convolutional Bi-LSTM model
Depth of the network	5	6	
How is the network helping the overall task? eg: feature engg or classification or regression or all	sequential models have achieving competitive	CNN has been a popular method for TSA in recent years. The present study, too, created a database consisting of tweets about COVID-19 for TSA to propose a new CNN-based hybrid approach. To this aim, tweets about COVID-19 were extracted from Twitter to create a large database and propose Twitter sentiment analysis using convolutional neural network optimized via arithmetic optimization algorithm (TSA-CNN-AOA). The proposed approach attempted to classify individuals' tweets about COVID-19 into three main categories: positive, negative, and neutral. Thus, it has become possible to reach significant conclusions about people's attitude toward the COVID-19 pandemic, which can help lessen and eliminate the impact of the disease on them. The experimental studies were performed to test the classification accuracy performances of TSA-CNN-AOA (Decision tree), TSA-CNN-AOA (SVM), and TSA-CNN-AOA (KNN) on the dataset, which yielded an accuracy rate of 92.533%, 95.007%, and 95.098%, respectively. Additionally, CNN, SVM, Naive Bayes, logistic regression, decision tree, and KNN approaches were also used for the testing process, and the highest classification accuracy rate was achieved by TSA-CNN-AOA (KNN)	A novel neural word embedding-based approach was suggested by for SA across several domains. They addressed the major limitations of existing methods, which did not perform well on using in other domains than the domain it was trained on. Their new technique outperformed the old one by achieving higher performance. However, these approaches required a large training corpus for gen erating accurate vectors. In 2019, another word embedding vector was proposed. They called it model Improved Word Vector (IWV), which was a combination of Parts of Speech (POS) tagging methods, word2vec/glove models, lexicon-based method, and word lo calization algorithm. Four datasets were used to conduct sentence-level classification using CNN classifier. However, the main downside of all these methods is the cost of finding the opinion/ emotion orientation of all word individually in the built-in vocabulary. Furthermore, it is possible that the emotion/opinion orientation terms may vary by domain

Loss function used	Softmax , Tanh	Softmax , Tanh	
Evaluation / Performance metric used	Accuracy	Accuracy, F1-score, precision and recall results	Accuracy, F1-score, precision, recall, AUC
Name of Dataset used. If a public dataset,			
provide the URL.	https://github.com/yangheng95/ABSADatasets/tree/v2	https://drive.google.com/drive/folders/1S3SFatKgOA0IzzITgfyNrB	https://www.kaggle.com/c/cifar-10/
Conclusion			

Paper 1 proposes T-MGAN model for aspect-based sentiment analysis. It combines Transformer Encoder and Tree Transformer Encoder with specific sub-tasks, utilize T-MGAN to model the word-level and phrase-level features of aspects and contexts. T-MGAN not only acquires semantic features, but also effectively obtains language hierarchy and syntax information. In addition, the dual-pooling method is used to extract the key features in the hidden layer features of aspect and context, and the attention mechanism is used multiple times to effectively obtain the fine-grained associated emotional features between the specific aspect and the context. Through the attention weight visualization, it is verified that the Transformer Encoder module in this model can effectively focus on the important words in the sentence, while the Tree Transformer Encoder module can learn syntactic information to effectively focus on important phrases

Paper 2 uses CNN model for TSA. The present study, too, created a database consisting of tweets about COVID-19 for TSA to propose a new CNN-based hybrid approach. To this aim, tweets about COVID-19 were extracted from Twitter to create a large database and propose Twitter sentiment analysis using convolutional neural network optimized via arithmetic optimization algorithm (TSA-CNN-AOA). The proposed approach attempted to classify individuals' tweets about COVID-19 into three main categories: positive, negative, and neutral. Thus, it has become possible to reach significant conclusions about people's attitude toward the COVID-19 pandemic, which can help lessen and eliminate the impact of the disease on them.

Paper 3 uses an enhanced feature extraction and classification model using BERT model and dilated convolutional Bi-LSTM model. A BERT-based CBRNN SA model has been proposed for sentence-level classification. The data were annotated using zero-shot BERT, then a pre-trained BERT model was employed to obtain sentence-level semantics and contextual features from the data. The obtained results are then compared with the most commonly used embedding models, such as glove and word2vec. The proposed model in paper 3 obtained significant improvement in f1-score 0.2%, accuracy 0.3% and AUC 0.4%

We can see all the 3 papers are focussing on social media data and sentiment analysis. There are below multiple reasons to use the paper 2 "Twitter sentiment analysis using CNN optimized via arithmetic optimization algorithm" as a paper of