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Networks and Dragons: A data-driven approach to procedural dungeon generation

Angel Simeonov

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Supervisor: Dr. Rob Alexander

<insert dedication>

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Executive Summary

>At most two (2) pages, aimed a non-specialist, knowledgeable authorial peer. The summary must: –state the aim of the reported work, –motivate the work, –state methods used, –state results found, and –highlight any legal, social, ethical, professional, and commercial issues as appropriate to the topic of study (if none, then this should be explicitly stated).

1 Introduction

1.1 Role-playing games

- Nature and goal RP is trying to create a compelling narrative in which players can be involved in.
 - Contextualise RPs by comparing different genres. Digital vs Tabletop.
- Interaction mechanisms. How the game is played and what is the role of the DM
- Problems: DM has to create the narrative. Can we devise an algorithm that helps DMs create a compelling narrative?

Role-playing game (RPG) is a broad term encompassing a multitude of different games with often distinct mechanics and platforms of interaction (the media). The common factor between all RPGs is that the player(s) portray a fictional character and is involved in a fictional world (or a subset of one). The interaction of the players with this world is governed by rules, defined by the media. To better understand the structure of RPGs, we can view it in terms of two sets:

1. Rules defining how we play the game. Describe the allowed actions for the player at any given moment in the game.
2. Narrative elements. Provide the *purpose* of the players to interact with the fictional world.

The first set can be viewed as "How I interact with the environment" (functional) and the second as "What is the meaning of the environment" (narrative).

Computer RPGs like the Action RPG (ARPG) Legend of Zelda, Diablo, Fallout and many alike have both sets of rules defined by the game designers. A functional rule in an RPG like Skyrim is that you can attack with the left mouse button and a story element is that you are a Dragonborn with a quest. This quest is defined by the writers and designers of the game and

as such exploring the narrative in a digital RPG can be comparable to an interactive reading of a fiction book. Good computer RPGs often have the ability to relax the narrative [1], allowing for the player to have more freedom in the exploration of the world and as such create the sensation that the player has some impact on this pre-programmed fictional environment.

The ability for a player to influence the narrative is one of the defining features of tabletop RPGs (TRPG a.k.a pen-and-paper PnP) like Dungeons and Dragons. In them, the functional rules are usually defined by a rulebook (like the Player's Handbook) and players verbally describe their interactions with the environment. The narrative is an ever evolving amalgam between the input of players and the Dungeon Master (DM). The DM's task is to create a narrative outline and guide the player interaction. Because of the verbal nature of the game, the narrative does not suffer the limitations of its digital counterparts. But because there are no hard constraints to how the narrative is told, the DM has the non-trivial task of introducing consistency and outlining a structure for the story that would result in a compelling and ideally immersive experience for the players. This is often achieved by focusing the adventure's act on a particular and detailed location. These locations are often referred to as Dungeons and in practice can be anything from the villain's mansion or a beast's cave to a city under siege. A good dungeon design is crucial for creating a compelling narrative. The creative task of making the Dungeon is a laborious process. To facilitate that, academics and the PnP community have been exploring different ways of automating Dungeon creation. Arguably the greatest problem posed by automating Dungeon creation is answering the question of "What is a compelling Dungeon?". In the next section we will review different generative methods for Dungeons and their associated limitations in an attempt to answer this question.

1.2 State of the Art Generative methods

- Review various different algorithms for PDG and discuss their take on solving the "is this dungeon interesting?" problem.
 - Non-digital: [[the Advanced DnD DM design kit book 3: Adventure Cookbook]]. Using dice tables to create different elements of the dungeon. Laborious process which requires the DM to remove any inconsistencies. Does not provide an actual Dungeon structure.
 - Cellular automata: donjon [2]. Issues associated with completely random topology and random content generation. Evaluated

- only on if the level is solvable (no evaluation of semantic content).
- Constraint Propagators and their related issues when formalising narrative as constraints.
- BPTs + semantic additions[[citation needed Thrall and Brown]]. Difficulty proving the 'goodness' of the generator. Discuss difficulties with an objective HCI evaluation for PnP (i.e. difficult to get a decent sample size, because playing a game is time consuming).
- Formal language approach. Grammars [3]–[5] as a natural way of describing narrative structures.
- Data-driven approaches. [6]–[8] Relate the notion of learning from human-made dungeons as a way to create good structure + narrative. Note the lack of data. Expand more in the Mission and Space → Aim sections

1.2.1 Non-digital Generators

The designers of the classic PnP modules acknowledge the issue of Dungeon creation and often incorporate *Loot* or *Encounter* tables in the rule-books. The tables are a reference over a set of treasures or monsters respectively, which can be chosen by rolling the specified die **need figure of loot/encounter table**. As noted before, a PnP RPG can evolve rapidly outside the planned narrative structure prepared by the DM and a simple way to quickly define new adaptive story elements via a loot table can be useful. Some early modules even provided a step-by-step guide for creating a full campaign from the plot, villains' obsessions and story setting to the various encounters and and treasures [9] entirely based on randomised content tables. An issue arises when using loot tables when a randomly selected element from one table is contradicted by an item selected from another. **show an example from the ADND module**. It is up to the DM to resolve such disparities. As can be seen, this process although providing some degree of creative assistance, it does little to facilitate the laborious nature of creating elements for a compelling narrative. Furthermore, these methods pay no attention to providing guidance of what would mean a good Dungeon as they are occupied with solving global narrative questions. **Argue that DMs know what they want to do, they just need a facilitator that will create a dungeon based on their requirements ? citation needed for what people use dungeon generators for.**

1.2.2 Digital Generators

It is important to note that PnP RPGs and the various genres of computer RPGs although differing in the mechanisms of interaction, they share the same narrative goal [10]. Therefore we will not limit ourselves to looking only at existing tabletop solutions.

Cellular Automata

Constraint Propagators

Binary Partitioning

Formal Languages

Data-driven approaches

1.3 Mission and Space

- Introduce the notion of Mission and Space
- The reason for separating Mission and Space. We can model player experience better [3], [6].

=== Discussion of Mission::Space separation ===

1.4 Aim

- The only "*dungeon generator*" that has been empirically proven to have the ability of creating a truly compelling narrative is the human designer
- Discuss attempts at the data-driven approaches from **Generation Methods** in detail
 - Deery's data inspired approach, but not data-driven
 - Summerville's Learning of Zelda

The dungeon's topology (Space) and contents (Mission) are correlated and form the narrative. As we have seen, dungeon generators usually im-

explored bottom up approaches in which they apply rules for topology and then introduce dungeon content. The objective goal of all dungeon generators is if not to provide a full end-to-end narrative, then at least set the baseline for a captivating one. I want to argue that from that definition we can say that the ultimate *dungeon generator* is the human designer as dungeons and therefore narratives produced by humans are the most compelling out of all created dungeons. Deery made the first steps by manually analysing submissions to the One Page Dungeon OPDC competition [[OPDC]] to extract a graph grammar for Mission generation [4]. The Mission graph was then mapped to a physical Space in a 1:1 ratio. The result was that each room was limited to a single Mission element. It is trivial to see that the originals in OPDC do not impose such a restriction. One room can have multiple Mission elements (e.g. the key to unlocking the door is on the bandit's waist, Key + Encounter). Furthermore, Deery's approach was to manually look at 10 competition winners and heuristically extract the grammar rules, which he highlights that they do not capture all the possible patterns. An automated approach to learning would potentially solve that issue. Programmatically learning a level from data has been an object of interest for computer based RPGs [6], but has not been applied to PnP RPGs, presumably because of the lack of a consistent dataset. In this paper we will investigate if we can use the OPDC dataset and apply a data-driven approach of sampling an Inference (Bayesian) Network to create the layout of a small (one session long) dungeon.

We will first explore the availability of data and the selections of parameters we want to learn. We will discuss the choice of using inference networks for our generator. Then we will do a comparative analysis of different graph structures and algorithms for fitting our parameters. The comparison will be based on our internal validation Scoring Rules [11] assessing which model has the greatest statistical predictive capabilities. We will conclude by externally validating our best model's capability to create human-like topologies with a user study.

2 Conclusion

A Some appendix

Use this section for graphical showing of the models. Nets, result tables (or tables should be inline?)

B Another appendix

Use this section for questionnaires and external validation support

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