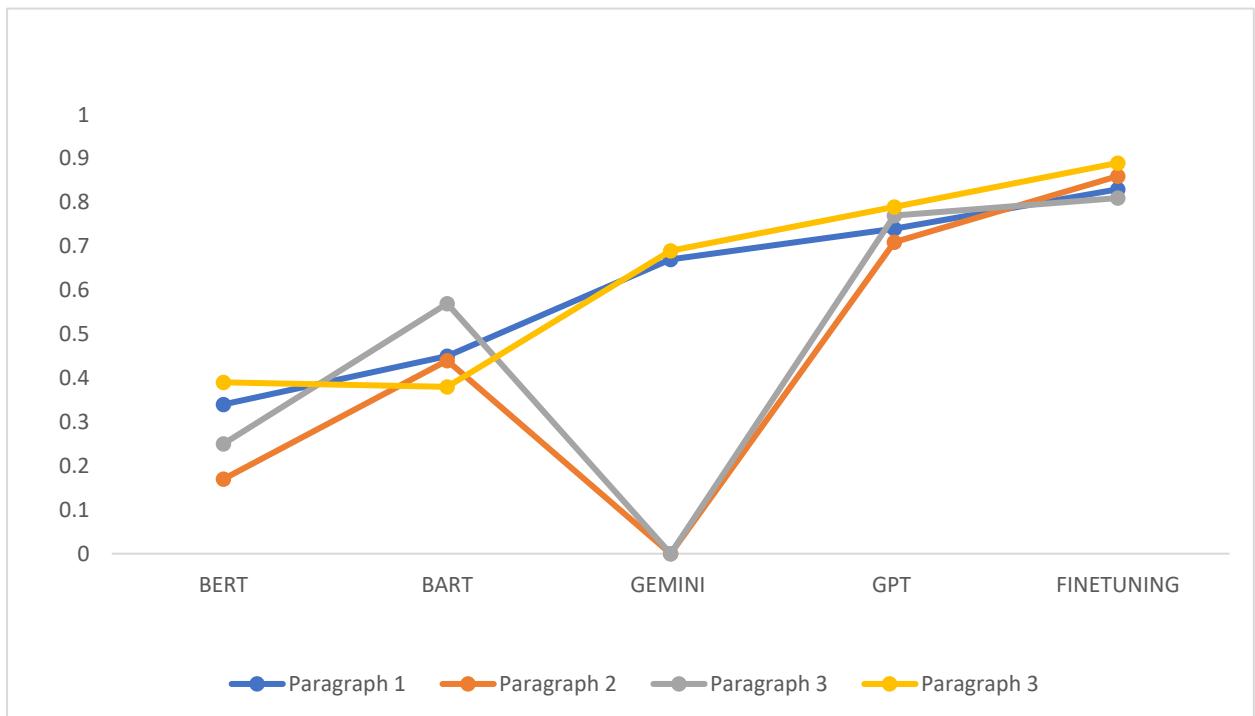


Multiclass Text Classification with Fine Tuning (F1- Score)

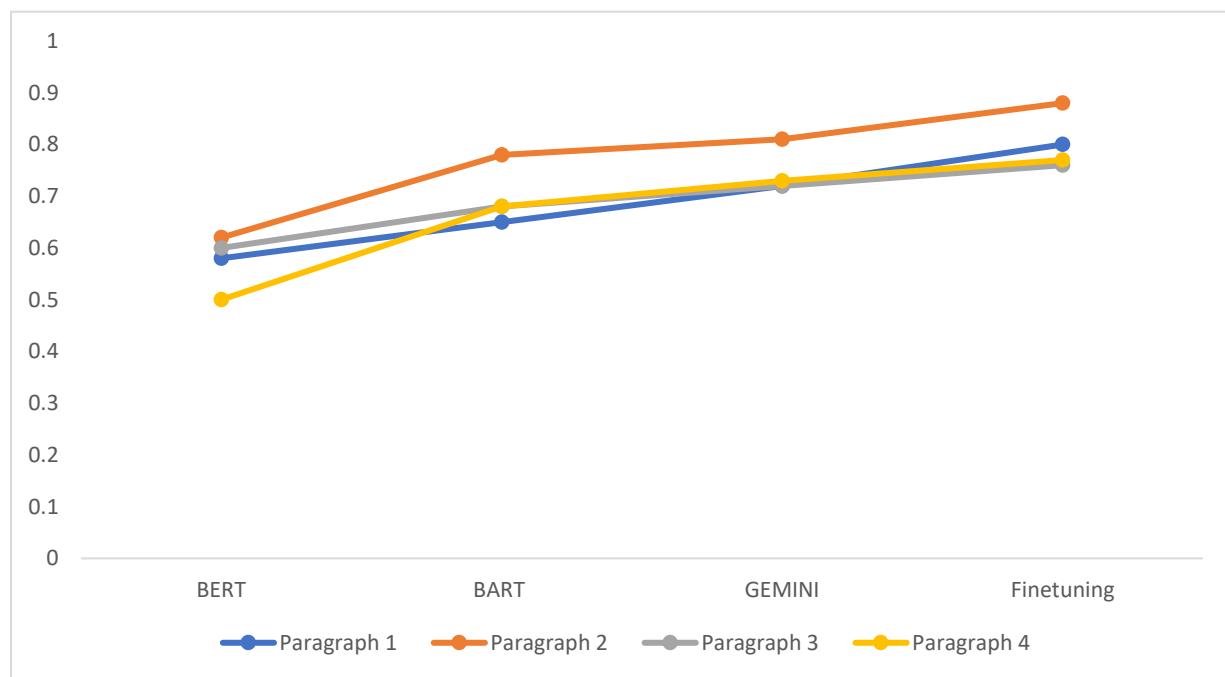
| PARAGRAPHS | BERT | BART | GEMINI | GPT | FINE TUNING (PIPELINE) |
|------------|------|------|--------|------|---------------------------|
| P1 | 0.34 | 0.45 | 0.67 | 0.74 | 0.83 |
| P2 | 0.17 | 0.44 | NA | 0.71 | 0.86 |
| P3 | 0.25 | 0.57 | NA | 0.77 | 0.81 |
| P4 | 0.39 | 0.38 | 0.69 | 0.79 | 0.89 |



The fine-tuning model demonstrates superior performance compared to existing NLP models such as BERT, BART, GEMINI, and GPT, achieving an average F1-Score of 0.85. Within the existing models, certain ones exhibit notably lower accuracy, while others fail to produce results altogether.

Relation Extraction and Summarization using Fine-tuning (F1-Score)

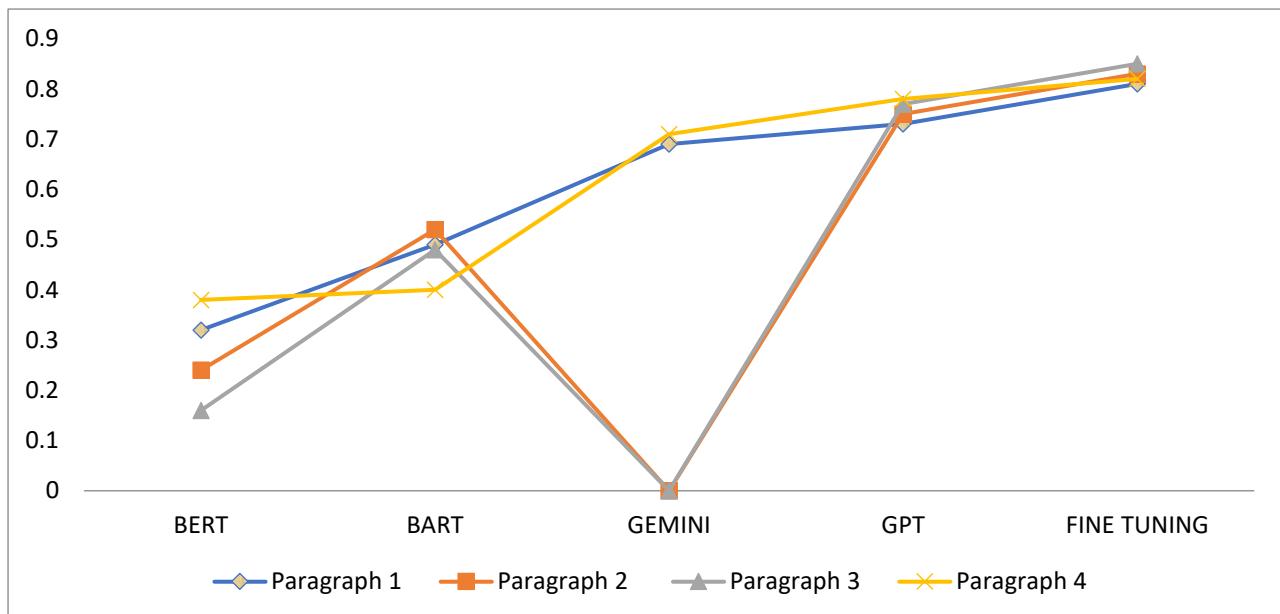
| Paragraph | BERT | BART | Gemini | Finetuning |
|-----------|------|------|--------|------------|
| P1 | 0.58 | 0.65 | 0.72 | 0.80 |
| P2 | 0.62 | 0.78 | 0.81 | 0.88 |
| P3 | 0.60 | 0.68 | 0.72 | 0.76 |
| P4 | 0.50 | 0.68 | 0.73 | 0.77 |



The fine-tuning model demonstrates superior performance compared to existing NLP models such as BERT, BART, GEMINI, and GPT, achieving an average F1-Score of 0.85. Within the existing models, certain ones exhibit notably lower accuracy, while others fail to produce results altogether.

Multi-Label text classification using Fine-tuning (F1-Score)

| Paragraphs | BERT | BART | GEMINI | GPT | FINE-TUNING |
|------------|------|------|--------|------|-------------|
| P1 | 0.32 | 0.49 | 0.69 | 0.73 | 0.81 |
| P2 | 0.24 | 0.52 | NA | 0.75 | 0.83 |
| P3 | 0.16 | 0.48 | NA | 0.77 | 0.85 |
| P4 | 0.38 | 0.40 | 0.71 | 0.78 | 0.82 |



The fine-tuning model demonstrates superior performance compared to existing NLP models such as BERT, BART, GEMINI, and GPT, achieving an average F1-Score of 0.87. Within the existing models, certain ones exhibit notably lower accuracy, while others fail to produce results altogether.