

First-Passage Percolation on a Width-2 Strip and the Path Cost in a VCG Auction

Abraham D. Flaxman, Microsoft Research
David Gamarnik, MIT
Gregory B. Sorkin, IBM Research

May 29, 2007

Outline

- 1 Introduction
 - What the title means
 - Width-2 strip
 - First-Passage Percolation
 - Path Cost in a VCG Auction
 - Fixed graphs with random edge weights
 - Minimum Spanning Tree
 - Minimum Perfect Matching
- 2 The width-2 strip
 - First-passage percolation
 - Path cost in a VCG auction

Width-2 Strip

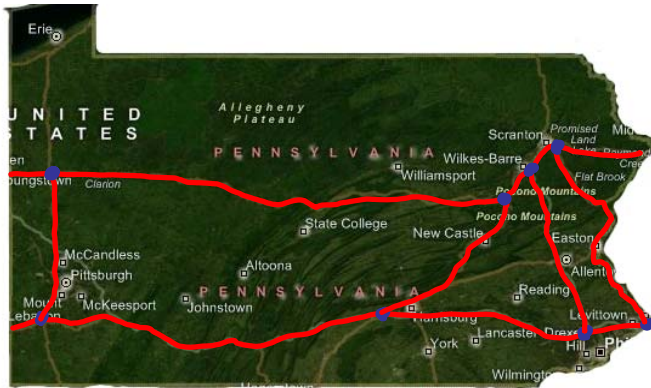


- The *infinite width-2 strip*:
 - Vertex set is $\{0, 1\} \times \mathbb{Z}$
 - edges join vertices at ℓ_1 distance 1
- The *n-long strip* is the (finite) subgraph induced by $\{0, 1\} \times \{0, \dots, n\}$.

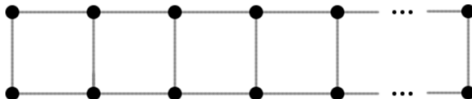
Width-2 Strip



Width-2 Strip



First-Passage Percolation



First-Passage Percolation:

First-Passage Percolation



First-Passage Percolation:

- Models the time it takes a fluid to spread through a random medium.

First-Passage Percolation



First-Passage Percolation:

- Models the time it takes a fluid to spread through a random medium.
- Each edge of graph has a i.i.d. random weight, find shortest edge-weighted (s, t) -path.

First-Passage Percolation



First-Passage Percolation:

- Models the time it takes a fluid to spread through a random medium.
- Each edge of graph has a i.i.d. random weight, find shortest edge-weighted (s, t) -path.
- The *time constant* is the limiting ratio of this length to the unweighted shortest path length n , as n tends to infinity.

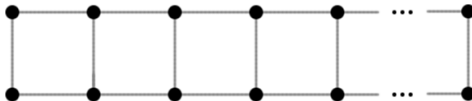
First-Passage Percolation



First-Passage Percolation:

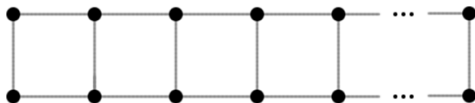
- Models the time it takes a fluid to spread through a random medium.
- Each edge of graph has a i.i.d. random weight, find shortest edge-weighted (s, t) -path.
- The *time constant* is the limiting ratio of this length to the unweighted shortest path length n , as n tends to infinity.
- Introduced in [Broadbent and Hammersley \(1957\)](#) and [Hammersley and Welsh \(1965\)](#).

Path Cost in a VCG Auction



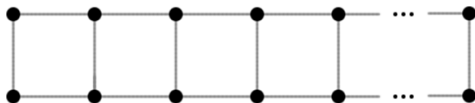
- VCG mechanism for buying an (s, t) -path:

Path Cost in a VCG Auction



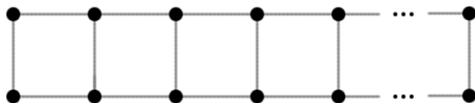
- VCG mechanism for buying an (s, t) -path:
 - Utility-maximizing agents each control an edge, e , of a graph, and can transmit a message at cost c_e .

Path Cost in a VCG Auction



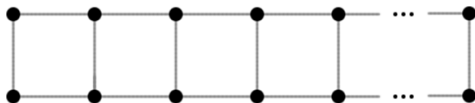
- VCG mechanism for buying an (s, t) -path:
 - Utility-maximizing agents each control an edge, e , of a graph, and can transmit a message at cost c_e .
 - Auctioneer finds a cheapest path, pays each edge-agent difference in cost of a cheapest path avoiding edge and cost of a cheapest path if edge cost were 0.

Path Cost in a VCG Auction



- VCG mechanism for buying an (s, t) -path:
 - Utility-maximizing agents each control an edge, e , of a graph, and can transmit a message at cost c_e .
 - Auctioneer finds a cheapest path, pays each edge-agent difference in cost of a cheapest path avoiding edge and cost of a cheapest path if edge cost were 0.
- First applied to the shortest-path problem explicitly by **Nisan and Ronen (1999)**.

Path Cost in a VCG Auction



- VCG mechanism for buying an (s, t) -path:
 - Utility-maximizing agents each control an edge, e , of a graph, and can transmit a message at cost c_e .
 - Auctioneer finds a cheapest path, pays each edge-agent difference in cost of a cheapest path avoiding edge and cost of a cheapest path if edge cost were 0.
- First applied to the shortest-path problem explicitly by [Nisan and Ronen \(1999\)](#).
- May require paying much more than the cost of the shortest path (more to say: [Archer and Tardos \(2002\)](#)).

Fixed graph with random edges weights

Today:

First passage percolation and path cost of VCG auction in the width-2 strip as specific examples of fixed graph with random edge weights.

Fixed graph with random edges weights, Ex 1: MST

Notable example of **fixed graph with random edge weights**:

- Complete graph K_n with edge weights independent, uniform in $[0, 1]$

Fixed graph with random edges weights, Ex 1: MST

Notable example of **fixed graph with random edge weights**:

- Complete graph K_n with edge weights independent, uniform in $[0, 1]$
 - Cost of **minimum spanning tree** in this network, as $n \rightarrow \infty$, cost \rightarrow

Fixed graph with random edges weights, Ex 1: MST

Notable example of **fixed graph with random edge weights**:

- Complete graph K_n with edge weights independent, uniform in $[0, 1]$
 - Cost of **minimum spanning tree** in this network, as $n \rightarrow \infty$,
cost $\rightarrow \zeta(3) = \frac{1}{1^3} + \frac{1}{2^3} + \frac{1}{3^3} + \dots$

Fixed graph with random edges weights, Ex 1: MST

Notable example of **fixed graph with random edge weights**:

- Complete graph K_n with edge weights independent, uniform in $[0, 1]$
 - Cost of **minimum spanning tree** in this network, as $n \rightarrow \infty$,
cost $\rightarrow \zeta(3) = \frac{1}{1^3} + \frac{1}{2^3} + \frac{1}{3^3} + \dots$
- Proof by studying a greedy algorithm for constructing MST
[Frieze (1985)]

Fixed graph with random edges weights, Ex 2: MM

Another notable example of fixed graph with random edge weights:

- Complete bipartite graph $K_{n,n}$, edges weights independent, uniform in $[0, 1]$

Fixed graph with random edges weights, Ex 2: MM

Another notable example of fixed graph with random edge weights:

- Complete bipartite graph $K_{n,n}$, edges weights independent, uniform in $[0, 1]$
 - Cost of minimum weight perfect matching in this network, as $n \rightarrow \infty$, cost \rightarrow

Fixed graph with random edges weights, Ex 2: MM

Another notable example of fixed graph with random edge weights:

- Complete bipartite graph $K_{n,n}$, edges weights independent, uniform in $[0, 1]$
 - Cost of minimum weight perfect matching in this network, as $n \rightarrow \infty$, cost $\rightarrow \zeta(2) = \frac{1}{1^2} + \frac{1}{2^2} + \frac{1}{3^2} + \dots = \frac{\pi^2}{6}$

Fixed graph with random edges weights, Ex 2: MM

Another notable example of fixed graph with random edge weights:

- Complete bipartite graph $K_{n,n}$, edges weights independent, uniform in $[0, 1]$
 - Cost of minimum weight perfect matching in this network, as $n \rightarrow \infty$, cost $\rightarrow \zeta(2) = \frac{1}{1^2} + \frac{1}{2^2} + \frac{1}{3^2} + \dots = \frac{\pi^2}{6}$
- Calculated non-rigorously via statistical physics [Mézard and Parisi (1987)]

Fixed graph with random edges weights, Ex 2: MM

Another notable example of fixed graph with random edge weights:

- Complete bipartite graph $K_{n,n}$, edges weights independent, uniform in $[0, 1]$
 - Cost of minimum weight perfect matching in this network, as $n \rightarrow \infty$, cost $\rightarrow \zeta(2) = \frac{1}{1^2} + \frac{1}{2^2} + \frac{1}{3^2} + \dots = \frac{\pi^2}{6}$
- Calculated non-rigorously via statistical physics [Mézard and Parisi (1987)]
- Rigorous proof limit exists [Aldous (1992)]

Fixed graph with random edges weights, Ex 2: MM

Another notable example of fixed graph with random edge weights:

- Complete bipartite graph $K_{n,n}$, edges weights independent, uniform in $[0, 1]$
 - Cost of minimum weight perfect matching in this network, as $n \rightarrow \infty$, $\text{cost} \rightarrow \zeta(2) = \frac{1}{1^2} + \frac{1}{2^2} + \frac{1}{3^2} + \dots = \frac{\pi^2}{6}$
- Calculated non-rigorously via statistical physics [Mézard and Parisi (1987)]
- Rigorous proof limit exists [Aldous (1992)]
- Rigorous proof of $\zeta(2)$ (not by analyzing known algorithm) [Aldous (2001)]

Ingredients for analyzing minimum perfect matching

Proof ingredients:

Ingredients for analyzing minimum perfect matching

Proof ingredients:

- **An infinite object**; fixed graph with random weights should converge to it; in this case, **Poisson Infinite Weighted Tree (PWIT)**

Ingredients for analyzing minimum perfect matching

Proof ingredients:

- **An infinite object**; fixed graph with random weights should converge to it; in this case, **Poisson Infinite Weighted Tree (PWIT)**
- A **Recursive Distributional Equation (RDE)** for a carefully chosen random variable of interest.

Ingredients for analyzing minimum perfect matching

Proof ingredients:

- **An infinite object**; fixed graph with random weights should converge to it; in this case, **Poisson Infinite Weighted Tree (PWIT)**
- A **Recursive Distributional Equation (RDE)** for a carefully chosen random variable of interest.
- A **proof** that the solution to the RDE on infinite object has something to do with the expectation for the finite object.

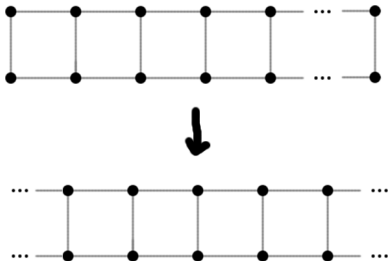
This present paper

Consider the present paper a simple example of that approach.

This present paper

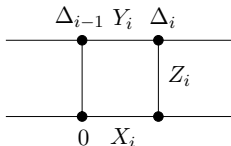
Consider the present paper a simple example of that approach.

- Infinite analog of n -long width-2 strip is the infinite width-2 strip



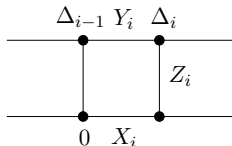
First passage percolation in the width-2 strip

- Recursive distributional equations



First passage percolation in the width-2 strip

- Recursive distributional equations

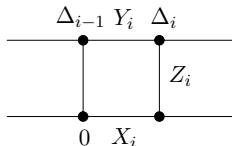


$$\ell(0, i) = \min \{ \ell(0, i-1) + X_i, \ell(1, i-1) + Y_i + Z_i \}$$

$$\ell(1, i) = \min \{ \ell(1, i-1) + Y_i, \ell(0, i-1) + X_i + Z_i \}$$

First passage percolation in the width-2 strip

- Recursive distributional equations



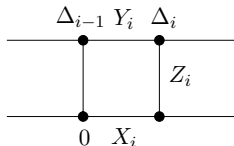
$$\ell(0, i) = \min \{ \ell(0, i-1) + X_i, \ell(1, i-1) + Y_i + Z_i \}$$

$$\ell(1, i) = \min \{ \ell(1, i-1) + Y_i, \ell(0, i-1) + X_i + Z_i \}$$

(not such a useful RDE)

First passage percolation in the width-2 strip

- Recursive distributional equations



$$\ell(0, i) = \min \{ \ell(0, i-1) + X_i, \ell(1, i-1) + Y_i + Z_i \}$$

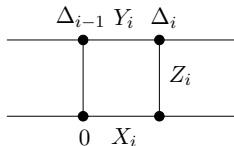
$$\ell(1, i) = \min \{ \ell(1, i-1) + Y_i, \ell(0, i-1) + X_i + Z_i \}$$

(not such a useful RDE)

Better to consider $\Delta_j = \ell(1, j) - \ell(0, j)$.

First passage percolation in the width-2 strip

Recursive distributional equation for $\Delta_i = \ell(1, i) - \ell(0, i)$.



$$\Delta_i = \begin{cases} -Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i < -Z_i; \\ \Delta_{i-1} + X_i - Y_i, & \text{if } \Delta_{i-1} + X_i - Y_i \in [-Z_i, Z_i]; \\ Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i > Z_i. \end{cases}$$

Proof that Δ has something to do with $\mathbb{E}[\ell_n]$

$$\Delta_i = \begin{cases} -Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i < -Z_i; \\ \Delta_{i-1} + X_i - Y_i, & \text{if } \Delta_{i-1} + X_i - Y_i \in [-Z_i, Z_i]; \\ Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i > Z_i. \end{cases}$$

For a concrete example, suppose $Y_i, X_i, Z_i \sim \text{Be}(p)$. Then

Proof that Δ has something to do with $\mathbb{E}[\ell_n]$

$$\Delta_i = \begin{cases} -Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i < -Z_i; \\ \Delta_{i-1} + X_i - Y_i, & \text{if } \Delta_{i-1} + X_i - Y_i \in [-Z_i, Z_i]; \\ Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i > Z_i. \end{cases}$$

For a concrete example, suppose $Y_i, X_i, Z_i \sim \text{Be}(p)$. Then

- Δ_i is a Markov chain on $\{-1, 0, 1\}$ with a unique stationary distribution.

Proof that Δ has something to do with $\mathbb{E}[\ell_n]$

$$\Delta_i = \begin{cases} -Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i < -Z_i; \\ \Delta_{i-1} + X_i - Y_i, & \text{if } \Delta_{i-1} + X_i - Y_i \in [-Z_i, Z_i]; \\ Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i > Z_i. \end{cases}$$

For a concrete example, suppose $Y_i, X_i, Z_i \sim \text{Be}(p)$. Then

- Δ_i is a Markov chain on $\{-1, 0, 1\}$ with a unique stationary distribution.
- $\gamma_i = \ell(0, i) - \ell(0, i-1)$ is, too.

Proof that Δ has something to do with $\mathbb{E}[\ell_n]$

$$\Delta_i = \begin{cases} -Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i < -Z_i; \\ \Delta_{i-1} + X_i - Y_i, & \text{if } \Delta_{i-1} + X_i - Y_i \in [-Z_i, Z_i]; \\ Z_i, & \text{if } \Delta_{i-1} + X_i - Y_i > Z_i. \end{cases}$$

For a concrete example, suppose $Y_i, X_i, Z_i \sim \text{Be}(p)$. Then

- Δ_i is a Markov chain on $\{-1, 0, 1\}$ with a unique stationary distribution.
- $\gamma_i = \ell(0, i) - \ell(0, i-1)$ is, too.
- $\lim_{n \rightarrow \infty} \frac{\mathbb{E}[\ell(0, n)]}{n} = \lim_{n \rightarrow \infty} \frac{\sum_{i=1}^n \mathbb{E}[\gamma_i]}{n} = \lim_{n \rightarrow \infty} \mathbb{E}[\gamma_n]$.

What you get

If cost of edge = $\begin{cases} 0 & \text{w. pr. } p \\ 1 & \text{w. pr. } 1 - p \end{cases}$ then shortest path from $(0, 0)$ to $(n, 0)$ tends to

$$\left(\frac{p^2(1+p)^2}{3p^2+1} \right)^n.$$

What you get

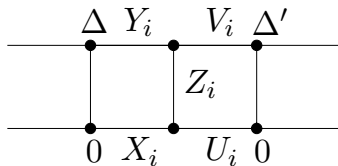
If cost of edge = $\begin{cases} 0 & \text{w. pr. } p \\ 1 & \text{w. pr. } 1 - p \end{cases}$ then shortest path from $(0, 0)$ to $(n, 0)$ tends to

$$\left(\frac{p^2(1+p)^2}{3p^2+1} \right) n.$$

If cost of edge is uniform in $[0, 1]$, then shortest path tends to $\approx (0.42\dots)n$.

Path cost in a VCG auction

Same general approach can find the VCG cost of a path in the width-2 strip:



Results

If cost of edge = $\begin{cases} 0, & \text{w. pr. } p; \\ 1, & \text{w. pr. } 1 - p; \end{cases}$ then get

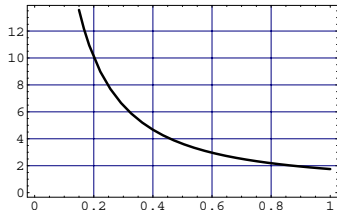
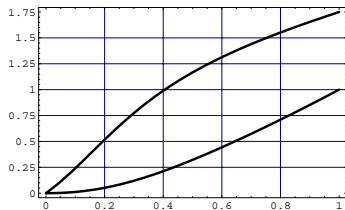


FIGURE 4. *Left:* VCG and usual shortest-path rates.
Right: Ratio of VCG cost to shortest-path cost.

Conclusion

Width-2 strip with random edge weights

- First-passage percolation
- VCG path auction

Extensions:

- Extend directly to Width-3 strip with no backtracking.
- Width- k strip?
- With backtracking?