

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

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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

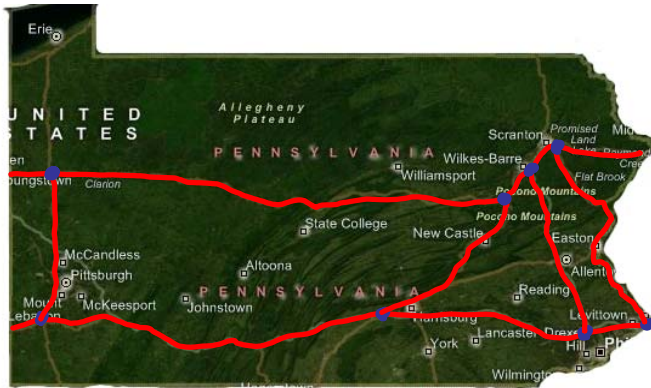


- The *infinite width-2 strip*:
 - Vertex set is $\{0, 1\} \times \mathbb{Z}$
 - edges join vertices at ℓ_1 distance 1
- The *n-long strip* is the (finite) subgraph induced by $\{0, 1\} \times \{0, \dots, n\}$.

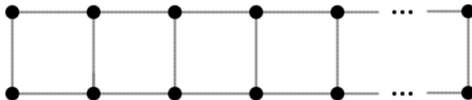
Width-2 Strip



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First-Passage Percolation



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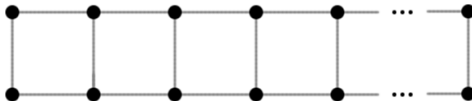
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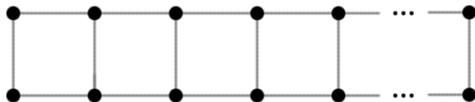
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- Each edge of graph has a i.i.d. random weight, find shortest edge-weighted (s, t) -path.
- The *time constant* is the limiting ratio of this length to the unweighted shortest path length n , as n tends to infinity.
- Introduced in [Broadbent and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

Path Cost in a VCG Auction



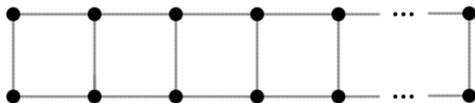
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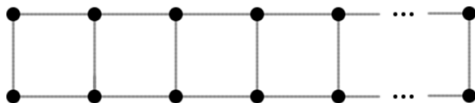
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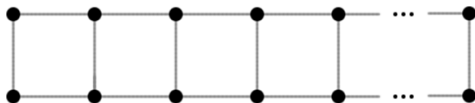
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- First applied to the shortest-path problem explicitly by [Nisan and Ronen \(1999\)](#).
- May require paying much more than the cost of the shortest path (more to say: [Archer and Tardos \(2002\)](#)).

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.

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- Proof by studying a greedy algorithm for constructing MST
[Frieze (1985)]

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- Rigorous proof of $\zeta(2)$ (not by analyzing known algorithm) [Aldous (2001)]

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- A **Recursive Distributional Equation (RDE)** for a carefully chosen random variable of interest.

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- **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.

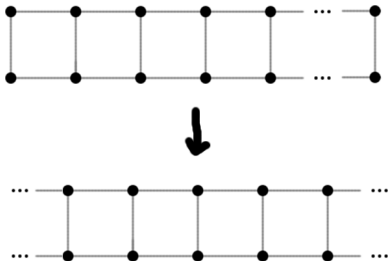
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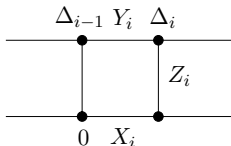
Consider the present paper a simple example of that approach.

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



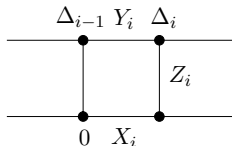
First passage percolation in the width-2 strip

- Recursive distributional equations



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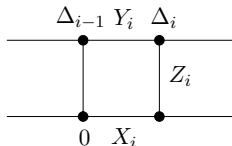


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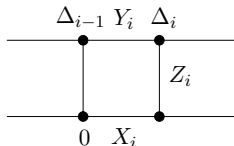
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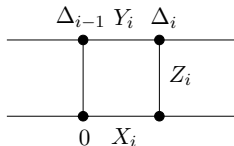
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Better to consider $\Delta_j = \ell(1, j) - \ell(0, j)$.

First passage percolation in the width-2 strip

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



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

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- $\gamma_i = \ell(0, i) - \ell(0, i-1)$ 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

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

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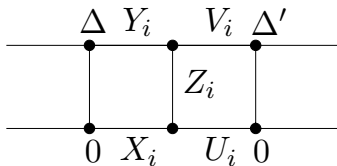
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If cost of edge is uniform in $[0, 1]$, then shortest path tends to $\approx (0.42\dots)n$.

Path cost in a VCG auction

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



Results

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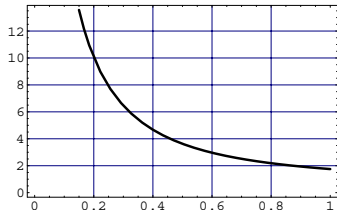
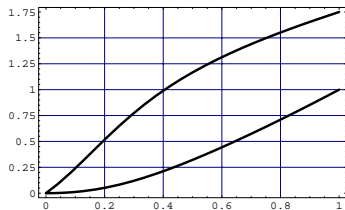


FIGURE 4. *Left:* VCG and usual shortest-path rates.
Right: Ratio of VCG cost to shortest-path cost.

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?