

Pólya's Theorem on Random Walks

Vilas Winstein

March 23, 2021

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“A drunk human will find their way home, but a drunk bird may get lost.”

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Since the events of first returning on step n are disjoint, we have

$$p = \sum_{n \geq 0} p_n.$$

Loop Decomposition

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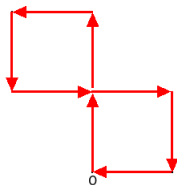
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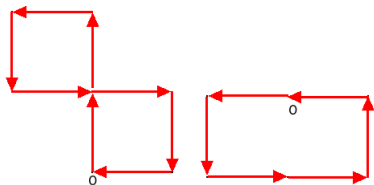
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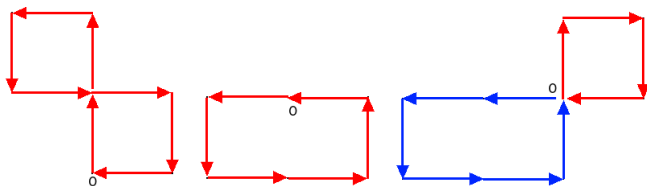
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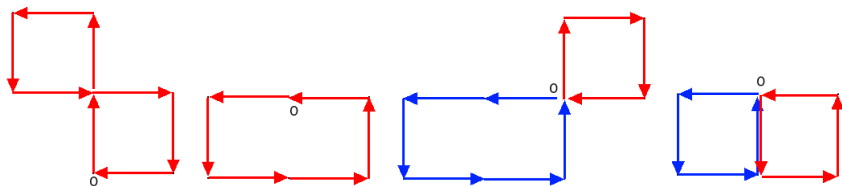
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The equation becomes

$$Q(z) - 1 = P(z)Q(z).$$

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Remember, we are trying to find out if the random walk in \mathbb{Z}^d is recurrent or transient.

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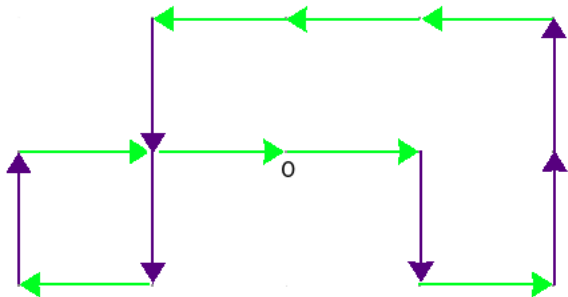
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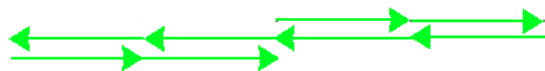
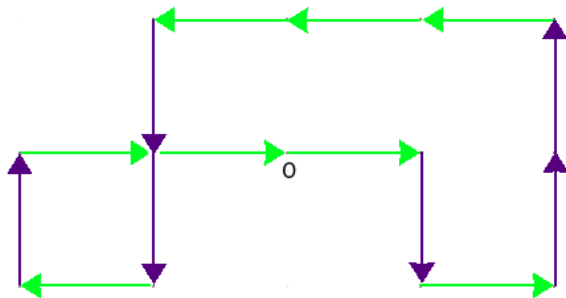
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Don't worry! The formulas that we will be using are a special case of these, and are a bit simpler.

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It will be handy to remember the integral expression for $I_0(z)$, which we will use later:

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Here's that integral formula for the Bessel function again:

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Approximation Time

Since we don't really care about the value of the limit (we just care about whether it is finite or not), it suffices to consider the tail integral

$$\int_N^\infty I_0\left(\frac{tz}{d}\right)^d e^{-t} dt$$

for large N . This is because the integrand doesn't blow up at any finite values. So the integral from 0 to N is always finite, regardless of z .

Here's that integral formula for the Bessel function again:

$$I_0\left(\frac{tz}{d}\right) = \frac{1}{\pi} \int_0^\pi e^{t\frac{z}{d} \cos \theta} d\theta.$$

We will estimate this, as $t \rightarrow \infty$, using an asymptotic analysis method which is called *Laplace's method* (and which uses *Laplace's principle*).

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To quantify this, let's expand $f(\theta)$ using it's second Taylor polynomial approximation:

$$f(\theta) \approx f(0) - |f''(0)| \frac{\theta^2}{2}.$$

Laplace's Method

Now we can use that approximation to approximate the integral:

$$\int_0^\pi e^{tf(\theta)} d\theta \approx e^{tf(0)} \int_0^\pi e^{-t|f''(0)|\frac{\theta^2}{2}} d\theta.$$

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So, hopefully, we have the following good approximation:

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as $t \rightarrow \infty$, where $C = \frac{d^{d/2}}{2^{d/2} \pi^{d-d/2}}$ is a constant. Of course, C depends on d , but that doesn't matter to us at this point, since we are considering the convergence of an integral in t and a limit in z .

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So we have proved Pólya's theorem!

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