# A Journal on Training an LLM from Scratch

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## Introduction

I starting working on this project with the interest to reproduce what code completion large language models are capable of doing. My goal is to optimize a tiny completions model, and discover where are the limits.

This journal serves as documentation for the design decisions, implementation choices, parameter changes, etc.

**Disclaimer:** This is an education project and there will be mistakes. Contributions are welcome!

## Week 22.09.25

Since we are building a new model from scratch, I wanted to also retrain a new tokenizer. In addition to learning how all this works, I wanted to make changes in the regex parsing, which groups tokens together.

#### Building the Tokenizer

I decided to reimplement BPE (byte-pair-encoding) tokenization. The main purpose was to learn the BPE algorithm and control it.

I know I needed a high-performance language for this take, so I started out implementing a version in C. However, I quickly realized that the algorithm depends on quite a lot of data structures (vectors, hash maps, heaps) in addition to the regex parsing. Therefore, I switched to C-style C++.

For the training data, I trained the tokenizer exclusively on Python code using the py150 dataset.

Since python code is *usually* written using ASCII characters, I removed UTF-8 characters before tokenizing, effectively excluding them from the vocabulary. This removes tokens that are very rarely used and allows the model to focus on the essential.

## Creating the Regex

I took inspiration from the cl100k\_base regex [1]:

```
'(?i:[sdmt]|l1|ve|re)|[^\r\n\p{L}\p{N}]?+\p{L}++\\p{N}{1,3}+| ?[^\s\p{L}\p{N}]+ + [\r\n]*+|\s+*|\s*[\r\n]|\s+(?!\S)|\s
```

and ended up on the following regex:

```
 ?[A-Za-z_{0}][A-Za-z_{0}]*|%(?:\.\d+)?[sdifFeEgGxXoc%]|[0-9]{1,3}| ?[^ %_A-Za-z0-9]+(?: ")?[\r\n]*|%|\s+$|\s+(?=\s)|\s
```

- I removed UTF-8 handling and compound expression grouping (ex: 've )
- %(?:\.\d+)?[sdifFeEgGxXoc%] groups together printf formats like %s and the rest says very similar

## Week 29.09.25

## **Building the Training Loop**

I used the torch along with lightning library [2] to implement a training loop.

For the model, instead of reimplementing from scratch I used the Qwen3 (without MoE) model from the transformers library [3].

I choose AdamW for the optimizer with a warm-up and cosine schedule. Below is a non exhaustive list of the parameters chosen:

```
split_ratio: 0.7
 vocab_size: 20260 # 256 byte tokens + 20000 BPE merges + 4 special tokens
(BOS, EOS, PAD, UNK)
 # Qwen3 architecture parameters
 hidden_size: 512
 num_hidden_layers: 4
 num_attention_heads: 16
 num_key_value_heads: 8
 intermediate_size: 1024
 max_position_embeddings: 2048
 rope_theta: 10000.0
 attention_dropout: 0.1
 rms_norm_eps: 0.000001
training:
 batch_size: 32
 epochs: 50
 lr: 0.0001
 weight_decay: 0.01
 grad_clip: 1.0
 gradient_accumulation_steps: 4
```

### Position IDs & Attention Mask

TODO

## Week 06.10.25

## **BOD Special Token**

I added this "beginning of sequence" special token to allow it to act as a "attention skin" [4].

The BOS token is added at the start of every input sequence to the model. Is is not like the EOS (end of sequence) token, where it is added between documents to delimit them.

```
input = [BOS, 234, 6236, 346, 4357, 347, ...] # where BOS is the token id for
BOS
```

### **Training Fixes**

I updated the scheduler to step on every step instead of epoch. I also scaled the training loss over the accumulated batches instead of using the loss of the step.

## Tokenizer Regex

I realized that it cannot hurt to join compound expression grouping. So I added the following capture group:

```
'(?i:[sdmt]|l1|ve|re)
```

#### **Data Distribution**

I also removed <code>max\_batches\_per\_epoch</code> , replacing it with <code>max\_tokens</code> such that we have a fair distribute. Before the training batch was shuffled and all data was included but limited to 750. However the evaluation batch was not shuffled but also limited to 750. This meant that the model was training on more that 70% training data, and the validation data was not representative.

## PAD Special Token

I've encountered some issues with incomplete batches. Either I drop the last batch or I pad it. I've decided to pad it with the PAD special token. Currently, the model does not see it very often when training so it might not work too well.

### Run #1

In progress

# **Bibliography**

- [1] OpenAI, "cl100k\_base tokenizer regular expression in tiktoken\_ext/openai\_public.py, line 89." 2025.
- [2] Lightning, "Lightning GitHub." 2025.
- [3] Huggingface, "Transformers GitHub." 2025.
- [4] G. Xiao, Y. Tian, B. Chen, S. Han, and M. Lewis, "Efficient Streaming Language Models with Attention Sinks." [Online]. Available: https://arxiv.org/abs/2309.17453