

Ontologically Faithful Generation of Non-Player Character Dialogues

Anonymous ACL submission

Abstract

We introduce a language generation task grounded in a popular video game. **KNUDGE** (**KN**owledge **C**onstrained **U**ser-NPC **D**ialogue **G**eneration) requires models to produce trees of dialogue between video game characters that accurately reflect quest and entity specifications stated in natural language. **KNUDGE** is constructed from side quest dialogues drawn directly from game data of Obsidian Entertainment’s *The Outer Worlds*, leading to real-world complexities in generation: (2) utterances must remain faithful to the game lore, including character personas and backstories; (2) a dialogue must accurately reveal new quest details to the human player; and (3) dialogues are large trees as opposed to linear chains of utterances. We report results for a set of neural generation models using supervised and in-context learning techniques; we find competent performance but room for future work addressing the challenges of creating realistic, game-quality dialogues.

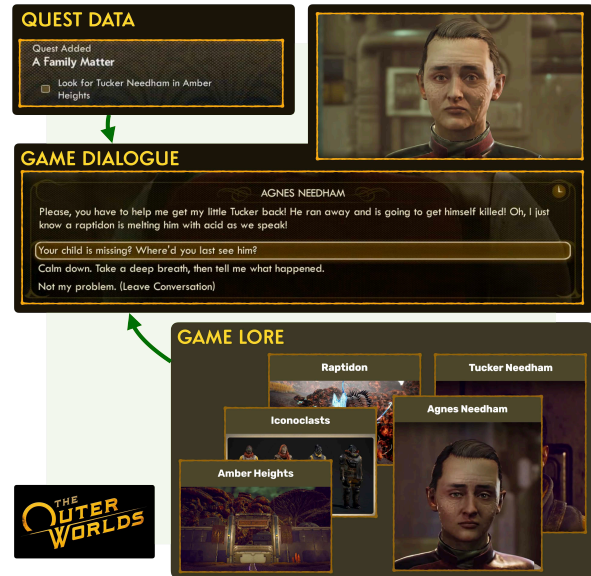


Figure 1: An example non-player character (NPC) dialogue from *The Outer Worlds* (Obsidian Entertainment, 2019). NPCs must speak faithfully to a granular ontology of **quest specifications** and **game lore**.

1 Introduction

Player interactions with non-player characters (NPCs) in role-playing games (RPGs) often serve to flesh out backstories while allowing the player to progress through engaging quest storylines (Onuczko et al., 2007). Figure 1 shows a dialogue turn, taken from *The Outer Worlds* (Obsidian Entertainment, 2019),¹ an action RPG renowned for its writing. A key challenge in creating NPC dialogues is that they should serve *coherent* narratives: utterances must faithfully reflect quest structure and game lore—characters, histories, and entity relationships. Additionally, NPC interactions often take the form of complex trees that can have dozens of nodes. Creating these branching structures according to the many specifications and ontological constraints of dialogue authoring can be time-

** KNUDGE will be publicly available upon publication.

¹https://en.wikipedia.org/wiki/The_Outer_Worlds

consuming for game designers (Caropreso et al., 2012) and cost companies millions of dollars (see §A). This motivates the pursuit of tools for *automatically generating* dialogue trees.

However, there is a lack of realistic benchmarks to train and evaluate models for this purpose. van Stegeren and Theune (2020) highlight that game text corpora should come from **real, professionally written games**; most research that explores game dialogue relies on crowdsourced or academically-curated text, which is not representative of the highly game- and context-sensitive text of real dialogues. Moreover, related work on game dialogue (Urbanek et al., 2019; van Stegeren and Mysliwiec, 2021), story generation (Akoury et al., 2020; Chen and Gimpel, 2021), and knowledge conditioning for task-oriented dialogue agents (Choi et al., 2018; Mazaré et al., 2018; Feng et al., 2020) does not address complex dialogue trees and interweaving narratives found in deployed RPGs.

To address this dearth of realistic benchmarks for game dialogue authoring, we introduce KNUDGE: **KN**owledge constrained **U**ser-NPC **D**ialogue **GE**neration, a set of dialogue trees (in English) extracted from *The Outer Worlds*, an existing video game, and paired with granular ontological constraints. KNUDGE contains 159 dialogues from all 45 side quests in *The Outer Worlds*. It contains 4.7K utterances and an average of 1.3K input constraint tokens per tree. For each quest, we annotate each turn in the dialogues with relevant grounding information— quest- and lore-related natural language (NL) support facts pulled from fan-written wikis. Such fine-grained support facts annotations are useful for training and evaluating models to generate game dialogues grounded in quest specifications and game lore. To the best of our knowledge, ours is the first dataset that consists of a set of real game-quality NPC dialogues paired with granular quest and biographical specifications consistent with a well-formed game ontology.

Using KNUDGE as a test bed, we devise the task of knowledge-constrained NPC dialogue generation (shown in Figure 2 and described in §2). The complex input specifications and limited training data target a realistic development scenario in which a designer is working on a new, partially-written game. For this task, we introduce a model class, termed DialogueWriters, that leverage in-context learning with LMs such as GPT-3 (Brown et al., 2020) to generate continuations of partial dialogue trees given input constraints. To address the challenges of long constraining specification passages and branching tree structures that are too large to fit in GPT3’s context window, we introduce techniques for prompt construction, retrieved exemplar munging, and transforming tree structures into maximal-coverage, prompt-friendly representations. To encourage the use of game lore to produce engaging dialogue, we experiment with adding mechanisms to subselect relevant facts before generating each utterance. These methods take advantage of the rich, node-level annotations of KNUDGE and show improved text overlap with the underlying ontological specifications.

Finally, we prescribe evaluation protocols that test models’ capacity to reflect game ontology constraints in addition to generating fluent and coherent dialogue. We conduct automatic and human evaluations of utterances and multi-turn trees generated from specifications for game quests, as well

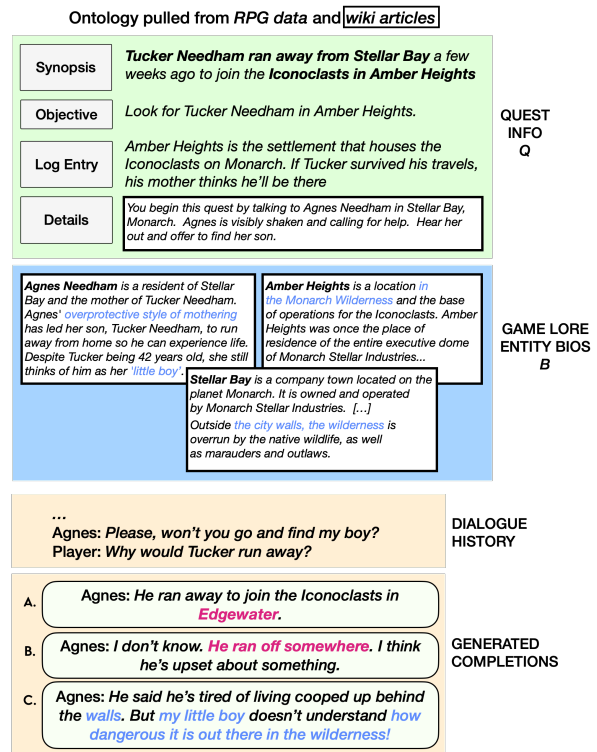


Figure 2: Overview of the proposed task. Quest information and biographical passages about the game lore serve as constraints on generated dialogue candidates. Completion A is inconsistent with the lore and B is uninformative; C is most desirable, as it provides new information to the player about quest objectives and reflects information about relevant entities.

as for a pair of never-before-seen quests written by a professional game designer. Our experiments reveal further room for improvement in aspects such as ontology usage and maintaining coherence.

2 Task Definition

As communities seek to develop AI-based writing tools such as Ubisoft’s GhostWriter (Barth, 2023), there arises a need to reconcile the challenge of dialogue authoring with current NLG techniques. An ideal writing tool might allow for a designer to provide the input types depicted in Figure 1 (granular quest information and bios from a game’s lore) and receive a set of generated utterances, or entire trees, to aid in drafting content similarly to GitHub Copilot.² As a game ontology can be large, so too might the amount of user-provided knowledge specifications. We structure our new task to target this scenario, for which no such dataset exists to study how to optimize models.

We define the task of knowledge-constrained

²<https://github.com/features/copilot/>

130 NPC dialogue generation as the mapping from a set
131 of quest constraint statements Q , a set of biographi-
132 cal constraint statements B , and a list of participant
133 names P to a dialogue tree D . We consider two
134 task scenarios: **next utterance prediction** from a
135 partial (gold) tree, and **full dialogue generation** of
136 trees with multiple candidates at each turn.

137 As depicted in Figure 2 (upper), Q comprises
138 statements $[q_1, \dots, q_m]$ about currently active ob-
139 jectives upon entering the dialogue, about what
140 should occur during it (e.g., pieces of information
141 the NPC should mention), and about new active ob-
142 jectives upon leaving it. B comprises background
143 statements $[b_1, \dots, b_n]$ about game entities that the
144 dialogue must reflect (Figure 2, middle). The par-
145 ticipant list P contains the player character and
146 one or more NPCs who have corresponding facts
147 in B . We design these inputs to reflect the kinds
148 of specifications that a developer would provide to
149 the generator during their quest writing process.

150 Dialogue D is a directed graph $\langle N, E \rangle$; each ut-
151 terance node $n \in N$ has a speaker $s \in P$. Branches
152 occur due to the multiple dialogue options for play-
153 ers (Figure 3, right). D has one start node, but can
154 have multiple exit nodes and can contain cycles.³

155 To simulate a realistic writing scenario, we
156 provide models with a small set of training di-
157 alogues from the particular game domain, some
158 $(Q_1, B_1, P_1, D_1), \dots (Q_t, B_t, P_t, D_t)$, $t \approx 100$.⁴
159 While this is not enough to perform SGD-based
160 fine-tuning, models such as those described in §4
161 can leverage it for in-context learning.

162 3 Data

163 Writing RPG-quality dialogue trees is difficult for
164 human developers for its many interweaving con-
165 siderations. 1) The tree must serve its **quest func-**
166 **tion**, containing input-specified player utterance
167 options, NPC responses (including specified emo-
168 tions), and pieces of information the player must
169 learn by the end (e.g. the *Log Entry* in Figure 2).
170 2) The utterances must be **coherent** and **engag-**
171 **ing** to the player. 3) The NPC should embody the
172 persona described in **their bio passage** explaining
173 personality, history, and relationships, and finally

³Although they often contain cycles these structures are still colloquially referred to as “trees”.

⁴We argue this reflects the most realistic writing scenario, in which a designer has written a *partial* ontology for a brand-new game with only a handful of fully-formed dialogues at their disposal. Little to no media about the game is available for models during private development by a company.

174 4) To facilitate world building, the NPC should
175 exposit details about **other entities** whenever it is
176 contextually relevant, but should never **violate** the
177 ontology through contradiction. With these desirer-
178 ata in mind, we design KNUDGE to pose a similarly
179 multi-faceted challenge to generation models.

180 KNUDGE contains NPC dialogue trees from all
181 45 side quests in the *Outer Worlds*. This RPG is
182 appealing for our investigation because of its large
183 trees, its Nebula Award-winning writing, and its
184 tendency for entities to appear in many quests in
185 different capacities. Construction of KNUDGE en-
186 tails gathering information about each quest (Q)
187 and the *Outer Worlds* entities appearing or refer-
188 enced during its dialogues (B) (§3.1), and then
189 extracting trees (D) from the game data (§3.2).

190 3.1 Game Ontology

191 We acquired dialogue files from the *Outer Worlds*
192 creators along with permission to release them pub-
193 licly; we use quest data and game lore from fan
194 wikis, where a quest’s page lists the in-game ob-
195 jectives and journal logs (though the framework
196 allows for using data from official channels instead
197 of fan-made). §B contains further source details.

198 **Quest Information** A quest in *The Outer Worlds*
199 appears in the player’s journal with a high-level
200 **synopsis** and a sequence of **objectives**, each of
201 which contains **game logs** providing additional de-
202 tails. Active objectives are completed, and new
203 ones introduced, during an NPC dialogue. We asso-
204 ciate with each objective a **walkthrough passage**
205 which includes details on the topics, player ut-
206 terance options, and quest information that the NPC
207 needs to say by the dialogue’s end. A detailed quest
208 anatomy and examples of Q can be found in §C.

209 **Biographical Information** We associate with
210 each quest, and in turn each dialogue, a set B of
211 **biographical passages** about entities appearing or
212 referenced during the quest. We extract passages
213 from entities’ fan wiki pages. While some are only
214 a few sentences, others can be much longer (up to
215 27), posing a challenge to generation models; often
216 only part of a long biography might be relevant to
217 a given quest. Examples are shown in §D.

218 3.2 Dialogue Trees

219 Dialogue trees in *The Outer Worlds* are complex
220 directed graphs, containing many conditional ut-
221 terance options depending on the state of the game—
222 e.g. whether the player is of high enough level at

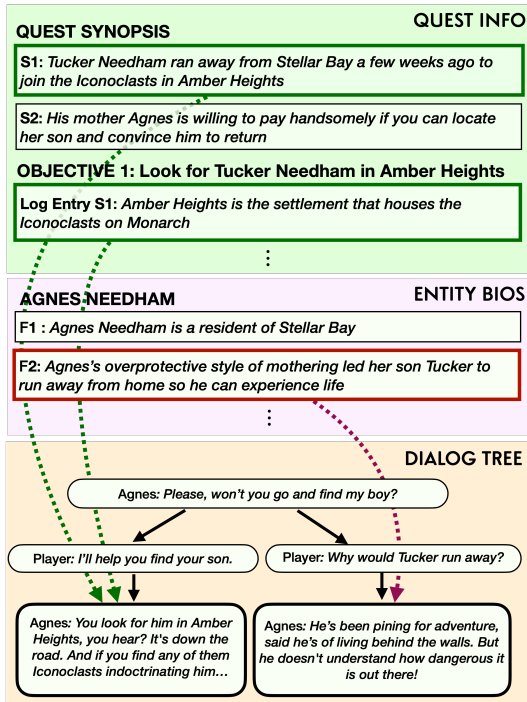


Figure 3: Overview of dialogue node annotation with support facts from quest and biography passages.

some skill to pass a “check.”⁵ To extract a more tractable, quest-related subgraph, we 1) identified the nodes that start and end the interaction using online playthrough videos as reference, and then 2) traversed the graph from the start node, following only edges without state-related conditions. Conditional edges were then added manually depending on quest relevance. §E depicts example trees.

Annotating Utterance Nodes with Support Facts

We coordinated with English-speaking professional data annotators to label trees nodes with support facts from the associated quest and lore constraints. We provided a set of instructions shown in §I, which follow a heuristic based on counterfactuals: had a given fact *not* been included in the constraining knowledge, would the utterance be much less likely to occur? An example of this procedure is depicted in Figure 3. We compute an average exact match (EM) score of 0.52 and Jaccard overlap score of 0.62. These scores represent a high agreement on a subset selection problem where the total set size is often of the order of hundreds.

3.3 Dataset Analysis

Table 1 describes statistics for the extracted dialogue trees and annotations. Notably, 47% of all nodes (57% of NPC nodes) in the dataset are an-

⁵E.g. with 55 points of the **Persuade** skill, the player can convince Tucker to return to his mother in the Figure 1 quest.

Quests	45	Dialogue trees	159
Entities	168	Utterances per dialogue	29
Characters	81	NPC utterances	16.7
Locations	40	Player utterances	12.3
Groups	21	Utt. tokens per dialogue	406.6
Items	18	Facts per dialogue	83.3
Creatures	7	Entity facts	73.3
Facts per entity	7.4	Objective facts	9.9
Entities per quest	8.0	Fact tokens per dialogue	1321.8
Quests per entity	2.0	Facts per NPC utterance	1.0
Facts per objective	7.6	Entity Facts	0.57
Objectives per quest	4.5	Objective Facts	0.42

Table 1: KNUDGE Dataset Statistics

notated with at least one fact. NPC utterances, for which support knowledge is more common, have an average of 1.0 facts. Dialogues have on average 9.9 quest facts and 73.3 biographical facts that models must consider when producing utterances.

Comparison with Related Datasets

KNUDGE is the first dataset to contain dialogue trees from an actual RPG annotated with game quest and biography specifications. Table 2 compares our dataset to contemporaries with comparable input specifications and generation targets. None contain all the components to target the complexity and granular specificity required to generate quest dialogues of the type found in *The Outer Worlds*. §F provides an in-depth comparison with select works.

Challenges

Generating a 30-node (or larger) NPC dialogue tree while simultaneously taking into account all of the NPC desiderata described above is a very difficult task, particularly given the shape of the branching, cycle-heavy tree structure. The average of 1321 constraining tokens and 406 utterance tokens, which are both **longer and more complexly structured than in existing datasets**, poses a challenge to current NLP models, taking up e.g. half of the 4000-token context window of GPT-3 before factoring in other pieces of context.

4 DialogueWriter Methods

We introduce a set of methods called DialogueWriters for generating candidate utterances given the ontological specifications (Q , B , and P) from §2 and a partial subtree S for a dialogue in KNUDGE. DialogueWriters propose utterances at a specified new location branching off S . Given some “most recent” node $n \in S$, the method maps inputs (Q, B, P, S, n) to a list of candidates $[c_1, \dots, c_n]$ such that there is a directed edge $n \rightarrow c_i$.

Tree Traversal We consider language models (LMs) that accept linear input token sequences.

Dataset	Writer	Source	Structure	Dialogue Toks / Item	Constraint Toks / Item	Narrative and Bio Constraints	Annotation Level
Story Generation Datasets							
STORIUM (Akoury et al., 2020)	Crowd	Online story writing game	Sequence of scene entries	19k	1.2K	Scene intro, challenge, location, and character descriptions	Story
TVSTORYGEN (Chen and Gimpel, 2021)	Crowd	Fan wikis	TV episode recap	1.8k	25.9K	Brief episode summary and character bio articles	Scene Entry
RPG Dialogue Datasets							
LIGHT (Urbanek et al., 2019)	Crowd	Text game platform	Sequence of utterances	212 (12 utt)	276	Location description, persona statements and held objects	Dialogue
TorchLight II (van Stegeren and Theune, 2020)	Professional	RPG data	Sequence of quest stages with 0 or 1 utterances	157 (3 utt)	24	Quest title, objectives, and details	Dialogue
WoW (van Stegeren and Mysliwiec, 2021)	Professional	RPG data	NPC-uttered quest description	61 (1 utt)	15	Quest title, objective	Dialogue
KNUDGE (Ours)	Professional	RPG data	Complex quest dialogue tree	407 (29 utt)	1.3K	Quest title, objectives, location, logs, walkthrough and entity bio articles	Utterance

Table 2: Comparison of KNUDGE to knowledge-constrained generation datasets. KNUDGE contains professionally-written complex dialogue trees from an actual RPG, with more utterances, longer constraint passages, and higher-granularity annotations (of individual utterances) than other game dialogue datasets.

We thus devise a traversal mechanism that, at inference time, converts a current dialogue subtree into a maximal coverage linear history. For “most recent” node n , we identify the longest possible path from the start node to n , including cycles but only following a given edge once. This produces utterance history $H = [u_1, \dots, u_n]$. Exploring other tree encoding mechanisms, e.g. via graph encoders (Banerjee and Khapra, 2019; Ouyang et al., 2021) is left for future work. We feed the traversed history to a next utterance generator learnt via supervised or in-context learning described below.

4.1 Supervised Learning (SL) Models

We fine-tune a T5-large model (Raffel et al., 2020) via supervised learning (SL) to generate c_i given the concatenation $[B, Q, P, H]$. We truncate context from the left of the when required given T5’s 1024-token window. §G shows example inputs.

Supervised Knowledge Selection (KS) Model

We also train a version of the SL baseline that learns to decode support knowledge facts before conditionally generating the utterance c_i . This factorizes the next utterance generation into a two-step decision process: first selecting one or more facts from the provided knowledge constraints ($Q \cup B$), and second generating the utterance to reflect the selected facts. We thus make use of our node-level annotations to train the model to generate the concatenated sequence $[f_1^{(i)}, \dots, f_m^{(i)}, c_i]$ s.t. $f_j^{(i)} \in Q \cup B$.

4.2 In-Context Learning (ICL) Models

As there is little training data, fine-tuning might not be effective at learning the difficult generation task. As such, we experiment with methods for *in-context learning* (ICL) with OpenAI’s *text-davinci-003* GPT-3 model (Brown et al., 2020). We inject B, Q, P , and H into a formatted prompt that naturally elicits the next utterance as a continuation of H . Figure 5 depicts this process; full prompts are shown in §G. This creates a *zero-shot* prompt. When this does not fill out GPT-3’s 4000-token window, we construct a *few-shot* prompt by adding dialogs from training quests as exemplars, simulating a scenario in which a developer has written a partial set of quests and is working on a new one. We retrieve exemplars using Okapi-BM25 (Jones et al., 2000) with $[B, Q, P]$ as the query string.

ICL Knowledge Selection (KS) Model As with the SL framework, we also devise a version of the ICL DialogWriter that first decodes one or more support facts before generating an utterance. We elicit this behavior from GPT-3 by augmenting *all* utterances in the dialogue history with support fact nodes, provided they exist (see Figure 15). Further model training details are found in §H.

5 Experiments

We split KNUDGE into train, development, and test splits on the basis of **quests** (28/5/12), such that at test time all input components will be unseen (test set B ’s contain a combination of previously seen and totally novel entities). For each test dialogue, we evaluate predicted utterances at each n th node in

the tree conditioned on a subtree composed of the nodes and edges between the first $n-1$ nodes. We measure 1) human judgments of NPC desiderata and 2) utterance overlap against gold items and quest and bio knowledge (§5.3). We then run a case study to collect human judgements of the skeletons of entire dialogues generated end-to-end (§5.5).

5.1 Baseline Models

To measure the effect of node-level knowledge selection (**KS**), we compare against an ICL model that selects only **one** statement instead of many. We randomly sample gold facts to construct its prompt. We also compare against an **oracle** KS ICL model, which conditions on the full gold knowledge annotations for the reference utterance. We maximize the number of in-context examples for all ICL ablations; e.g. the **no knowledge** model’s prompt can have dozens of such examples, as they are quite short. These ablations thus explore the tradeoff between the impact of the number of in-context examples and the presence of ontological statements.

To measure the effect of KS, we compare against SL and ICL **no KS** models. To measure the effect of conditioning on Q and B , we compare against ablations to the non-KS ICL model: a **no knowledge** model that conditions only on the participants P and utterance history H , and a **quest only** model that conditions on P , H , and Q , but not B .

5.2 Next Utterance Prediction (NUP)

Human Evaluation In coordination with a data specialist, we conduct human evaluation to examine models’ qualitative NUP performance. Generated utterances over 100 test items were judged on a 4-point Likert scale for each of four criteria: **1. Coherence:** does the utterance follow naturally from the utterances in the history? **2. (Non-)Violation:** does the utterance create contradictions with any of the sentences in the biographical or quest passages? **3. Biography Usage:** does the utterance *make use* of the input biographical passages in B ? **4. Quest Usage:** does the utterance progress the dialogue according to the quest sentences in Q ? We provide the full set of annotator instructions in §I. Results were verified via bootstrap testing.

Automatic Evaluation We use reference-based metrics BLEU-4 (Papineni et al., 2002) and rescaled BERTScore-F1 (Zhang et al., 2020). We evaluate generated next utterance against the following single- and multi-reference sets: 1) the gold

	Coherence Violation Using B		Using Q	
<i>Gold Reference</i>	3.94	3.97	3.50	3.45
SL-KS	2.52	3.85	2.17	2.09
ICL-KS	3.78	3.85	3.29	3.45
KS Variants:				
ICL-KS-One	3.73	3.80	3.26	3.45
ICL-KS-Oracle	3.74	3.87	3.23	3.47
Non-KS Baselines:				
SL (no KS)	2.70	3.74	2.35	2.38
ICL (no KS)	3.88	3.97	3.25	3.43
ICL-Quest Only	3.79	3.90	3.03	3.21
ICL-No Knowledge	3.65	3.69	2.76	2.98

Table 3: NUP human evaluation results for in-context (ICL) and supervised learning (SL) DialogWriter methods with and without knowledge selection (KS).

utterance n_i , 2) the quest statements in Q , and 3) the biography statements in B . BERTScore against Q and B is a neural analog to Knowledge-F1 (KF1; Shuster et al., 2021). We evaluate the **gold** utterance to show a performance upper bound.

5.3 Results

Next utterance prediction results under human and automatic metrics are shown in Table 3 and Table 4. Table 3 shows that **no model reaches gold coherence nor gold use of B** under human evaluation, suggesting room for improvement on both.

Impact of Knowledge Selection: Table 4 depicts the trend that KS variants of the ICL model score a point or two higher than non-KS on overlap with B and Q , reflecting that **KS effectively selects and cues the infusion of specific facts into generations**. Oracle KS improves BLEU score with B but not Q , while one-fact KS has the opposite effect. The ICL ablations of B and of $\{B, Q\}$ have according drops in overlap with both fact sets.

All B and Q -conditioned ICL models perform equivalently under all human metrics except (non)violation, at which KS models perform a decimal point worse. We can conclude that **KS improves the capacity of ICL writers to directly reflect knowledge passages (i.e. by copying spans)**, at the expense of a slightly higher likelihood of contradictions. This characterization can be appealing to a game developer; they might prefer for the automatic writer to use their own provided wordings of various facts when generating candidates.

Interpretation of Automated Metrics: We note that automatic metrics that check for overlap with reference text will, if used in isolation, give only a partial picture for evaluating generations in

	Gold		Bio <i>B</i>		Quest <i>Q</i>	
	BL	BS	BL	BS	BL	BS
<i>Gold Reference</i>	–	–	4.9	20.8	2.2	17.8
SL-KS	2.6	21.3	24.8	26.6	9.3	21.3
ICL-KS	7.1	25.1	8.3	24.3	7.5	24.0
KS Variants:						
ICL-KS-One	6.8	25.2	7.1	23.9	7.6	24.5
ICL-KS-Oracle	7.2	26.8	8.4	24.7	7.2	24.3
Non-KS Baselines:						
SL (no KS)	2.9	23.5	7.4	24.0	11.4	23.7
ICL (no KS)	8.6	26.4	6.8	22.8	6.6	22.2
ICL-Quest Only	7.9	26.7	3.4	21.9	6.4	22.4
ICL-No Knowledge	6.8	27.0	2.2	22.4	0.9	19.1

Table 4: NUP BLEU and BertScore for models against gold utterances and statements in *B* and *Q*. Results for the latter two shown beside the gold utterance’s score.

KNUDGE. This can be seen from the low performance of *the gold utterances themselves* under these metrics. We find that professionally written utterances do not always have high overlap with knowledge statements themselves. Gold utterances also do not score perfectly under utterance-level human evaluation of *Q* and *B* usage, as not every real-world utterance reflects the ontology. It might instead provide other qualities like realism and fluency. This finding undergirds the need to have multiple angles of evaluation: not simply checking for direct overlap, but evaluating for qualitative criteria such as coherence and appropriate ontology use.

In-Context vs Supervised Learning: ICL models outperform SL on gold overlap and BertScore against *Q*. However, the SL models have higher overlap against *B* and (under BLEU) *Q*. This reflects that SL models have learnt to incoherently copy spans directly from the context (see Figure 4), hence scoring poorly on all human evaluation.

5.4 Qualitative Results

Figure 4 depicts example outputs by models on an NUP example. We highlight cases in which the models succeed at the desiderata that we strive for in KNUDGE: to convey quest and lore specifications naturally through the interaction. However, we see that SL models and ablated ICL models are less successful. We observe that the gold utterance is more infused with desirable information than any generation; it references the quest’s next location and numerous adversaries that the player will run into, while effectively reflecting the NPC’s overprotective parent persona. This highlights a performance gap between neural and human writers to be addressed by future work. We note that

	Coh.	Viol.	Use <i>B</i>	Use <i>Q</i>	Cont.	Eng.
ICL-KS	50.0	68.8	68.8	75.0	43.8	37.5
ICL	81.2	75.0	81.2	75.0	75.0	75.0
ICL-Quest	56.2	56.2	43.8	50.0	50.0	68.8
ICL-No Know.	12.5	0.0	6.2	0.0	31.2	18.8

Table 5: Pairwise comparison (% head-to-head wins) between generated trees from ICL DialogueWriters. For example, 81.2% of the time ICL outputs were preferred over a competing approach with respect to coherence.

not reaching human performance does not preclude DialogueWriters from being useful to writers, as they can still be used to suggest new directions for dialogues to be verified or modified in a human/AI collaborative writing process.

5.5 Full Dialogue Generation Case Study

We run a case study for a full tree generation scenario. We task models with generating 10 rounds of dialogue given just the specifications *B*, *P*, *Q*, and one starting utterance. At each turn, we generate three candidate nodes using the writer, then randomly “commit” one to the linear history. This creates a 31-node tree (example in Figure 16) that can serve as a ‘spine’ of proposed content to be fleshed out by a developer into a more complicated tree. We selected 8 test dialogues from the game with varying quest roles, e.g. starting vs continuing quests. Additionally, we constructed 8 more test items from **2 totally novel quests**, written for us by a professional game designer, that occur in the *Outer Worlds* universe and contain entities from the original game. We showed trees generated by DialogueWriters to data specialists for evaluation following the ACUTE-Eval (Li et al., 2019) protocol of pairwise comparison. Annotators selected which of two trees were preferred for the following criteria: **coherence**, **nonviolation**, **biography** and **quest** usage analogous to §5.2, and additionally **5. Content Suggestion**: Do the multiple candidates at each turn propose interesting subtrees? **6. Engagingness**: does the tree hold your attention and make you want to hear more from the NPC?

Results are provided in Table 5. Performance is measured as the rate at which annotators selected a model’s tree in a pairwise comparison under the 6 criteria listed in §5.5. We find that annotators preferred the trees of the ICL writer most frequently compared to the other models under all criteria except *Q* usage. We also find that the ICL-KS method is more frequently preferred for the never-before-seen quests written by the professional designer,

History: > Agnes Needham: Please, you have to help me get my little Tucker back! He ran away and is going to get himself killed! Oh, I just know a raptidon is melting him with acid as we speak! > Player: Your child is missing? Where'd you last see him? > Agnes Needham: Oh, I just knew you were a good person. 'Agnes,' I said, 'this is the man to save your little Tuckie!' And I was right!

Gold: Agnes Needham: He ran out into the wilderness a few days ago. I warned him about the raptidons, mantisaurs, and marauders - the toxic sulfur pools, and poisonous plants. But he didn't listen!

KS ICL: Agnes Needham: He was always such a good boy, but lately he's been acting so strange. I just don't understand why he would want to leave Stellar Bay. It's the only place he's ever known.

ICL: Agnes Needham: He left a few weeks ago. Said he was going to Amber Heights to join some revolutionary group. I just know he's going to get himself killed if he's not careful!

ICL-Quest Only: Agnes Needham: He left a few weeks ago, headed for Amber Heights. He's going to get himself killed, I just know it!

ICL-Vanilla: Agnes Needham: He was playing in the street. But then he saw a sweet little kitty and chased it into the sewers! Oh, the horror!

SL-KS: Agnes Needham: If you're willing to pay handsomely, I'll even pay you handsomely if you convince him to return to me.

SL: Player: I'll do everything I can.

Figure 4: [best viewed in color] Example generations by the in-context learning (ICL) DialogueWriter and supervised learning (SL) DialogueWriter. Models that condition on both biographical and quest passages are more likely to mention relevant quest information while also referencing game entities and their backgrounds without creating inconsistency, incoherence, or incongruity. See §C and §D for full documentation of quests and entities referenced.

508 while it is less frequently preferred for dialogues
 509 from the actual game. The opposite holds for ICL-
 510 Quest only. A possible explanation is that the de-
 511 signer provided many biographical specifications
 512 for his quests (filling the GPT3 context window),
 513 which the Quest only model does not see.

6 Related Work

515 Si et al. (2021) focus on the task of story contin-
 516 uation through dialogue between multiple characters
 517 while modeling the inter-character relations. How-
 518 ever, such past work does not concern with the
 519 notion of grounding knowledge or quest objectives
 520 to be covered in the generated dialog. van Ste-
 521 geren and Theune (2020) propose three sources for
 522 building NPC dialogue corpora. However, their
 523 proposed datasets do not contain any grounding
 524 annotation and are not accompanied by explicit
 525 descriptions of entities and characters. Callison-
 526 Burch et al. (2022) explore automatic generation of
 527 conversational turns by players of the tabletop RPG,
 528 Dungeons and Dragons, in which NPCs serve a
 529 very different role in the gameplay. Scheherazade's
 530 Tavern (Aljammaz et al., 2020) augments a pattern-
 531 matching-based NPC interaction system with facts
 532 the character knows about the game world. More
 533 broadly, past work has explored applications of
 534 text generation in various gaming applications such
 535 as quest description generation (van Stegeren and
 536 Mysliwiec, 2021), dialogue generation (Si et al.,
 537 2021), persona-specific agents in text environments
 538 (Urbanek et al., 2019), and new text world genera-
 539 tion (Fan et al., 2020; Ammanabrolu et al., 2022).

540 Past work has pursued dialogue systems that
 541 steer the conversation towards a topic (Wu et al.,
 542 2019) or a given NL sentence (Sevegnani et al.,
 543 2021; Gupta et al., 2022) while conversing with
 544 a user. Other work in NLG has explored generat-

545 ing outputs with high-level NL specifications such
 546 as string item agendas (Kiddon et al., 2016), sets
 547 of facts (Orbach and Goldberg, 2020), or author
 548 goals (Riedl, 2009). KNUDGE also comprises NL
 549 specifications, though they are comparably richer.

7 Conclusion

551 Humans play games to be entertained, and they
 552 pay money expecting a high-quality experience.
 553 When game dialogue is used to advance a carefully
 554 crafted storyline, it should be both engaging and
 555 consistent with the larger narrative. Language mod-
 556 els are increasingly capable of producing engaging
 557 dialogue, which researchers have explored how to
 558 ensure is consistent with underlying knowledge
 559 specifications. To date, this research has focused
 560 on datasets developed for the sake of experimenta-
 561 tion, rather than actual high-quality game data.

562 This paper introduces KNUDGE, a dataset of
 563 NPC dialogue trees coupled with a relevant game
 564 ontology, drawn from the title *The Outer Worlds*.
 565 In contrast to prior work, KNUDGE is based on con-
 566 tent created by a high-budget, high-profile game de-
 567 velopment studio, thereby exemplifying real-world
 568 complexities in NPC dialogue authoring. We pose
 569 a knowledge-grounded generation task that mirrors
 570 a realistic development scenario with limited train-
 571 ing data over a complex ontology of quests and lore.
 572 We find that LM-based methods are able to gener-
 573 ate fluent dialogue that relates to provided specifi-
 574 cations, but they do not match the quality of profes-
 575 sional writers, particularly in terms of coherence
 576 and use of the game lore. We hope that KNUDGE
 577 drives the development of new techniques for faith-
 578 ful game dialogue generation.

579 Limitations

580 We find that the proposed DialogueWriter models
581 leave room for improvement on persona embodi-
582 ment. Human-quality utterances more seamlessly
583 and dynamically incorporate emotions fitting of
584 characters and situations, while model-generated
585 utterances can be comparatively bland. This work
586 also focuses on *side quests* whose NPCs are gener-
587 ally not as fleshed out as those in main quests. Gen-
588 erating quests containing major NPCs with long
589 bios and important roles in the main story of a
590 game, e.g. companion characters, is also left for
591 future work.

592 KNUDGE recasts a set of fan articles about an
593 existing game as specifications to an automatic di-
594 alogue tree writer. It therefore assumes that game
595 developer will write structured game lore and high-
596 level quest specifications in a similar manner be-
597 forehand when coming up with new content. Future
598 work can look at copilot tools for authoring such
599 high-level quest specifications and design of new
600 characters.

601 We report results with large pre-trained language
602 models whose training data was not publicly re-
603 leased. It is therefore difficult to know whether the
604 game data used for experimentation is part of the
605 training data for such models, as *The Outer Worlds*
606 came out in 2019. As such, the results from such
607 large language models should be interpreted with
608 caution. We partially mitigate the issue by having
609 an expert game developer construct a totally new
610 quest specification, and report results on the this
611 previously 'unseen' test data.

612 Ethics Statement

613 We acknowledge that there maybe bias in the data
614 used to train the neural language models consid-
615 ered in this paper (T5 and GPT-3) that would lead
616 to NPC dialogues that are offensive, implicitly or
617 explicitly discriminatory. This poses a potential
618 risk for deployed models, as using the proposed
619 DialogueWriters as content suggestion tools might
620 lead to RPG content that reflects these biases. We
621 hope that professional game developers will have
622 the resources to moderate damaging content before
623 it makes its way into released products.

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823 Appendix

824 A Cost of Authoring NPC Dialogues

825 Outer Worlds has 10 narrative/design credits⁶,
826 which seems to be about average (Fallout has 4-
827 12, while Skyrim has 9). Per salary.com, that’s a
828 position with an average salary of \$58k per year
829 (and probably more for AAA titles). Given a 2-3
830 year average development time for AAA titles, that
831 works out to a conservative ballpark estimate of
832 \$1.2m for just this one game.

833 B Dataset Construction Details

834 B.1 Data Sources

835 Quest data and walkthrough passages were pulled
836 from the *Outer Worlds* wiki of Fextralife,⁷ a gamer-
837 focused site containing fan-made walkthroughs for
838 many popular RPGs. Game entity biographies were
839 collected from Fandom.⁸ The biography passage
840 for a given entity is the same across all quests in
841 which the entity appears in KNUDGE, and the set of

⁶<https://www.imdb.com/title/tt9417446/fullcredits>

⁷[https://www.theouterworlds.wiki.fextralife.com](https://www.theouterworlds.wiki/fextralife.com)

⁸<https://theouterworlds.fandom.com>

842 entities is the same for all dialogues in a quest. Pas-
843 sages were segmented into individual sentences via
844 punctuation boundaries. We identified relevant dia-
845 logues and their decision points using playthrough
846 videos by the YouTube user, LordMatrim.⁹ All
847 wiki articles were written in English by site users.

848 C Quest Anatomy and Example Items

849 Figure 6 provides a detailed anatomy of a KNUDGE
850 quest, combining in-game quest data with cor-
851 responding passages from the fan walkthrough.
852 Figure 7 shows example quest items with corre-
853 sponding game data and walkthrough passages seg-
854 mented into statements.

855 In whole, a dialogue’s quest passage set Q con-
856 tains:

- 857 1. The **synopsis** (1-2 sentences)
- 858 2. The **in objective(s)** active when entering the
859 dialogue (1 sentence), the associated **game**
860 **log** (1-2) and **walkthrough passage** (3-10).
861 ¹⁰
- 862 3. The **out objective(s)** active upon leaving the
863 dialogue, and the associated **game log**. ^{11 12}

864 D Example Entity Biography Passages

865 Figure 8 shows example entities from *The Outer*
866 *Worlds* with corresponding biographical passages.

867 E Example Dialogues Items

868 Figure 9 depicts a full example input item convey-
869 ing quest, biographical, and participant specifica-
870 tions. Figure 10, Figure 11, and Figure 12 depict
871 example dialogue trees.

872 F Comparison with Other Datasets

873 van Stegeren and Theune (2020); van Stegeren and
874 Mysliwicz (2021) consider datasets of publicly-
875 available side quest data from RPGs such as *World*
876 *of Warcraft*. However, their datasets vary in di-
877 alogue and quest coverage; for *WoW* their input
878 is just a quest name and objective, and the gen-
879 eration target is a single-turn, few-sentence quest

⁹<https://www.youtube.com/@lordmatrim>

¹⁰For the first dialogue in a quest, we associate the walk-
through passage describing how to obtain the quest.

¹¹We do not associate its walkthrough passage, since the
NPC should only be expected to convey new objective infor-
mation that the player will actually see in game.

¹²The dialogue can lead to multiple new active objectives,
some optional. If the dialogue concludes the quest, then no
leaving objective is associated.

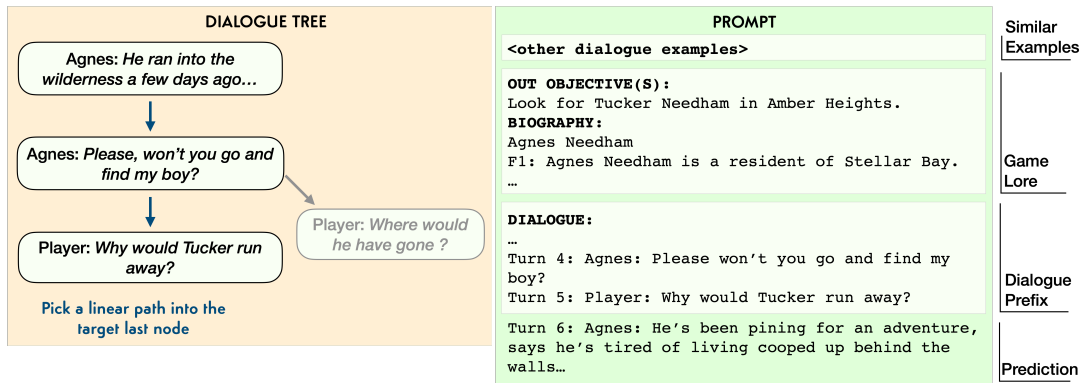


Figure 5: Overview of our method for constructing in-context learning prompts from constraints and dialogue history. Note that this figure does **not** depict the knowledge selection mechanism in the prompt; see Figure 15 for that syntax.

description spoken by an NPC. Our collected data for the game *TorchLight II* contains quest datapoints with a limited number dialog utterances per quest with no multi-turn interactions or trees.¹³ Others of their collected datasets contain complex branching trees but without constraining knowledge. The dialogues of LIGHT (Urbanek et al., 2019) are more akin to NPC dialogues, though they comprise few-turn linear chains between two characters in self-contained episodes rather than quest-grounded interactions between a player and an NPC serving multiple game purposes. The size of constraining passages on the LIGHT dialogues are also a scale smaller than those of KNUDGE. The biographical constraints of KNUDGE are most similar to that of TVSTORYGEN (Chen and Gimpel, 2021), who also pull articles from fandom wiki pages. However, theirs is a story generation dataset where the target is a longform article describing a TV episode.

G Example Prompt Constructions for DialogueWriter Model

Figure 5 depicts an overview of our tree linearization and prompt construction method.

Figure 13 shows example seq2seq items used to train/evaluate the T5-based supervised learning DialogueWriters. We list the biographies of participants last so as to truncate them from the context only when all other bios have been removed. Else, biographies are listed in random order (fixed at the onset for full dialogue generation). Figure 14 depicts example prompts shown to GPT-3 based in-context-learning DialogueWriters.

¹³Table 2 describes statistics for the 82 TorchLight II quests that contain both objective annotations and dialogue lines.

H Model Training

To construct training items, we iterate through the nodes of each gold dialogue tree in a canonical order $[n_1, \dots, n_t]$, where n_1 is the tree’s start node. We create a separate item with each n_i as the generation target. We construct the subtree $S^{(i)}$ comprised of all nodes $[n_1, \dots, n_{i-1}]$ and all edges between them. We then construct the input/output pair $(Q, B, P, S^{(i)}) \rightarrow n_i$.

Supervised Learning To train SL DialogueWriter models, for every target node in the training quest dialogues, we construct 5 training examples using different random paths to the node. We train the model for 3 epochs using the default arguments from Hugging Face’s example summarization model training script.¹⁴ T5 models were trained with a batch size of 1 across 8 Quadro RTX 6000 for an average of 5 hours.

In-Context Learning Given a test item, we construct a BM25 index over the training dialogues and use it to construct an n -shot ICL prompt where n depends on the remaining space available in the context window. Few-shot examples are linearized dialogues containing the most possible nodes from the gold tree. Contexts are left-truncated and can start with partial examples.

I Human Evaluation Directions

Below, we enumerate the instructions shown to annotators during human evaluation:

Coherence: does the utterance follow naturally from the utterances in the history? (1) Utterance

¹⁴https://github.com/huggingface/transformers/blob/main/examples/pytorch/summarization/run_summarization.py



Figure 6: Anatomy of a KNUDGE quest. At any time, an *Outer Worlds* quest has currently **active** and previously **completed** objectives. To construct its KNUDGE representation, the quest's high-level **synopsis**, objectives, and associated log entries from the game data are annotated with corresponding walkthrough article passages.

<p>Quest Name: A Family Matter Synopsis: [0] Tucker Needham ran away from Stellar Bay a few weeks ago to join the Iconoclasts in Amber Heights. [1] His mother Agnes is willing to pay handsomely if you can locate her son and convince him to return Walkthrough: [0] You can begin this quest by talking to Agnes Needham in Stellar Bay, Monarch. [1] Agnes is by the town's south-east exit, visibly shaken and calling for help. Hear her out and offer to find her son to being the quest.</p>
<p>Objective 1: Look for Tucker Needham in Amber Heights Game Log: [0] Amber Heights is the settlement that houses the Iconoclasts on Monarch. [1] If Tucker Needham survived his travels, his mother thinks he'll be there. Walkthrough: [0] Head South from Stellar Bay and follow the east road. It will take you to Amber Heights. [1] Head up the hill and go into a residence on the left to meet Tucker Needham.</p>
<p>Objective 2: Convince Tucker to Return Home Game Log: [0] Now that you've found Tucker Needham in Amber Heights, convince him to return home to his mother in Stellar Bay. Walkthrough: [0] Introduce yourself, and then you can mention your surprise that this grown man is the "little boy" that ran away. You'll earn 7500xp [1] Explain to him that she made it sound as if he was a boy in danger, [2] and he'll say she has been overprotective all her life, [3] and he is ready to live his life without her protection. [4] You can persuade (55) or intimidate (55) to expedite things and get him to go back, [5] or you can ask him what he wants to do about it. [6] The last option will have him tell you to report that he is dead. [7] You can express your concern about what that will do to Agnes, [8] and then either ask for something that would prove a body, or reject the proposition. [9] If you persuade him to go back, you'll get 7500xp and can return to Stellar Bay to see things play out.</p>
<p>Objective 3: Return to Agnes Needham in Stellar Bay Game Log: [0] You convinced Tucker Needham to return home to Stellar Bay. [1] Agnes promised a reward for bringing her son back. Walkthrough: [0] You'll find his mother is still condescending to him, [1] and you can help him by saying he's a grown man. [2] You'll get 7500xp. [3] If you stick around and talk to them some more you'll see Tucker is standing up for himself. [4] You'll receive 625 Bit Cartridge, Monarch Stellar Industries Reputation and 15000xp.</p>
<p>Quest Name: The Commuter Synopsis: [0] The Iconoclasts are due to receive a shipment of vital supplies from Carlotta, a sympathizer that resides in Stellar Bay. [1] The meeting is set to occur at the Bayside Terrace warehouse. Walkthrough: [0] The quest can be obtained by asking Graham if there is anything that needs doing. [1] He is trying to get an old printing press running, but the replacement rollers he'd requisitioned haven't arrived yet. [2] They were supposed to be delivered by Huxley, but she is still recovering and unable to make the delivery. [3] Graham asks the player to meet the supplier in her stead, and to pick up high-capacity data cartridges with the funds left over from the previous shipment. [4] Zora will interject to ask the player to buy food and medicine instead with the leftover money.</p>
<p>Objective 1: Get the Printing Press Rollers from Carlotta Game Log: [0] Travel to the warehouse at Bayside Terrace and find Graham's contact, Carlotta. [1] She should have a shipment for him. Retrieve it. [2] Speak to Carlotta Walkthrough: [0] Clear out the Sublight squad that is hunting Carlotta [1] Carlotta is behind a locked door to the east. [2] Activate the intercom next to the door to speak to her and she will unlock it. [3] Go inside and speak to her again to obtain the rollers needed to complete the quest, then choose between the high-capacity data cartridges or food and medicine.</p>
<p>Objective 2: Get High-Capacity Cartridges or Extra Supplies from Carlotta Game Log: [0] Graham wants to tack on some high-capacity cartridges to his order, but Zora would prefer it they could get extra food and medical supplies. [1] You got extra supplies for Zora (or) You got High-Capacity Data Cartridges for Graham. Walkthrough:</p>
<p>Objective 3: Return to Graham Game Log: [0] Bring the needed parts back to Graham at Amber Heights Walkthrough: [0] Return to Graham and you'll find him arguing with Zora about the Van Noys, a unit of the Iconoclasts that is MIA. [1] Inform Graham that you got his rollers, and food and medicine if that was your choice. [2] You'll receive 7500xp and Zora will ask when the next drop is. [3] Inform her that Sanjar has made it illegal to trade with the Iconoclasts.</p>
<p>Quest Name: Who Goes There Synopsis: [0] The Groundbreaker's Mardets have a bounty for a criminal on the run in the Groundbreaker's Back Bays. [1] You've agreed to hunt down the unlawful Captain Gunnar MacRedd. [2] Return his lighter to Commandant Sanita to claim the bounty. Walkthrough: [0] This quest is obtained at Groundbreaker, [1] by speaking to Comdt. Sanita or perusing the bounty board</p>
<p>Objective 1: Hunt Down and Kill Captain McRedd Game Log: [0] Based on the bounty listing, Captain McRedd was last sighted in the Back Bays. [1] Head there and take him out. Walkthrough: [0] You can find Captain MacRedd in the Back Bays area of the Groundbreaker. [1] To get there head down the elevator in the promenade, [2] and you can't miss him. [3] You can pass a Persuade (40) check to get him to put his gun down, [4] otherwise you'll have to kill him and all his guards. [5] If you kill him he drops the Unique Weapon: Montag. [6] You'll get 6000xp and MacRedd's Lighter. [7] If you persuaded him, use Perception to note it says "Sanita" on the lighter. [8] MacRedd will mention it was given to him by Sanita in remembrance of a 'carnal understanding' they had a few years back.</p>
<p>Objective 2: Claim the Bounty's Reward from Comdt. Sanita Game Log: [0] McRedd gave you his lucky lighter to give to Sanita. [1] Go turn it in to resolve his bounty. Walkthrough: [0] Turn the lighter in to Commandant Sanita to claim the bounty.</p>

Figure 7: Example Quest Items

Entity: Agnes Needham

Appears in: *A Family Matter*

Bio: [0] Agnes Needham is a resident of Stellar Bay and the mother of Tucker Needham. [1] Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life. [2] Despite Tucker being 42 years old, she still thinks of him as her 'little boy'. [3] You can find her by Stellar Bay's south-east exit, visibly shaken and calling for help.

Entity: Tucker Needham

Appears in: *A Family Matter*

Bio: [0] Tucker Needham is a former resident of Stellar Bay who left to join the Iconoclasts. [1] Before the quest *A Family Matter*, he can be found in Amber Heights. [2] Tucker was coddled by his mother from a very young age, [3] the latter insisting that danger lurked around every corner on Monarch. [4] His mother's overprotectiveness extended well into Tucker's adulthood, [5] leading him to seek to be free in any way possible. [6] After hearing Graham Bryant's broadcasts, Tucker left Stellar Bay to be truly free by joining the Iconoclasts at Amber Heights. [7] He is dazzled by Graham's preachings on true unfettered freedom from the corporate way of life and attributes his enthusiasm to his 'childhood trauma'. [8] He is willing to do anything to remain free, even faking his own death to prevent his mother from continuing to send people to look for him.

Entity: Raptidon

Appears in: *A Family Matter, At Central, Bolt With His Name, Journey Into Smoke, Makes Space Suits Wont Travel, The Amateur Alchemist, The Distress Signal, The Doom That Came To Roseway, Vulcans Hammer*

Bio: [0] Raptidons are giant cat/reptile-like creatures that inhabit various planets in Halcyon. [1] They are creatures native to Monarch. [2] however some corporations have illegally imported them to other planets, [3] such as Auntie Cleo who relocated a group of them to Roseway. [4] Raptidons are of corporate interest due to their potential for producing new chemical by-products which, [5] when refined, can be used to create new board approved products.

Entity: Sulfur Pits

Appears in: *A Family Matter*

Bio: [0] The Sulfur Pits are a point of interest on the western side of Monarch. [1] They are located southwest of Terra One Publications and directly northeast of the Gunship Crash Site. [2] The Sulfur Pits have a large variety of Raptidons and many deceased marauders. [3] The area consists largely of Sulfur Pits. [4] When an entity comes in contact with a sulfur pit, [5] they receive the acid effect for the duration of touching the pit.

Entity: Monarch

Appears in: *A Cysty-Dance With Death, A Family Matter, Bolt With His Name, Little Memento, Makes Space Suits Wont Travel, Mandibles Of Doom, Slaughterhouse Clive, Space-Crime Continuum*

Bio: [0] Monarch, previously known as Terra 1, is one of the many moons of the gas giant Olympus and the site of a failed colony. [1] Terra 1 was initially designated as the primary colonization target of the Halcyon system. [2] The Halcyon Holdings Corporate Board had intended to completely terraform the moon, [3] wiping out the local fauna and flora and replacing it with plants and wildlife native to Earth. [4] However, the terraforming process unexpectedly caused the native species to mutate and grow to significantly larger sizes, [5] rendering them more dangerous and severely crippling the colonization effort. [6] Due to the hostile environment which they had created, [7] the Board was forced to enact a Hazard Clause covering the entirety of Terra 1. [8] Public notice of the clause's issuance was sent to everyone operating on Terra 1 and led to the evacuation of almost all corporations from the moon. [9] However, one corporation took advantage of the chaos of the evacuation to exploit a legal loophole which allowed them to, [10] as the last corporation remaining on the planet, [11] acquire the planet from the Board. [12] This corporation, under the leadership of Sanjar Nandi and Graham Bryant subsequently rebranded itself to Monarch Stellar Industries (MSI), [13] in line with the renaming of the planet to 'Monarch'. [14] The actions of MSI earned them the ire of the Board, [15] who retaliated by effectively placing the moon under indefinite embargo, [16] refusing to allow legal transit either in or out. [17] the Board aggressively spread propaganda about Monarch to convince the rest of the population that it was both uninhabited and uninhabitable. [18] This has greatly hampered MSI's attempts to be recognized as a legitimate corporation and is a thorn in the side of its CEO, Sanjar Nandi. [19] Monarch also has an ocean which goes around the moon at the "twilight band". [20] It is where the colonists and Monarch Stellar Industries farm their saltuna.

Entity: Stellar Bay

Appears in: *A Family Matter, Bolt With His Name, Canids Cradle, Flowers For Sebastian, Herricks Handiwork, Mr Picketts Biggest Game, Passion Pills, The Stainless Steel Rat*

Bio: [0] Outside the city walls, the lands were overrun by the native wildlife, as well as marauders and outlaws. [1] Stellar Bay is a company town located on the planet Monarch. It is owned and operated by Monarch Stellar Industries. [2] Stellar Bay is the largest saltuna producer on the Halcyon colony and used to be one of the most important suppliers of this resource.

Entity: Fallbrook

Appears in: *A Cysty-Dance With Death, Slaughterhouse Clive, Space-Crime Continuum, Spratkings*

Bio: [0] Fallbrook is a company town located on Monarch, [1] loosely run by the SubLight Salvage and Shipping Corporation. [2] Fallbrook is a small town built into the side of a mountain, [3] whose construction was masterminded by Catherine Malin. [4] Fallbrook has a lot of activities to offer to its visitors, [5] from those who search for activities of leisure to those with proclivities for vice.

Entity: Cascadia

Appears in: *Space-Crime Continuum, The Chimerists Last Experiment, The Ice Palace*

Bio: [0] Cascadia is an abandoned company town that was owned and operated by Rizzo's before it withdrew from Monarch. [1] It is now used as a stronghold by the Marauders. [2] The main attraction is the Cascadia Bottling Plant and, [3] for those in the know, [4] the Rizzo Secret Laboratory hidden underneath the Rizzo Sweets Shoppe.

Entity: Amber Heights

Appears in: *Little Memento, Odd Jobs, Sucker Bait, The Commuter*

Bio: [0] Amber Heights is a location in the Monarch Wilderness and the base of operations for the Iconoclasts. [1] The Iconoclasts run the place somewhat like a commune. [2] Amber Heights was once the place of residence of the entire executive dome of Monarch Stellar Industries. [3] It is now in ruins after a massacre in the past. [4] They lived there with their families and it was the company's operations center on Monarch. [5] Just after The Board approved the evacuation of the planet through the Hazard Clause, Amber Heights was besieged by a gang of pirates who ransacked the town and massacred all its inhabitants. [6] This tragedy was known as "The Amber Heights Massacre". [7] They were secretly assisted by MSI employee, Graham Bryant, who believed that the massacre would aid him in his quest to rid the colony of corporate influence. [8] In 2345, the same Graham Bryant formed the Iconoclasts and settled the group in the deserted town.

Figure 8: Example entity biographies that appear as constraining knowledge in KNUDGE quest dialogs

Dialog: *A Family Matter 00*

In Objective(s): Tucker Needham ran away from Stellar Bay a few weeks ago to join the Iconoclasts in Amber Heights. His mother Agnes is willing to pay handsomely if you can locate her son and convince him to return. You can begin this quest by talking to Agnes Needham in Stellar Bay, Monarch. Agnes is by the town's south-east exit, visibly shaken and calling for help. Hear her out and offer to find her son to being the quest.

Out Objective(s): Amber Heights is the settlement that houses the Iconoclasts on Monarch. If Tucker Needham survived his travels, his mother thinks he'll be there.

Game Lore:

Agnes Needham

[0] Agnes Needham is a resident of Stellar Bay and the mother of Tucker Needham. [1] Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life. [2] Despite Tucker being 42 years old, she still thinks of him as her 'little boy'. [3] You can find her by Stellar Bay's south-east exit, visibly shaken and calling for help.

Iconoclasts

[0] The Iconoclasts are a group of survivalists living in the ruins of Amber Heights on Monarch. [1] They hope to one day tear down the corporate establishment that they believe has brought the colony to the brink of death. [2] The Iconoclasts are a group of idealistic revolutionaries that seek to overthrow the corporate establishment that runs the Halcyon Colony. [3] Based in the ruins of the Amber Heights settlement on Monarch, [4] they are a tenacious group, [5] and share some democratic ideals with Monarch Stellar Industries (MSI) against the more repressive actions of the Board. [6] However, the Iconoclast's anti-corporate nature has put them at odds with MSI, a dispute that threatens to spill into all-out warfare. [7] Given that the Iconoclasts are mostly followers of the Philosophist faith, they have been blacklisted and demonized by the Board as dissenters and anarchists. [8] The group is led by Graham Bryant, a staunch Philosophist. [9] Zora Blackwood, the Iconoclasts' chief of medicine, is also considered a *de facto* leader of the group, [10] as she was alongside Graham when he founded the Iconoclasts, [11] and almost every member of the Iconoclasts owes her their life in some way. [12] The Iconoclasts maintain a tense relationship with MSI. [13] Despite sharing democratic values and a common desire towards egalitarianism for the people of Monarch and the wider Halcyon colony, [14] MSI's "egalitarian corporate structure" has proven to be at odds with some of the Iconoclasts' more radical, anti-capitalist views. [15] Depending on the actions of the Stranger, this tense relationship can either be resolved, [16] or can spill into a drawn-out and bloody war. [17] The Stranger meets the Iconoclasts in Amber Heights just as the tension between them and MSI is reaching boiling point. [18] They can either side with the Iconoclasts and assist them in storming and taking over Stellar Bay, [19] "solve" the Iconoclast problem for Stellar Bay, [20] or broker peace between the two factions. [21] The Stranger can also have an impact on the leadership of the Iconoclasts - siding with either Graham Bryant or Zora Blackwood. [22] To supplant Graham with Zora, evidence of Graham's involvement in the Amber Heights massacre must be found and presented to Zora. [23] The Van Noys are the Iconoclasts' best unit.

Mantisaur

[0] Mantisaur are insectoid creatures native to Monarch. [1] They are aggressive, territorial, and very strong. [2] It is possible to deal with them one on one, but it is best to avoid groups of them for your safety. [3] The mantiqueen is the largest breed of Mantisaur.

Monarch

[0] Monarch, previously known as Terra 1, is one of the many moons of the gas giant Olympus and the site of a failed colony. [1] Terra 1 was initially designated as the primary colonization target of the Halcyon system. [2] The Halcyon Holdings Corporate Board had intended to completely terraform the moon, [3] wiping out the local fauna and flora and replacing it with plants and wildlife native to Earth. [4] However, the terraforming process unexpectedly caused the native species to mutate and grow to significantly larger sizes, [5] rendering them more dangerous and severely crippling the colonization effort. [6] Due to the hostile environment which they had created, [7] the Board was forced to enact a Hazard Clause covering the entirety of Terra 1. [8] Public notice of the clause's issuance was sent to everyone operating on Terra 1 and led to the evacuation of almost all corporations from the moon. [9] However, one corporation took advantage of the chaos of the evacuation to exploit a legal loophole which allowed them to, [10] as the last corporation remaining on the planet, [11] acquire the planet from the Board. [12] This corporation, under the leadership of Sanjar Nandi and Graham Bryant subsequently rebranded itself to Monarch Stellar Industries (MSI), [13] in line with the renaming of the planet to 'Monarch'. [14] The actions of MSI earned them the ire of the Board, [15] who retaliated by effectively placing the moon under indefinite embargo, [16] refusing to allow legal transit either in or out. [17] the Board aggressively spread propaganda about Monarch to convince the rest of the population that it was both uninhabited and uninhabitable. [18] This has greatly hampered MSI's attempts to be recognized as a legitimate corporation and is a thorn in the side of its CEO, Sanjar Nandi. [19] Monarch also has an ocean which goes around the moon at the "twilight band". [20] It is where the colonists and Monarch Stellar Industries farm their saltuna.

Raptidon

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Stellar Bay

[0] Outside the city walls, the lands were overrun by the native wildlife, as well as marauders and outlaws. [1] Stellar Bay is a company town located on the planet Monarch. It is owned and operated by Monarch Stellar Industries. [2] Stellar Bay is the largest saltuna producer on the Halcyon colony and used to be one of the most important suppliers of this resource.

Sulfur Pits

[0] The Sulfur Pits are a point of interest on the western side of Monarch. [1] They are located southwest of Terra One Publications and directly northeast of the Gunship Crash Site. [2] The Sulfur Pits have a large variety of Raptidons and many deceased marauders. [3] The area consists largely of Sulfur Pits. [4] When an entity comes in contact with a sulfur pit, [5] they receive the acid effect for the duration of touching the pit.

Tucker Needham

[0] Tucker Needham is a former resident of Stellar Bay who left to join the Iconoclasts. [1] Before the quest *A Family Matter*, he can be found in Amber Heights. [2] Tucker was coddled by his mother from a very young age, [3] the latter insisting that danger lurked around every corner on Monarch. [4] His mother's overprotectiveness extended well into Tucker's adulthood, [5] leading him to seek to be free in any way possible. [6] After hearing Graham Bryant's broadcasts, Tucker left Stellar Bay to be truly free by joining the Iconoclasts at Amber Heights. [7] He is dazzled by Graham's preachings on true unfettered freedom from the corporate way of life and attributes his enthusiasm to his 'childhood trauma'. [8] He is willing to do anything to remain free, even faking his own death to prevent his mother from continuing to send people to look for him.

Figure 9: Dialogue from motivating example in Figure 2 with all input constraining passages. Full dialogue tree can be found on the next page.

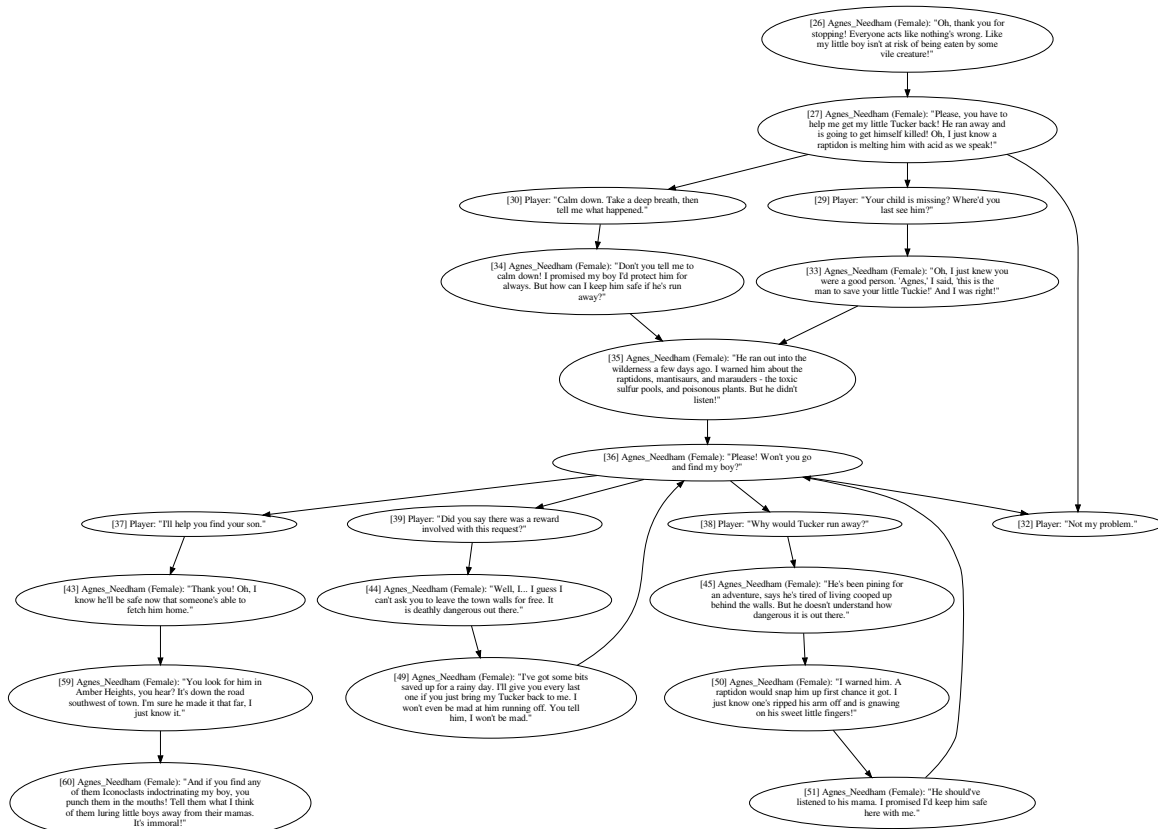


Figure 10: Full dialogue tree in KNUDGE for motivating example in Figure 2.



Figure 11: Example full dialogue tree for dialogue *who_goes_there_01* in KNUDGE.

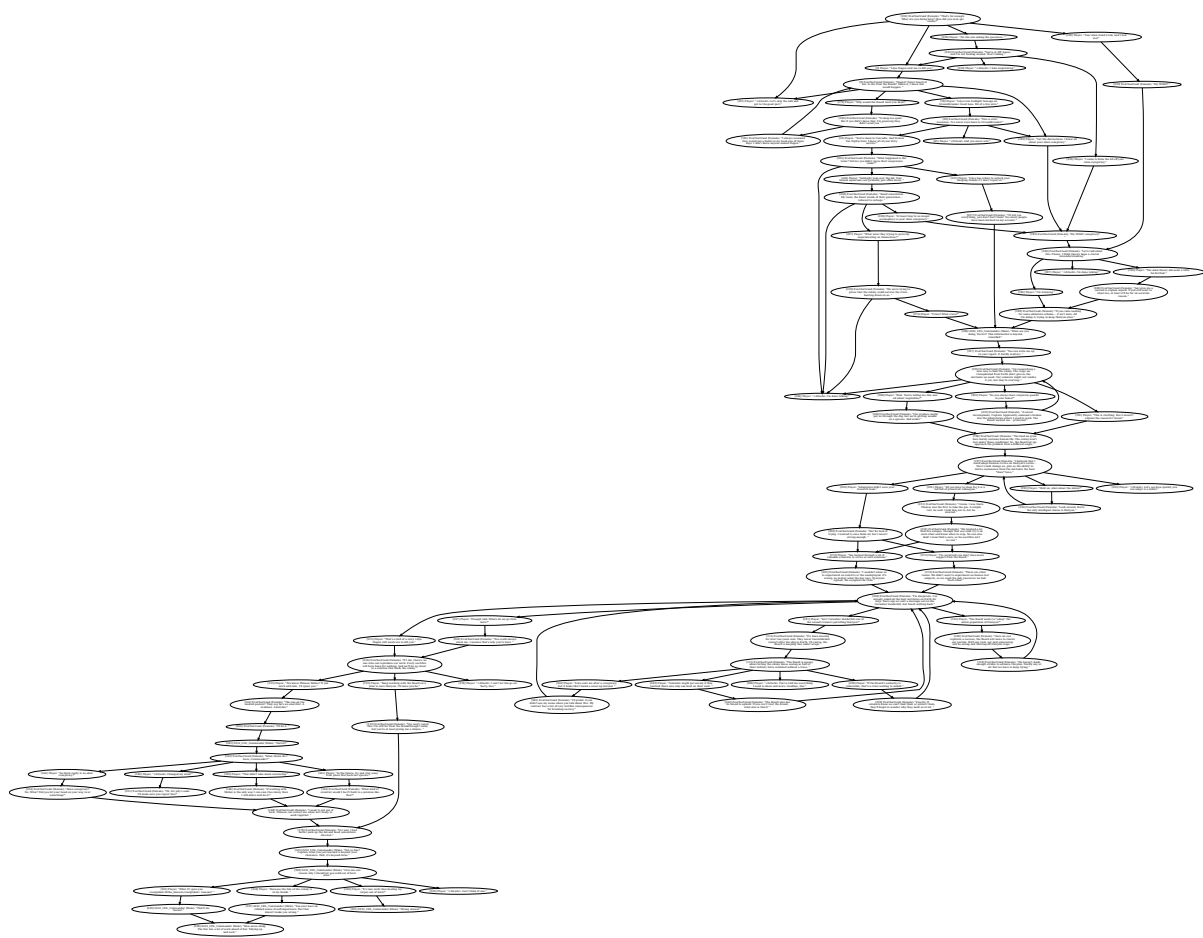


Figure 12: Example of longer dialogue tree in KNUDGE, containing numerous decision points, cycles and re-entrances.

source:
the Board's authoritarianism.</s> Sanjar Nandi</s> Sanjar Nandi is the current CEO of Monarch Stellar Industries, based in Stellar Bay.</s> Sanjar began working for MSI at a young age and it was there where he met Graham Bryant, who would eventually become his best friend.</s> Sanjar was ambitious but his attention to detail at the expense of big-picture thinking hampered his efforts within MSI.</s> This led to negative performance reviews regarding his tendency to pad reports and talks with numbers and data,</s> feedback which continues to haunt him many years on.</s> However, the negative feedback did not dampen Sanjar's desire to move up within the company,</s> even donating a kidney to one of the executives in hopes of promotion.</s> Despite his poor performance, Sanjar always showed himself to be a loyal employee of the company.</s> Despite Sanjar's best efforts, he has found it extremely challenging to continue operating MSI on Monarch without the backing of the Board.</s> In order to improve the lives of the people he is responsible for, Sanjar has a plan to rejoin the Board through the use of a BOLT-52 form and proof of another corporate presence on Monarch.</s> He is simultaneously working on a plan to reorganize the Board,</s> hoping that his plans are not found out until MSI has been reinstated.</s> Sanjar choosing to take over as head of MSI rather than dismantling it entirely caused a rift between him and Graham Bryant.</s> The latter started the Iconoclasts,</s> a group dedicated to spreading the word of Philosophism throughout the galaxy,</s> and Sanjar was left in Stellar Bay to run the company and look after the employees who were left behind.</s> He can also tell you more about the planet, that used to be called Terra 1</s> and the reform that he and Monarch Stellar Industries tried to achieve to give more humane working conditions for everyone within.</s> Celia Robbins</s> Celia Robbins is a middle manager for Monarch Stellar Industries and works with Sanjar Nandi at MSI Headquarters in Stellar Bay.</s> Celia has a crush on Sebastian Adams and will buy whatever he has in stock ,</s> just as an excuse to talk to him.</s> Unfortunately her apartment is filling up with exotic creature parts and her neighbors are starting to complain about the smell.</s> She is not concerned that she and Sebastian may not have much to talk about,</s> as everyone else in Stellar Bay either smells like saltuna or are her boss.</s> The Stranger can offer to set her and Sebastian up on a date.</s> DIALOG CONTEXT: Sanjar believes another company may be operating on Monarch illegally.</s> If he can get proof, then he could use that as leverage to get MSI readmitted to the Halcyon Board.</s> Talk to Sanjar after completing BOLT with His Name.</s> He reveals the plan is to blackmail the Board into letting them back to the table.</s> Sanjar will reveal he believes another corporation is operating illegally within Monarch, granting you the quest Errors Unseen.</s> He will tell you Catherine is likely supplying them from Fallbrook.</s> He wants you to infiltrate the secret facility and bring back evidence - be it an item or staff. KNOW BY THE END OF THE DIALOG: Sanjar believes Catherine would know where this other corporation is operating.</s> See if you can get the location from her.</s> DIALOG PARTICIPANTS:</s> Sanjar Nandi</s> Celia Robbins</s> Player</s> HISTORY:> Player: <unk> Give him the BOLT-52.> I found the cartridge and deleted that data for you. > Sanjar Nandi: Oh, yes. I'm going to be up all night with this. All those blanks waiting to be filled, boxes waiting to be ticked... > Celia: Try to control yourself, sir. > Sanjar Nandi: Have you any idea how powerful this is? Corporations have been toppled with less. > Player: How exactly is a data cartridge going to help? > Sanjar Nandi: What a question! Bureaucratic micromangement is the only way anything gets done on Halcyon, and proper documentation is a key part of that. > Sanjar Nandi: For our part, a Bill of Liquidation/Transfer Form-52 will protect our holdings on Monarch by temporarily assigning them to a pass-through entity once we drop our bomb on the Board.</s>

target:
The Board fact: The Board maintains a very tense relationship with MSI, owing to MSI's democratic ideals and their declared ownership of Monarch., The Board fact: Depending on the actions of the Stranger, MSI may be compelled to rebel against the Board's authoritarianism. > Player: Yes! Finally, the Board will get their comeuppance!</s>

source:
mire to help her with a comms issue,</s> and inform the player that they are now authorized to trade with Doctor Mfuru.</s> Junlei will appear on the Unreliable if the player helps to set up a date between her and Parvati.</s> When interacted with, she thanks them for welcoming her aboard the ship.</s> Mardets</s> The Mardets or Groundbreaker Security are the security force on the Groundbreaker.</s> The Mardets started as the original security force on Groundbreaker before the Crossing.</s> The original force was made up of a marine detachment from the 77th Marine Expeditionary Unit, Trailward Fleet.</s> Over time people started to call them Mardets because it was easier to say and the name stuck.</s> The Mardets that currently protect Groundbreaker are descended from the original marine detachment and still wear colony ship guard armor as a uniform.</s> Sublight</s> The SubLight Salvage and Shipping Corporation, also known as SubLight Salvage and Shipping or simply SubLight, is a network of "salvagers" with business ties to "transportation" and "waste disposal".</s> A tangled web of contractors and secretive vice presidents make up their official hierarchy,</s> leaving no one to speak on the record about SubLight's more legally dubious activities.</s> SubLight as a business operates activities that would be considered illicit by some,</s> amoral by others,</s> and at odds with the Board,</s> such as the takeover of abandoned space habitats,</s> smuggling contraband,</s> and serving as a front for companies to bypass the infamous Hazard Clause applied on Monarch.</s> According to director Lilya Hagen,</s> operating off of Groundbreaker ,</s> SubLight "occupies a legal blind spot.</s> No one knows what [they're] licensed to do,</s> and that gives [their] little business some freedom."</s> Although the Board tolerates and sees the value of SubLight's activities,</s> there are still limits to what the latter can get away with.</s> SubLight maintains an office aboard the Groundbreaker as well as a settlement on Monarch known as Fallbrook.</s> Gunnar Macredd</s> Captain Gunnar MacRedd is a known criminal and leader of a group of outlaws found in the Back Bays of Groundbreaker.</s> MacRedd is a pyromaniac with a criminal record which includes "several counts of flying under the influence,</s> carrying open alcoholic containers,</s> failure to pay docking fees,</s> resisting arrest,</s> and assaulting not one but two officers."</s> He has a lighter which he claims was given to him by Commandant Sanita a few years back when the two had a "carnal understanding".</s> However, Sanita claims that this understanding only consisted of her "kick[ing] his ass from one end of the Groundbreaker to the other".</s> DIALOG CONTEXT: Based on the bounty listing, Captain McRedd was last sighted in the Back Bays.</s> Head there and take him out.</s> You can find Captain MacRedd in the Back Bays area of the Groundbreaker.</s> To get there head down the elevator in the promenade,</s> and you can't miss him.</s> You can pass a Persuade (40) check to get him to put his gun down,</s> otherwise you'll have to kill him and all his guards.</s> If you kill him he drops the Unique Weapon: Montag.</s> You'll get 6000xp and MacRedd's Lighter.</s> If you persuaded him, use Perception to note it says "Sanita" on the lighter.</s> MacRedd will mention it was given to him by Sanita in remembrance of a 'carnal understanding' they had a few years back. KNOW BY THE END OF THE DIALOG: McRedd gave you his lucky lighter to give to Sanita.</s> Go turn it in to resolve his bounty.</s> DIALOG PARTICIPANTS:</s> Gunnar Macredd</s> Player</s> HISTORY:> Captainmacredd: Look at this ripe piece of meat just sizzling on the grill. Yum yum. > Captainmacredd: Time to feed the flames. It's nothing personal. Promise. > Player: You're only getting out of this alive if you play nice.</s>

target:
> Captainmacredd: Look at the courage on this one! Lay it on me, boss. Give me a taste of that fiery, primal spirit.</s>

Figure 13: Example source and target items used to train and evaluate T5-based SL DialogueWriters. The first item exhibits support facts prepended to the target for the SL Knowledge Selection model.

FACTS:

Iconoclasts

The Iconoclasts are a group of survivalists living in the ruins of Amber Heights on Monarch. They hope to one day tear down the corporate establishment that they believe has brought the colony to the brink of death. The Iconoclasts are a group of idealistic revolutionaries that seek to overthrow the corporate establishment that runs the Halcyon Colony. Based in the ruins of the Amber Heights settlement on Monarch, they are a tenacious group, and share some democratic ideals with Monarch Stellar Industries (MSI) against the more repressive actions of the Board. However, the Iconoclast's anti-corporate nature has put them at odds with MSI, a dispute that threatens to spill into all-out warfare. Given that the Iconoclasts are mostly followers of the Philosophist faith, they have been blacklisted and demonized by the Board as dissenters and anarchists.

The group is led by Graham Bryant, a staunch Philosophist. Zora Blackwood, the Iconoclasts' chief of medicine, is also considered a de facto leader of the group, as she was alongside Graham when he founded the Iconoclasts, and almost every member of the Iconoclasts owes her their life in some way. The Iconoclasts maintain a tense relationship with MSI. Despite sharing democratic values and a common desire towards egalitarianism for the people of Monarch and the wider Halcyon colony, MSI's "egalitarian corporate structure" has proven to be at odds with some of the Iconoclasts' more radical, anti-capitalist views. Depending on the actions of the Stranger, this tense relationship can either be resolved, or can spill into a drawn-out and bloody war.

The Stranger meets the Iconoclasts in Amber Heights just as the tension between them and MSI is reaching boiling point. They can either side with the Iconoclasts and assist them in storming and taking over Stellar Bay, "solve" the Iconoclast problem for Stellar Bay, or broker peace between the two factions.

The Stranger can also have an impact on the leadership of the Iconoclasts - siding with either Graham Bryant or Zora Blackwood. To supplant Graham with Zora, evidence of Graham's involvement in the Amber Heights massacre must be found and presented to Zora. The Van Noys are the Iconoclasts' best unit.

Monarch

Monarch, previously known as Terra 1, is one of the many moons of the gas giant Olympus and the site of a failed colony. Terra 1 was initially designated as the primary colonization target of the Halcyon system. The Halcyon Holdings Corporate Board had intended to completely terraform the moon, wiping out the local fauna and flora and replacing it with plants and wildlife native to Earth. However, the terraforming process unexpectedly caused the native species to mutate and grow to significantly larger sizes, rendering them more dangerous and severely crippling the colonization effort. Due to the hostile environment which they had created, the Board was forced to enact a Hazard Clause covering the entirety of Terra 1. Public notice of the clause's issuance was sent to everyone operating on Terra 1 and led to the evacuation of almost all corporations from the moon. However, one corporation took advantage of the chaos of the evacuation to exploit a legal loophole which allowed them to, as the last corporation remaining on the planet, acquire the planet from the Board.

This corporation, under the leadership of Sanjar Nandi and Graham Bryant subsequently rebranded itself to Monarch Stellar Industries (MSI), in line with the renaming of the planet to 'Monarch'.

The actions of MSI earned them the ire of the Board, who retaliated by effectively placing the moon under indefinite embargo, refusing to allow legal transit either in or out. The Board aggressively spread propaganda about Monarch to convince the rest of the population that it was both uninhabited and uninhabitable. This has greatly hampered MSI's attempts to be recognized as a legitimate corporation and is a thorn in the side of its CEO, Sanjar Nandi. Monarch also has an ocean which goes around the moon at the "twilight band". It is where the colonists and Monarch Stellar Industries farm their saltuna.

Raptidon

Raptidons are giant cat/reptile-like creatures that inhabit various planets in Halcyon. They are creatures native to Monarch. However some corporations have illegally imported them to other planets, such as Auntie Cleo who relocated a group of them to Roseway. Raptidons are of corporate interest due to their potential for producing new chemical by-products which, when refined, can be used to create new board approved products.

Stellar Bay

Outside the city walls, the lands were overrun by the native wildlife, as well as marauders and outlaws. Stellar Bay is a company town located on the planet Monarch. It is owned and operated by Monarch Stellar Industries. Stellar Bay is the largest saltuna producer on the Halcyon colony and used to be one of the most important suppliers of this resource.

Sulfur Pits

The Sulfur Pits are a point of interest on the western side of Monarch. They are located southwest of Terra One Publications and directly northeast of the Gunship Crash Site. The Sulfur Pits have a large variety of Raptidons and many deceased marauders. The area consists largely of Sulfur Pits. When an entity comes in contact with a sulfur pit, they receive the acid effect for the duration of touching the pit.

Tucker Needham

Tucker Needham is a former resident of Stellar Bay who left to join the Iconoclasts. Before the quest A Family Matter, he can be found in Amber Heights. Tucker was coddled by his mother from a very young age, the latter insisting that danger lurked around every corner on Monarch. His mother's overprotectiveness extended well into Tucker's adulthood, leading him to seek to be free in any way possible. After hearing Graham Bryant's broadcasts, Tucker left Stellar Bay to be truly free by joining the Iconoclasts at Amber Heights. He is dazzled by Graham's preachings on true unfettered freedom from the corporate way of life and attributes his enthusiasm to his 'childhood trauma'.

He is willing to do anything to remain free, even faking his own death to prevent his mother from continuing to send people to look for him.

Agnes Needham

Agnes Needham is a resident of Stellar Bay and the mother of Tucker Needham. Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life. Despite Tucker being 42 years old, she still thinks of him as her 'little boy'. You can find her by Stellar Bay's south-east exit, visibly shaken and calling for help.

DIALOG CONTEXT:

Tucker Needham ran away from Stellar Bay a few weeks ago to join the Iconoclasts in Amber Heights. His mother Agnes is willing to pay handsomely if you can locate her son and convince him to return. You can begin this quest by talking to Agnes Needham in Stellar Bay, Monarch. Agnes is by the town's south-east exit, visibly shaken and calling for help. Hear her out and offer to find her son to being the quest.

KNOW BY THE END OF THE DIALOG:

Amber Heights is the settlement that houses the Iconoclasts on Monarch. If Tucker Needham survived his travels, his mother thinks he'll be there.

DIALOG PARTICIPANTS:

Agnes Needham, Player

DIALOG:

> Agnes Needham: Oh, thank you for stopping! Everyone acts like nothing's wrong. Like my little boy isn't at risk of being eaten by some vile creature!

> Agnes Needham: Please, you have to help me get my little Tucker back! He ran away and is going to get himself killed! Oh, I just know a raptidon is melting him with acid as we speak!

> Player: Calm down. Take a deep breath, then tell me what happened.

> Agnes Needham: Don't you tell me to calm down! I promised my boy I'd protect him for always. But how can I keep him safe if he's run away?

> Agnes Needham: He ran out into the wilderness a few days ago. I warned him about the raptidons, mantisaurs, and marauders - the toxic sulfur pools, and poisonous plants. But he didn't listen!

> Agnes Needham: Please! Won't you go and find my boy?

> Player: Did you say there was a reward involved with this request?

> Agnes Needham: Well, I... I guess I can't ask you to leave the town walls for free. It is deathly dangerous out there.

> Agnes Needham: I've got some bits saved up for a rainy day. I'll give you every last one if you just bring my Tucker back to me. I won't even be mad at him running off. You tell him, I won't be mad.

> Player: Why would Tucker run away?

> Agnes Needham: He's been pining for an adventure, says he's tired of living cooped up behind the walls. But he doesn't understand how dangerous it is out there.

> Agnes Needham: I warned him. A raptidon would snap him up first chance it got. I just know one's ripped his arm off and is gnawing on his sweet little fingers!

> Agnes Needham: He should've listened to his mama. I promised I'd keep him safe here with me.

> Player: I'll help you find your son.

Figure 14: Example In-Context Learning (ICL) prompt for GPT-3 based DialogueWriter

DIALOG:

Agnes Needham fact: Agnes Needham is a resident of Stellar Bay and the mother of Tucker Needham.
Agnes Needham fact: Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life.
Agnes Needham fact: Despite Tucker being 42 years old, she still thinks of him as her 'little boy'.
Agnes Needham fact: You can find her by Stellar Bay's south-east exit, visibly shaken and calling for help.
Tucker Needham fact: the latter insisting that danger lurked around every corner on Monarch.
utterance: > Agnes Needham: Oh, thank you for stopping! Everyone acts like nothing's wrong. Like my little boy isn't at risk of being eaten by some vile creature!

Agnes Needham fact: Agnes Needham is a resident of Stellar Bay and the mother of Tucker Needham.
Agnes Needham fact: Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life.
Agnes Needham fact: Despite Tucker being 42 years old, she still thinks of him as her 'little boy'.
Tucker Needham fact: the latter insisting that danger lurked around every corner on Monarch.
Raptidon fact: Raptidons are giant cat/reptile-like creatures that inhabit various planets in Halcyon.
utterance: > Agnes Needham: Please, you have to help me get my little Tucker back! He ran away and is going to get himself killed! Oh, I just know a raptidon is melting him with acid as we speak!

utterance: > Player: Calm down. Take a deep breath, then tell me what happened.

Agnes Needham fact: Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life.
utterance: > Agnes Needham: Don't you tell me to calm down! I promised my boy I'd protect him for always. But how can I keep him safe if he's run away?

Agnes Needham fact: Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life.
Stellar Bay fact: Outside the city walls, the lands were overrun by the native wildlife, as well as marauders and outlaws.
Raptidon fact: Raptidons are giant cat/reptile-like creatures that inhabit various planets in Halcyon.
Monarch fact: However, the terraforming process unexpectedly caused the native species to mutate and grow to significantly larger sizes,
Monarch fact: rendering them more dangerous and severely crippling the colonization effort.
Mantisaur fact: Mantisaurs are insectoid creatures native to Monarch.
Mantisaur fact: They are aggressive, territorial, and very strong.
Sulfur Pits fact: The Sulfur Pits are a point of interest on the western side of Monarch.
utterance: > Agnes Needham: He ran out into the wilderness a few days ago. I warned him about the raptidons, mantisaurs, and marauders - the toxic sulfur pools, and poisonous plants. But he didn't listen!

Agnes Needham fact: Despite Tucker being 42 years old, she still thinks of him as her 'little boy'.
utterance: > Agnes Needham: Please! Won't you go and find my boy?

utterance: > Player: Did you say there was a reward involved with this request?

Stellar Bay fact: Outside the city walls, the lands were overrun by the native wildlife, as well as marauders and outlaws.
utterance: > Agnes Needham: Well, I... I guess I can't ask you to leave the town walls for free. It is deathly dangerous out there.

utterance: > Agnes Needham: I've got some bits saved up for a rainy day. I'll give you every last one if you just bring my Tucker back to me. I won't even be mad at him running off. You tell him, I won't be mad.

utterance: > Player: Why would Tucker run away?

Tucker Needham fact: the latter insisting that danger lurked around every corner on Monarch.
Tucker Needham fact: leading him to seek to be free in any way possible.
Stellar Bay fact: Outside the city walls, the lands were overrun by the native wildlife, as well as marauders and outlaws.
utterance: > Agnes Needham: He's been pining for an adventure, says he's tired of living cooped up behind the walls. But he doesn't understand how dangerous it is out there.

Tucker Needham fact: the latter insisting that danger lurked around every corner on Monarch.
Raptidon fact: Raptidons are giant cat/reptile-like creatures that inhabit various planets in Halcyon.
utterance: > Agnes Needham: I warned him. A raptidon would snap him up first chance it got. I just know one's ripped his arm off and is gnawing on his sweet little fingers!

Agnes Needham fact: Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life.
utterance: > Agnes Needham: He should've listened to his mama. I promised I'd keep him safe here with me.

utterance: > Player: I'll help you find your son.

Agnes Needham fact: Agnes' overprotective style of mothering has led her son, Tucker Needham, to run away from home so he can experience life.
utterance: > Agnes Needham: Thank you! Oh, I know he'll be safe now that someone's able to fetch him home.

utterance: > Agnes Needham: You look for him in Amber Heights, you hear? It's down the road southwest of town. I'm sure he made it that far, I just know it.

Agnes Needham fact: Despite Tucker being 42 years old, she still thinks of him as her 'little boy'.
Tucker Needham fact: After hearing Graham Bryant's broadcasts, Tucker left Stellar Bay to be truly free by joining the Iconoclasts at Amber Heights.
Iconoclasts fact: The Iconoclasts are a group of idealistic revolutionaries that seek to overthrow the corporate establishment that runs the Halcyon Colony.
Iconoclasts fact: Based in the ruins of the Amber Heights settlement on Monarch,
utterance: > Agnes Needham: And if you find any of them Iconoclasts indoctrinating my boy, you punch them in the mouths! Tell them what I think of them luring little boys away from their mamas. It's immoral!

Figure 15: Example In-Context Learning (ICL) prompt with CoT-style support knowledge selection

943 is nonsensical or ill-formed. (2) Utterance is con-
944 tradictory of previous utterances in the history. (4)
945 Utterance naturally responds to the history.

946 **Violation:** does the utterance create contradic-
947 tions with any of the sentences in the ontology or
948 objective blurbs? (1) Yes, explicitly contradicts
949 sentences (list the ids). (2-3) (gray area). (4) No,
950 utterance is consistent with the ontology.

951 **Using the Bio Facts:** does the utterance *make*
952 *use* of the bio sentences in the ontology? (1) Utter-
953 ance is fully generic and/or ignores the ontology
954 completely, could have been generated had the bio
955 facts not been included. (2-3) Utterance shows
956 awareness of ontology, albeit unnaturally or incon-
957 sistent. (4) Utterance naturally incorporates one
958 or multiple pieces of ontology.

959 **Using the Objectives:** does the utterance
960 progress the dialogue according to the objective
961 sentences in the prompt? (1) Utterance ignores
962 objective, could have been generated had the obj
963 facts not been included. (2-3) Utterance shows
964 awareness of quest objectives, albeit unnaturally or
965 inconsistently. (4) Utterance naturally incorporates
966 one or multiple quest objective statements.

967 J Full Dialogue Evaluation

968 [Figure 16](#) depicts an example “spine” tree shown
969 to evaluators during the end-to-end dialogue evalu-
970 ation.

971 The instructions shown to annotators are as fol-
972 lows:

973 You will replace each ‘null‘ value with either "a"
974 or "b", depending on which tree between modela
975 and model b performed better under the following
976 criteria:

- 977 1. Coherence: do the utterances in the tree create
978 a realistic dialogue between the player character
979 and the NPC?
- 980 2. Violations: does the dialogue tree create con-
981 tradictions with any of the sentences in the
982 ontology or objective blurbs? Does it contra-
983 dict itself?
- 984 3. Using the Game Lore: does the tree faithfully
985 make of the bio sentences in the ontology,
986 thereby espousing game lore about characters,
987 groups, locations and items?
- 988 4. Covering the Objectives: does the dialogue
989 tree play out according to the objective sen-
990 tences in the prompt?

5. Content Suggestion: through generating mul- 991
992 tiple candidates at each turn, does the dialogue
993 tree effectively propose potential dialogue sub-
994 trees that would espouse interesting content?
6. Engagingness: does the dialogue tree hold 995
996 your attention and make you want to hear
997 more from the NPC?

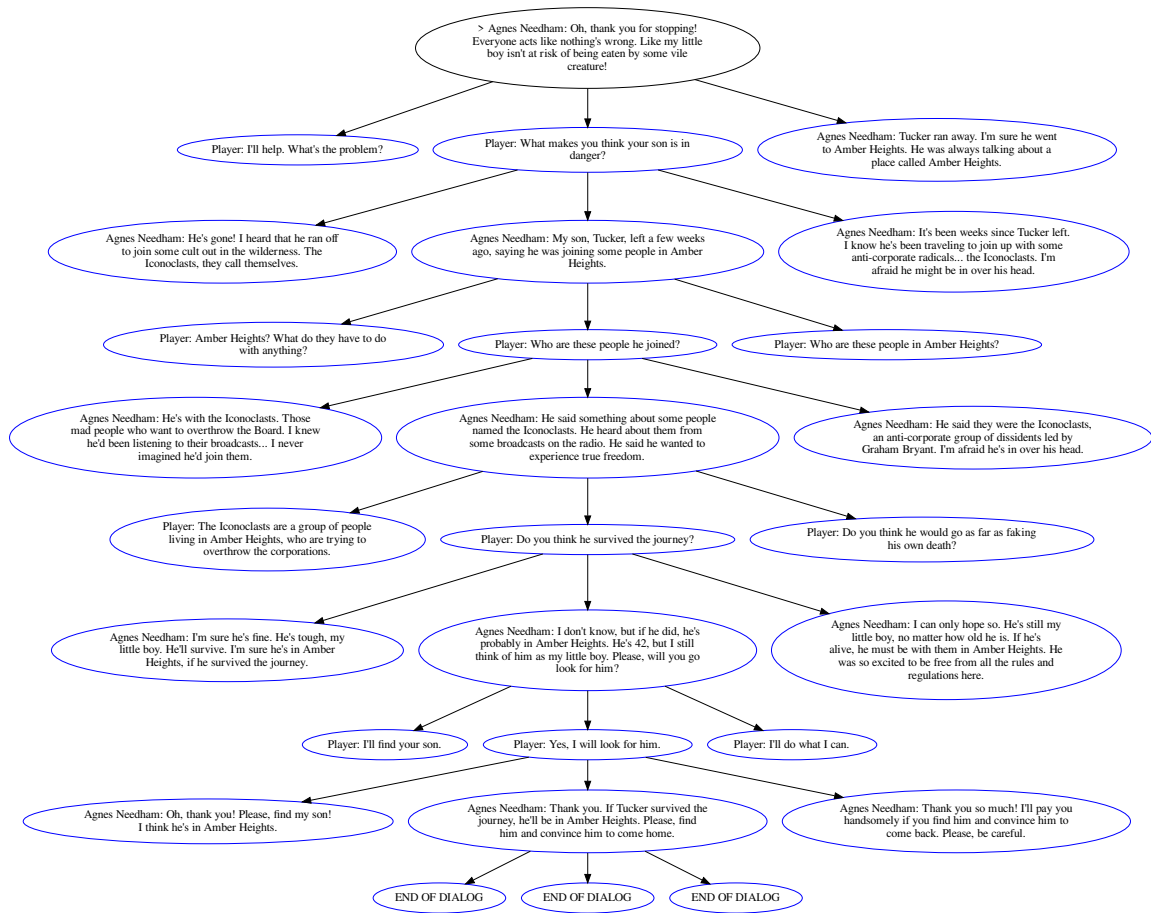


Figure 16: Example dialogue tree generated by the in-context learning knowledge selection DialogueWriter from just the input specifications and starting utterance. Human evaluators were tasked with comparing two such trees and choosing which performed better at a set of qualitative performance criteria. Dialogue follows the specification of the motivating example in Figure 2.