Contents

1	\mathbf{Pro}	oblem: 2	•
	1.1	Part- a	4
	1.2	Part- b and c	4
	1.3	Part_ d	6

1 Problem: 2

NOTE: I am running the code on MacBook GPU named 'mps'.

1.1 Part- a

Completed all the TODO sections in the provided code to create a functioning transformer model with the ability to autoregressively generate 'n' tokens. It is done with proper comments in the code, present in the zip file.

1.2 Part- b and c

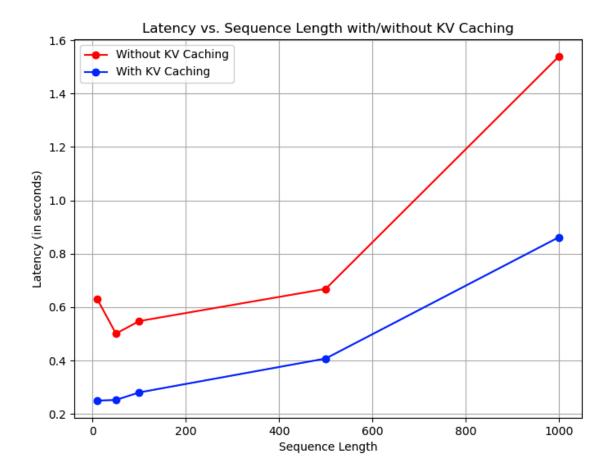
These two parts also well implemented in the code file.

```
    (base) neha@Meghas-MacBook-Air sysForML_assign4 % python transformer.py --mode evaluate Successfully loaded model state dictionary from model_state_dict.pt
Test cases loaded from test_cases.json
Evaluation Results:
    Num test cases: 10
    All tests passed: True
    Pass rate: 100.00% (10/10)
    (base) neha@Meghas-MacBook-Air sysForML_assign4 % python transformer.py --mode kv_evaluate Successfully loaded model state dictionary from model_state_dict.pt
Test cases loaded from test_cases.json
Evaluation Results:
    Num test cases: 10
    All tests passed: True
    Pass rate: 100.00% (10/10)
```

1.3 Part- d

```
(base) neha@Meghas-MacBook-Air sysForML_assign4 % python transformer_skeleton.py --mode benchmark
Seq Len: 10 | No KV: 0.6300s | With KV: 0.2501s
Seq Len: 50 | No KV: 0.5011s | With KV: 0.2526s
Seq Len: 100 | No KV: 0.5480s | With KV: 0.2809s
Seq Len: 500 | No KV: 0.6686s | With KV: 0.4078s
Seq Len: 500 | No KV: 1.5388s | With KV: 0.8618s
Seq Len: 1000 | No KV: 1.5388s | With KV: 0.8618s
2025-03-28 22:38:22.712 python[49765:1950394] +[IMKClient subclass]: chose IMKClient_Modern
2025-03-28 22:38:22.712 python[49765:1950394] +[IMKInputSession subclass]: chose IMKInputSession_Modern
Successfully loaded model state dictionary from model_state_dict.pt
Benchmarking...
Results:
Without KV cache: 0.0088 seconds
With KV cache: 0.0100 seconds
Speedup: 0.88x
```

- Faster Generation with KV Caching: KV caching significantly reduces latency, especially for longer sequences.
- Minimal Difference for Short Sequences: For shorter sequences (10, 50), the performance gain is less noticeable.
- Exponential Growth Without Caching: Without KV caching, latency increases almost exponentially as sequence length increases.
- Efficient Handling of Long Sequences: KV caching eliminates redundant computations, leading to faster token generation in longer sequences.
- Conclusion: KV caching greatly enhances the efficiency of decoder-only transformers, particularly in long-sequence scenarios.



References

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- [4] M. Shoeybi, M. Patwary, R. Puri, P. LeGresley, J. Casper, and B. Catanzaro, "Megatron-LM: Training Multi-Billion Parameter Language Models Using Model Parallelism," arXiv preprint arXiv:1909.08053, 2019.