Heteroglossia: In-Situ Story Ideation with the Crowd

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ABSTRACT
Ideation is essential for creative writing. Many authors struggle to come up with ideas throughout the writing process, yet modern writing tools fail to provide on-the-spot assistance for writers when they get stuck. This paper introduces Heteroglossia, an add-on for Google Docs that allows writers to elicit story ideas from the online crowd using their text editors. Writers can share snippets of their working drafts and ask the crowd to provide follow-up story ideas based on it. Heteroglossia employs a strategy called “role play”, where each worker is assigned a fictional character in a story and asked to brainstorm plot ideas from that character’s perspective. Our deployment with two experienced story writers shows that Heteroglossia is easy to use and can generate interesting ideas. Heteroglossia allows us to gain insight into how future technologies can be developed to support ideation in creative writing.

Author Keywords
Crowdsourcing; Creative Writing; Ideation; Role Play; Story

CCS Concepts
•Information systems → Crowdsourcing; •Human-centered computing → Human computer interaction (HCI); User studies;

INTRODUCTION
Storytelling is one of the oldest known human activities [39]. People engage in storytelling to communicate, teach, entertain, establish identity, or simply relate to each other in meaningful ways [33]. Storytelling is important, but writing a good story is a challenging and complicated task, and many creative writers struggle to come up with ideas throughout the process. Roland Barthes said: “A creative writer is one for whom writing is a problem.” Despite this common experience, research into writing support systems has long been focused on business and technical writing. Researchers have developed systems that can automatically generate follow-up text in an auto-complete manner [8]; decompose and re-compose complicated writing tasks [21]; outsource writing jobs to online crowds [3, 14], collaborators [36], or writers themselves [37]; or even allow the user to write a paper solely using a smartwatch [29]. However, these prior works were largely developed and tested for producing technical reports [36, 37, 29], Wikipedia-like essays [21], or business documents [13], rather than short stories or novels. One of the few exceptions is the work done by Kim et al., who created Ensemble [18] and Mechanical Novel [19]. These two systems pushed the boundaries of collaborative story writing, but did not focus on helping creative writers, who mostly write alone [26, 12, 32].

This paper introduces Heteroglossia¹, a crowd-powered system that allows writers to elicit story ideas simply using their text editors. Figure 1 overviews the Heteroglossia system, which we built as an add-on for Google Docs. A writer can select a part of a story draft as the prompt and ask the online crowd to provide follow-up story ideas based on it. Heteroglossia employs an ideation strategy called “role play,” where each worker is assigned a fictional character in the story and asked to brainstorm plot ideas from the character’s perspective. This work is motivated by the fact that role-playing and acting traditionally have had a role in the creative writing process [9]. Some professional novelists also use role play to help writing. Also relevant is the well-known “six hats” method, which asks people to wear metaphorical hats representing different thinking perspectives [10]. Teevan et al. proposed to use the six hats schema to assign different thinking roles to the authors themselves in order to promote self-reflection from different angles [38]. Chou et al. showed that perspective-taking can

¹Heteroglossia: a diversity of voices, styles of discourse, or points of view in a literary work and especially a novel [27].
RELATED WORK
This work is related to (i) crowd ideation, (ii) crowd writing, (iii) supporting creative writing, and (iv) crowd feedback.

Crowd Ideation
Prior work has used the online crowd as a source of new ideas, primarily for problem-solving and product design. Chan et al. introduced IdeaGens, an ideation system where a group of workers proposes ideas in real-time and the expert monitors the incoming ideas and provides instant feedback to the crowd [4]. Yu et al. explored using a schematic representation for the target design problem to guide the crowd to “think outside of the box” [42]. Online crowds were also used to provide real-time creative input during early-stage design activities [2].

Crowd Writing
Heteroglossia also builds upon the work in crowd writing, which aims to allow a group of people, including experts and non-experts, to work together to write an article. Many crowd writing projects focused on decompose and recompose complicated writing tasks. For example, the Knowledge Accelerator used a complex workflow where each worker contributes small amounts of effort to synthesize online information, generating a Wikipedia-like article for open-ended questions [14]. Soylent used a Find-Fix-Verify workflow to allow crowd workers to identify problems in a draft, propose solutions, and select the best solution for each identified problem [3]. MicroWriter decomposed the task of writing into three subtasks: idea generation, labeling, and writing [36]. Meanwhile, some other work has pushed the boundaries of the classic workflow approach for crowd writing. For example, WearWrite explored using wearable devices, such as smart watches, to guide a group of crowd workers to write articles [29]. Agapie et al. explored using local crowds to generate event reports [1]. However, these projects all focused on business or technical writing, rather than creative writing.

One of the few exceptions is the work done by Kim et al., who created Ensemble [18] and Mechanical Novel [19]. Ensemble is a volunteer-based collaborative story competition platform where Leaders set high-level creative goals and constraints for a story and Contributors participate in low-level tasks, such as drafting, commenting, and voting. Mechanical Novel embodies a more organic workflow, the “Reflect-and-Revise loop,” that allows crowd workers to revisit and revise their writing goal. These works pushed the boundaries of creative writing and helped to answer why collaborative novels at a scale similar to that of Wikipedia do not exist.

While collaborative writing has opened new possibilities, most writers still write alone. Professional novelists write alone [26], freelance writers write alone [12], and, even within an industry with a collaborative culture, many TV screenwriters still write alone [32]. Our goal is to assist creative writers, who often write alone, without drastically changing the way they work.

Supporting Creative Writing
A few researchers have developed technologies to support creative writing. Most of them focused on lower-level text generation or proofreading. For example, the Creative Help system used a recurrent neural network model to generate suggestions for creative writing [34]. The Scheherazade system was developed for interactive narrative generation [24]. InkWell produced stylistic variations on texts to assist creative writers [11]. More recently, Clark et al. studied machine-in-the-loop story writing and suggested that machine intervention should balance between generating coherent and surprising suggestions [8].

Crowd Feedback Systems
Researchers have also attempted to use online crowds to generate critiques and feedback. Xu et al. created Voyant, a system that used non-expert crowd workers to generate structured feedback on visual designs [40]. Their classroom study further demonstrated the effectiveness of using crowd feedback in the design process [41]. As for visual designs, Luther et al. also created CrowdCrit, a system that aggregated multiple critiques from non-expert crowd workers [25]. Luther showed in experiments that the critiques generated by CrowdCrit could help designers improve their design processes. On the other hand, some other researchers focused on generating writing feedback. For example, Huang et al. used workers from Amazon Mechanical Turk (Mturk), who are often fluent in English, to provide structural feedback for ESL writing [15].

HETEROGLOSSIA SYSTEM
Heteroglossia incorporates a web site to manage information and a Google Docs add-on for writing. Figure 2 shows the screenshot of each page of Heteroglossia website. Users start by creating characters (Figure 2A) and forming teams of characters (Figure 2B and 2C) on the Heteroglossia website. After setting up the characters and teams, users start writing the story in Google Docs. When they get stuck, users can select a story snippet to initiate an ideation task through Heteroglossia (Figure 2D) and acquire follow-up story ideas (Figure 2E).

Creating Characters
Figure 2A shows that to create a new character, users specify an image, name, and description. Notice that only the name and description will be shown to workers. Users can provide a detailed setting for a character in the description, such as inner goal, outer goal, and personality, to help workers understand the story background and come up with new story ideas. Editing and deleting an existing character can be done through the setting button in the upper right corner.

Forming a Team of Characters
A team represents a group of characters used in role play ideation. Figure 2B shows the interface of editing a team. Available characters are listed in the “Team Members” block.
Figure 2. The system overview of Heteroglossia. Heteroglossia incorporates a web site to manage information and a Google Docs add-on for writing. Users start by creating characters (A) and forming teams of characters (B and C) on the Heteroglossia website. After setting up the characters and teams, users start writing their own story in Google Docs. When they get stuck, users can select a story snippet to initiate an ideation task through Heteroglossia (D) and acquire follow-up story ideas (E).
are implemented on the worker interface: a 30-second time lock for HIT submission, a reach-to-the-bottom check for the story prompt, and a prohibition of copy-paste functionality in the idea pane.

Dynamic Payment for Workers

We proposed a formula to dynamically estimate working time and set up corresponding payment for workers. The estimation is based on two factors: reading comprehension and writing. The average reading speed of English native speakers is 200-300 words per minute with reasonable comprehension when using LCD monitors [35]. We empirically estimated that writing a fifty-word-long story idea in Heteroglossia takes approximately 5-6 minutes. Aiming at providing a $10 hourly wage, we implemented the formula as follows:

\[
\text{Cost(HIT)} = \text{Cost(Reading)} + \text{Cost(Writing)} = \$\left(\frac{\text{#words}}{1000}\right) + \$1.0
\]

where \#words refers to the word count of the story prompt. Heteroglossia then creates HITs with the reward dynamically computed according to the designed formula.

STUDY 1: THE EFFECTS OF ROLE PLAY STRATEGY

The goal of Heteroglossia is to provide inspiring ideas to creative writers, especially when they get stuck during writing. Heteroglossia particularly uses an ideation strategy called “role play.” To understand the effects of the role-play strategy and inform the design of Heteroglossia, we conducted two sets of experiments. We would like to answer these two questions that are motivated by literature: (i) can the role play strategy produce more useful story ideas? and (ii) what are some trade-offs of using this strategy?

Role Play Produces Semantically-Far Story Ideas

Per Chan et al. [6, 5], when a creator reaches an impasse, ideas that are semantically far from current working ideas are more helpful than those that are nearer. Chan’s work is powered by the Search for Ideas in Associative Memory (SIAM) theory [30] and verified with crowd ideation experiments. SIAM [30] assumes that idea generation is proceed in two stages, knowledge activation and idea production. In the first stage, an image will be retrieved according to the problem. The given image is assumed to have several features that is then used to generate ideas. Chan et al. [6, 5] showed that in the idea production stage, relevant stimulations help generate more ideas. However, after exhausting the related ideas, semantically far stimulations would help people to change the category of images and thus generate more ideas. Applying Chan’s conclusion to our system says that the theoretical prerequisites for resolving writer’s block are to come up with story ideas that with greater semantic distance from the current working draft. In this subsection, we conducted a set of experiments to examine if role-play strategy results in semantically distant ideas.

Pilot Study: The overview procedure of the pilot study is shown in Figure 4. We first conducted a pilot study using five Taiwanese folk stories\(^2\). These stories are unfamiliar to many crowd workers in order to simulate the workers’ sense of freshness when reading the stories. We then took the first 30% and first 60% of each story (based on word count) to simulate a writer’s working story drafts. For each of these ten (5 × 2) story drafts, we manually labeled the fictional characters in the story. Each story contained two or three characters. For each \((\text{story draft}, \text{character})\) tuple, we recruited five workers from MTurk to read the story draft and provide a follow-up story idea in free text from the character’s perspective. For comparison, we also recruited \((\#\text{character} \times 5)\) workers for each draft without assigning any characters, and asked workers to write story ideas. We paid $0.5 for 30% story drafts and $0.8 for 60% story drafts. Each worker was allowed to work on each story once, i.e., five was the maximum. In total, 105 workers participated in our pilot study.

The pilot study leads to four main findings:

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\(^2\)Taiwanese folk stories were retrieved from TaiwanDC (https://www.taiwandc.org/folk.htm). We used “The Legend of Sun-Moon Lake,” “The Legend of Muddy River,” “The Lake of the Sisters and the Tree Brothers,” “Ban Pin Shan,” and “The Tigress Witch (Hoko Po)”
1. **Paragraph-level semantic distance measurement is needed.** Chan et. al’s work either focused on short text [6] or large-scale ideation data [5]. However, in Heteroglossia, the prompts and ideas will be in paragraphs, necessitating a paragraph-level semantic distance measurement. We introduce using doc2vec to measure semantic distances automatically, which we will describe in later subsections.

2. **The role play strategy resulted in longer semantic distance.** Using doc2vec (Wiki), we estimated the semantic distance between the story draft and ideas. The ideas that came from workers with a role measured 0.490 and without a role measured 0.468.

3. **There’s a need to know where the writer actually got stuck.** The story drafts used in the pilot study were not actually interrupted where the writer got stuck. Some of the drafts were even segmented at where the follow-up plot is straightforward.

4. **Story prompts could miss critical context.** Workers sometimes provided ideas that conflicted with the core character setting because they simply did not know it. Prior works have demonstrated that maintaining context is critical for designing efficient crowdsourcing workflows. We will allow users to supplement the background information of each character in Heteroglossia.

**Data Preparation:** In response to the need of knowing where the writer actually got stuck, we acquired the data collected by the Creative Help system [34] for further study. Creative Help is an online writing application where users can freely write stories. When the writer explicitly requests for help, the system automatically generates suggestions for the next sentence in a story. Users can modify, delete, or adopt the suggestions. We considered a user request in Creative Help as a strong signal indicating the writer gets stuck. Creative Help collected 1,078 stories during its deployment between 2015 to 2018. A story contains one or more “instances”, each represents a help request sent from the writer to the Creative Help system. We removed stories with fewer than 20 sentences or fewer than three requests, resulting in a total of 107 stories. Each story on average contains 37.8 (SD=17.4) sentences and 510.9 (SD=252.8) words. We further removed stories that are obviously copied from the Internet or generated entirely by Creative Help without any human-written parts. One co-author and one collaborator then labeled the characters that appeared in each story, respectively. Only 14 stories, whose characters were totally agreed by two annotators, were used in the following experiment. We segmented each story at the second last request and used it as the story prompt.

**Story Ideation Using Role Play Strategy:** We used the same interface as Heteroglossia (Figure 3) to collect story ideas. Identical to the pilot study, two conditions, [role] and [no-role], were subject to experiment. In the [role] condition, workers were instructed to imagine that they were one of the main characters in the story and provide a follow-up plot idea in free text. In the [no-role] condition, we asked workers to provide ideas without any constraints. A total of 330 stories were collected contributed by 101 workers.

**Human Evaluation:** For each received story idea, we recruited another five workers from MTurk to rate the semantic distance to the story prompt. We collected the rating scores using a 5-point Likert scale of agreement with the statement, “This story idea is a relevant follow-up of the original story prompt.” (1 = Strongly Disagree, 5 = Strongly Agree.) Table 2 shows the results. The story ideas collected in the [role] condition had an average relevance score of 3.869, while the ideas in the [no-role] condition had an average relevance score of 3.998. The difference is statistically significant (paired t-test, $p < 0.05, N = 14$). Namely, based on human evaluation, the role play strategy generated semantically further ideas.

Researchers used MTurk to evaluate creative works and resulted in high inter-annotator agreements [6, 25]. However, some prior works also raised concerns about using non-experts to assess creative works such as graphic designs [16] and poems [17]. To further examine our findings, in the following subsection, we measured the semantic distance between text snippets using vector representations.

**Automatic Evaluation:** Most of the automated distance measures require first representing text snippets as numeric vectors, called “document representation.” Previous studies [6], where ideas were short pieces of text, simply summed up the corresponding GloVe vectors [31] and used cosine similarity for distance measurement. However, in our study, the collected ideas on average contains 78.0 words (SD=27.8) and thus require a paragraph-based representation. To this end, we experimented using the following six document representations to measure semantic distance:

1. **GloVe:** We used pretrained GloVe vectors (glove.6B.300d) [31]. The document vector was obtained by summing up the corresponding word vectors and $1 - \text{cosine similarity}$ was applied for distance measurement.

2. **Doc2Vec (Wikipedia):** We used the Doc2Vec model [23, 22], which was trained on the Wikipedia dataset (github.com/jhlau/doc2vec), and applied $1 - \text{cosine similarity}$ as a function for distance measurement.

3. **Doc2Vec (News):** Same as #2 but was trained on the Associated Press News dataset.

4. **Doc2Vec (Story):** Same as #2, but was trained on ROCStories [28].

5. **Skip-thought Vector:** We used the pretrained skip-thought model [20] to encode the document and applied $1 - \text{cosine similarity}$ as a function for distance measurement.

6. **Sentence-level Skip-thought Vector (Mean):** We segmented a document into sentences first and encoded each sentence using the pretrained skip-thought model. Thus, the document can be represented as a set of vectors $V_{sent} = \{v_1, v_2, ..., v_n\}$. When computing the distance between two sentence-level skip-thought vectors, we computed $1 - \text{cosine similarity}$ among pairs. The mean over the distances of each pair was used as the distance measure.
8. Sentence-level Skip-thought Vector (Median): Same as #6, but used the median over the distances of every pair as the distance measure.

8. Sentence-level Skip-thought Vector (Min): Same as #6, but used the min over the distances of every pair as the distance measure.

In order to evaluate how well these methods reflect human judgements, we calculated the correlation coefficients between the automatic score and human scores collected above. Note that a good semantic distance indicator should negatively correlate with relevance scores. Table 1 shows that only GloVe and Doc2Vec-Story generate scores that are negatively correlated to relevance in both Pearson (ρ) and Kendall (τ) correlation coefficients. Doc2Vec-Story is a stronger indicator than GloVe because it yields higher correlation scores.

Finally, we used GloVe and Doc2Vec-Story to measure the semantic distance automatically. Table 3 shows that both methods suggest the story ideas collected in the [role] condition had an longer semantic distance to the story prompt. These automatic measurement can be used in Heteroglossia to automatically rank the usefulness of received story ideas, or to filter out ideas that are abnormally similar to each other.

### Trade-offs Between Task Structures and Creativity
Per Kim et al. [18], in collaborative story writing, task structures and creativity have some trade-offs. Too little structure leads to “unfocused, sprawling narratives”, and too much structure “stifles creativity.” The role play strategy enforces a schema of characters for ideation and could possibly sacrifice the quality or creativity of story ideas submitted by workers.

To understand the effect of role play strategy thoroughly, we conducted experiments to examine this possible trade-off.

For each story idea received in Study 1, we recruited five workers from MTurk to rate the quality of the idea in various aspects3, using a 5-point Likert scale: Legitimate, Creative, Interesting, Willing-to-Read, and Surprising. (*: p < 0.05, **: p < 0.01). Paired t-test. N = 14. Cohen’s $d$ reported as [no-role] - [role]. Medium effect size: $|d| > 0.5$. Large effect size: $|d| > 0.8$.

To understand how writers would use Heteroglossia, we conducted a three-day deployment study with two experienced creative writers and held pre-study and post-study interviews.

### Participants
Two experienced creative writers, P1 and P2, participated in our deployment study. Both participants are women and native speakers of American English. They were recruited through our personal networks. P1 has been writing since she could pick up a pencil and has always written stories. She wrote a lot of fan fiction in middle school and high school. P1 received a minor in creative writing. She has participated in National Novel Writing Month (NaNoWriMo, www.nanowrimo.org) nine times and succeeded five times. (NaNoWriMo is an annual event where writers set goals to finish novels within the month.) P2 has never done technical writing, and her main genre focus is fantasy. P2 (age=32) has a minor in creative writing. She has participated in National Novel Writing Month (NaNoWriMo, www.nanowrimo.org) nine times and succeeded five times.

Table 1. Correlation between human-rated relevance and automatic evaluation scores. Note that a good semantic distance indicator should negatively correlate with relevance scores. Only GloVe and Doc2Vec-Story generate scores that are negatively correlated to relevance in both Pearson (ρ) and Kendall (τ) correlation coefficients. Doc2Vec-Story is a stronger indicator than GloVe because it yields higher correlation scores.

| Metric   | GloVe | D2V-Wiki | D2V-News | D2V-Story | ST | S-ST Mean | S-ST Min | S-ST Median |
|----------|-------|----------|----------|-----------|----|-----------|----------|-------------|
| ρ        | 0.142 | 0.159    | 0.141    | 0.125     | -0.153 | -0.021    | 0.202    | 0.129       |
| τ        | 0.053 | 0.075    | 0.011    | 0.128     | 0.080 | 0.012     | 0.080    | 0.080       |

Table 2. Relevance of ideas, rated by human judges. The role play strategy (Role) generated semantically further (i.e., less relevant) ideas. (*: $p < 0.05$, Paired t-test. N = 14. Cohen’s $d$ reported as [no-role] - [role]. Large effect size: $|d| > 0.8$.)

| No-Rule | Role | d  |
|---------|------|----|
| Mean    | 95% CI Mean | 95% CI |
|         |        |      |
| Relevance | *3.998 | [3.925, 4.070] | 3.869 | [3.789, 3.948] | 0.89 |

Table 3. Automatic evaluation metrics of semantic distance. Both D2V-Story and GloVe methods suggested that the story ideas collected in the [Role] condition had an longer semantic distance to the story prompt. (Cohen’s $d$ reported as [no-role] - [role]. Small effect size: $|d| < 0.2$.)

| No-Rule | Role | d  |
|---------|------|----|
| Mean    | 95% CI Mean | 95% CI |
|         |        |      |
| D2V-Story | 0.840 | [0.819, 0.860] | 0.848 | [0.829, 0.866] | -0.23 |
| GloVe   | 0.039 | [0.034, 0.044] | 0.045 | [0.037, 0.053] | -0.47 |

Table 4. Trade-offs between task structures and creativity. Five human judges on MTurk rate each story ideas on the following five aspects, using a 5-point Likert scale: Legitimate, Creative, Interesting, Willing-to-Read, and Surprising. (*: $p < 0.05$; **: $p < 0.01$). Paired t-test. N = 14. Cohen’s $d$ reported as [no-role] - [role]. Medium effect size: $|d| > 0.5$. Large effect size: $|d| > 0.8$.

3Legitimate (“This story idea makes sense given the story prompt.”), Creative (“This story idea is creative.”), Interesting (“This story idea is interesting.”), Willing-to-Read (“I’m willing to read the final story that is written based on this story idea.”), and Surprising (“This is a surprising story idea.”)
Figure 5. User activity logs of P1 shown in the cumulative word count with respect to time. When different numbers of characters were used, the resulting number of ideas varied. Therefore, the total number of ideas is shown in the legends. As we can see, participants usually requested ideas and paused writing. After hours, when most of ideas had appeared, they came back and resumed writing.

Figure 6. User activity logs of P2 shown in the cumulative word count with respect to time. P2 requested multiple tasks at the same time.

great deal. She has done technical writing before and mainly focuses on science fiction and science fantasy. P2 uses Google Docs as her primary text editor.

Study Protocol
Before the study, we had a semi-structured pre-study interview via Skype with the participants to understand their backgrounds, needs, and creative writing processes. At the end of the pre-study interview, one of the authors gave a brief tutorial of Heteroglossia and demonstrated how to use the system. Note that we explicitly informed the participants that Heteroglossia is a crowd-powered system and that their stories would be viewed by online crowd workers. In the study, the participants were asked to use Heteroglossia to write a story of approximately 1,000 words. We asked the participants to finish the story in a time span of three days, during which they needed to use Heteroglossia’s ideation function at least three times when writing their stories. After the study, we had a semi-structured post-study interview via Skype with the participants to understand their experience and feedback. The pre- and post-study interviews were both around thirty minutes long. The audio was recorded and transcribed by the authors. Each participant was compensated with $50 after the post-study interview. Table 5 shows one example of crowd ideation created by P2.

How Did the Participants Use Heteroglossia?
To capture how P1 and P2 used the system, we plotted the evolution of the cumulative word count to visualize their writing progress, aligned with the time they requested ideas. Figure 5 and 6 show the logs from P1 and P2, respectively. Both participants usually requested ideas and paused writing, which might signal getting stuck. After a few hours, the participants came back, read all the returned story ideas and continued writing. We also asked participants about how they interacted with Heteroglossia in the post-study interview. Both participants wrote sequentially without any outlines.

“... I would write until I didn’t know what to do next and then I would use the tool. The next day, I would read over everyone’s responses and then write until I got stuck and then use the tool.” (P1)

P2 finished all the writing within a day, so she tried various lengths of story prompts and launched several requests at the same time, as shown in Figure 6. Note that we allowed the participants to write freely (see the “Study Protocol” Section) and did not enforce any writing processes.

Findings
We summarize our findings of the study below, supported by quotes from the participants.

The output of Heteroglossia is interesting. Both P1 and P2 expressed that the ideas are interesting and fun:

“Yeah, there are some very, very creative answers in there. Some people would just be like ... “well, if I was this character, I would do this, this, and this” ... Some people would write a whole paragraph continuing the story. And I thought that was really interesting.” (P1)

“I really like it; it’s pretty fun... that it came up with interesting stuff. There’s one... “Oh my gosh, that weirdo. I don’t like her, booo.” And it’s just so funny... One of my favorite comments was like, [P2 read one idea] I was like, Oh, that’s really interesting.... I thought that was really fun.” (P2)

Heteroglossia is useful in generating inspiration. Both P1 and P2 think Heteroglossia can be useful for getting inspiring ideas from the crowd.

“Yeah, it’s helpful, even if I don’t use their ideas.” (P1)

“It was nice when I got stuck on what to do next to be able to ask people. That gave me more inspiration to continue and also more insight into what people’s expectations were for the story.” (P1)
Writers benefit from Heteroglossia in different ways. We also agreed that Heteroglossia can capture the personality proceed, even inspiring the next part of the story. P1 did not have ideas, people helped her figure out how to use it. When noticed that, although both participants think the system is relevant story ideas:

P2 said that, as an idea generator, Heteroglossia produces meaningful story ideas:

“It’s like an idea generator. I was really surprised by how much the ideas were actually related to the story.” (P2)

P2 also mentioned that she would like to use the system for her next NaNoWriMo:

“NaNoWriMo is coming up ... where you have to write a whole novel in a month ... sometimes it can be tricky to come up with ideas.” (P2)

Writers benefit from Heteroglossia in different ways. We noticed that, although both participants think the system is useful, the way they used it was slightly different. When P1 did not have ideas, people helped her figure out how to proceed, even inspiring the next part of the story.

“Very helpful for when I don’t have any ideas ... then I can ask a lot of other people and they’ll help me figure it out. Even if I don’t take any of their ideas, ... it might inspire something else that I will think of for the next part of the story.” (P1)

P1 also agreed that Heteroglossia can capture the personality of the role that is assigned to it.

“Again, I think it depends on how much information the writer ... gives the people taking on the roles. And I think it can also (help) even if people don’t get the correct personality, it still helps you learn ... “well, they wouldn’t do that, but they would do this instead.” So it’s still it’s helpful to rule things out in that way too.” (P1)

For P2, the ideas can be relevant to the characters: either they matched the character’s personality or figured out a personality that had not been provided by P2.

“...There’s also... some things that are actually relevant to the characters. One of the characters ... was very dramatic. And then it (Heteroglossia) came up with this idea that she would go make a lot of money and go to Vegas and like, sip martinis on an island somewhere... That’s exactly like that!” (P2)

“There was another one... the surgeon says “I’m going to do nothing. Except taking aspirin for my headache.” And I was like, wait, Heteroglossia remembered that he had a headache!... I didn’t give them that much to go on. And it still had some personality figured out.” (P2)

Table 5. An ideation example created by P2. Three characters are involved and their descriptions are listed accordingly.
Waiting for story ideas can overlap with other tasks:

“... because I had other things that I was doing. So, I would write, and then I would âÁE do the other things during the day that I needed to do... I kind of had a schedule so it didn’t really affect my time.” (P1)

“Probably, but it wasn’t too bad... when you’re coming up with stuff, it’s always... a long process anyway... it’s not really like, “okay, I got the whole idea down now”... some of the ideas, it’s kind of like “that’s really interesting; maybe I’ll go back and change it.” But it’s not that big of a deal to go back and change it because you have to... make like four or five drafts anyway.” (P2)

The role play strategy does not fit some use cases. Both participants pointed out some problems they encountered, some of which were caused by the nature of the role play strategy. P1 would like to use the baseline strategy (no-role) and dynamic team management in some cases:

“I’m conflicted between wanting to be able to assign a certain task to just one character versus the whole team because sometimes characters don’t fall in... I would imagine if you have a larger cast of characters, the teams would overlap quite a bit, so it might be easier to be able to... assign tasks to single characters.” (P1)

P1 also pointed out that some scenarios might be hard to use the role play strategy, since the structure of the story will be too complex:

“... it depends on how detailed you are when you write the character, because I only wrote a couple sentences for my characters, but if you wrote... a whole biography, then maybe. I think it depends on the complexity of the story and the complexity of the characters.” (P1)

Working with stranger workers have trade-offs. We asked the participants to compare working with strangers versus working with friends, families, or colleagues. P1 explained the trade-offs between them.

“(Using Heteroglossia) It’s less pressure because you don’t know the people, but it’s also a little more nerve-wracking because you don’t know the people. So there’s good and bad... It’s better to have someone that you know and have a good relationship with. It’s hard to trust strangers with a story, especially a story as complex as a book.” (P1)

Copyright issues were also raised if users were to use Heteroglossia for their own professional work.

“... because I had other things that I was doing. So, I would write, and then I would âÁE do the other things during the day that I needed to do... I kind of had a schedule so it didn’t really affect my time.” (P1)

“Depending on how big it gets, it might be overwhelming to have to read through all those responses.” (P1)

“It’s not that there are too many it’s just that I didn’t realize how many.” (P2)

DISCUSSION
In this section we discuss topics that are broader than the scope of the Heteroglossia system.

Differences Between Technical and Creative Writing
P2 in Experiment 2 had experience in both technical writing and creative writing. In order to better inform our future system design, we asked P2 in pre-study interview what are differences between these two. P2 said she thinks technical writing is much easier because the goal and style is more clear.

“Often (in technical writing) you have a style guide, and you have the a goal. The goal with technical writing is to make something that’s confusing understandable. Whereas the goal with creative writing is usually to give some kind of feeling to the reader.” (P2)

She further explained that, in creative writing, the writers sometimes need to intentionally avoid clear explanations, whereas technical writing is all about making things clear and understandable.
What Do Writers Do to Resolve Writer’s Block

P1 struggles with plot the most. She said that knowing the characters more can help with the situation since characters and plots are intertwined.

“Plot is my weakest skill when it comes to writing... So I find that if I really get to know my characters really well and understand the choices that they would make in any given situation, then the plot can kind of unravel itself from there.” (P1)

P2 usually let the characters talk when getting stuck. Even if the conversations are deleted afterwards, it helps her understand the characters more.

“When I get stuck, what I usually do is just have characters talk about dumb stuff... just have two characters talk to each other... And then sometimes you figure out more about what you want to do by that conversation... It helps you understand the characters more, if you have them talk to each other. And then knowing “now I know that this person wants to do this.”” (P2)

How Do Writers Write

P1 stated that her writing process was to come up with an idea, create characters, and finally, design a plot. P1 also said that she would write one draft first and later revise for plot and character.

“The idea always comes first. So I always have “what if this happened; that would be an interesting story.” And then I create the characters for that idea. The plot comes last. I will write one draft and then revise for plot and character. But it’s always idea, characters, plot.” (P1)

P2 usually thinks about the message she wants to send, picks overall “concepts of things,” thinks about characters and sets up the inner/outer goals, and figures out what she needs in the plot to satisfy the characters’ needs.

“So first, I think about what message I want to send. What...is something that I want to talk about or discuss? And I’ll pick...the setting and the time period. You know, overall concepts of things like social change that I want to talk about. Then I’ll think about the character. Because for me, the character drives the plot. Usually characters have two goals. The first is the outer goal and the second is an inner goal. ... You figure out what you want in the plot based on how it’ll satisfy the characters’ needs.” (P2)

Limitations

The need of non-role strategy. Heteroglossia currently only supports a role-playing strategy, so we do not observe any cases where the user prefers to request ideas without role schema. We will add new features to Heteroglossia to allow users to request ideas using different strategies.

Scalability. Our experiment focused on short stories (under 1,000 words) with a few characters who have relatively simple backstories. When working on long stories, the structures and characters may become complicated, raising two issues: handling a large number of characters and conveying complicated backstories. Heteroglossia currently requires users to define characters and teams before writing stories. However, when a story has more characters, it can become difficult to handle all the characters and teams. Features such as automatically suggesting teams based on context might help. We will also explore automatic summarization technologies to produce or update character backstories automatically.

Insufficient amount of participants. Only two participants were recruited in this study, as it is hard to recruit creative writers who are willing to participate in a multiple-day study. A one-day study might be too short for creative writers to come up with good ideas for stories. In the future, a deployment several months long with more users would allow us to better understand how people interact with Heteroglossia.

Evidence for relieving writer’s block. In this paper, we showed that the role-playing strategy produces semantically far story plot ideas (Study 1), and participants were satisfied with the ideas provided by Heteroglossia (Study 2). However, we did not directly examine whether the system helped relieve writer’s block. Evaluating the usefulness of an idea is challenging because an idea can still be considered useful or inspiring even if it is not directly adopted. A large-scale deployment will allow us to observe whether writers use the ideation feature frequently, which could better validate the usefulness of Heteroglossia.

CONCLUSION AND FUTURE WORK

This paper introduces Heteroglossia, a crowd-powered system that applied crowd ideation to help creative writers. We built Heteroglossia as a Google Docs add-on, and writers can simply use the editor to elicit story ideas with the online crowd. Heteroglossia adopts the role play strategy for story ideation. In controlled experiments, this strategy produced story ideas that are semantically more distant to the working story draft, which is known to be more useful to creator who reaches an impasse. We also conducted a deployment study with two experienced creative writers. In the deployment study, we found that the outputs of Heteroglossia is generally interesting and useful, while two participants benefit from the system in different ways. In the future, we will relax the definition of “characters” in our system, allowing writers to use these “roles” in Heteroglossia in a more general way. For example, each role could be a “thinking hat” that represents one perspective [10].

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REFERENCES

[1] Elena Agapie, Jaime Teevan, and Andrés Monroy-Hernández. 2015. Crowdsourcing in the field: A case study using local crowds for event reporting. In Third AAAI Conference on Human Computation and Crowdsourcing.

[2] Salvatore Andolina, Hendrik Schneider, Joel Chan, Khalil Klouche, Giulio Jacucci, and Steven Dow. 2017. Crowdboard: augmenting in-person idea generation with real-time crowds. In Proceedings of the 2017 ACM SIGCHI Conference on Creativity and Cognition. ACM, 106–118.

[3] Michael S Bernstein, Greg Little, Robert C Miller, Björn Hartmann, Mark S Ackerman, David R Karger, David Crowell, and Katrina Panovich. 2010. Soylent: a word processor with a crowd inside. In Proceedings of the 23rd annual ACM symposium on User interface software and technology. ACM, 313–322.

[4] Joel Chan, Steven Dang, and Steven P Dow. 2016. Improving crowd innovation with expert facilitation. In Proceedings of the 19th ACM Conference on Computer-Supported Cooperative Work & Social Computing. ACM, 1223–1235.

[5] Joel Chan, Steven P Dow, and Christian D Schunn. 2018. Do the best design ideas (really) come from conceptually distant sources of inspiration? In Engineering a Better Future. Springer, 111–139.

[6] Joel Chan, Pao Siangliulue, Denisa Qori McDonald, Ruixue Liu, Reza Moradinezhad, Safa Aman, Erin T Solovey, Krzysztof Z Gajos, and Steven P Dow. 2017. Semantically Far Inspirations Considered Harmful?: Accounting for Cognitive States in Collaborative Ideation. In Proceedings of the 2017 ACM SIGCHI Conference on Creativity and Cognition. ACM, 93–105.

[7] Yung-Yi Juliet Chou and Barbara Tversky. 2017. Finding Creative New Ideas: Human-Centric Mindset Overshadows Mind-Wandering.. In CogSci.

[8] Elizabeth Clark, Anne Spencer Ross, Chenhao Tan, Yangfeng Ji, and Noah A. Smith. 2018. Creative Writing with a Machine in the Loop: Case Studies on Slogans and Stories. In 23rd International Conference on Intelligent User Interfaces (IUI ’18). ACM, New York, NY, USA, 329–340. DOI: http://dx.doi.org/10.1145/3172944.3172983

[9] Colette Daiute. 1989. Play as thought: Thinking strategies of young writers. Harvard educational review 59, 1 (1989), 1–24.

[10] Edward De Bono. 2017. Six thinking hats. Penguin UK.

[11] Richard P Gabriel, Jilin Chen, and Jeffrey Nichols. 2015. InkWell: A Creative Writer’s Creative Assistant. In Proceedings of the 2015 ACM SIGCHI Conference on Creativity and Cognition. ACM, 93–102.

[12] John L Geiger and Howard Suber. 2019. Creativity and Copyright: Legal Essentials for Screenwriters and Creative Artists. University of California Press.

[13] Nick Greer, Jaime Teevan, and Shamsi T Iqbal. 2016. An introduction to technological support for writing. Technical Report. Technical Report. Microsoft Research Tech Report MSR-TR-2016-001.

[14] Nathan Hahn, Joseph Chang, Ji Eun Kim, and Aniket Kittur. 2016. The Knowledge Accelerator: Big picture thinking in small pieces. In Proceedings of the 2016 CHI Conference on Human Factors in Computing Systems. ACM, 2258–2270.

[15] Yi-Ching Huang, Jiunn-Chia Huang, Hao-Chuan Wang, and Jane Yung-jen Hsu. 2017. Supporting ESL Writing by Promp ting Crowdsourced Structural Feedback. In Fifth AAAI Conference on Human Computation and Crowdsourcing.

[16] Karl K Jeffries. 2017. A CAT with caveats: is the Consensual Assessment Technique a reliable measure of graphic design creativity? International Journal of Design Creativity and Innovation 5, 1-2 (2017), 16–28.

[17] James C Kaufman, John Baer, Jason C Cole, and Janel D Sexton*. 2008. A comparison of expert and nonexpert raters using the consensual assessment technique. Creativity Research Journal 20, 2 (2008), 171–178.

[18] Joy Kim, Justin Cheng, and Michael S Bernstein. 2014. Ensemble: exploring complementary strengths of leaders and crowds in creative collaboration. In Proceedings of the 17th ACM conference on Computer supported cooperative work & social computing. ACM, 745–755.

[19] Joy Kim, Sarah Sterman, Allegra Argent Beal Cohen, and Michael S Bernstein. 2017. Mechanical novel: Crowdsourcing complex work through reflection and revision. In Proceedings of the 2017 ACM Conference on Computer Supported Cooperative Work and Social Computing. ACM, 233–245.

[20] Ryan Kiros, Yukun Zhu, Ruslan R Salakhutdinov, Richard Zemel, Raquel Urtasun, Antonio Torralba, and Sanja Fidler. 2015. Skip-thought vectors. In Advances in neural information processing systems. 3294–3302.

[21] Aniket Kittur, Boris Smus, Susheel Khankar, and Robert E Kraut. 2011. Crowdforge: Crowdsourcing complex work. In Proceedings of the 24th annual ACM symposium on User interface software and technology. ACM, 43–52.

[22] Jey Han Lau and Timothy Baldwin. 2016. An empirical evaluation of doc2vec with practical insights into document embedding generation. arXiv preprint arXiv:1607.05368 (2016).

[23] Quoc Le and Tomas Mikolov. 2014. Distributed representations of sentences and documents. In International conference on machine learning. 1188–1196.

[24] Boyang Li and Mark Riedl. 2015. Scheherazade: Crowd-powered interactive narrative generation. In Twenty-Ninth AAAI Conference on Artificial Intelligence.
[25] Kurt Luther, Jari-Lee Tolentino, Wei Wu, Amy Pavel, Brian P Bailey, Maneesh Agrawala, Björn Hartmann, and Steven P Dow. 2015. Structuring, aggregating, and evaluating crowdsourced design critique. In Proceedings of the 18th ACM Conference on Computer Supported Cooperative Work & Social Computing. ACM, 473–485.

[26] James Maxey. 2007. You Never Write Alone. (Nov 2007). http://jamesmaxey.blogspot.com/2007/11/you-never-write-alone.html

[27] Merriam-Webster. 2014. Heteroglossia. (2014). https://www.merriam-webster.com/dictionary/heteroglossia

[28] Nasrin Mostafazadeh, Nathanael Chambers, Xiaodong He, Devi Parikh, Dhruv Batra, Lucy Vanderwende, Pushmeet Kohli, and James Allen. 2016. A corpus and evaluation framework for deeper understanding of commonsense stories. arXiv preprint arXiv:1604.01696 (2016).

[29] Michael Nebeling, Alexandra To, Anhong Guo, Adrian A de Freitas, Jaime Teevan, Steven P Dow, and Jeffrey P Bigham. 2016. WearWrite: Crowd-assisted writing from smartwatches. In Proceedings of the 2016 CHI Conference on Human Factors in Computing Systems. ACM, 3834–3846.

[30] Bernard A Nijstad and Wolfgang Stroebe. 2006. How the group affects the mind: A cognitive model of idea generation in groups. Personality and social psychology review 10, 3 (2006), 186–213.

[31] Jeffrey Pennington, Richard Socher, and Christopher Manning. 2014. Glove: Global vectors for word representation. In Proceedings of the 2014 conference on empirical methods in natural language processing (EMNLP). 1532–1543.

[32] Eva Novrup Redvall. 2012. European TV Drama Series Lab. Summary of Module 1. (2012).

[33] Melissa Roemmele. 2018. Neural Networks for Narrative Continuation. Ph.D. Dissertation. University of Southern California.

[34] Melissa Roemmele and Andrew S Gordon. 2015. Creative help: a story writing assistant. In International Conference on Interactive Digital Storytelling. Springer, 81–92.

[35] Eva Siegenthaler, Yves Bochud, Per Bergman, and Pascal Wurtz. 2012. Reading on LCD vs e-Ink displays: effects on fatigue and visual strain. Ophthalmic and Physiological Optics 32, 5 (2012), 367–374.

[36] Jaime Teevan, Shamsi T Iqbal, and Curtis Von Veh. 2016. Supporting collaborative writing with microtasks. In Proceedings of the 2016 CHI conference on human factors in computing systems. ACM, 2657–2668.

[37] Jaime Teevan, Daniel J Liebling, and Walter S Lasecki. 2014. Selfsourcing personal tasks. In CHI’14 Extended Abstracts on Human Factors in Computing Systems. ACM, 2527–2532.

[38] Jaime Teevan and Lisa Yu. 2017. Bringing the Wisdom of the Crowd to an Individual by Having the Individual Assume Different Roles. In Proceedings of the 2017 ACM SIGCHI Conference on Creativity and Cognition. ACM, 131–135.

[39] Polly W Wiessner. 2014. Embers of society: Firelight talk among the Ju’hoansi Bushmen. Proceedings of the National Academy of Sciences 111, 39 (2014), 14027–14035.

[40] Anbang Xu, Shih-Wen Huang, and Brian Bailey. 2014. Voyant: generating structured feedback on visual designs using a crowd of non-experts. In Proceedings of the 17th ACM conference on Computer supported cooperative work & social computing. ACM, 1433–1444.

[41] Anbang Xu, Huaming Rao, Steven P Dow, and Brian P Bailey. 2015. A classroom study of using crowd feedback in the iterative design process. In Proceedings of the 18th ACM conference on computer supported cooperative work & social computing. ACM, 1637–1648.

[42] Lixiu Yu, Aniket Kittur, and Robert E. Kraut. 2016. Encouraging “Outside-The-Box” Thinking in Crowd Innovation Through Identifying Domains of Expertise. In Proceedings of the 19th ACM Conference on Computer-Supported Cooperative Work & Social Computing (CSCW ’16). ACM, New York, NY, USA, 1214–1222. DOI: http://dx.doi.org/10.1145/2818048.2820025