

Scheherazade: Crowd-Powered Interactive Narrative Generation

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Abstract

Interactive narrative is a form of storytelling in which users affect a dramatic storyline through actions by assuming the role of characters in a virtual world. This extended abstract outlines the SCHEHERAZADE-IF system, which uses crowdsourcing and artificial intelligence to automatically construct text-based interactive narrative experiences.

Introduction

Interactive narrative is a form of storytelling in which users affect a dramatic storyline through actions by assuming the role of characters in a virtual world. The simplest interactive narratives are Choose-Your-Own-Adventure books and hypertexts in which each plot point has branching options. More complex systems use artificial intelligence to determine available options for the user.

Prior AI-driven interactive narrative systems (see Riedl and Bulitko (2013) for a survey) rely on an *a priori* known domain model—a description of a fictional world, including characters, objects, places, and the actions that entities can perform to change the world. Once the domain model has been engineered, a story generation system can tell a potentially infinite number of stories involving these characters, places, and actions known to the system.

SCHEHERAZADE (Li et al. 2012; 2013) uses crowdsourcing to automatically learn the domain knowledge needed to construct construct and understanding stories about everyday activities such as going to a restaurant or going to a movie theater. The SCHEHERAZADE-IF system (Li, Lee-Urban, and Riedl 2012) creates interactive narrative experiences by allowing a human user to assume the role of one of the characters in the domain.

Scheherazade-IF

The SCHEHERAZADE-IF system attempts to create a novel, playable interactive fictions about simple, user-provided scenarios. For example, a human designer may request an interactive experience about a “bank robbery”. If the system

does not have a model of the domain, the system uses crowdsourcing to rapidly acquire a number of linear narrative examples about typical ways in which the topic might occur. In other words, the system collects human experiences and learns a generalized model—a *plot graph*—about the topic domain. Figure 1 shows the system architecture.

A plot graph is a representation in story generation systems that models the author-intended logical flow of events in the virtual world as a set of precedence constraints between plot events (Weyhrauch 1997; Nelson and Mateas 2005). In our work, a plot graph is a tuple $G = \langle E, P, M \rangle$ where E is the set of plot events, P is a set of temporal ordering constraints between events, and M is a set of unordered mutual exclusion relations that indicate which events can never co-occur in the same narrative experience. Mutual exclusion relations indicate *branches* where alternative narratives can unfold. See Figure 3 for a plot graph model of a bank robbery.

By learning plot graphs from the collected examples, our system can generate an interactive narrative about any topic for which a crowd of people can generally agree on the main events that should occur and the sequencing of the events.

Plot Graph Learning

The plot graph learning process is described in detail in Li et al. (2012) and Li et al. (2013) and summarized below.

The process begins with a user request for an interactive narrative on a particular topic. The system generates a query to Amazon’s Mechanical Turk to solicit example narratives of the given topic, provided in natural language. To simplify the complexity of natural language processing, crowd workers are asked to segment their narratives such that each sentence contains one event. Crowd workers are instructed to use one verb per sentence and to avoid complexities such as conditionals, compound sentences, and pronouns.

Second, the system analyzes the simplified natural language narrative examples to discover the fundamental plot points on which people agree. Sentences from different narrative examples are clustered together according to semantic similarity to create plot events. The simplified language use allows clustering algorithms to discover plot events with relatively high accuracy.

Third, we identify the precedence constraints between plot events. Crowd workers produce noisy and sometimes er-

