

TUM EI 70360: MACHINE LEARNING AND OPTIMIZATION
FALL 2023

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Problem Set 12

Issued: Tuesday, Jan. 16, 2024

Due: Thursday, Jan. 25, 2024

Problem 1 (Number of parameters and compute of the GPT model). Consider the decoder-only transformer considered in class with the following parameters. For all tasks omit sub-leading terms such as non-linearities, biases, and normalization.

`vocab_size`: size of the vocabulary

`dim_embd`: dimension or size of the embeddings

`n_heads`: number of heads

`n_layers`: number of layers

`context_length`: the (maximal) context length, that is the maximal number of input tokens

1. Compute the total number of non-embedding trainable parameters N of the transformer, i.e., the total number of trainable parameters excluding the number of parameters associated with embeddings.
2. Now compute the total number of non-embedding compute in floating point operations (FLOPs) per tokens for a forward pass. Make the following assumption: The multiplication of two matrices \mathbf{A} and \mathbf{B} , where \mathbf{A} is $i \times j$ and \mathbf{B} is $j \times k$ gives a matrix \mathbf{C} of dimension $i \times k$. For each element of \mathbf{C} , you perform j multiply-accumulate operations (since you're multiplying and then summing up j pairs of elements). Each of these involves one multiplication and one addition. Therefore, for each element of \mathbf{C} , you have $2j$ FLOPs. Extending this to the entire matrix \mathbf{C} (which has ik elements), the total number of FLOPs is $2ijk$.
3. Assume that the cost of a backward pass is roughly two times a forward pass, and assume that we have a transformer in the regime `dim_embd` \gg `context_length`/12. Justify the common statement 'the non-embedding compute per token for training is roughly $6N$ '.