PERCOLATION AND THE FIREFIGHTER PROBLEM

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1. Introduction

The Firefighter Problem was first discussed by Hartnell in 1995 [5] and models an outbreak of fire with a firefighter strategically blocking its path. We will discuss this original formulation of the problem and explain how it can be used in applications from the spread of disease to the spread of viral content on social media.

The other topic we will address is Percolation Theory, and how it proves very promising in answering questions about Firefighter. We will outline the original conception and the context it was considered and then explain how a slight variant will prove far more useful for our purposes in considering the Firefighter Problem.

2. Background

2.1. Firefighter. The Firefighter Problem, which we refer to as simply 'Firefighter,' is the focus and motivation of our discussion; we, in particular, hope to show that Percolation can provide many answers to questions asked about Firefighter. We first formalise Firefighter as follows: at t = 0, fire breaks out at some vertex v_0 of graph G. The firefighter then 'protects' another vertex¹ of G. A protected vertex is protected for the remainder of the game; it is inflammable ad infinitum. Similarly, a vertex that has been on fire is 'burnt' for the rest of the game and cannot ignite again. The fire spreads to any immediate neighbouring vertices that are neither protected nor burnt. Then, the firefighter may protect another vertex, the fire spreads again and so on.² Common questions to ask include: how do we minimise the number of vertices that will be burnt? In a given class of trees, what is the average number of burnt vertices? Is the problem NP-complete?

There are many natural contextualisations of Firefighter - for instance, let each vertex be an individual and let the edges between them represent social contact. This is a simple model for disease infection. If instead we think of these edges representing virtual contact between individuals on social media, we have a model for the spread of viral internet memes [8].

¹ This has been extended to n vertices in some research, but here we wish to illustrate for simplicity the original form of the problem.

² It may be tempting to put the firefighting move and the fire spread into two distinct time steps, t and t+1 - this is beneficial in some applications (such as two player versions of Firefighter) but in general will make computation more time consuming.

2.2. **Percolation Theory.** Widely known and used in physics, statistics and mathematics, Percolation theory involves modelling scenarios as n-dimensional graphs, so application to Firefighter is not entirely unexpected. The edges between vertices in the graph can be either 'open' or 'closed' with probability p and 1-p respectively. We can think of percolation problems as liquid being poured onto a porous material and whether there is a path from hole to hole along open paths through the material. Note that removing more and more edges mov†es us towards a critical point at which removing further edges would cause the graph to fall apart into smaller clusters of vertices and edges that have no access to each other [3]. This is known as 'bond' percolation, as edges correspond to bonds in many of its applications.

Several authors have suggested percolation as a possible approach to Firefighter [2]. In this context, we could determine the critical point to see how we might contain the fire to a smaller cluster that cannot spread to the wider graph. For Firefighter, site percolation is more applicable: rather than considering open or closed *edges* ('bonds') between vertices as in bond percolation, we consider each vertex ('site') as being 'occupied' or 'unoccupied' with probability p and 1-p respectively.

Formally, we consider a point lattice \mathbb{L} and denote the open cluster as C(x), where $x \in \mathbb{L}$ is the local origin of the cluster. This cluster C(x) is defined as the set of all vertices that can be reached from open paths beginning at the nucleation site, x. Then, we are particularly interested in the *percolation probability*:

$$\theta(p) = \mathbb{P}_p(|C(0)| = \infty),$$

and the critical probability (or percolation threshold):

$$p_c = \sup\{ p \mid \theta(p) = 0 \}.$$

Here, \mathbb{P}_p is the product measure given by:

$$\mathbb{P}_p = \prod_{v \in \mathbb{L}^d} \mu_v$$

where μ_v is the *Bernoulli measure*, which returns p when v is open and 1-p when v is closed [7, p. 28]. Analytically, others have shown that in the case of a two-dimensional regular point lattice, the critical probability is $p_c = 1/2$ [6].

3. Potential Uses of Percolation in The Firefighter Problem

We have identified two main avenues that may be pursued in Firefighter using Percolation: the firefighter may use percolation in order to defend the graph, or the fire spreads with percolation probability p. The former may be more useful when the firefighter can save more than one vertex at each turn; the latter may be more useful when modelling disease spread where the reproduction rate is introduced as the percolation probability. For instance, we can model how a fire spreads given a certain propagation probability or consider forest fires and how fire can spread in densely in regularly populated forests and forests with some degree of percolation to remove some of the population, to more accurately represent the less regular tree density of real forests. We could extend this latter thinking to our earlier contextualisations, where we wish to consider vertices as

individuals and edges as the connections between them. In that case, percolation may give us a more useful model for disease spread when we do not assume the population is well mixed and instead introduce probability functions to correspond to the likelihood one vertex is connected to another.

3.1. **Better than random.** One potential use for the firefighter using percolation as a method of defence would be as a baseline test: in most scenarios, a method for obtaining defence strategies should be at least as effective as a random defence sequence. We could find such a random sequence using percolation for comparison purposes. Consider a sequence of vertices in graph G, written as d_1, d_2, \ldots, d_t . An optimal defence sequence could be found using integer programming as provided by Finbow and MacGillivray [2]:

Maximise
$$\sum_{v \in V(G)} d_v w(v) \qquad \text{for each level } i$$
 subject to
$$d_v + \sum_{\text{level}(v)=i} d_v \leq 1 \text{ for each level } i$$

$$d_v + \sum_{u \succ v} d_u \leq 1 \qquad \text{for every outer vertex } v \text{ of } T,$$

$$d_v \in \{0,1\}.$$

where $u \succ v$ indicates that u is an ancestor of v. The optimal strategies provided for different classes and densities of graphs here will provide an upper bound (which may indeed be impossible to attain in some cases) for success of a given strategy. We can find a lower bound using percolation, and so we have a range of success values as a starting point: if some strategy is better than random percolation, then it is worth considering, but below the particular expected optimal solution from integer programming and we can improve or find a better strategy.

We conjecture that, at the lowest graph densities, the random strategy will be very close to the optimal strategy and thus finding an improvement is at once difficult and lacking in great utility. At the very highest graph densities, while random strategies will have a very low expected best-case scenario, as will most strategies as the constraint on the firefighter that they have only one vertex to save per turn does not go as far when vertices are much better connected. Thus, finding improvements on random will again not prove very useful as random and best-case scenarios serve to show us that there is a small range of possible, not good outcomes.

- 3.2. Reproduction rate. In contrast to the previous section, we now focus our attention on the fire spread being determined by percolation (rather than the firefighter's defence sequence). Diseases, when there is a large enough sample size, have a basic reproduction rate associated with them, denoted R_0 : for instance, measles has a basic reproduction rate $12 \le R_0 \le 18$ [4] and the Influenza strain responsible for the 1918 pandemic has a basic reproduction rate of $1.4 \le R_0 \le 2.8$ [1]. These values correspond to a basic expectation
- 3.3. Irregular population density. A great deal of the literature surrounding Fire-fighter assumes a regular graph that is, in the context of disease we assume a well-mixed

population where everyone has equal probability of coming into contact with their neighbours. Of course, this is a large simplification of the reality: some individuals are very well connected and have lots of contact with others, other people have significantly less contact with others. Similarly, in the context of a forest fire, the density of a forest is irregular and there is a probability that fire can spread between two trees, depending on their proximity (among other factors). Hence, percolation on regular grids to more closely resemble populations or forest density could lead to far more useful and realistic modelling results.

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