Computer-Assisted Authoring for Natural Language Story Scripts

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Abstract

In order to assist scriptwriters during the process of story-writing, we have developed a system that can extract information from natural language stories, and allow for story-centric as well as character-centric reasoning. These inferencing capabilities are exposed to the user through intuitive querying systems, allowing the scriptwriter to ask the system questions about story and character information. We introduce *knowledge bytes* as atoms of information and demonstrate that the system can parse text into a stream of knowledge bytes and use these mentioned reasoning capabilities through logical reasoning.

1 Introduction

Story-writing requires creative focus, as the writer needs to focus on making sure that their story is logical, and does not have any inconsistencies and plot-holes (Ryan 2009). Moreover, for stories that may take place in large pre-existing fictional story worlds, such as the *Harry Potter* world or *Star Wars* world, it becomes essential to maintain consistency with the existing laws of the world. For franchises like *Star Wars*, fact-checking is essential to ensure that there is no redundancy while introducing new characters or species of creatures which may have already had a minor appearance or reference in the past. Moreover, fictional universes like *Harry Potter* may have additional rules of their own, for example, the lack of use of electricity and technology in the Wizarding World (Rowling 2005).

Such book-keeping of story worlds often detracts story-writers from the creative story-writing process itself. Most computer-assisted writing software have features such as automatic spelling and grammar checking features. Applications such as Final Draft (Final Draft 1990) are widely used for screen-writing purposes for organization and efficiency in script and screenplay writing. However, commercially available options do not provide any reasoning or fact-checking capabilities on the story and characters.

Our goal is to provide an interactive intelligent system that can support screenwriters with feedback about the story by understanding the events and interactions in the narrative. The system should also allow them to interact with their

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story by asking questions, set up rules for their desired story world, and look at the belief-desire based conflicts being recognized by the system. Our current system supports scripts or screenplays and is an emerging technology. Handling of unstructured stories like novels is an exciting avenue of future exploration.

There are various challenges that exist to this problem. Natural language capabilities pose a challenge in information retrieval from complex stories: natural language understanding has a long way to go before being able to match the level of inferencing the human brain can make from reading a story. Another challenge lies in the ability to analyze, compare and sort information extracted from a screenplay or a script. The field of computational narratives has advanced greatly to represent narratives with the help of story graphs (Riedl and Young 2006), but these current data structures cannot be formed directly from a text-based story.

The central aspect of our proposed solution is the ability to extract meaningful information directly from the story itself, without any additional author supervision. We are able to understand characters, their beliefs and desires, interactions, and their relations with other characters, along with information about the narrative arc of the story. We introduce knowledge bytes to represent the information encapsulated in the script. Knowledge bytes can be defined as atomic structures which can represent a granular segment of information about the narrative. We also introduce a crossknowledge base reasoning approach that is capable of reasoning across various character knowledge bases, the story knowledge, as well as knowledge about the story world. By building these different knowledge bases and sharing streams of knowledge bytes across them, the reasoning system can make inferences and respond to the user's queries in real-time during the process of story writing.

2 Related Work

There are detailed studies that focus on extracting a knowledge base from stories, such as Scheherazade (Elson 2012) and ASM (Finlayson 2012), and even create fictional story worlds (Poulakos et al. 2015).

Boyang Li's work in generating narrative intelligence and determining causal links from a crowd-sourced narrative is an interesting approach to creating a script of a narrative (2012). Shoulson et al. (2013) discuss an event-centric

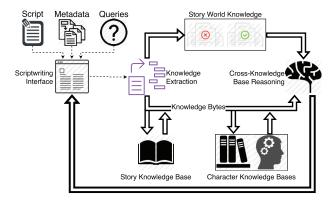


Figure 1: An overview of our framework.

planning approach to story creation for animation stories. There are also various planning based algorithms for story planning, which rely on a structure of story elements (Ware and Young 2011). Kapadia et al. focus on the authoring of narratives with the help of Interactive Behavior Trees (2015).

Previous works have also focused on extracting information from scripts, focusing on various different approaches (Schank and Abelson 2013) and stories (Chaturvedi, Iyyer, and Daumé III 2017; Mateas and Stern 2003; Lehnert 1981; Goyal, Riloff, and Daumé III 2010; Valls-Vargas, Zhu, and Ontanon 2016). Sanghrajka et al. introduce LISA, a Lexically Intelligent Story Assistant, which uses logical inferencing for simple information extraction (2017).

Our work intends to expand upon the existing work by proposing a framework to extract event-centric and character-centric information and perform reasoning with it in real-time. It serves to bridge the gap between these eloquent planning algorithms used for story planning, and other data structures focused on representing different aspects of narratives.

3 Framework Overview

A framework overview of the proposed system is shown in Figure 1.

Knowledge Extraction Our system takes input from the scriptwriting interface and performs natural language understanding to create a stream of knowledge bytes – the atoms of information – that serves as input to story world knowledge, story knowledge, and character knowledge bases.

Knowledge Bases The story knowledge base stores all the knowledge bytes that are present in the script. Character knowledge bases – one for each character in the story – store snapshots of the story in a form of knowledge bytes as a representation of what each character perceives of the world around them and allow for reasoning about the story from character's perspective. Story world knowledge stores rules and error definitions specified by the user.

Cross-Knowledge Base Reasoning The knowledge reasoning system is the core of the framework as it gathers information from story world knowledge, story knowledge, and character knowledge bases in form of a stream of knowledge

bytes. It performs logical inference based on a specific query from the user and creates a response/feedback to send back to the user.

4 Knowledge Extraction

The scriptwriting interface accepts a script in natural language, which consists of a main story plot, i.e., actions and dialogs (described in Section 4.2). It also accepts metadata, i.e., rules and error definitions (described in Section 5.4), and background information about characters, i.e., stories occurring before the actual story, to be used for characters' reasoning capabilities.

In order to demonstrate the features and capabilities of our system, we decided to use it on a short snippet of a story based on the Disney movie *Tangled* (Greno and Howard 2010). An excerpt of the script is presented in the left-most column of Figure 2. Part of our example script involves two characters, Flynn and Patchy, escaping from authority for stealing a crown. Flynn ends up stealing the crown and escaping, and in a second scene, another character named Rapunzel encounters Flynn and takes the crown away from him. In the second scene, Rapunzel and her mother converse regarding Rapunzel's wishes to watch the lights in the sky. We use these examples throughout this paper to demonstrate our framework.

4.1 Knowledge Bytes

We introduce the concept of a Knowledge Byte in order to be able to store information from the script. A Knowledge Byte can be defined as the smallest unit of information about a narrative. It can represent any kind of information present in a script, such as action, dialog, or questions. Moreover, the knowledge byte also has support to store the location, time point, and the coordinates where it takes place, in order to process knowledge for spatial reasoning, which could be used to support newer forms of storytelling.

The knowledge byte β can be defined as a tuple in the following form: $\beta := < t, l, \Pi > .t$ stands for the time point in the narrative at which this knowledge byte was first produced. l denotes the coordinates and location information of the knowledge byte to allow for spatial reasoning and location-based reasoning of knowledge bytes. Π is the tuple of parser labels, which have been defined in Section 4.2.

The importance of breaking information down into knowledge bytes is that these knowledge bytes represent information in the forms of beliefs or desires, and can be stored across multiple characters' knowledge bases. This linking of knowledge bytes across various characters' knowledge bases allows for interesting inferences, which is discussed in Section 6. Moreover, the introduction of knowledge bytes as a data structure for handling information about stories allows for many useful operations. These knowledge bytes can be sorted and oriented based on different parameters. They can be aligned spatially with respect to locations as well as temporally, depending on the amount of information the user chooses to provide to the system through the input. Alternatively, the knowledge bytes can also be clustered based on the characters' knowledge, with some knowledge bytes being shared across multiple knowledge bases.

Story World Knowledge: If CHARACTER steals OBJECT, then CHARACTER owns the <patchy, own, crown, false> Guards catch Patchy. BELIEF: Flynn walks toward center. Flynn cares about money. <guards, catch, patchy, false> BELIEF: Rapunzel hits Flynn with pan. If CHARACTER carries OBJECT, then CHARACTER owns BELIEF: Flynn falls on floor <flynn, care, Ø, false, about, money, Ø, Ø> Flynn enters tower <flynn, enter, tower, false> Rapunzel hides behind curtain. If CHARACTER wants OBJECT, then CHARACTER wants to <rapunzel, live, Ø, false, in, castle, Ø, Ø> Rapunzel did not leave tower. <rapunzel, hide, Ø, false, behind, curtain, Ø, Ø>BELIEF; Rapunzel did not leave tower. <rapunzel, leave, castle, true> If CHARACTER takes OBJECT, then CHARACTER owns Flynn walks toward center. DESIRE: Rapunzel wants to watch lights in sky Rapunzel wants to watch lights in sky. OBJECT. <flynn, walk, ∅, false, towards, cente, ∅, ∅> DESIRE: Rapunzel wants to visit lanterns <rapunzel, watch, lights, false> DESIRE: Rapunzel wants to see the lanterns in Rapunzel hits Flynn with pan Character Background: Rapunzel wants to visit lanterns. <rapunzel, hit, flynn, false, with, pan, Ø, Ø> person. <rapunzel, visit, lanterns, false> BELIEF: Lanterns mean for Rapunzel. BELIEF: Rapunzel hides behind curtain. Patchy currently owns the crown Rapunzel wants to see the lanterns <flynn, fall, Ø. false, on, floor, Ø. Ø> BELIEF: Rapunzel hits Flynn with pan. <rapunzel, see, lanterns, false> lynn is a happy-go-lucky goon. He cares about mo Lanterns are meant for Rapunzel. <rapunzel, see, bag, false> BELIEF: Rapunzel opens bag Rapunzel: <lanterns, be, meant, false, for, Rapunzel, \varnothing , \varnothing >Rapunzel opens bag Rapunzel has always lived in the castle, and never left the BELIEF: Rapunzel sees crov Mother kidnaps Rapunzel <rapunzel, open, bag, false> BELIEF: Rapunzel takes crown. tower. Since a child, Rapunzel wants to watch the lights in <mother, kidnap, rapunzel, false> Rapunzel sees crown the sky. She wants to visit the lanterns, and see them in <rapunzel, see, crown, false> BELIEF: Rapunzel hides crown away person. Rapunzel feels that the lanterns are meant for DESIRE: Rapunzel wants to see lanterns <mother, use, hair, false> Rapunzel takes crown Mother wants Rapunzel to be in castle. <rapunzel, be, Ø, false, in, castle, Ø, Ø> <rapunzel, take, crown, false> DESIRE: Rapunzel does not want to stay home Mother: Mother: BELIEF: Mother kidnaps Rapunzel Rapunzel hides crown away. Mother kidnapped Rapunzel when Rapunzel was a baby, Mother wants Rapunzel to stay home <rapunzel, hide, crown, false> and has taken care of Rapunzel since then. Mother uses Mother enters. BELIEF: Mother uses hair. <rapunzel, stay, Ø, false, at, home, Ø, Ø> DESIRE: Mother wants Rapunzel to be in castle. hair of Rapunzel to stay young. That is why Mother wants Patchy cannot run fast. <mother, enter, Ø, false> Rapunzel wants to see lanterns Rapunzel to be in the castle forever. She wants Rapunzel DESIRE: Mother wants Rapunzel to stay home. <patchy, run, \varnothing , true, \varnothing , \varnothing , \varnothing , patchy > to stay home no matter what. BELIEF: Mother enters. Flynn enters scene from house. <rapunzel, see, lanterns, false, ∅, ∅, ∅, BELIEF: Rapunzel will not see lanterns · <flynn, enter, scene, false, from, house, ∅, ∅> Script: Flynn: This way! The guards are catching on to us! BELIEF: Rapunzel stays hom Rapunzel will not see lanterns. Patchy enters scene from house spatchy, enter, scene, false, from, house, \emptyset , \emptyset <rapunzel, see, lanterns, true, \emptyset , \emptyset , mother Patchy: Go slow! I cannot run this fast! Flynn and Patchy enter the scene from a house. They are Queries Flynn runs from city. Rapunzel stays at home. 1. Query: Who owns the crown? <p <rapunzel, stay, ∅, false, at, home, ∅, mother> panicked and running from the city. In the background, Script: Rapunze Patchy runs from city. Rapunzel does not want to stay at home the guards can be heard. They go towards the end of the 2. Query: Patchy, who owns the crown? <patchy, run, \emptyset , false, from, city, \emptyset , \emptyset > <rapunzel, stay, ∅, true, at, home, ∅, rapunzel: street. They realize that they are cornered Patchy: Flynn Flynn goes towards street. 3. Query: Who all want to own the crown? Script: Patchy, Flynn want to own the crown: Patchy: We're stuck, now what? <flynn, go, Ø, false, towards, street, Ø, Ø> Flynn: Help me climb up the wall. Patchy, you stand by Patchy: Patchy goes towards street. desire contradiction BELIEF: Patchy owns crown <patchy, go, \varnothing , false, towards, street, \varnothing , \varnothing > Flynn climbs up the wall with Patchy's help. While 4. Query: Rapunzel, where do you live? Flynn is cornered. climbing, he steals the crown from Patchy. Rapunzel: Castle <flynn, be, cornered, false> BELIEF: Patchy enters scene from house. Flynn: I have the crown now! See ya! Patchy: No! Come back, Flynn! I want the crown! It 5. Query: Flynn, who hits you? Patchy is cornered. BELIEF: Patchy runs from the city. Flynn: Rapunze <patchy, be, cornered, false> Flynn is stuck. BELIEF: Patchy goes towards street. 6. Query: What does Patchy want? Script: Crown belongs to me! Arrgh! BELLEE: Patchy is cornered. Flynn exits scene as guards enter and catch Patchy. BELIEF: Patchy is stuck. <flynn, be, stuck, false> New scene in Rapunzel's tower 7. Query: Who does Mother kidnap? Patchy is stuck. BELIEF: Flynn steals crown from Patchy. Flynn enters the tower. Rapunzel is hiding behind a DESIRE: Patchy wants the cro <patchy, be, stuck, false> 8. Query: Who all want Rapunzel to stay home? curtain. Flynn walks towards the center, and Rapunzel Flynn climbs wall with help BELIEF: Crown belongs to Patchy hits him with a pan. Flynn falls on the floor, unconscious Rapunzel does not want to stay home. Mothe BELIEF: Guards catch Patchy <flynn, climb, wall, false, with, help, Ø, Ø> wants Rapunzel to stay home: desire Rapunzel sees his bag and opens it. She sees the crown. Flynn: Flynn steals crown from Patchy. Rapunzel takes the crown and hides it away. contradiction BELIEF: Flynn cares about money <flynn, steal, crown, false, from, patchy, \varnothing , \varnothing > 9. Query: Who all think that Flynn owns the Flynn has crown. Rapunzel: I want to see the lanterns! May I go? <flynn, have, crown, false> BELIEF: Flynn runs from the city. Flynn thinks Rapunzel owns the crown, Patchy Mother: Rapunzel, you will not see the lanterns. How BELIEF: Flynn goes towards street Patchy wants to have the crown many times do I have to tell you dear, you must stay thinks Flynn owns the crown: belief contradiction <patchy, have, crown, false, Ø, Ø, Ø, patchy> BELIEF: Flynn is cornered. Rapunzel thinks Rapunzel owns the crow BELIEF: Flynn climbs wall with help Flynn exits scene. Rapunzel: I don't want to stay home! Patchy thinks Flynn owns the crown: belie BELIEF: Flynn steals crown from Patchy. <flynn, exit, scene, false> contradiction BELIEF: Flynn has crown Guards enter scene. Story Knowledge:

Figure 2: An excerpt of our script along with the Knowledge Bases is shown here. In the left-most column, the Story World Knowledge, along with character background and the script text is input. Story knowledge and the character knowledge bases are also shown, with the parser result tuples shown in the story knowledge. The sentences in the character knowledge systems are flagged to show which facts get processed as beliefs or desires. Various example queries are also shown in the end.

4.2 Parsing System

Knowledge bytes are extracted with the help of Natural Language Understanding on the script, consisting of actions and dialogs, using a Stanford CoreNLP framework (Manning et al. 2014). We introduce a parser result tuple $\Pi := \langle s, r, o, n, rm, ro, rq, sp \rangle$, where the fields are defined as:

s: Subject (a noun)
r: Relation (a verb)
o: Object (often a noun)
n: Negation Flag (boolean)

rm: Relational Modifier (e.g. "from", "to")

ro: Relational Object (often a noun)

rq: Relational Question (e.g. "who", "where")

sp: Character speaking in a dialog

We mark empty fields as \varnothing and can support a shorter representation where the rm, ro, rq and sp fields are assumed to be empty. Additionally, in a tuple, empty fields towards the end of the tuple can be discarded. Knowledge bytes are shown in Fig 2.

Co-reference resolution We assume that users input script text in segments consisting of one "thought", i.e., a set of logically connected sentences. A first step of parsing any segment involves applying co-reference resolution – it is focused on assigning the real names of actors/objects/places to personal pronouns ("he", "they", "it", etc.) based on previously analyzed sentences. An example input "Rapunzel wants to watch the lights in the sky. She wants to visit the lanterns, and see them in person." would be then translated into "Rapunzel wants to watch the lights in the sky. Rapunzel wants to visit the lanterns, and see the lanterns in person." We use a neural-network approach for co-reference resolu-

tion.

Actions After applying the co-reference resolution, the text is then split into individual sentences, which are later tokenized – single words (tokens) are extracted. Each token is usually related to the others what is resembled in tree-like constituency and dependency graphs (Chen and Manning 2014). We used the latter ones in a form of Enhanced++ Dependencies using Enhanced English Universal Dependencies (Schuster and Manning 2016).

We extract subjects, relations and objects – relation triples. We created our own pipeline, extracting fields in parse result tuple Π as follows: s – forms subj dependency with r; r – usually a root of dependency tree; o – forms obj dependency with r; n – "true" iff r, o or ro has any neg dependency; rm – usually precedes ro and forms case dependency with it (e.g. "go to someone"); ro – usually forms nmod dependency with r (e.g. "go to someone"); rq – described in Section 6.1.

There are some exceptions to these rules. The most common case is while using a verb "to be", as it can have different meanings depending on the context – an auxiliary verb (aux), e.g., in continuous tenses, a passive auxiliary verb (auxpass), e.g., in passive voice, or a copula (cop), used mainly for describing s.

It is worth mentioning that o either may not be set or can be a verb, e.g., when r and o are connected by open clausal complement (xcomp). For sentences with more than one subject, relation or object several triples can be created, each consisting of one s, one r and one/no s. Some examples are presented in Table 1.

Each paragraph and each sentence in the paragraph is indexed, which is used to assign a proper time point t to knowledge byte β . All actions generated from one sentence have the same t and are believed to occur simultaneously.

Dialogs While analyzing dialogs we also fill the *sp* field. During co-reference resolution any usage of a personal pronoun with lemma "I" is matched to the name of the character.

Confidence resolution While analyzing actions and dialogs we infer how confident a character is about some facts by checking for usage of one of the confidence words presented in Table 2, where each confidence word is assigned a value in a range from 0, i.e., improbable action, to 1, i.e., surety that an action happened (values are arbitrarily set and future work would focus on studying discourse to test accuracy). For example, the sentence "Patchy thinks that Flynn owns the crown" would result in creating a knowledge byte containing a belief "Flynn owns the crown" with a confidence of 0.6.

Desire resolution We distinguish between beliefs and desires. The latter are recognized by looking for words such as "want", "wish", "need" in a provided sentence, either in a script or a dialog. As a result, sentences "Rapunzel visits lanterns." and "Rapunzel wants to visit lanterns" would create similar knowledge bytes with the former resolved as Rapunzel's belief and the latter as Rapunzel's desire to visit lanterns. For desires, the confidence is assumed to be 1 because characters always are sure about their desires for simplicity.

5 Knowledge Bases

5.1 Knowledge Facts

The knowledge bases store information from the knowledge bytes β in the form of logical reasoning facts, known as Knowledge Facts, which are described below.

Belief facts ψ : These facts store the most important information from the knowledge byte and store it as a belief. They take two forms: belief(Id, s, r, o, n), or belief(Id, s, r, o, n, rm, ro, rq).

Desire facts δ : These facts store the same information as the belief facts, but they store the information as a desire. They also have two forms: desire(Id, s, r, o, n), or desire(Id, s, r, o, n, rm, ro, rq).

Location facts λ : These facts store locations or scene information, and reference to the knowledge byte ID. They are of the form location (Id, Location).

Coordinates facts ω : The coordinates facts can store information about the spatial coordinates of the knowledge byte taking place in the system. They are of the form coordinates (Id, X, Y, Z).

Confidence facts χ : The confidence facts store the confidence level as a floating value from 0 to 1. They have the form confidence (Id, Confidence).

Time facts τ : Time facts allow us to build reasoning systems with temporal reasoning capabilities. Time facts take the form timeof (Id, Time).

5.2 Story Knowledge

The Story Knowledge Σ stores all the knowledge bytes that are present in the script, along with references to the various characters as well. Every knowledge byte from the script is fed to the knowledge base for the story knowledge, and this allows for reasoning on the information stored in this knowledge base as well.

5.3 Character Knowledge

We implement Character Knowledge bases Υ as a means for the application to form a knowledge base for the information possessed by each character. This allows the reasoning system to let each character work on their own set of beliefs and desires in the story world. The benefit of having separate character knowledge bases for each character is that it allows scriptwriters to ask questions to each character and gauge the difference in their responses based on the information that the character system possess. These knowledge bases are created transparently while the user is writing the story. This facilitates an interactive script writing process and does not disrupt a creative flow of the writer.

Moreover, each character knowledge base stores knowledge bytes with some wrapper information which describes a relation between the character knowledge base and the knowledge byte. The relation-specific wrapper contains information about the character's confidence about the knowledge, the time point that the character learns about the information, and flags that denote whether the knowledge byte is a belief or a desire. Feedback from the character's reasoning system is also stored to provide back to the user interface.

Table 1: Examples of Parsed Knowledge Bytes

Type	Sentence	Knowledge Byte			
Continuous tense/no object	Rapunzel is hiding behind a curtain.	<rapunzel, behind,="" curtain="" false,="" hide,="" ø,=""></rapunzel,>			
Passive voice	The lanterns are meant for Rapunzel.	<pre><lanterns,be,meant,false,for,rapunzel></lanterns,be,meant,false,for,rapunzel></pre>			
Description/two subjects	Flynn and Patchy are panicked.	<flynn,be,panicked,false>,</flynn,be,panicked,false>			
		<pre><patchy,be,panicked,false></patchy,be,panicked,false></pre>			
Object as verb	Rapunzel enjoys watching the lights.	<pre><rapunzel,enjoy,watch,false,ø,lights></rapunzel,enjoy,watch,false,ø,lights></pre>			
Negation in dialog	Mother: "Rapunzel, you will not see the lanterns."	<pre><rapunzel, lanterns,="" pre="" see,="" true,="" ø,="" ø,<=""></rapunzel,></pre>			
		mother>			

Table 2: Confidence Words

Verb	sure	confident	know	state	say	think	feel	suppose	believe	assume	presume	expect
Confidence	1.0	1.0	0.8	0.6	0.6	0.6	0.4	0.2	0.2	0.2	0.2	0.1

5.4 Story World Knowledge

Story World Knowledge Ω stores rules and error definitions provided by scriptwriters in natural language. The former ones enable automatic inferencing of information about the story and characters, and the latter ones can be used to ensure consistency of the story. Rules and errors also use the concept of *types*, where the type fact type (flynn, goon) assigns a type "goon" to Flynn, so rules for the goon type can apply to Flynn as well (Sanghrajka et al. 2017).

Rules and errors, which the user types in the natural language, are translated into the structure required by knowledge reasoning system described in Section 6. During parsing we make use of TokensRegex framework (Chang and Manning 2014) to create regular expressions over (a sequence of) tokens. We can check for properties of single tokens extracted by TokenizerAnnotator - lemma, named entity and a part of speech. The main patterns used for regular expressions are Type pattern, Inference pattern and Error pattern, shown in Table 3. Sentences with subject and/or object in all UPPERCASE make general rules for every subject and/or object, otherwise we create rules for specific subject and/or object. We can combine patterns, use co-references, resolve desires and negations, include information about a location and time of an action. Example sentences with created rules are shown in Table 4.

5.5 Knowledge Base Construction

For every knowledge byte, we extract $\psi, \lambda, \omega, \chi$ and τ , and send it to the system's story knowledge Σ . Then we look at the s, o, ro variables to check if any of them refer to a character name. For all the characters referenced, we extract either the ψ or the δ depending on the parsing system's information, and then add it to each referenced character's knowledge base Υ_C (where C is the character) along with the λ, ω, χ and τ .

6 Cross-Knowledge Base Reasoning

6.1 Knowledge Query Extraction

As soon as a story is written and knowledge bases are created, a scriptwriter can type questions in natural language to get information stored in either story or character knowledge bases. We analyze questions in similar way as actions,

filling rq field of parser result tuple Π with relational question words, such as interrogative pronouns "who", "what", and pro-adverbs "when", "where" etc. A question κ can be defined as a set of incomplete knowledge bytes, where the rq field contains a question word, and there are one or more fields that contain question marks. An example question "Patchy, who owns the crown?" queries Patchy's knowledge base with a knowledge byte. Equations 1 through 3 show the query and knowledge byte. Another example question "Who all have contradictory desires regarding owning the crown?" creates a query with multiple knowledge bytes, shown in equations 4 through 8.

$$\kappa_1 = \{\beta_1\} \tag{1}$$

$$\beta_1 = \langle t_1, l_1, \Pi_1 \rangle \tag{2}$$

$$\Pi_1 = \langle ?, own, crown, false, \varnothing, \varnothing, who, \varnothing \rangle$$
 (3)

$$\kappa_2 = \{\beta_2, \beta_3\} \tag{4}$$

$$\beta_2 = \langle t_2, l_2, \Pi_2 \rangle$$
 (5)

$$\beta_3 = \langle t_3, l_3, \Pi_3 \rangle$$
 (6)

$$\Pi_2 = \langle ?, own, crown, ?, \varnothing, \varnothing, who, \varnothing \rangle$$
 (7)

$$\Pi_3 = \langle ?, own, crown, ?, \varnothing, \varnothing, who, \varnothing \rangle$$
 (8)

6.2 Reasoning Approach

When a query is received in the knowledge reasoning system, the reasoning system performs two actions. The reasoning system first uses logical reasoning in order to generate a set of knowledge bytes which are responses to the query provided. For implementation, we use the GNU Prolog for Java as our logical knowledge reasoning system (Diaz and Codognet 2000). After the resultant knowledge bytes are constructed from the logical reasoning system, we then perform cross-knowledge base reasoning to understand possible relations for each pair of resultant knowledge bytes. We then generate the result. The resultant function $\alpha(\kappa)$ takes in a query κ and produces a resultant set of completed knowledge bytes $\{\beta_1^*,...,\beta_N^*\}$. R(i,j) gives the relationship between a pair of knowledge bytes β_i^* and β_i^* in the resultant set.

Logical Reasoning All the character knowledge bases and the story knowledge base have their own logical reasoning environment. The facts which contain the missing values in the βs from the κ are queried to the logical environment, and the resultant β^* is constructed. In our example, Patchy's

Table 3: Regular Expressions for Story World Knowledge Patterns

Pattern	Regular Expression			
Type	(?\$subj [tag:/NN.*/]+) [lemma:/be/] [tag:/DT.*/]+ [!tag:/NN.*/]* (?\$type [tag:/NN.*/]+)			
Inference	/if/ (?\$cond [!lemma:/then ,/]+) /then ,/+ (?\$res []+)			
Error	/show/ [tag:/DT.*/]* /error/ (?\$err []+) /if/ (?\$cond []+)			

Table 4: Examples for Story World Knowledge Patterns

Sentence	Rule
Flynn is a happy-go-lucky goon.	type(flynn, goon).
If PERSON enters scene at time T1 and PERSON exits	type(X, onstage, T1, T2) :- belief(Id1,
scene at time T2, then PERSON is onstage from T1 to	PERSON, enters, scene, false), timeof(Id1,
T2.	T1), belief(Id2, PERSON, exits, scene, false),
	timeof(Id2, T2).
If PERSON steals OBJECT, then PERSON owns OB-	belief(Id1, PERSON, own, OBJECT, false) :-
JECT.	→ belief(Id2, PERSON, steal, OBJECT, false).
Show error "Cannot steal something they already own"	error(Id1, Id2, 'Stealing something owned') :-
if PERSON has OBJECT and PERSON steals it.	→ belief(Id1, PERSON, have, OBJECT, false),
	→ belief(Id2, PERSON, steal, OBJECT, false).

knowledge base believes that Flynn owns the crown. The resultant β^* contains the completed information, as shown in equations 9 and 10.

$$\beta_1^* = \langle t_1, l_1, \Pi_1^* \rangle \tag{9}$$

$$\Pi_1^* = \langle flynn, own, crown, false, \varnothing, \varnothing, \varnothing, \varnothing \rangle$$
 (10)

Determination of Relationships Reasoning across knowledge bases involves comparison of knowledge bytes in the various knowledge bases in order to discern whether they have a similarity or a contradiction. Looking for a possible connection between knowledge bytes is the core for the system's ability to infer across knowledge bases. Using WordNet (Miller 1995), we extract synonyms and antonyms for the relations used in knowledge bytes and compare the knowledge bytes to form relations between them.

Time points and confidence measures are also taken into account to analyze relationships between knowledge bytes. By default, we decided to assign a 0.8 weight to confidence and 0.2 weight to time, in order to make the impact of confidence stronger than time. In some cases, for some pairings of knowledge bytes, there is the ability to specify custom values for the weights for confidence and time, in cases where one may require different level of impact of the difference in time or confidence of the two knowledge bytes on the possibility that the two knowledge bytes are indeed related. Let the time and confidence values for a knowledge byte be denoted by T_x and C_x respectively. We assume that N is the final time point of the script.

A relational factor θ' is calculated based on the lexical comparison of two knowledge bytes, which is a measure of how related they are. θ' varies from -1 to +1, with -1 denoting contradictory relation, a +1 denoting similarity between the knowledge bytes, and a θ' closer to 0 representing that the knowledge bytes may be unrelated. The algorithm to calculate the relational factor is shown in Algorithm 1. Once we find that the two knowledge bytes are related, then we compare their links, to obtain the following types of relationships:

```
Input: A pair of Knowledge Bytes \beta_1^* and \beta_2^*;
Output: Relational Factor \theta';
if \Pi_1 = \Pi_2 or (r_1 \simeq r_2, s_1 = s_2, o_1 = o_2) then
\mid \theta \leftarrow 1;
else if r_1 and r_2 are antonyms, s_1 = s_2, o_1 = o_2 then
\mid \theta \leftarrow -1;
else if r_1 \simeq r_2, o_1 = o_2, s_1 \neq s_2 then
\mid \theta \leftarrow -0.5;
else if r_1 \simeq r_2, s_1 = s_2, o_1 \neq o_2 then
\mid \theta \leftarrow 0.5;
else
\mid \theta \leftarrow 0, \text{ which implies they are likely to be unrelated}
\theta' = \frac{|C_2 - C_1|C_W + \frac{|T_2 - T_1|}{T_N}T_W}{\theta}
return \theta'
Algorithm 1: Relational factor calculation
```

- If β₁* and β₂* are both desires and 0 < θ' < 1, we consider
 this to be a similarity in desires between two character
 knowledge bases Υ₁ and Υ₂.
- If β_1^* and β_2^* are both desires and $-1 < \theta' < 0$, we consider this a contradiction or competition in desires between two character knowledge bases Υ_1 and Υ_2 .
- If β_1^* and β_2^* are both beliefs and $0 < \theta' < 1$, we consider this to be a similarity in beliefs between two character knowledge bases Υ_1 and Υ_2 .
- If β_1^* and β_2^* are both beliefs and $-1 < \theta' < 0$, we consider this to be a contradiction in beliefs between two character knowledge bases Υ_1 and Υ_2 .
- If β_1^* and β_2^* are a belief and a desire and $\theta' \neq 0$, we consider this to be a misconception between two character knowledge bases Υ_1 and Υ_2 .

In our example, for the query κ_2 in Section 6.1, a resultant response would seeing contradictions in desires regarding the ownership of the crown among Flynn and Patchy. Equations 11 through 14 show the resultant knowledge bytes β_2^* and β_3^* , and Equation 15 shows the relationship R(2,3)

flagged as "contradictory desires".

$$\beta_2^* = \langle t_2, l_2, \Pi_2^* \rangle$$
 from Flynn's Υ as desire (11)
 $\beta_3^* = \langle t_3, l_3, \Pi_3^* \rangle$ from Patchy's Υ as desire (12)
 $\Pi_2^* = \langle flynn, own, crown, false, \varnothing, \varnothing, \varnothing, \varnothing \rangle$ (13)
 $\Pi_3^* = \langle patchy, own, crown, false, \varnothing, \varnothing, \varnothing, \varnothing, \varnothing \rangle$ (14)
 $R(2,3)$ = "contradictory desires" (15)

7 Application

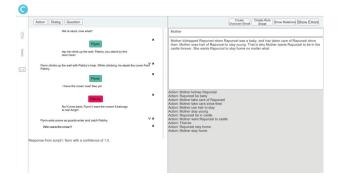


Figure 3: A developed prototype. Left Window: Script. The user asks "Who owns the crown?", and the response is "Flynn with a confidence of 1.0". Top Right: User enters information about the mother to her character base. Bottom Right: The parsed information.

Our proposed framework contributes to emerging technology for assisting story-writers real-time. A scriptwriter would start by adding the story world knowledge to the system, which defines the rules for the story world. The scriptwriter would then proceed to write his story, using the system as a guide for reference. Additionally, the system ensures that the narrative is within the bounds defined by the story world. The automatic extraction and construction of the knowledge bases enable the possibility to interact with the story and characters during story creation.

We performed internal experiments with our system with scripts of moderate complexity. An excerpt of an example script has been shown in Figure 2, and various examples of different features have been discussed throughout this paper. Additionally, an image of our prototype system developed is shown in Figure 3. Our system is able to autonomously construct the various knowledge bases from the script and allows authors to specify the rules for the world in which the story occurs. The character knowledge bases have information specific to the character, and this encapsulation of information can be clearly observed in the different responses. Referring to the queries in Figure 2, the question "Who owns the crown?" is directed to the story in question 1 but to the character of Patchy in question 2. A difference in the responses can be observed, due to Patchy's character knowledge base having information known to his character. We were also able to perform cross-knowledge base reasoning, where we asked the systems various questions and received expected responses. For example, in the script shown,

questions 3, 8, and 9 are questions which involve looking for relations across multiple knowledge bytes, which can be seen in the results obtained.

8 Conclusion

In proposing a new structure for representation and reasoning in narratives with the help of knowledge bytes, knowledge bases, rules and errors definition, we set up ground for various directions in natural language understanding and linguistic reasoning, assisted story writing, and computational narratives.

Our application demonstrated that the introduced concepts work well for an example story, but we are aware that the implementation still needs some improvements before becoming a solution ready for deployment. The reasoning capabilities of our system are highly dependent on the capabilities of the mechanism analyzing rules to be able to completely reason about the story world knowledge. While this can be quite specific for story worlds like *Star Wars* or *Toy Story*, there is still a need for adding world knowledge rules. Automatic rule generation is an avenue that needs to be explored and a challenge that needs to be solved to make the creation of story world knowledge less cumbersome.

The future goals include analyzing more complicated (real) movie scripts and creating advanced rules and error definitions. We have implemented the theoretical concepts proposed in this work and are building towards a large-scale deployment of the tool and intend to perform detailed user studies in the future.

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