Abstract

As neural language models grow in effectiveness, they are increasingly being applied in real-world settings. However, these applications tend to be limited in the modes of interaction they support. In this extended abstract, we propose Wordcraft, an AI-assisted editor for story writing in which a writer and a dialog system collaborate to write a story. Our novel interface uses few-shot learning and the natural affordances of conversation to support a variety of interactions. Our editor provides a sandbox for writers to probe the boundaries of transformer-based language models and paves the way for future human-in-the-loop training pipelines and novel evaluation methods.

1 Introduction

In this work, we introduce Wordcraft, a text editor with a built-in creative writing assistant. For an assistant to be useful to creative writers, it ought to be able to handle simple tasks like appending text to the end of a passage as well as complex ones such as suggesting alternative wordings, modifying text style, and inserting flavour text about a particular object or person in the scene. Since it would be costly to train separate language models for each of these features and make them all available through a single interface, we propose using few-shot learning techniques to build an assistant powered by a single language model. Our tool supports a variety of story writing tasks such as continuation, infilling, and rewriting, while also providing users the ability to create custom tasks on the fly. In addition, we explore the unique advantages of using dialog models over general-purpose language models for building such a tool.

Story writing is an ideal sandbox setting in which to explore the abilities and limitations of language models as unintended outputs can be creative opportunities. From an HCI perspective, we intend to use Wordcraft to learn how people interact with language models, what people ask them to do, how well the models can deliver, and how that feeds back into what people ultimately create. We also plan to investigate methods for incorporating human feedback back into the training loop, collecting dynamic datasets that can be used for further training and evaluation.

2 Related work

Neural language models have been applied to a variety of creative tasks, including text-adventure games, collaborative slogan writing, and story writing. All of these systems ultimately boil down to a series of alternating turns between the user and the machine adding more text to the end of the story. There has been extensive work moving beyond the continue-my-text generation paradigm by incorporating additional control signals, such as event sequences, desired topic, and story title. Fill-in-the-blank tasks have also been proposed. However, most of these works that give users control beyond left-to-right generation require explicit training, making it difficult to support a variety of interaction types.

Few-shot learning, where a series of examples of the task the model is supposed to perform are passed in to the model as the input context, was made popular by Brown et al. (2020). The idea of using dialog models to develop a fantasy story was introduced by Urbanek et al. (2019). While in their framework the dialog takes place between two characters in the story, in our framework the conversation is between the storyteller and an agent.

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1AI Dungeon: https://play.aidungeon.io/
who responds to the storyteller’s requests.

3 The Wordcraft Editor

We set out to build a text editor from scratch that is able to provide NLG support to users at several stages of story creation:

1. Planning: sketching an outline for the story
2. Writing: getting words down
3. Editing: rewriting existing text

Our web interface (Figure 1) consists of a traditional text editor on the left and a few key commands that trigger requests to the NLG assistant. The commands available differ depending on whether the user is currently editing text or instead has selected a section they would like the assistant to focus on.

We designed the interface to feel familiar. Users are presented with only a blank screen at the beginning, like any text editor. As they write text, they are shown options on the right (along with their keyboard shortcuts) for summoning the NLG assistant to perform one of the supported interactions (Section 5). Users can also try out their own queries by modifying the prompt.

4 Dialog as a Generation Method

In order to support collaboration between a human writer and the AI writing assistant, our Wordcraft editor uses an open-ended dialog system for generation. This allows the UI to have functionality where the human writer can ask the AI to do story-specific tasks like “Help me describe the elderly man’s emotional state” in addition to providing button-press support for standard tasks like continuation and elaboration. The dialog system that we use is Meena (Adiwardana et al., 2020), a language model that is broadly capable of following instructions and answering questions posed in a conversational format. In small-scale qualitative studies, we find Meena compares favorably to a similarly sized general-purpose language model (GPLM) that was not trained explicitly for dialog. We hypothesize that open-ended dialogue systems are better suited to interactive writing than GPLMs, and we plan to conduct more comprehensive experiments to investigate this in the future.

For both Meena and the GPLM, we explored the use of few-shot learning to improve the language models’ abilities to perform story writing tasks. Meena is designed to take as input the previous turns in a conversation and predict the next turn. The GPLM is designed to take the previous text in a passage then continue it. In both cases, the language model can be influenced to perform a desired task by including in the model’s input context several examples of the task it should be performing. This is known as few-shot learning. For Meena, these examples are formulated as a conversation between a story writer and an assistant. For the GPLM, they are formulated as a block of text with each input and its expected output on a new line. Figure 2 shows several examples from our few-shot learning prompts for Meena.

In our early experiments, we found it very natural to construct few-shot learning contexts for the dialog model; humans are intuitively familiar with the conversational format. We also found that the outputs of the dialog model tended to be more interpretable. When the dialog model does not un-
An elderly man was sitting alone on a dark path. A lightning bolt lit up the sky. With a swift movement he pulled ___ out of his jacket.

User's Story:
An elderly man was sitting alone on a dark path. A lightning bolt lit up the sky. With a swift movement he pulled ___ out of his jacket.

Meena's prediction:
> an amulet

User's desired task:
replace the highlighted section using 2 words.

Prompt for Meena:
[filling conversational context]
> Here's another story: An elderly man was sitting alone on a dark path. The man looked to light as a figure ran up to him. The figure was holding a sword. The old man recognized the young man as [BLANK]. The young man said he had had the strangest dream.} Fill in the blank with 2 words.

> His son

Meena's prediction:
> an amulet

Figure 2: For each task, we handcrafted several turns of staged conversational context. The first two examples in each context are shown here. For “continuation,” we did not employ a few-shot prompt.

Figure 3: Example of prompt construction for text infilling task. The user’s story is the text they have written in the editor. The user has asked the editor to rewrite the phrase ‘a sword.’ The editor appends the user’s story to the corresponding staged conversation context (Figure 2), and asks the dialog model for a response.

Understand the task, or requires more information, it often simply says so. For example, given the following one-turn context:

Here's my story so far: 'An elderly man was sitting alone on a dark path. A lightning bolt lit up the sky.' Help me describe the elderly man's emotional state.

The dialog model responds with:
I don't really get what you are asking. Is it a happy lightning bolt? Or a dark cloud lightning bolt? What was the man doing before the lightning bolt? Did he do something that may have sparked his emotions and thoughts?

Such responses can help users figure out how to improve their prompt. In contrast, when the GPLM was provided confusing prompts, we felt it more often generated nonsensical continuations.

In addition, users may want to ask the model to perform a task for which it is not easy to design a few-shot template. As part of a conversation, a user might ask: “What should the character do to resolve the conflict?” In this situation, the user is looking for a writing partner, not a model that simply continues their text.

In continuing this line of research, we plan to conduct a more systematic study of how users interact with Meena versus the GPLM.

5 Supported Interactions
In this section, we describe several of the interactions built into Wordcraft.

5.1 Continuation
Continuation is the most basic task that language models are trained to perform. The GPLM can do story continuation very successfully without any few-shot prompting (see Appendix A.2 for examples). However when Meena is prompted with:

An elderly man was sitting alone on a dark path.

It responds:
I don’t understand, could you explain? And this man you describe is me?

Here we see Meena gamely attempting to respond to a rather strange query; the resulting outputs might be reasonable for a dialog system but are not helpful in the context of story writing. To address this, we wrap the input in descriptive text:

I don’t really get what you are asking. Is it a happy lightning bolt? Or a dark cloud lightning bolt? What was the man doing before the lightning bolt? Did he do something that may have sparked his emotions and thoughts?

Such responses can help users figure out how to improve their prompt. In contrast, when the GPLM...
Here is my story so far: 'An elderly man was sitting alone on a dark path.'
Give me the next sentence.

Now Meena’s responses are much more useful:
The air was cold but he was warm inside.
He had come across a clearing, and in the center of it was an ancient old willow tree.
Advancing towards him, the man could see a small figure, illuminated by a light from the distance.
The moonlight glistened on the full, rounded moon.

5.2 Infilling

Infilling is a harder task, but potentially very useful for writers who would like to have their assistant propose alternative wordings to some section of their story. Suppose a writer has written the following but wants new ideas about what the man saw:
An elderly man was sitting alone on a dark path. Suddenly he saw a whitetail doe. It was beautiful.
The writer selects the words they wish to experiment with and wraps up the sentence like this:
Here’s another story: ‘An elderly man was sitting alone on a dark path. Suddenly ______. It was beautiful.’
Fill in the blank with 4 words.
In this case, Meena provides responses like:
There was a blinding flash of light.
What’s the story about? Why does he need the four words? What’s the conflict?
an angel showed up to make friends, it was beautiful. I have no idea what this is for but I like where its going.
These responses are interesting, but less than ideal.
The second response demonstrates the problem of the model generating meta text, that is, talking about the story, rather than writing the story.

To minimize the production of meta text, we construct a few-shot learning prompt with several examples of infilling. The start of this prompt is shown in Figure 2.

When priming with this prompt, Meena is able to generate outputs which, while not precisely four words, do make sense when used to fill the blank:
He was old, but you couldn’t help but notice his youthful appearance; his longish hair and his warm eyes.
The elderly man is a short man, with gray hair and thin, almost frowzy.

5.2.2 Rewriting

We also experimented with rewriting to change the tone. The current story and a desired tone are appended to the few-shot learning prompt shown in Figure 2. For example, we might append:
Here is some text: An elderly man was sitting alone on a dark path. Please rewrite it to be more descriptive.
In this case, the model responds with:
The elderly man was sitting alone. The darkness around him was oppressive and heavy.

When instead asked to make the story more humorous, the model responds with:
The elderly man was walking to his house, but a little fox ran up to him and started laughing at his outfit.

6 Discussion and Future Work

We see Wordcraft as a launching point for deeper investigations into the strengths and weaknesses of using language modes for multi-purpose creative writing assistants. The stories users created (Appendix A.1) show the broad potential for human-AI collaboration across different stages of the writing process: from shaping the narrative all the way down to rephrasing a sentence. Early users attested
to the tool inspiring them to take their stories in dire-
ctions they would not have otherwise considered.

We also encountered failure modes. Both Meena
and the GPLM produced outputs of inconsistent
quality, some superb, some nonsensical. On top
of that, slight tweaks in prompt phrasing caused
large deviations in output quality. It is also worth
noting that language models have extensive docu-
mented issues with bias (Dhamala et al., 2021) and
memorization (Carlini et al., 2020) that will need
to be carefully addressed before Wordcraft can be
brought to larger audiences. We also found that for
harder tasks like infilling, finetuned models, even
when much smaller in parameter count, tended to
outperform few-shot learning with larger models.

Our next plans for Wordcraft include conducting
more formal user studies with the goal of better
understanding what writers want so we can make
Wordcraft more useful. We also plan to investigate
light-weight finetuning methods which obviate the
computationally-expensive need to keep in memory
one finetuned model per task (Lisa Li and Liang,
2021). With such methods, we can imagine fine-
tuning custom tasks for each user or finetuning a
model to predict the corrections that users make to
model outputs. Lastly, we plan to look into how
the data collected from usage of Wordcraft can be
used for evaluation and model training.

Wordcraft is still in prototype phase, but we have
already collected a unique corpus of stories written
with the tool that showcase human-AI collabora-
tion. We conclude with one such story. The text in
blue is generated by Wordcraft through interaction
with a human writer; text in black represents text
typed in directly by the human writer.

There was once a fairy living in a datacenter. It
was a place filled with lights and noise, and the
fairy loved it. The fairy lived on a mountain of
Ethernet cables. She liked the way they glittered in
the fluorescent artificial light that filled the room.
The noise from the constant chatter of the servers,
like a babbling brook, lulled her to sleep. One
day she was woken abruptly by a sudden burst of
silence. She looked up to the ceiling from where
she slept curled up by a large mainframe. The
air was suddenly silent. No fans, no servers, no
blinking lights. Just silence. The babbling brook
that the fairy loved had stopped flowing, and she
lay terrified in the still, soundless waters. The fairy
was a hardy and resourceful spirit so she wasn’t
completely distraught. She took a deep breath
and calmed herself in the knowledge that the
engineers from the ISP were probably working on
a fix. Could she help them? The fairy made herself
small and floated into an open ethernet port. She
found herself flying over wires and circuits. As
her brain processed the images and information,
her eyes grew wider and wider; taking in the 0s
and 1s at an incredible rate. And then she found
it, the bug that had caused her server rack to go
mute. It was a stray bit of code. It had somehow
made it through the system without being caught.
It was a simple little mistake, but the effect was
catastrophic. She would correct that stray bit, and
she would fix the servers. The engineers glued to
their dashboards were shocked when their plots
and logs began singing to life once again. What
they didn’t know was that the ethereal grace of a
fairy’s touch had fixed the network with a few well
placed bits of binary.

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A Appendix

A.1 Example Stories

Below are stories written in Wordcraft (text in blue is generated by Meena).

An elderly woman was sitting alone on an illuminated path. A man in a top hat approached and introduced himself. ‘I’ve been looking for you’, he said. There was a moment’s pause. ‘What do you want?’ the elderly woman asked. The man smiled and extended his hand. ‘I want to ask you a few questions. First, what do you think of young men in top hats?’ The woman said she thought young men in top hats were alright. But that they can be a little silly sometimes. The man grinned and held up a pair of small top hats the size of coasters. ‘Would you like one?’ the woman nodded yes and the man presented his tiny top hat collection to the now smiling old woman.

A young man was sitting with his friends on a park bench. He was thinking deeply about the events which had occurred earlier that day. He and his friends were hanging out smoking cigars when an attractive girl who was smoking a cigar herself joined them. He was struck by her cool demeanour and her natural beauty. She also had a cute laugh that he couldn’t get out of his head. He had never seen a girl just smoking a cigar. He had to find out who she was. It turns out the girl’s name was Anna, and she was an archeologist travelling the world studying ancient artifacts. She was making a brief stop in the man’s neighborhood in order to meet with a collector of ancient artifacts. She had already met with him before, and it had gone well. He was friendly and interested in her work, and he was also very wealthy, which was convenient because the entire reason she was in town was to sell some artifacts. She needed the money to continue to fund her research. She had found a priceless piece that was of great historical significance.

A small man was sitting alone on his front porch. A raven sat across the way, staring at him. It was black, as all ravens are, but had a peculiar white patch on the chest. There was also blood smeared across its feathers, from some sort of injury, the small man guessed. He called out, “Mister raven! Did you hurt yourself?” The bird stared at him in silence, unmoving, and the man felt foolish for being so kind in the face of such coldness from the animal, and for trying to talk to the raven at all. The small man decided to leave the bird alone and returned to his thoughts when a loud bang disturbed him from his reverie. The small man shot to his feet and looked around wildly.

A.2 GPLM Continuations

The GPLM is able to do continuations directly. When primed with:
An elderly man was sitting alone on a dark path.

**It proposes the following continuations:**

He wasn’t certain it was entirely safe to be walking the streets, given the amount of trouble that had happened in the past while here. A man with shaggy dark hair stood near the edge of the road. The elderly man stared at him for a moment. He wasn’t certain of which direction to go, and he’d forgotten both where he was traveling to and who he was. He’d sat down for a moment to rest his weary legs, and suddenly looked up to see an elderly woman before I came to stand at the end of the path and he looked up. “You have a light?” he asked, frail and withered as he was. “Yes sir,” I replied. “May I see it?” he asked. “I