

# Random Walk and Other Lattice Models

Christian Beneš

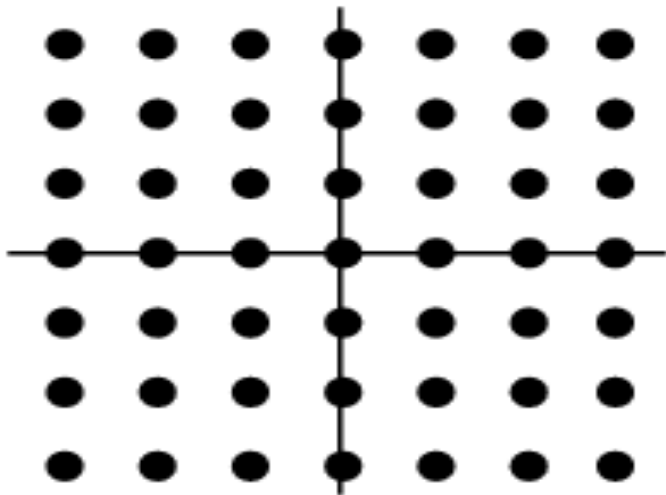
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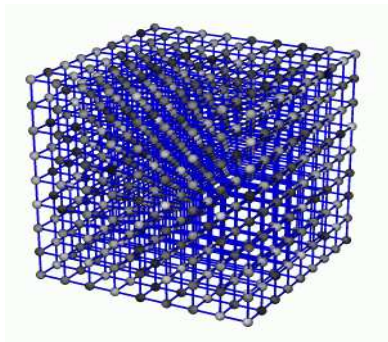
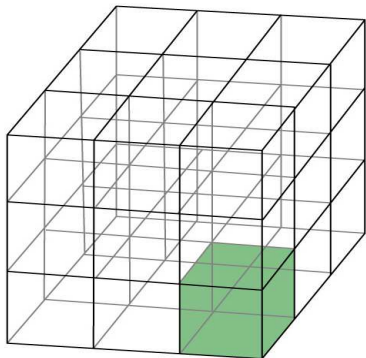
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# Outline

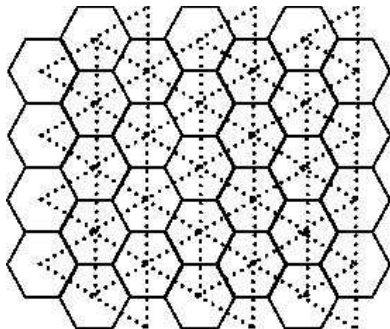
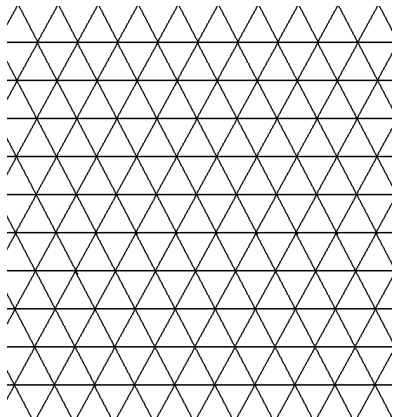
- 1 Lattices
- 2 Random Walk
- 3 Percolation

Lattices:  $\mathbb{Z}$ ,  $\mathbb{Z}^2$ 

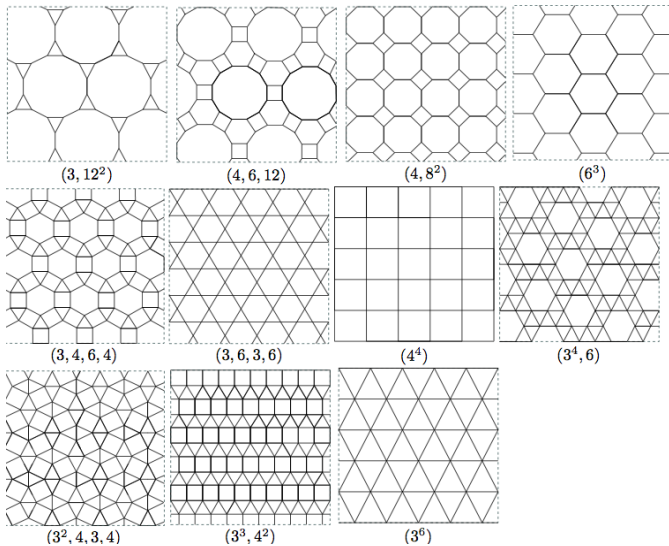
# Lattices: $\mathbb{Z}$ , $\mathbb{Z}^2$ and $\mathbb{Z}^3$



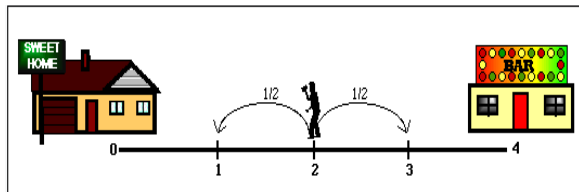
## More lattices: triangular and honeycomb



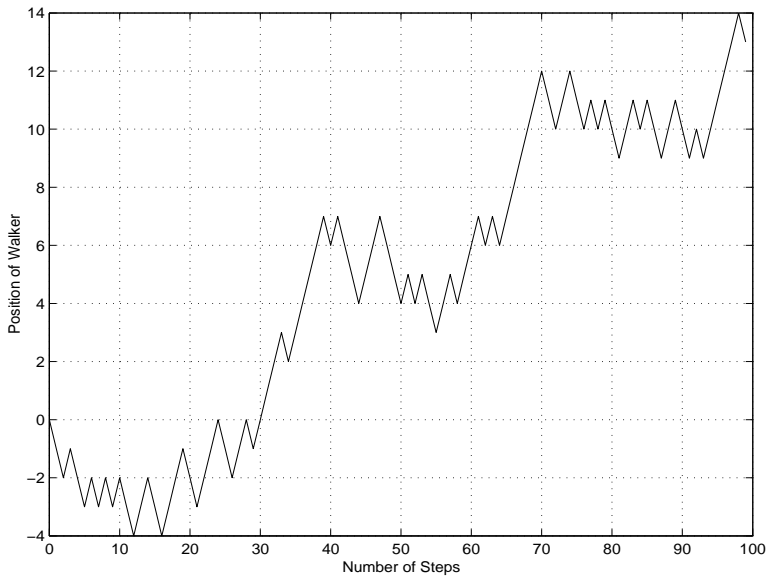
# More lattices: Archimedean Lattices



# Random Walk



Let  $X_i$  be independent random variables satisfying  $\mathbb{P}\{X_i = 1\} = p$ ,  $\mathbb{P}\{X_i = -1\} = 1 - p$ . Then  $S_n = \sum_{i=1}^n X_i$  is called a **simple random walk** on the one-dimensional integer lattice. We will define  $S_0 = 0$  (meaning that the random walk starts at the origin).





A random walk is called **recurrent** if

$$\mathbb{P}\{S_n = 0 \text{ for some } n > 0\} = 1.$$

Let

$$r = \mathbb{P}\{\text{returning to the origin}\} = \mathbb{P}\{S_n = 0 \text{ for some } n > 0\}$$

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$$\mathbb{P}\{\text{returning to the origin exactly } k \text{ times}\} = r^k(1 - r).$$

$$\begin{aligned} m &= \mathbb{E}[\text{number of returns to the origin}] \\ &= 1 \cdot \mathbb{P}\{\text{returning to the origin exactly 1 time}\} \\ &+ 2 \cdot \mathbb{P}\{\text{returning to the origin exactly 2 times}\} \\ &+ 3 \cdot \mathbb{P}\{\text{returning to the origin exactly 3 times}\} \\ &+ 4 \cdot \mathbb{P}\{\text{returning to the origin exactly 4 times}\} \\ &+ \dots \\ &= r(1 - r) + 2r^2(1 - r) + 3r^3(1 - r) + 4r^4(1 - r) + \dots \\ &= (1 - r) \left( r + 2r^2 + 3r^3 + 4r^4 + \dots \right) = \frac{r}{1 - r} \end{aligned}$$

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Since

$$m = \frac{r}{1-r},$$

we see that  $r = 1$  if and only if  $m$  is infinite. So the walker is certain to return to the origin if and only if  $m$  is infinite.

Let's compute  $m$ :

We define  $u_n = \mathbb{P}\{S_n = 0\}$ ,

$$A_n = \begin{cases} 1, & \text{if } S_n = 0 \\ 0, & \text{otherwise} \end{cases}$$

and  $A = \sum_{i \geq 0} A_i$ .

Note that  $\mathbb{E}[A_n] = 1 \cdot u_n + 0 \cdot (1 - u_n) = u_n$  and that  $A$  is the total number of times the random walk is at the origin.

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The random walk can only be at the origin at even times, so

$$m = u_0 + u_2 + u_4 + u_6 + \dots = \sum_{n \geq 0} u_{2n}.$$

$$u_{2n} = \binom{2n}{n} p^n (1-p)^n = \frac{(2n)!}{n!n!} p^n (1-p)^n.$$

Stirling's formula ( $n! \sim \sqrt{2\pi n} e^{-n} n^n$ ) implies

$$u_{2n} \sim \frac{\sqrt{2\pi 2n} e^{-2n} (2n)^{2n} p^n (1-p)^n}{(\sqrt{2\pi n} e^{-n} n^n)^2} = \frac{(4p(1-p))^n}{\sqrt{\pi n}}.$$

Therefore,

$$m = \sum_{n \geq 0} u_{2n} < \infty \iff \sum_{n \geq 0} \frac{(4p(1-p))^n}{\sqrt{\pi n}} < \infty.$$

This is infinite if  $p = \frac{1}{2}$  and finite if  $p \neq \frac{1}{2}$ .

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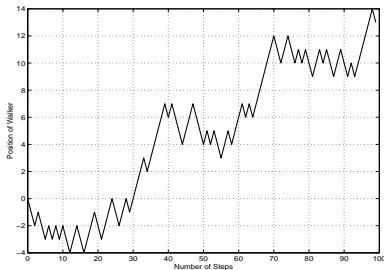
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This means that  $r = 1$  if and only if  $p = \frac{1}{2}$ , and so the random walk is certain to come back to the origin if and only if  $p = \frac{1}{2}$ . This gives the following result due to Pólya:

### Theorem

If  $p = \frac{1}{2}$ , the random walk returns to the origin infinitely often. If  $p \neq \frac{1}{2}$ , the random walk returns to the origin only finitely many times.

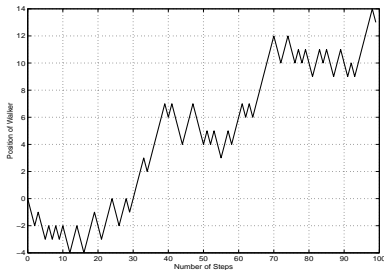


Simple Random Walk of 100 steps

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# Implications for (American) Roulette





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Expected gains if you bet one dollar on red:

$$\frac{18}{38} \cdot 1 + \frac{20}{38}(-1) = -\frac{1}{19}\$.$$

Thus on average, you will lose  $\frac{1}{19}$  dollars per game! You will of course have the privilege of losing more (on average) if you bet more.

Polya's theorem implies that eventually, your gains (a random walk with  $p = \frac{18}{38}$ ) will never be 0 again.

# Gambler's Ruin

Suppose you go to Las Vegas with \$200 and play roulette by betting on red at every spin. You do this until you win \$200 or lose all your money. What is the chance that you'll win \$200?

## Theorem

If you start a random walk (with probability  $p$  of going up,  $q = 1 - p$  of going down by 1) at an integer  $x > 0$  and decide to let the walk run until it reaches a pre-determined value of  $y > x$  or 0, the probability of reaching  $y$  before 0 is

- $\frac{x}{y}$  if  $p = \frac{1}{2}$ .
- $\frac{\left(\frac{q}{p}\right)^x - 1}{\left(\frac{q}{p}\right)^y - 1}$  if  $p \neq q$ .

So all we have to do to answer the question is use  $x = 200$ ,  $y = 400$ ,  $p = \frac{9}{19}$ ,  $q = \frac{10}{19}$ . This gives a probability of

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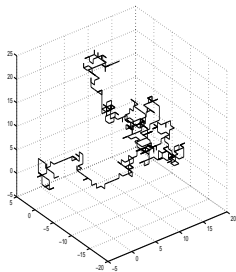
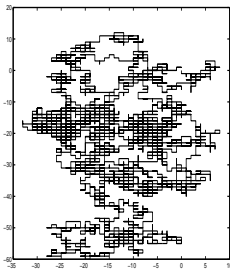
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$$\approx 7 \cdot 10^{-10}$$

## 2 and 3 Dimensions

Let  $X_i$  be independent random vectors in  $\mathbb{Z}^2$  satisfying  $\mathbb{P}\{X_i = \pm\langle 1, 0 \rangle\} = \mathbb{P}\{X_i = \pm\langle 0, 1 \rangle\} = \frac{1}{4}$ . Then  $S_n = \sum_{i=1}^n X_i$  a **symmetric random walk** in  $\mathbb{Z}^2$ . Similarly, if  $X_i$  are independent random vectors in  $\mathbb{Z}^3$  satisfying  $\mathbb{P}\{X_i = \pm\langle 1, 0, 0 \rangle\} = \mathbb{P}\{X_i = \pm\langle 0, 1, 0 \rangle\} = \mathbb{P}\{X_i = \pm\langle 0, 0, 1 \rangle\} = \frac{1}{6}$ . Then  $S_n = \sum_{i=1}^n X_i$  a **symmetric random walk** in  $\mathbb{Z}^3$ .



## 2 and 3 Dimensions

<http://stat.math.uregina.ca/~kozdron/Simulations/LERW/LERW.html>

## 2 and 3 Dimensions

We can use the same analysis as for the one-dimensional case:  
In the 2-d case,

$$\begin{aligned}
 u_{2n} &= \sum_{k=0}^n \frac{(2n)!}{k!k!(n-k)!(n-k)!} \left(\frac{1}{4}\right)^{2n} \\
 &= \left(\frac{1}{4}\right)^{2n} \frac{(2n)!}{n!n!} \sum_{k=0}^n \frac{n!n!}{k!k!(n-k)!(n-k)!} \\
 &= \left(\frac{1}{4}\right)^{2n} \binom{2n}{n} \sum_{k=0}^n \binom{n}{k}^2 = \left( \left(\frac{1}{2}\right)^{2n} \binom{2n}{n} \right)^2.
 \end{aligned}$$

This is the square of the 1-d case, and so

$$u_{2n} \sim \frac{1}{\pi n},$$

the sum of which diverges.



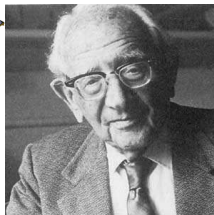
## 2 and 3 Dimensions

For the 3d case, we get (for some constant  $K$ )

$$U_{2n} \leq \frac{K}{n^{3/2}},$$

### Theorem (Pólya)

In two dimensions, the symmetric random walk returns to the origin infinitely often. In three dimensions, the symmetric random walk returns to the origin only finitely many times.



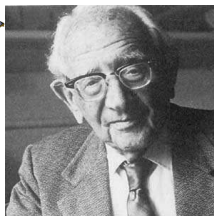
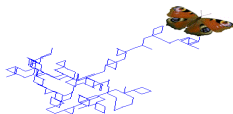
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# Percolation

Consider one of the lattices seen at the beginning of this talk. It is composed of vertices and edges.

Imagine that you paint each vertex in one of two colors, randomly, and independently for each vertex. This is called *site percolation*.

The same thing can be done with edges rather than vertices. This is called *bond percolation*.

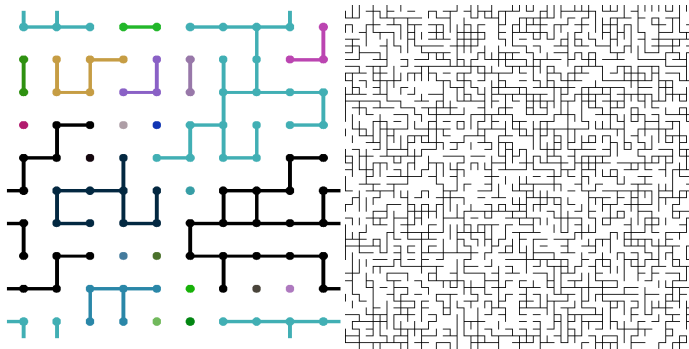
Often, one of the colors is the “invisible” color, that is, we either keep or remove the vertices (or edges).



# Percolation



# When is there a path from left to right?



# Questions

A related (but hard) question is the following:

When is there an infinite open cluster? The answer depends on the probability  $p$  of keeping an edge.

There is a value of  $p$  called  $p_c$  ( $p$ -critical) such that

- If  $p < p_c$ , there is no infinite cluster.
- If  $p > p_c$ , there is an infinite cluster.

The big questions of percolation theory are:

- 1 For a given lattice, what is  $p_c$ ?
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# Simulations

<http://www.physics.buffalo.edu/gonsalves/Java/Percolation.html>

<http://www.svengato.com/forestfire.html>

<http://stat.math.uregina.ca/~kozdron/Simulations/Perc.html>

<http://www.ibiblio.org/e-notes/Perc/perc.htm>

# Answers

lattice	$p_c$ (site)	$p_c$ (bond)
1 square	0.592746	$\frac{1}{2}$ ( <i>Kesten, 80's</i> )
triangular	$\frac{1}{2}$	$2 \sin\left(\frac{\pi}{18}\right)$
honeycomb	0.6962	$1 - 2 \sin\left(\frac{\pi}{18}\right)$

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