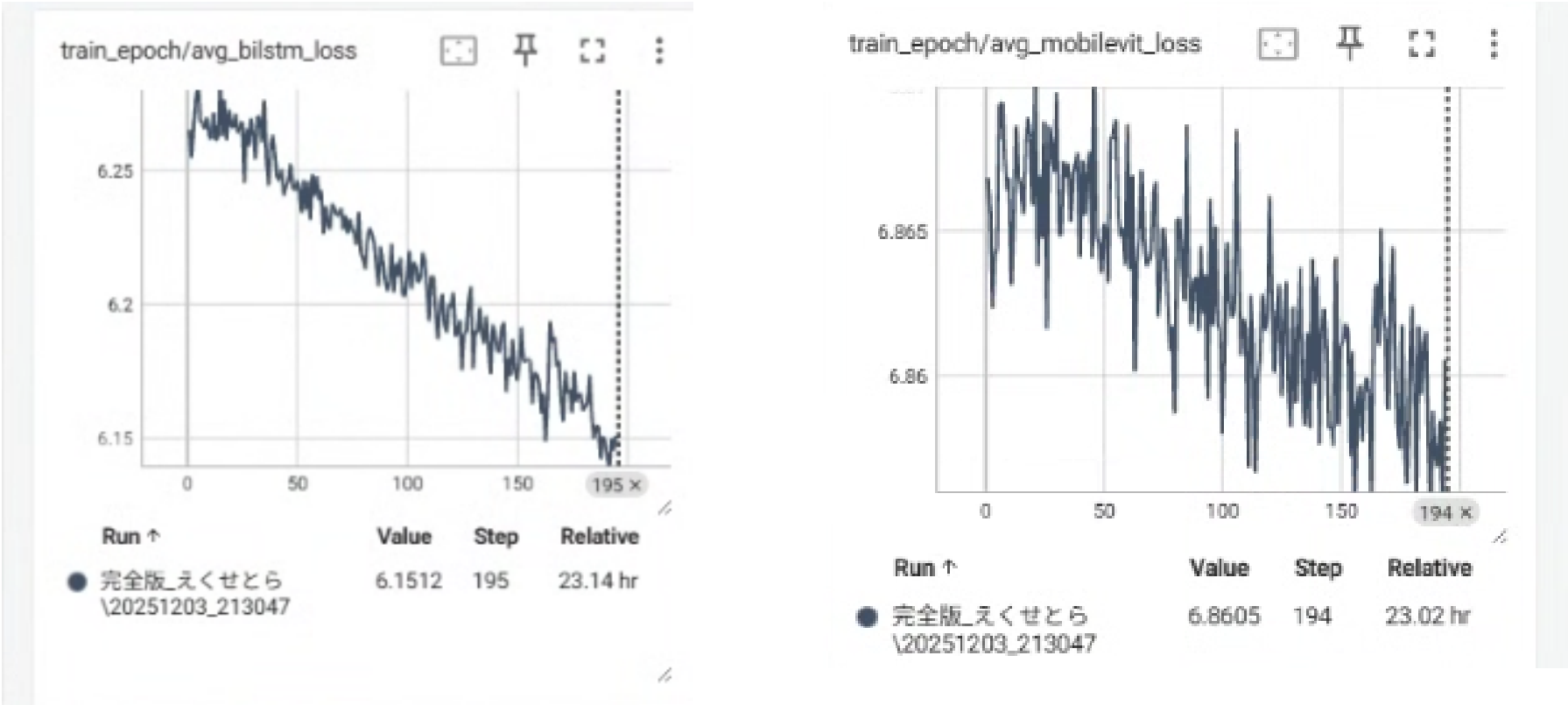
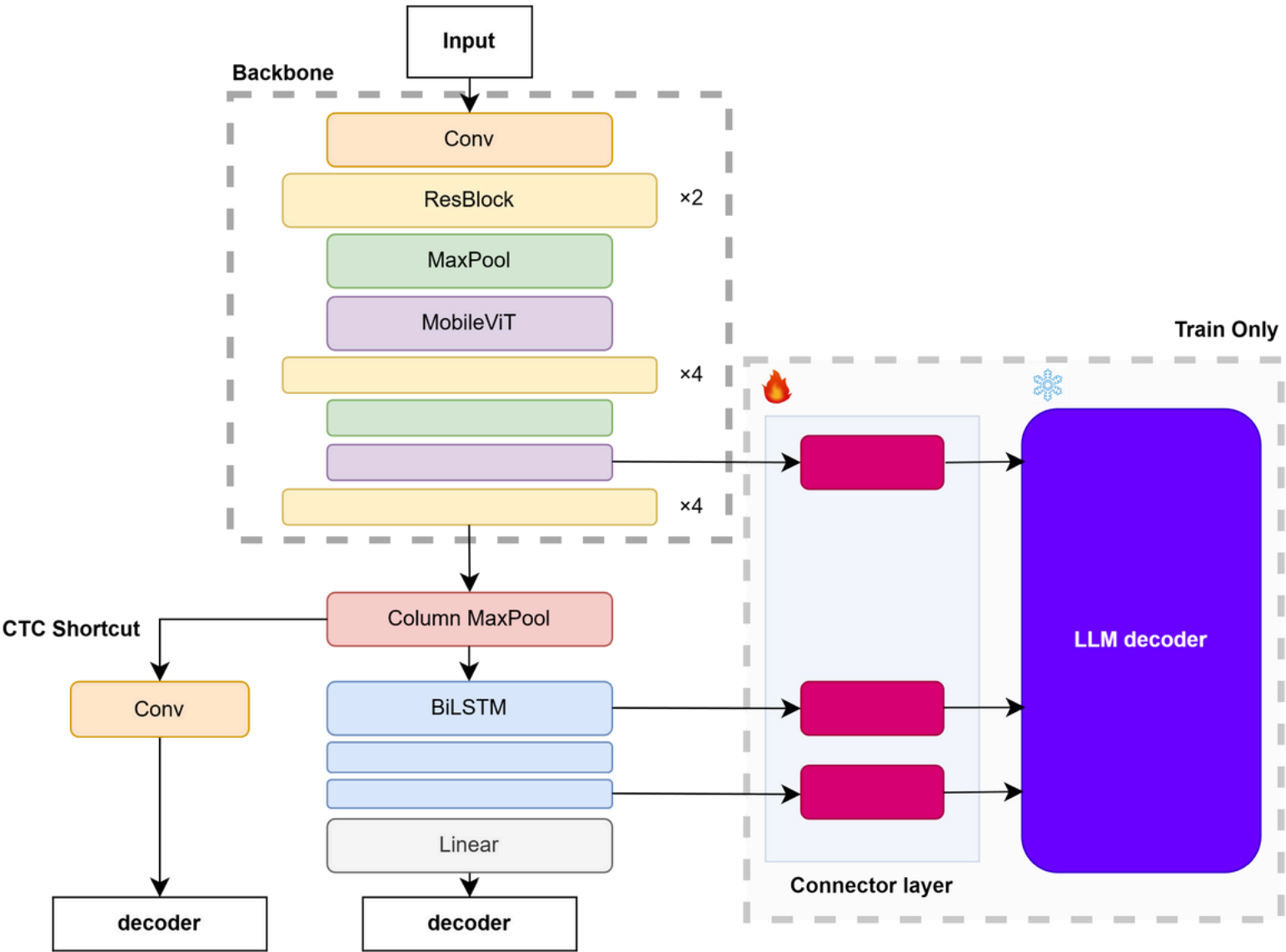


卒研にむけて②

# 提案手法前確認



lossの比較



前のモデル

# なにをしているのか

正しくファインチューニングできるのか（自分の工夫含まず）

→ファインチューニングできたら自分で書いた学習ループが正しいことが示せる

上記が確認できたら自分の工夫入れていこうと思う

# Trocrファインチューニング編

Model	Architecture	Training Data	External LM	CER
TrOCR <sub>SMALL</sub>	Transformer	Synthetic + IAM	No	4.22
TrOCR <sub>BASE</sub>	Transformer	Synthetic + IAM	No	3.42
TrOCR <sub>LARGE</sub>	Transformer	Synthetic + IAM	No	2.89

目標値

smallで精度確認→CER：70超



≡ TrOCR (small-sized model, fine-tuned on IAM)

TrOCR model fine-tuned on the [IAM dataset](#). It was introduced in the paper [TrOCR: Transformer-based Optical Character Recognition with Pre-trained Models](#) by Li et al. and first released in [this repository](#).

学習済み重み

もうこれはすでにIAMでファインチューニング済なので同じデータとモデルを使ってるおりかつ正しいモデル設定、データ前処理がなされていたら精度がでるはず

# Trocrファインチューニング編

## モデル設定見直し

```
model.config.pad_token_id = 1
model.config.eos_token_id = 2
model.config.decoder_start_token_id = 2
```

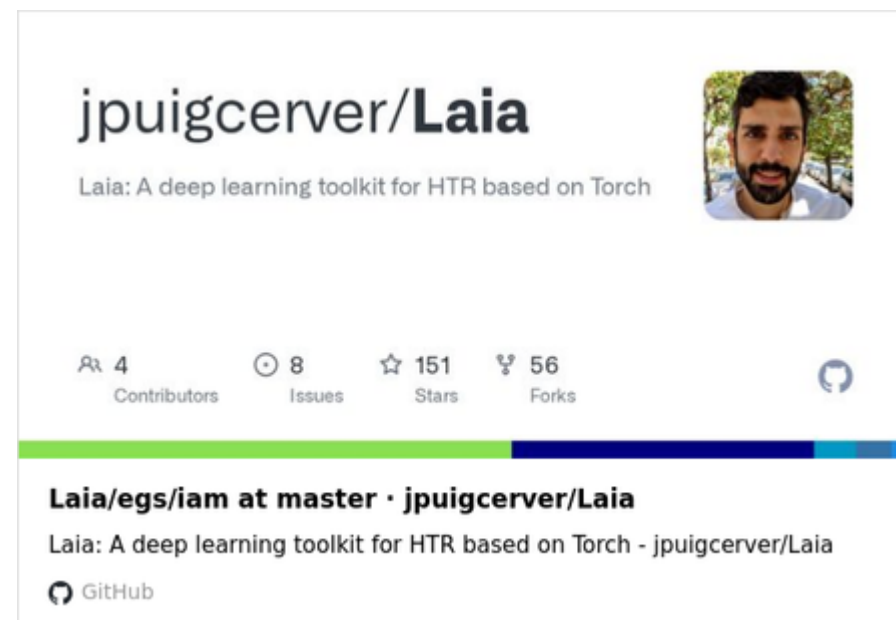
←CER:5.38

## 上記＋データ見直し

- train,valのデータ分布
- リサイズ
- 論文記載のデータ処理を使用

←CER:4.33

**(目標値にほぼ到達で提案手法に移れる！！)**



# Trocrファインチューニング編

モデル設定見直し

```
model.config.pad_token_id = 1
model.config.eos_token_id = 2
model.config.decoder_start_token_id = 2
```

←CER:5.38

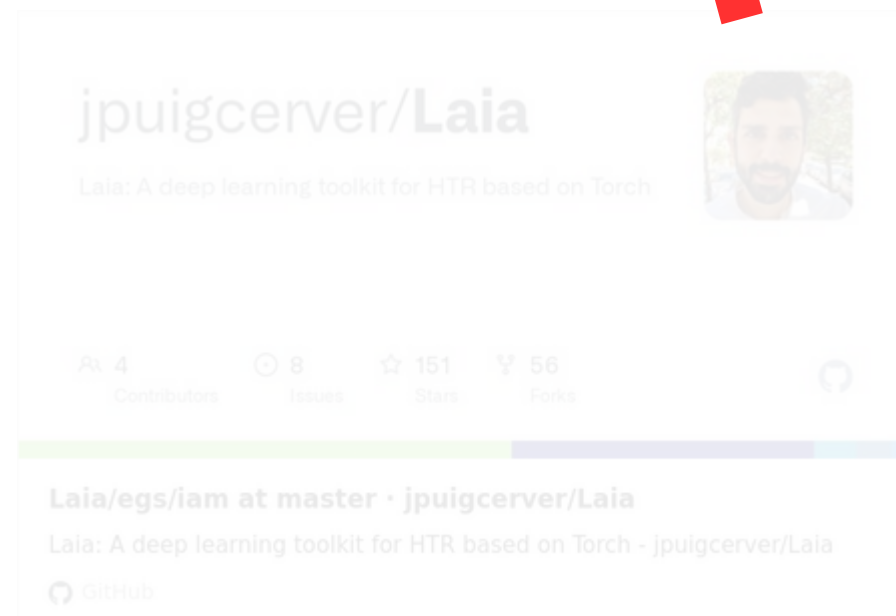
データ見直し

- train,valのデータ分割
- リサイズ
- 論文記載のデータ処理を使用

←CER:4.33

(目標値にほぼ到達で提案手法に移れる！！)

問題発生！！



# Trocrファインチューニング編

```
=====
TensorBoard logging to: ./logs/runs\20251208_205704_12-07_TOCR-small
Start TensorBoard with: python -m tensorboard --logdir="./logs/runs\20251208_205704_12-07_TOCR-small"
=====
Epoch 1: 0%|                                                    | 0/1541 [00:00<?, ?it/s]C
:\Users\user\AppData\Roaming\Python\Python310\site-packages\transformers\integrations\sdpa_attention.py:54: UserWarning: 1Torch was not compiled with flash attention. (Triggered internally at ..\aten\src\ATen\native\transformers\cuda\sdp_utils.cpp:263.)
  attn_output = torch.nn.functional.scaled_dot_product_attention(
Epoch 1: 100%|                                                    | 1541/1541 [02:42<00:00, 9.49it/s, loss=2.9129]
Epoch 2: 100%|                                                    | 1541/1541 [02:42<00:00, 9.51it/s, loss=1.6902]
Epoch 3: 100%|                                                    | 1541/1541 [02:41<00:00, 9.52it/s, loss=1.3823]
Epoch 4: 100%|                                                    | 1541/1541 [02:41<00:00, 9.53it/s, loss=1.1408]
Epoch 5: 100%|                                                    | 1541/1541 [02:41<00:00, 9.54it/s, loss=1.0002]
Epoch 6: 100%|                                                    | 1541/1541 [02:42<00:00, 9.51it/s, loss=0.8833]
Epoch 7: 100%|                                                    | 1541/1541 [02:41<00:00, 9.53it/s, loss=0.7851]
Epoch 8: 100%|                                                    | 1541/1541 [02:41<00:00, 9.54it/s, loss=0.7385]
Epoch 9: 100%|                                                    | 1541/1541 [02:42<00:00, 9.51it/s, loss=0.6279]
Epoch 10: 100%|                                                    | 1541/1541 [02:41<00:00, 9.52it/s, loss=0.5730]
Test epoch 10: 100%|                                                    | 365/365 [01:57<00:00, 3.10it/s]
Test CER: 0.3085
```

ログ

# Trocrファインチューニング編

## 課題

$$\text{loss} = -\frac{1}{|\Omega|} \sum_{(b,t) \in \Omega} \log p_{\theta}(y_{b,t} \mid \text{画像, 文の過去トークン})$$

学習を回すと精度が悪化する

- 現状：epoch:10、CER: 30.85
- 違和感：ロスは減少している、まず、valとtestで差が10倍

## 試したこと

- 1部の層のみ学習可能にする（encoderの最終層とdecoder 2層） → CER：5.90
- ロスを自作（内部で何してるかわからん！） → CER：8.06
- 上二つを組み合わせた：CER：5.16

```
Test CER: 0.3085
=====
TensorBoard logging to: ./logs/runs/20251208_212930_12-07_TOCR-small
Start TensorBoard with: python -m tensorboard --logdir=./logs/runs/20251208_212930_12-07_TOCR-small
=====
Epoch 1: 0% | 0/1541 [00:00<7, 71it/s]C
./Users\user\AppData\Roaming\Python\Python310\site-packages\torch\tensorboard\tensorboard.py:54: UserWarning: Torch was not compiled with flash attention. (Triggered internally at ..\aten\src\ATen\native\tensorboard\tensorboard.py:263.)
  attn_output = torch.nn.functional.scaled_dot_product_attention(
Epoch 1: 100% | 1541/1541 [01:14:00:00, 20.64it/s, loss=0.6422]
Epoch 2: 100% | 1541/1541 [01:15:00:00, 20.45it/s, loss=0.1520]
Epoch 3: 100% | 1541/1541 [01:15:00:00, 20.53it/s, loss=0.1244]
Epoch 4: 100% | 1541/1541 [01:16:00:00, 20.66it/s, loss=0.1063]
Epoch 5: 100% | 1541/1541 [01:16:00:00, 20.59it/s, loss=0.0930]
Epoch 6: 100% | 1541/1541 [01:15:00:00, 20.54it/s, loss=0.0851]
Epoch 7: 100% | 1541/1541 [01:14:00:00, 20.57it/s, loss=0.0798]
Epoch 8: 100% | 1541/1541 [01:15:00:00, 20.47it/s, loss=0.0692]
Epoch 9: 100% | 1541/1541 [01:15:00:00, 20.47it/s, loss=0.0657]
Epoch 10: 100% | 1541/1541 [01:14:00:00, 20.78it/s, loss=0.0621]
Test epoch 10: 100% | 365/365 [01:46:00:00, 3.43it/s]
Test CER: 0.0517
```

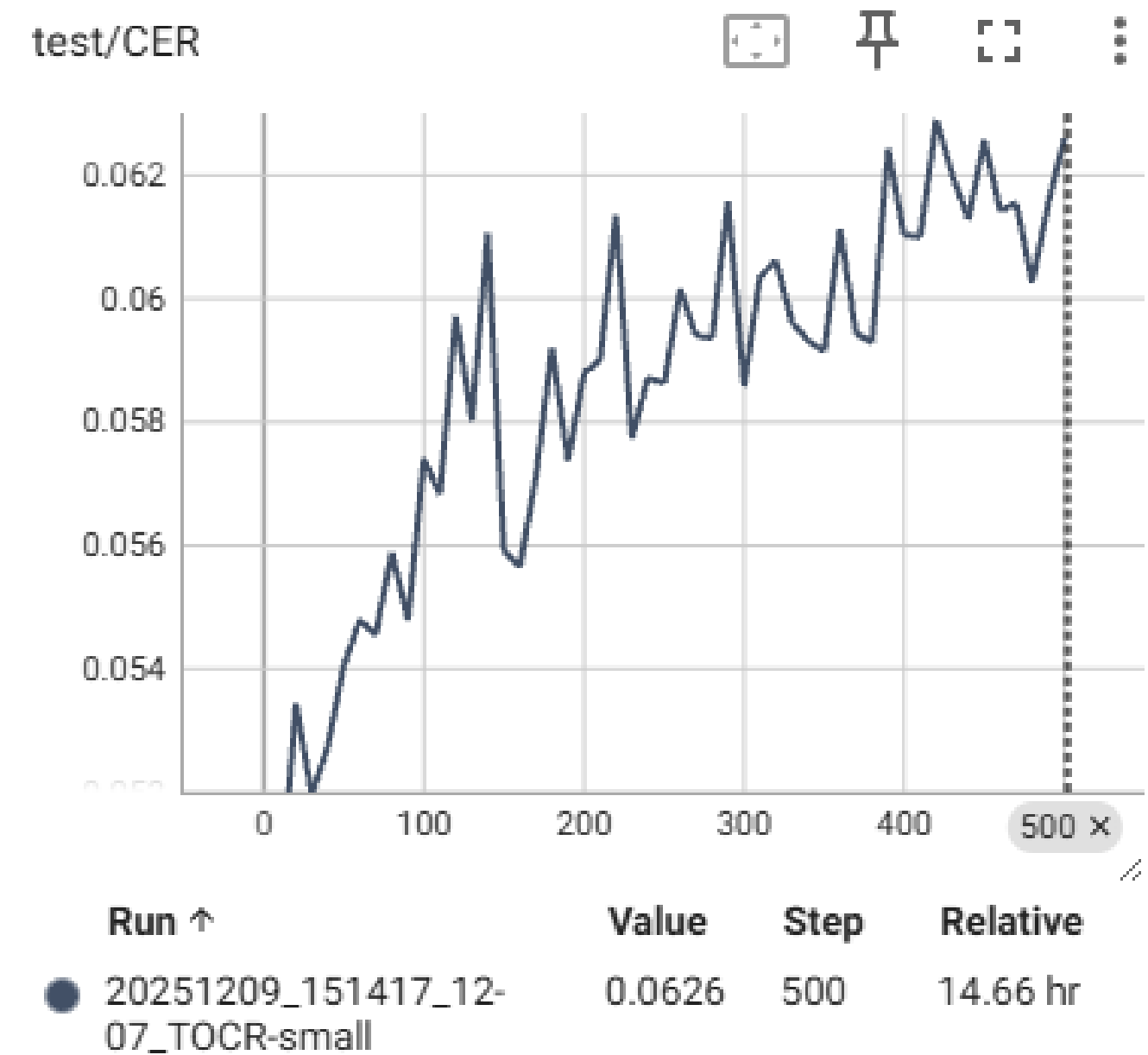
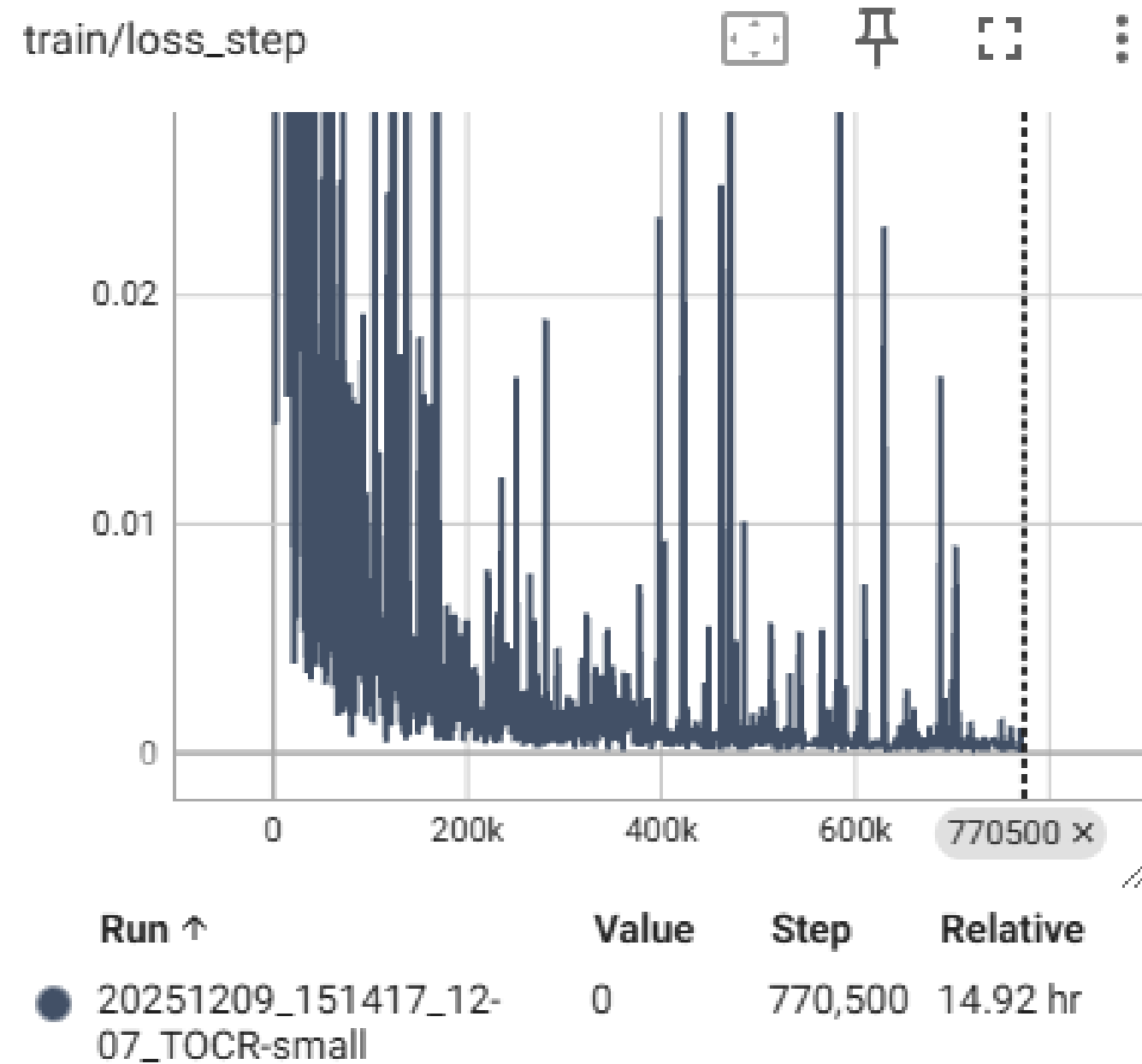
学習なしの評価→**CER:4.33**でいずれも精度悪化

lossは下がってるしもう少し学習ステップを増やしてみる→



# Trocrファインチューニング編

## ログ



ステップごとの値とってしまった見にくいですよね...lol

# モデル構造

```
(encoder): ViTModel(
  (embeddings): ViTEmbeddings(
    (patch_embeddings): ViTPatchEmbeddings(
      (projection): Conv2d(3, 768, kernel_size=(16, 16), stride=(16, 16))
    )
    (dropout): Dropout(p=0.0, inplace=False)
  )
  (encoder): ViTEncoder(
    (layer): ModuleList(
      (0-11): 12 x ViTLayer(
        (attention): ViTAttention(
          (attention): ViTSelfAttention(
            (query): Linear(in_features=768, out_features=768, bias=False)
            (key): Linear(in_features=768, out_features=768, bias=False)
            (value): Linear(in_features=768, out_features=768, bias=False)
          )
          (output): ViTSelfOutput(
            (dense): Linear(in_features=768, out_features=768, bias=True)
            (dropout): Dropout(p=0.0, inplace=False)
          )
        )
        (intermediate): ViTIntermediate(
          (dense): Linear(in_features=768, out_features=3072, bias=True)
          (intermediate_act_fn): GELUActivation()
        )
        (output): ViTOutput(
          (dense): Linear(in_features=3072, out_features=768, bias=True)
          (dropout): Dropout(p=0.0, inplace=False)
        )
        (layernorm_before): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
        (layernorm_after): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
      )
    )
  )
  (layernorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
  (pooler): ViTPooler(
    (dense): Linear(in_features=768, out_features=768, bias=True)
    (activation): Tanh()
  )
)
```

```
(decoder): TrOCRForCausalLM(
  (model): TrOCRDecoderWrapper(
    (decoder): TrOCRDecoder(
      (embed_tokens): TrOCRScaledWordEmbedding(50265, 1024, padding_idx=1)
      (embed_positions): TrOCRLearnedPositionalEmbedding(514, 1024)
      (layernorm_embedding): LayerNorm((1024,), eps=1e-05, elementwise_affine=True)
      (layers): ModuleList(
        (0-11): 12 x TrOCRDecoderLayer(
          (self_attn): TrOCRAttention(
            (k_proj): Linear(in_features=1024, out_features=1024, bias=True)
            (v_proj): Linear(in_features=1024, out_features=1024, bias=True)
            (q_proj): Linear(in_features=1024, out_features=1024, bias=True)
            (out_proj): Linear(in_features=1024, out_features=1024, bias=True)
          )
          (activation_fn): GELUActivation()
          (self_attn_layer_norm): LayerNorm((1024,), eps=1e-05, elementwise_affine=True)
          (encoder_attn): TrOCRAttention(
            (k_proj): Linear(in_features=768, out_features=1024, bias=True)
            (v_proj): Linear(in_features=768, out_features=1024, bias=True)
            (q_proj): Linear(in_features=1024, out_features=1024, bias=True)
            (out_proj): Linear(in_features=1024, out_features=1024, bias=True)
          )
          (encoder_attn_layer_norm): LayerNorm((1024,), eps=1e-05, elementwise_affine=True)
          (fc1): Linear(in_features=1024, out_features=4096, bias=True)
          (fc2): Linear(in_features=4096, out_features=1024, bias=True)
          (final_layer_norm): LayerNorm((1024,), eps=1e-05, elementwise_affine=True)
        )
      )
    )
  )
  (output_projection): Linear(in_features=1024, out_features=50265, bias=False)
)
```

)' loaded.