

ReaLLM ASIC

Make Your Own Lightweight LLMs

Outline of the Tutorial

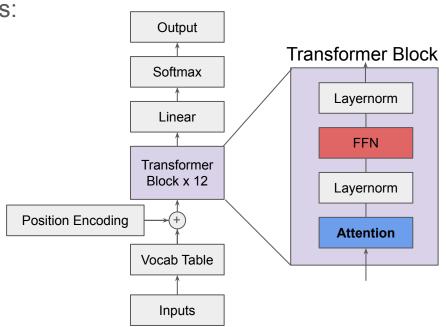
- Intro
 - LLMs and Applications
 - Recap of LLM architecture
- Hands-on Al From Scratch:
 - Building and training a custom lightweight LLM
 - Data preparation and preprocessing
 - Model training options and optimization
- Checkpoints and Finetuning

Applications Limited Only by Datasets

LLM Transformer Architecture

One architecture, limited only by datasets:

- Translation
- Poetry
- Music
- Robotics Motion
- Cooking Recipes
- Editing and Writing Reports

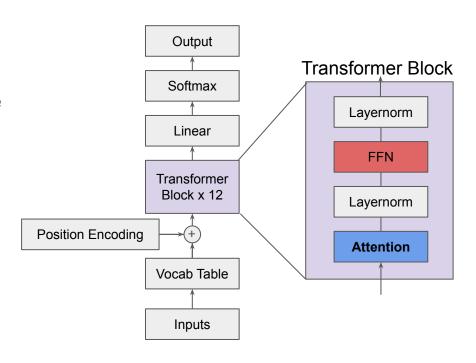


Main Hyperparameters

- "Height" Number of Layers
 - Deep Networks -> Abstract Knowledge
 - Linearly increases size of network

- "Width" Dimensions per Token
 - Better Per Token Understanding
 - Non-linear increase in size

LLM Transformer Architecture



Datasets

We are limited by what we can train on.

- Music
 - We'll work with JS Bach
- Language
 - Example with Japanese and English
- Generation
 - Example with Shakespeare

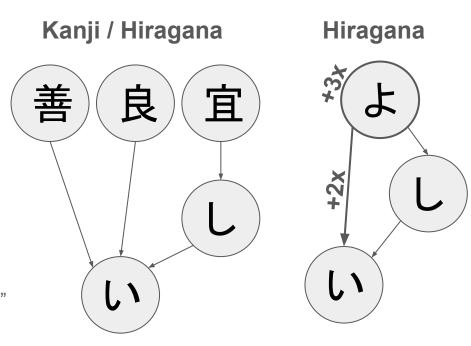
Tokenization Speed-Up Example

Tokens need to gain "Experience Points":

 Higher frequency of a token in the dataset then better the LLM will learn about it.

Language Example:

- Preprocessing Kanji -> Hiragana:
 - Faster training (smaller model)
 - Faster accumulation of "experience points"



Colab

Workshop Colab

- Workshop colab
 - based on popular "nanoGPT" framework

 We'll migrate to the Colab for the remainder of the workshop.

```
Run GPU Training
    !python3 data/shakespeare char/prepare.py
→ length of dataset in characters: 1,115,394
    all the unique characters:
     !$&',-.3:;?ABCDEFGHIJKLMNOPQRSTUVWXYZabcdefghijklmnopgrstuvwxyz
    vocab size: 65
    train has 1,003,854 tokens
    val has 111,540 tokens
   !python3 train.py --device="cuda" --dtype="float16" --max iters=
→ 2024-05-13 06:59:01.957995: E external/local xla/xla/stream exec
    2024-05-13 06:59:01.958075: E external/local xla/xla/stream exec
    2024-05-13 06:59:02.106433: E external/local xla/xla/stream exec
    2024-05-13 06:59:02.395367: I tensorflow/core/platform/cpu featu
    To enable the following instructions: AVX2 FMA, in other operations
    2024-05-13 06:59:04.921801: W tensorflow/compiler/tf2tensorrt/ut
    seed: 1337
    seed offset: 0
    number of parameters: 2.98M
    num decayed parameter tensors: 16, with 3,072,384 parameters
    num non-decayed parameter tensors: 13, with 4,992 parameters
    using fused AdamW: True
    step 0: train loss 4.2075, val loss 4.2067
    iter 0: loss 4.2097, time 14979.17 ms, mfu -100.00%
    iter 10: loss 3.3180, time 130.96 ms, mfu 1.00%
    iter 20: loss 3.2074, time 132.72 ms, mfu 1.00%
    iter 30: loss 2.9536, time 131.67 ms, mfu 1.00%
    iter 40: loss 2.7902, time 132.85 ms, mfu 1.00%
    iter 50: loss 2.6889, time 132.28 ms, mfu 1.00%
    iter 60: loss 2.6815, time 132.55 ms, mfu 1.00%
```