NLP Project: Woby's Spooky Tales



George Washington University
Natural Language Processing - DATS 6312
Group 4
Joshua Ting

https://github.com/justjoshtings/Final-Project-Group4

Introduction

- Recent language models have the ability to produce text in various genres and domains where humans are not aware they are computer generated.
- Horror stories have been an integral part of humanity's outlet to explore and bring to life our collective deepest fears and imagination.

Research Question

Can we create a language model that can generate coherent horror stories for readers who enjoy a good scare?

What Makes a Good Horror Story?

A well written horror story will need to understand:

- The semantics and structure of a particular language
- Different nuanced elements that elicit a response from our primal fight or flight instincts

Prior Similar Works

 In 2017, a group of people leveraged deep learning to create Shelley, a Twitter bot to complete scary stories from Twitter users.

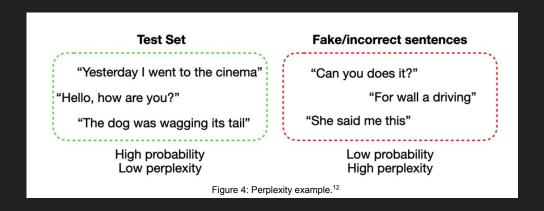
 The actual architecture of Shelley was never released to the public but since this was developed in/prior to 2017, it is likely some sort of RNN architecture.



Corpus Source

Number	SubReddit Name	Link		
1	r/nosleep	<u>Link</u>		
2	r/stayawake	Link		
3	r/DarkTales	Link		
4	r/LetsNotMeet	<u>Link</u>		
5	r/shortscarystories	Link		
6	r/Thetruthishere	Link		
7	r/creepyencounters	<u>Link</u>		
8	r/truescarystories	Link		
9	r/Glitch_in_the_Matrix	Link		
10	r/Paranormal	<u>Link</u>		
11	r/Ghoststories	Link		
12	r/libraryofshadows	Link		
13	r/UnresolvedMysteries	<u>Link</u>		
14	r/TheChills	Link		
15	r/scaredshitless	Link		
16	r/scaryshortstories	<u>Link</u>		
17	r/Humanoidencounters	Link		
18	r/DispatchingStories	<u>Link</u>		

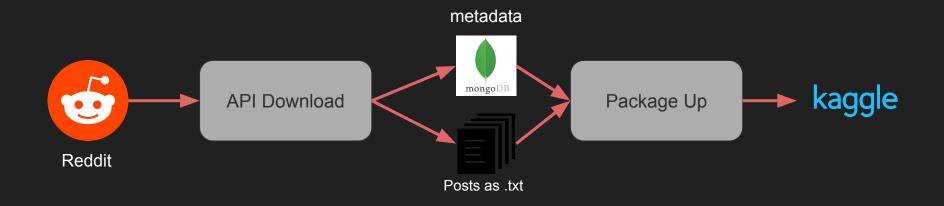
Metrics: Perplexity



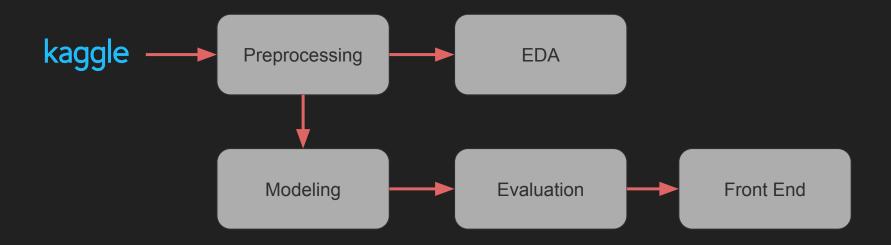
$$P(M) = e^{H(L,M)}$$

where, H is cross entropy loss, M is the true language model, L is generated data

Code Architecture: Data Acquisition Pipeline



Code Architecture: Preprocessing & Modeling Pipeline



Data Acquisition

- Python Reddit API Wrapper (PRAW)⁶
 - Official Reddit API Wrapper
 - Runs into issues of 1000 posts limit per SubReddit available
- Python Pushshift.io API Wrapper (PSAW)⁷
 - Community open-sourced API wrapper
 - Collects more posts than PRAW but can run into issues where certain hosting shards are not online
- We used both in order to have the widest coverage then removed any duplicates afterwards
- ~90 MB, ~14k posts

Data Storage: Local Disk/MongoDB

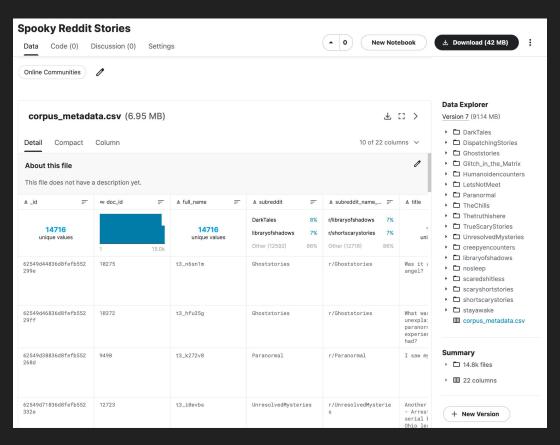
Metadata Saved to MongoDB

```
data_dict = {
            'doc id': ,
            'full_name': ,
            'subreddit': ,
            'subreddit_name_prefixed': ,
            'title': ,
            'little_taste': ,
            'selftext': ,
            'author': ,
            'upvote_ratio': ,
            'ups': ,
            'downs': ,
            'score': ,
            'num_comments': ,
            'permalink': ,
            'kind': ,
            'num_characters': ,
            'num_bytes': ,
            'created_utc': ,
            'created human readable': ,
            'filepath': ,
            'train test':
```

Post Text Saved as .txt to Disk

```
FINAL-PROJECT-GROUP4
   Code
    Woby_Log
    Corpus
        -nosleep
            1_t3_diuucz.txt
            2_t3_dyqd5e.txt
         -creepyencounters
            5931_t3_i31009.txt
            5931_t3_i31009.txt
            . . .
         -Ghoststories
            9845_t3_jdedeb.txt
            9846_t3_hvu2ko.txt
```

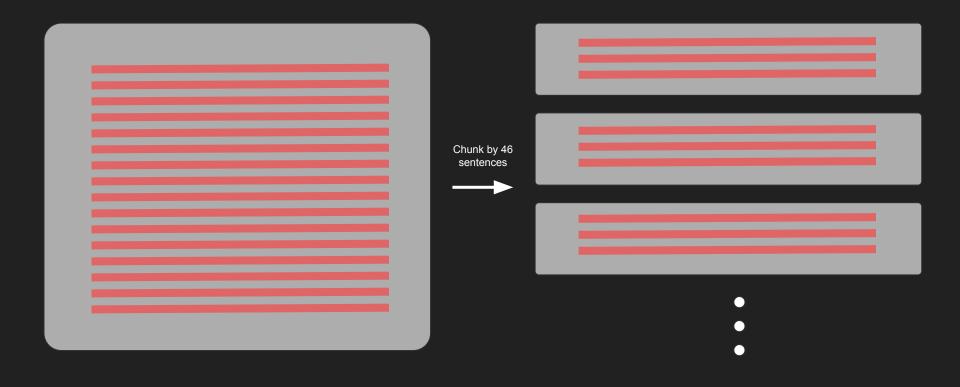
Data Storage: Kaggle



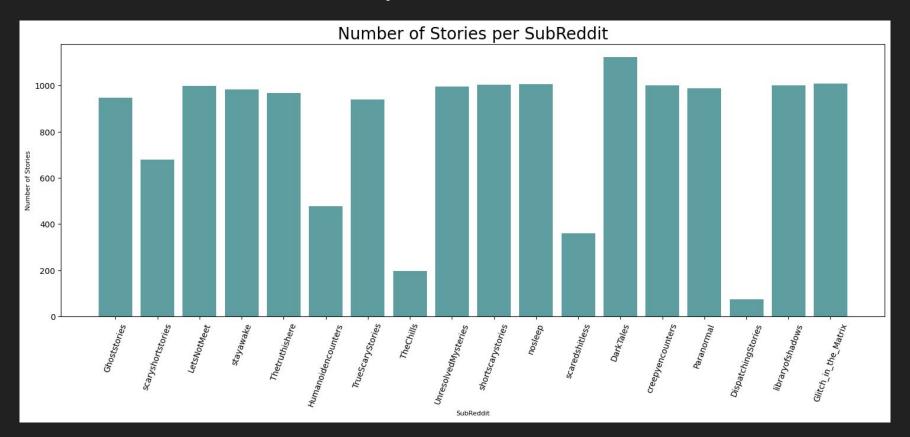
Data Preprocessing: Cleaning

Step	Cleaning Task	Description			
1	Remove all text after: "TLDR", "TLDR:", "TL;DR", "TL DR", "TL DR:".	TLDR stands for Too Long Didn't Read and the text that follows often is not part of the actual story.			
2	Remove all links	Not relevant for stories.			
3	Remove "&" and " "	These are HTML elements that are not needed in our corpus.			
4	Remove "***" or more *.	These are often used for formatting purposes and are not needed for our corpus.			

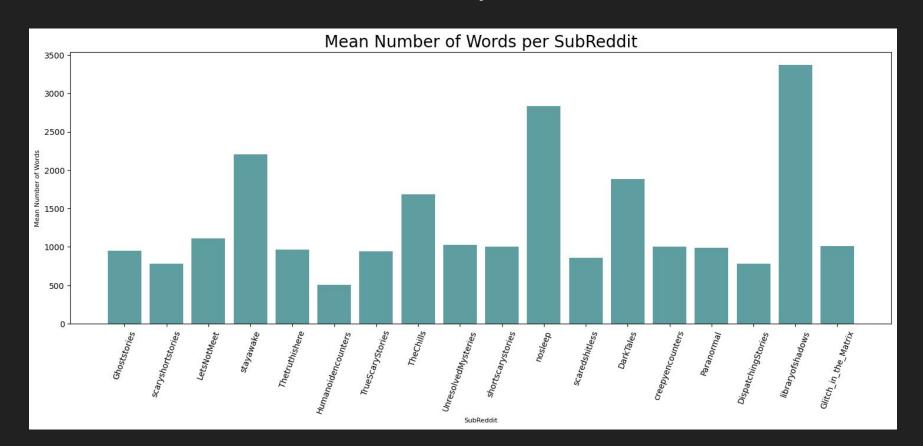
Data Preprocessing: Sentence Chunking



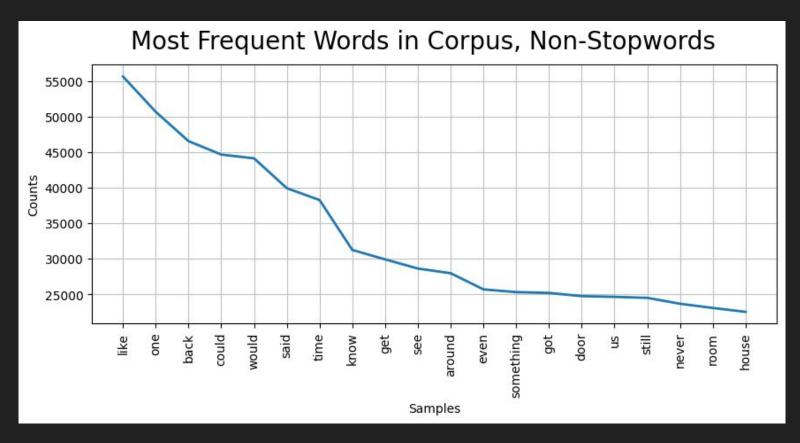
EDA: Number of Stories per SubReddit



EDA: Mean Number of Words per SubReddit



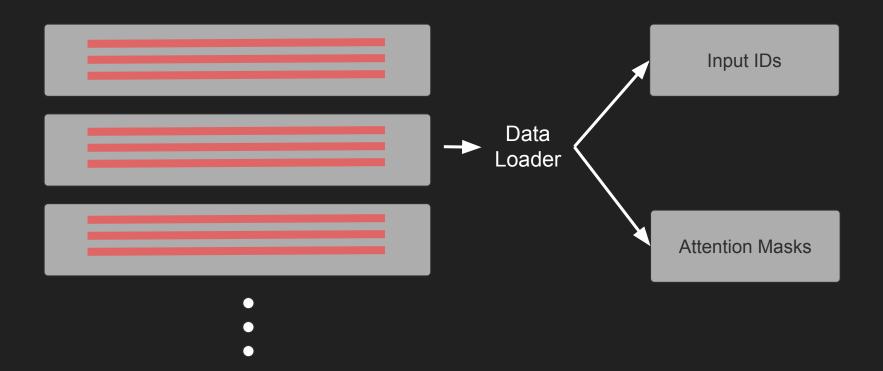
EDA: Most Frequent Words in Corpus



Modeling Options

- 1. **GPT2** fine tuned on our corpus for text generation.8
- 2. **GPT-NEO** fine tuned on our corpus for text generation.⁹
- 3. A **custom GPT2 variant** pretrained on our corpus then again fine tuned on our corpus for text generation, GPT2Spooky

Data Loader



GPT2

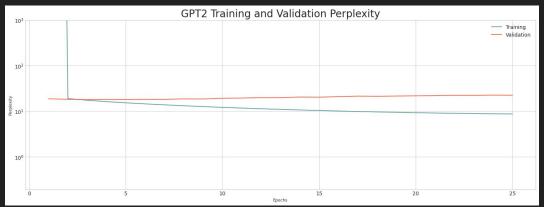
- OpenAl
- 1.5 billion parameters
- Trained on a dataset of 8 million web pages
- Predict the next word
- Hugging Face's 'gpt2' configuration: 117M parameters

GPT2 Fine Tuning

Configuration	Value				
Tokenizer	from transformers import GPT2Tokenizer				
Model Head	from transformers import GPT2LMHeadModel				
Optimizer	from torch.optim import AdamW				
Custom Tokens	bos_token='< startoftext >', eos_token='< endoftext >', pad_token='< pad >'				
Number of Epochs	25				
Learning Rate	5e-5				
Learning Rate Scheduler	Linear				
Batch Size	1				
Max Input Length	768 Tokens				
Model Type	'gpt2'				
Seed	42				
Loss	Cross Entropy				
Metric	Perplexity, Equation (3)				
Number of Parameters	117M				
Pretrained On	8 million web pages				

GPT2 Training





GPT-Neo

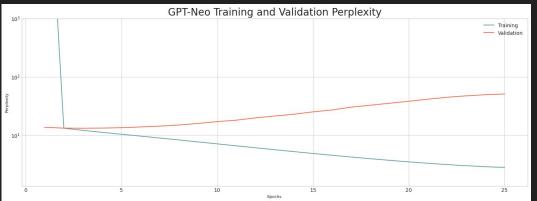
- Released in the <u>EleutherAl/gpt-neo</u> repository by Sid Black, Stella Biderman, Leo Gao, Phil Wang and Connor Leahy
- Similar to GPT2 except that GPT Neo uses local attention in every other layer with a window size of 256 tokens
- Trained on the <u>Pile</u> dataset (>800GB)
- GPT-Neo has several model versions with the largest being 2.7B parameters

GPT-Neo Fine Tuning

Configuration	Value				
Tokenizer	from transformers import GPT2Tokenizer				
Model Head	from transformers import GPTNeoForCausalLM				
Optimizer	from torch.optim import AdamW				
Custom Tokens	bos_token='< startoftext >', eos_token='< endoftext >', pad_token='< pad >'				
Number of Epochs	25				
Learning Rate	5e-5				
Learning Rate Scheduler	Linear				
Batch Size	1				
Max Input Length	1024 Tokens				
Model Type	'EleutherAl/gpt-neo-125M'				
Seed	42				
Loss	Cross Entropy				
Metric	Perplexity, Equation (3)				
Number of Parameters	125M				
Pretrained On	Pile dataset				

GPT-Neo Training





GPT2Spooky

- Custom pretrained model on our own corpus
- Uses the same architecture as the first model of GPT2, except it is completely pretrained on just our own custom corpus.

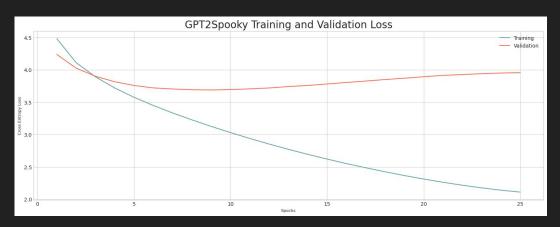
GPT2Spooky Custom Corpus Pretraining

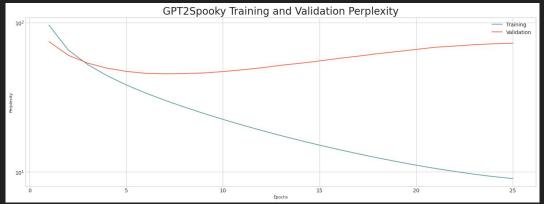
Configuration	Value				
Tokenizer	from tokenizers import ByteLevelBPETokenizer from tokenizers.implementations import ByteLevelBPETokenizer from transformers import GPT2TokenizerFast				
Custom Tokens	bos_token='< startoftext >', eos_token='< endoftext >', pad_token='< pad >'				
Model Head	from transformers import GPT2LMHeadModel				
Vocab Size	8000				
Number of Attention Heads	12				
Number of Hidden Layers	6				
Optimizer	from torch.optim import AdamW				
Number of Epochs	5				
Batch Size	64				
Max Input Length	512 Tokens				
Seed	42				

GPT2Spooky Fine Tuning

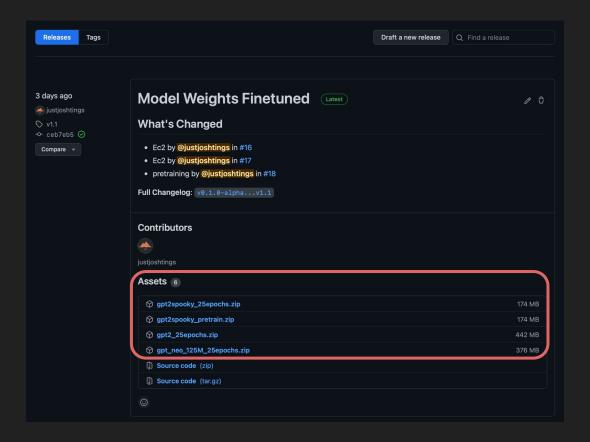
Configuration	Value				
Tokenizer	from transformers import GPT2Tokenizer				
Model Head	from transformers import GPT2LMHeadModel				
Optimizer	from torch.optim import AdamW				
Custom Tokens	bos_token='< startoftext >', eos_token='< endoftext >', pad_token='< pad >'				
Number of Epochs	25				
Learning Rate	5e-5				
Learning Rate Scheduler	Linear				
Batch Size	1				
Max Input Length	512 Tokens				
Model Type	'gpt2spooky'				
Seed	42				
Loss	Cross Entropy				
Metric	Perplexity, Equation (3)				
Number of Parameters	5M				
Pretrained On	Custom Corpus (Scary Stories)				

GPT2Spooky Training





Model Distribution



Model Evaluation: Perplexity

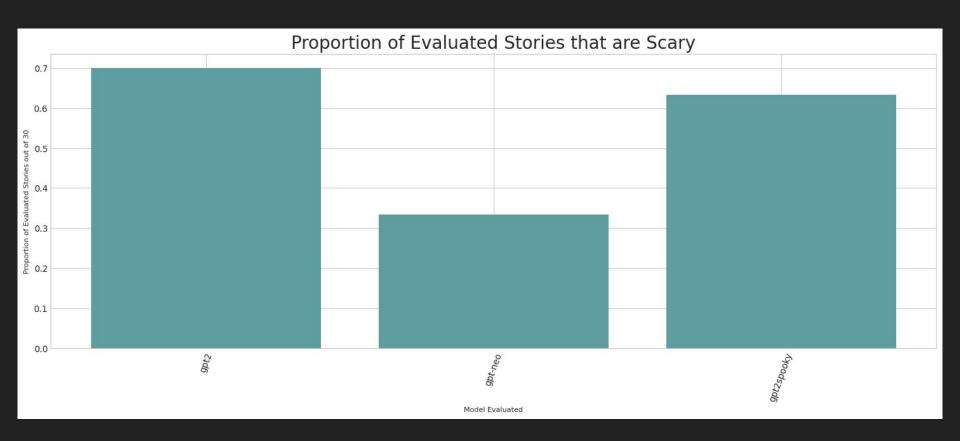
After 25 epochs:

- 1. GPT2 = 22.5
- 2. GPT-Neo = 50.7
- 3. GPT2Spooky = **72.7**

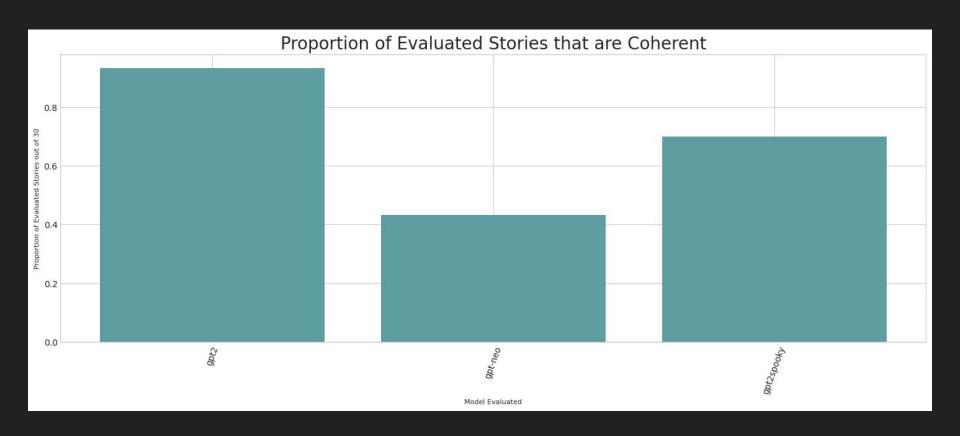
Model Evaluation: Human Evaluation

prompts	gpt2_25_generate	gpt_neo_25_generate	gpt2spooky_generate	gpt2_25_scary	gpt_neo_25_scary	gpt2spooky_scary	gpt2_25_choherent	gpt_neo_25_coherent	gpt2spooky_coherent
Lorem ipsum	Lorem ipsum	Lorem ipsum	Lorem ipsum	1	0	0	0	1	1
Lorem ipsum	Lorem ipsum	Lorem ipsum	Lorem ipsum	0	1	1	1	0	1

Model Evaluation: Human Evaluation



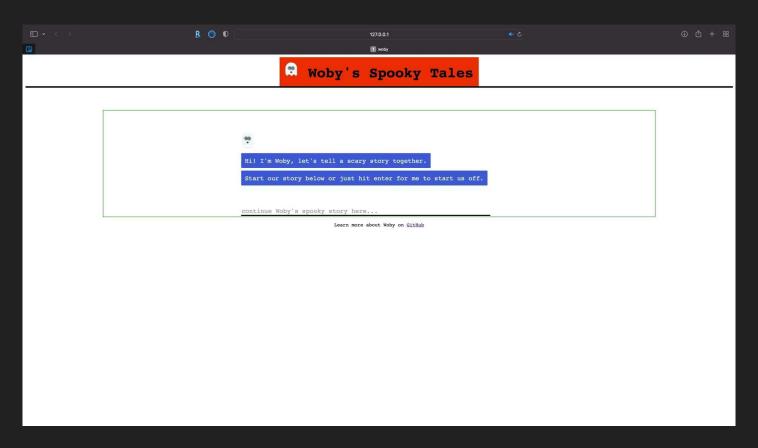
Model Evaluation: Human Evaluation



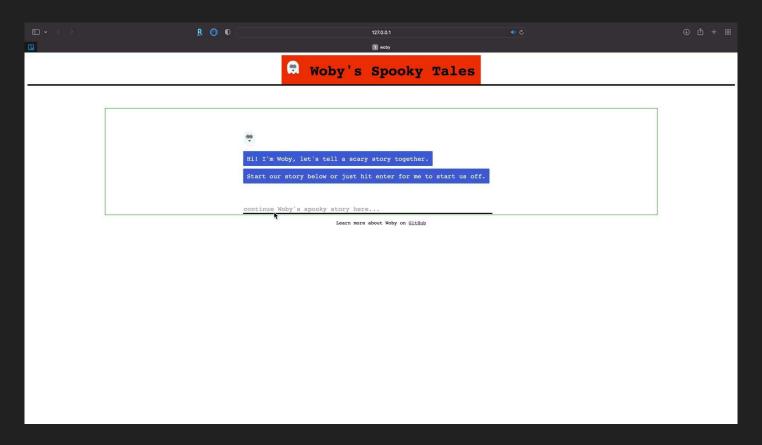
Conclusion

- Succeeded in our objective to train and compare several models to perform the downstream task of horror story generation
- Best model: GPT2 fine tuned
- Performance of GPT2 and GPT2Spooky are not too far off
- More tweaking and fine tuning, GPT2Spooky may perform close to GPT2 while being much smaller

Front End Chat Bot: Demo



Front End Chat Bot: Demo



References

- 1. Vaswani et. al (2017) Attention is All You Need
- 2. Glue Benchmark
- 3. Devlin et. al (2018) BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding
- 4. Radford et. al (2018) Improving Language Understanding by Generative Pre-Training
- 5. Shelley: Human-Al Collaborated Horror Stories
- 6. Python Reddit API Wrapper
- 7. Python Pushshift.io API Wrapper
- 8. Hugging Face GPT2
- 9. <u>Hugging Face GPT-Neo</u>
- 10. <u>Hugging Face Transformers</u>
- 11. PyTorch
- 12. Perplexity in Language Models
- 13. <u>Hugging Face Transformers Fine Tuning</u>
- 14. Hugging Face How to Text Generation
- 15. <u>Hugging Face Text Generation</u>
- 16. Fine Tune GPT2
- 17. Flask Chatbot
- 18. <u>Conditional Text Generation for Harmonious Human-Machine Interaction, Guo et al., 2020</u>
- 19. Transformers: State-of-the-Art Natural Language Processing, Wolf et al., 2020
- 20. <u>Transformer-based Conditional Variational Autoencoder for Controllable Story Generation, Fang et al., 2021</u>
- 21. <u>Improving Neural Story Generation by Targeted Common Sense Grounding, hhmao et al., 2019</u>
- 22. Evaluation of Text Generation: A Survey, Celikyilmaz et al., 2021