Sheet 9

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a)

- Sentence Order Prediction (SOP): SOP directly teaches the model to grasp sentence relationships
- Token Boundary Detection (TBD): It can benefit from learning more explicitly where natural word boundaries occur (especially for german)

b)

- Prefix Sequence Reconstruction (PSR): ability to encode earlier parts of a sequence conditioned on later context
- Future Span Prediction (FSP): FSP forces the model to reason about future sequences more holistically instead of predicting token-by-token

2 Under the hood of LLMs: Llama 2.7B

Initilazation of the model

```
from transformers import AutoTokenizer, AutoModelForCausalLM
In [ ]:
        import torch
        device = torch.device('cuda:0' if torch.cuda.is_available() else 'cpu')
        #with open("access_token.txt", "r") as f:
            access_token = f.read().strip()
        access_token = "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"
        model = "meta-llama/Llama-2-7b-chat-hf"
        tokenizer = AutoTokenizer.from_pretrained(model, token=access_token)
        model = AutoModelForCausalLM.from pretrained(model, token=access token, torch dt
        print(model)
In [2]: for id in range(5100, 5110): #Printing the tokens 5100 up to 5110 for example
            print(f"{id=}, {tokenizer.decode([id])}")
        print("\ntokenizer length:", len(tokenizer)) #the whole model contains 32000 to
        sun_id = tokenizer.encode("sun", return_tensors="pt")[-1] #find the token number
        print(f"\n{sun_id=}")
```

```
print(tokenizer.decode(sun_id))
        emb = model.get_input_embeddings()(sun_id.to(device))
        print("embedding shape:", emb.shape) #shape of the token
       id=5100, compet
       id=5101, pair
       id=5102, inglés
       id=5103, Response
       id=5104, Fig
       id=5105, grad
       id=5106, documentation
       id=5107, cant
       id=5108, appreci
       id=5109, ån
       tokenizer length: 32000
       sun_id=tensor([ 1, 6575])
       <s> sun
       embedding shape: torch.Size([2, 4096])
In [3]: sequence = "My favorite composer is"
        model_inputs = tokenizer(sequence, return_tensors="pt").to(device) #transforms i
        print(tokenizer.decode(model_inputs["input_ids"].tolist()[0])) #view decoded tok
        with torch.no_grad():
            outputs = model(**model_inputs) #the model gets the input and generates the
        logits = outputs['logits'][0, -1, :] #gets unnormalized probability for every to
        print("\nlogits shape:", logits.shape)
        probabilities = torch.nn.functional.softmax(logits, dim=-1) #normalizes
        top_prob, top_ind = torch.topk(probabilities, top_k)
        print("\nOutputs:\n")
        for i in range(top k):
            print(f"{tokenizer.decode(top ind[i].tolist())}: {top prob[i]:.2f}") #prints
       <s> My favorite composer is
       logits shape: torch.Size([32000])
       Outputs:
       Moz: 0.25
       Ch: 0.11
       Be: 0.09
       Ludwig: 0.08
       Fr: 0.03
       Wolfgang: 0.02
       Ig: 0.02
In [4]: import regex as re
        sequence = ""
        model_inputs = tokenizer(sequence, return_tensors="pt").to(device) #input is emp
        generated_answer = ""
        for _ in range(30): #iterative generation of the next 30 tokens
```

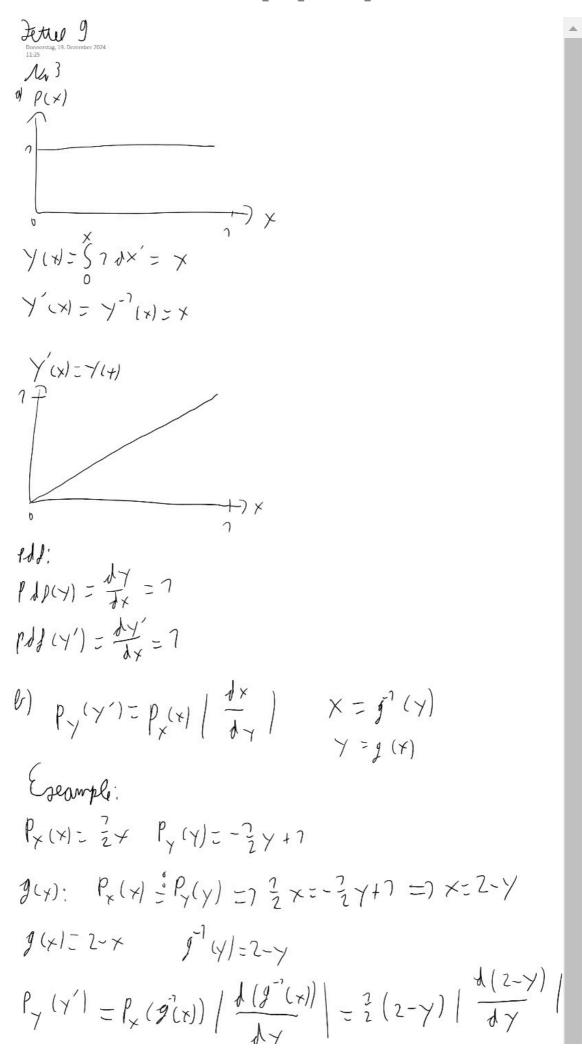
```
with torch.no_grad():
    outputs = model(**model_inputs) #generate output bases on input
    logits = outputs['logits'][0, -1, :] #unnormalized probabilities for the
    probabilities = torch.nn.functional.softmax(logits, dim=-1) #normalizes
    next_token_id = torch.argmax(probabilities).unsqueeze(0) #take the token
    model_inputs["input_ids"] = torch.cat([model_inputs["input_ids"], next_t

#printing output/generated answer in readable form for every iteration
    next_word = tokenizer.decode(next_token_id.tolist())
    next_word = re.sub(r"[^a-zA-Z0-9.?!]", "", next_word)
    generated_answer += next_word
    generated_answer += ""

print(generated_answer)
```

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3 Flow-based modeling



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(b)

```
import numpy as np
import matplotlib.pyplot as plt

# load the 1d samples:
samples = np.load("data/samples_1d.npy")

x_lin = np.linspace(0, 2, 1000)
plt.hist(samples, bins=50, density=True)
plt.plot(x_lin, 1/2 * x_lin, label="pdf(x) = 1/2*x, x in [0,2]")
plt.legend()
plt.show()
```

