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# Self-avoiding polygons in confined geometries

Nicholas Beaton, Jeremy Eng and Chris Soteros Department of Mathematics and Statistics University of Saskatchewan, Saskatoon

#### Retreat for Young Researchers in Stochastics 24 September 2016







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Introduction I:	Motivation			

DNA molecules can be packed incredibly tightly in cell nuclei. For example, human DNA can be 2 m long but must fit inside a cell nucleus of diameter  $10 \,\mu\text{m}$ . Similarly, bacteriophage DNA is packed into a hard capsid until it is injected into the host cell.



<sup>&</sup>lt;sup>1</sup>Arsuaga et al, PNAS 99 (2002), 5373-5377.

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Some DNA molecules (like mitochondrial DNA) have a natural ring structure, while linear DNA can cyclise (the ends stick together) in the nucleus or after being released from confinement.

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Some DNA molecules (like mitochondrial DNA) have a natural ring structure, while linear DNA can cyclise (the ends stick together) in the nucleus or after being released from confinement.

The tight packing within a cell or capsid may result in a high level of tangling, with lots of knots and/or links. Knotting rates of up to 95% have been observed for DNA released from certain bacteriophages.<sup>1</sup>

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<sup>&</sup>lt;sup>2</sup>Arsuaga et al, PNAS **102** (2005), 9165–9169.

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Moreover, it has also been observed that the knot types of randomly cyclised DNA from bacteriophages do not appear to be completely randomly distributed.<sup>2</sup> In particular, chiral knots appear more frequently than in random equilateral polygons.

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It was suggested that it is the writhe of the DNA within the capsid which induces this chirality, but the equilateral polygon models used to test this do not incorporate any excluded volume effects.

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Goal: Investigate the thermodynamic and topological properties of a model of tightly packed polymers which incorporates the excluded volume effect.

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Introduction II: Self-avoiding walks & polygons					

A self-avoiding walk (SAW)  $\omega$  on a graph is a sequence  $(\omega_0, \ldots, \omega_n)$  of distinct vertices with consecutive vertices adjacent on the graph.



When the graph is infinite and has translational symmetry (i.e. a lattice), define SAWs up to translation.

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When the graph is infinite and has translational symmetry (i.e. a lattice), define SAWs up to translation.

For a given lattice,  $c_n$  is the number of SAWs of length n (n edges  $\iff n+1$  vertices).

On 
$$\mathbb{Z}^2$$
,  $\{c_n\}_{n>0} = 1, 4, 12, 36, 100, 284, \dots$  Known up to  $n = 79$ .

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## Theorem (Hammersley 1957)

The limit

$$\lim_{n\to\infty}\frac{1}{n}\log c_n=\kappa$$

exists and is equal to  $\inf_{n\geq 0} \frac{1}{n} \log c_n$ .

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 $\kappa$  is known as the connective constant of the lattice, and  $\mu=e^{\kappa}$  is known as the growth constant.

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# Corollary

$$c_n = e^{o(n)} \mu^n.$$

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 $\mu$  is known exactly only for 2-dimensional honeycomb lattice. For the square  $\mathbb{Z}^2$  and cubic  $\mathbb{Z}^3$  lattices,

$$\begin{split} \mu_{\mathbb{Z}^2} &\approx 2.63815853031 \\ \mu_{\mathbb{Z}^3} &\approx 4.684039931 \end{split}$$

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 $\mu_{\mathbb{Z}^2} \approx 2.63815853031$  $\mu_{\mathbb{Z}^3} \approx 4.684039931$ 

Also interested in the geometric properties of SAWs. Various measures of size, e.g. mean squared end-to-end distance, radius of gyration, etc. are believed to obey a power law:

 $\langle d^2_{
m end-end} 
angle_n \sim {
m const.} imes n^{2
u}$ 

where  $\nu$  depends only on dimension. In 2D, expect  $\nu=$  3/4, while in 3D  $\nu\approx$  0.587597.

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Why use SAWs?

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Why use SAWs?

3D SAWs do a good job of modelling geometric properties of polymers in a good solvent (e.g. mean squared end-to-end distance)





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SAWs incorporate the excluded volume effect.

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Introduction	The model	Large forces	Hamiltonian polygons	Random sampling
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A SAP of n edges can be associated with a SAW of n-1 edges by selecting a vertex and a direction. There are 2n ways to do this.

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Let  $p_n$  be the number of SAPs of length n.

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Let  $p_n$  be the number of SAPs of length n.

Theorem (Hammersley 1961)

The limit

$$\lim_{n\to\infty}\frac{1}{n}\log p_n$$

exists and is equal to  $\kappa$ , the connective constant of the lattice, where the limit is taken through even values of n.

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In 3D SAPs can be knotted:





Introduction	The model	Large forces	Hamiltonian polygons	Random sampling
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In 3D SAPs can be knotted:





In fact, very long polygons are almost always knotted:

#### Theorem (Sumners & Whittington 1988)

All except exponentially few sufficiently long SAPs on the cubic lattice are knotted.



Let  $\mathbb{T}_{L,M} \equiv \mathbb{T}$  be an  $L \times M$  semi-infinite tube of  $\mathbb{Z}^3$ :

$$\mathbb{T} = \{(x, y, z) : x \ge 0, 0 \le y \le L, 0 \le z \le M\}.$$

(Assume  $L \ge M$ .)





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Let  $\mathcal{P}_{\mathbb{T}}$  be the set of SAPs confined within  $\mathbb{T},$  counted up to translation in the x direction.





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Let  $p_{\mathbb{T},n}$  be the number of polygons in  $\mathcal{P}_{\mathbb{T}}$  of length n.

	The model	Large forces	Hamiltonian polygons	Random sampling
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## Theorem (Soteros & Whittington 1989)

The limit

$$\kappa_{\mathbb{T}} = \lim_{n \to \infty} \frac{1}{n} \log p_{\mathbb{T},n}$$

exists.

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#### Theorem (Soteros & Whittington 1989)

The limit

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exists.

Note: Unlike in  $\mathbb{Z}^2$  or  $\mathbb{Z}^3$ , in general SAWs and SAPs in the tube have different growth rates. We will not consider SAWs in  $\mathbb{T}$ .

	The model	Large forces	Hamiltonian polygons	Random sampling
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Compressing/pu	Illing force			

To examine polygons which are tightly packed in a small space, we introduce a force. If  $\pi$  is a polygon in  $\mathbb{Z}^3$  or  $\mathbb{T}$ , let  $s(\pi)$  be its span in the x-direction.

To model a force f acting on polygons, we associate a weight of  $e^{f_2(\pi)}$  with each polygon. The partition function of polygons of length n in  $\mathbb{T}$  is then

$$Z_{\mathbb{T},n}(f) = \sum_{\substack{\pi \in \mathbb{T} \ |\pi| = n}} e^{fs(\pi)} = \sum_{s} p_{\mathbb{T},n}(s) e^{fs}$$

where  $p_{\mathbb{T},n}(s)$  is the number of polygons in  $\mathbb{T}$  of length *n* and span *s*.

	The model	Large forces	Hamiltonian polygons	Random sampling
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where  $p_{\mathbb{T},n}(s)$  is the number of polygons in  $\mathbb{T}$  of length *n* and span *s*.

Theorem (Atapour, Soteros & Whittington 2009)

The free energy

$$\mathcal{F}_{\mathbb{T}}(f) = \lim_{n \to \infty} \frac{1}{n} \log Z_{\mathbb{T},n}(f)$$

exists for all f. It is a continuous, convex function of f, and is almost-everywhere differentiable.

Introduction	The model	Large forces	Hamiltonian polygons	Random sampling		
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Boltzmann dist	Boltzmann distribution					

The partition function is the normalising constant for the Boltzmann distribution, where a polygon  $\pi$  has probability proportional to  $e^{f_s(\pi)}$ .

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If  $f \gg 0$  then polygons with large span (i.e. "stretched") are favoured, while if  $f \ll 0$  then polygons with small span (i.e. "compressed") are favoured.

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If  $f \gg 0$  then polygons with large span (i.e. "stretched") are favoured, while if  $f \ll 0$  then polygons with small span (i.e. "compressed") are favoured.

The expected span of a polygon of length n at a given f (under the Boltzmann distribution) is

$$rac{d}{df}\log Z_{\mathbb{T},n}(f)$$

so that the expected "span density" (span per unit length) in the limit of long polygons is

$$\frac{d}{df}\mathcal{F}_{\mathbb{T}}(f).$$

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Large forces: f	$\rightarrow \infty$			

Interested in the behaviour of the free energy as  $f \to \pm \infty$ .
	The model	Large forces	Hamiltonian polygons	Random sampling
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Large forces: f	$\rightarrow \infty$			

Interested in the behaviour of the free energy as  $f \to \pm \infty$ .

Theorem (NRB, Eng & Soteros 2016)
As $f  o \infty$ , the free energy $\mathcal{F}_{\mathbb{T}}(f)$ is asymptotic to $f/2$ . That is,
$\lim_{f\to\infty}\left(\mathcal{F}_{\mathbb{T}}(f)-f/2\right)=0.$

Note: This result also holds if polygons in  $\mathbb{T}$  are replaced by all polygons in  $\mathbb{Z}^d$ .

	The model	Large forces	Hamiltonian polygons	Random sampling
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Large forces: f	$\rightarrow \infty$			

Interested in the behaviour of the free energy as  $f \to \pm \infty$ .

Theorem (NRB, Eng & Soteros 2016) As  $f \to \infty$ , the free energy  $\mathcal{F}_{\mathbb{T}}(f)$  is asymptotic to f/2. That is, $\lim_{f \to \infty} (\mathcal{F}_{\mathbb{T}}(f) - f/2) = 0.$ 

Note: This result also holds if polygons in  $\mathbb{T}$  are replaced by all polygons in  $\mathbb{Z}^d$ .

Lower bound is straightforward: the maximum span for a polygon of length n is (n-2)/2, and there is always at least one with this span, so

$$Z_{\mathbb{T},n}(f) \ge e^{f(n-2)/2} \qquad \Rightarrow \qquad F_{\mathbb{T}}(f) \ge f/2.$$

	The model	Large forces		Random sampling
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Upper bound (sketch): divide polygons of length *n* into *m* pieces of size  $r = \lfloor n/m \rfloor$  (maybe with leftover piece of length q < r).

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Upper bound (sketch): divide polygons of length *n* into *m* pieces of size  $r = \lfloor n/m \rfloor$  (maybe with leftover piece of length q < r).

If a polygon has span at least t then it has at least 2t edges in the x direction. Pigeonhole principle  $\Rightarrow$  a minimum number of the m pieces contain only edges in the x direction. The number of possibilities for the other pieces is bounded above by counting self-avoiding walks.

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Let  $t = \alpha n$ . The  $f \to \infty$  limit of the free energy is connected (in a non-trivial way) with the limits  $n \to \infty$ ,  $\alpha \to 1/2$ . As  $\alpha \to 1/2$ , the "other" pieces become negligible, and the only contribution to the upper bound is by polygons with (almost) all x-steps, whose free energy  $\to f/2$ .

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Large forces: f	$\rightarrow -\infty$			

	The model	Large forces	Hamiltonian polygons	Random sampling
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Large forces: f	$\rightarrow -\infty$			

An *s*-block of  $\mathbb{T}$  is the section of any polygon between planes x = k + 1/2 and x = k + s + 1/2 for some k (with at least one vertex in each plane  $x = k + 1, k + 2, \dots, k + s$ ). The length is the total number of occupied vertices.



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An s-block is full if it has length Ws, ie. if it occupies every vertex.

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Let  $b_{\mathbb{T},s}$  be the number of *s*-blocks in  $\mathbb{T}$ , and  $b_{\mathbb{T},s}^{\mathsf{F}}$  the number of full *s*-blocks.

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#### Lemma

The following limits exist and are finite:

$$\beta_{\mathbb{T}} = \lim_{s \to \infty} \frac{1}{s} \log b_{\mathbb{T},s} \quad and \quad \beta_{\mathbb{T}}^{\mathsf{F}} = \lim_{s \to \infty} \frac{1}{s} \log b_{\mathbb{T},s}^{\mathsf{F}}$$

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## Theorem (NRB, Eng & Soteros 2016)

The free energy  $\mathcal{F}_{\mathbb{T}}(f)$  is asymptotic to  $(f + \beta_{\mathbb{T}}^{\mathsf{F}})/W$  as  $f \to -\infty$ , ie.

 $\lim_{f\to-\infty} \left( \mathcal{F}_{\mathbb{T}}(f) - f/W \right) = \beta_{\mathbb{T}}^{\mathsf{F}}/W.$ 

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A lower bound is obtained by showing that any full block can be "completed" into a polygon (by adding edges on the left and right) without changing the length or span too much.

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A lower bound is obtained by showing that any full block can be "completed" into a polygon (by adding edges on the left and right) without changing the length or span too much.

The upper bound is similar to the  $f \to \infty$  case, except instead of dividing the polygons up into disjoint subwalks, we divide them into disjoint blocks. As  $f \to -\infty$ , the PHP implies that most must be full.

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Why blocks and	d not polygons?			

The  $f\to-\infty$  asymptote is written in terms of  $\beta_{\mathbb{T}}^{\mathsf{F}}$ , the growth rate for full s-blocks. Why not a growth rate of "full" polygons?

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Why blocks and	d not polygons?	)		

The  $f \to -\infty$  asymptote is written in terms of  $\beta_T^F$ , the growth rate for full *s*-blocks. Why not a growth rate of "full" polygons?

A polygon in  $\mathbb{T}$  is Hamiltonian if it has span s and length n = W(s+1). Equivalently, it occupies every vertex in a  $L \times M \times s$  box of  $\mathbb{Z}^3$ .



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Why blocks and not polygons?				

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Let  $p_{\mathbb{T},n}^{\mathsf{H}}$  be the number of Hamiltonian polygons in  $\mathbb{T}$  of length *n*. Note that  $p_{\mathbb{T},n}^{\mathsf{H}} = 0$  if *n* is not a multiple of *W*.

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Why blocks and	d not polygons?	)		

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#### Theorem (Eng 2014)

The limit

$$\kappa_{\mathbb{T}}^{\mathsf{H}} = \lim_{n \to \infty} \frac{1}{n} \log p_{\mathbb{T},n}^{\mathsf{H}}$$

(taken through values of n which are multiples of W) exists and is finite.

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	The model	Large forces	Hamiltonian polygons	Random sampling
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But while every Hamiltonian polygon is comprised of full blocks, many full blocks cannot form part of a Hamiltonian polygon.



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Hamiltonian polygons only exist when n is a multiple of W. For other n, there are "minimum span" polygons.

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But we don't even know if they have a well-defined growth rate!

#### Conjecture

The growth rates of Hamiltonian polygons and full s-blocks (counted by length instead of span) are the same, ie.

$$\kappa_{\mathbb{T}}^{\mathsf{H}} = \frac{\beta_{\mathbb{T}}^{\mathsf{F}}}{W}.$$

	The model	Large forces	Hamiltonian polygons	Random sampling
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Knots				

### Theorem (Soteros 1998; Atapour, Soteros & Whittington 2009)

For any  $L \times M$  tube  $\mathbb{T}$  with  $L \ge 2$ ,  $M \ge 1$ , and for any finite f, the probability of a random n-step polygon in  $\mathbb{T}$  (sampled from the Boltzmann distribution) being knotted approaches 1 as  $n \to \infty$ .

	The model	Large forces		Random sampling
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Knots				

### Theorem (Soteros 1998; Atapour, Soteros & Whittington 2009)

For any  $L \times M$  tube  $\mathbb{T}$  with  $L \ge 2$ ,  $M \ge 1$ , and for any finite f, the probability of a random n-step polygon in  $\mathbb{T}$  (sampled from the Boltzmann distribution) being knotted approaches 1 as  $n \to \infty$ .

#### Theorem (Eng 2014)

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Knots				

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Like the result in  $\mathbb{Z}^3$ , both proofs use a pattern theorem: there are patterns which guarantee knotting, and which are found in all but exponentially few long polygons.



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Define a transfer matrix  $M_{\mathbb{T}}$  for 1-blocks:  $M_{\mathbb{T}}(i,j) = 1$  iff 1-block j can follow 1-block i in a polygon.

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Let  $S_{\mathbb{T}}$  be the vector with  $S_{\mathbb{T}}(i) = 1$  iff *i* is a 1-block which can start a polygon, and likewise  $E_{\mathbb{T}}$  is the vector for 1-blocks which can end polygons.

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$$p_{\mathbb{T},W(s+1)} = S_{\mathbb{T}} \cdot (M_{\mathbb{T}})^{s-1} \cdot E_{\mathbb{T}}.$$

Can do likewise for Hamiltonian polygons.

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Can do likewise for Hamiltonian polygons.

But the transfer matrices cannot be used to characterise all knotted or unknotted polygons.

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Introduction	The model	Large forces	Hamiltonian polygons	Random sampling

The transfer matrices can be used to generate random polygons of a given span, built up one 1-block at a time. Idea (adapted from [Alm & Janson 1990]):

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# Random sampling via the transfer matrix

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$$rac{1}{\lambda_{\mathbb{T}}} imes rac{\xi_{\mathbb{T}}(b_i)}{\xi_{\mathbb{T}}(b_{i-1})}.$$

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Most of the  $\xi_{\mathbb{T}}$  factors cancel, so we choose polygon  $(b_0, b_1, \ldots, b_s)$  with probability

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$$\frac{\xi_{\mathbb{T}}(b_{s-1})}{(\lambda_{\mathbb{T}})^{s-2}N_{\mathsf{start}}N(b_0)\xi_{\mathbb{T}}(b_1)N_{\mathsf{end}}(b_{s-1})}.$$

To accommodate this, we re-weight each sample by a factor of

$$\frac{N(b_0)\xi_{\mathbb{T}}^{\mathsf{F}}(b_1)N_{\mathsf{end}}(b_{s-1})}{\xi_{\mathbb{T}}^{\mathsf{F}}(b_{s-1})}.$$
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(Some) results				

Probability of unknot (01) for Hamiltonian polygons in  $3\times 1$  tube (horizontal axis is span):



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$$P(0_1) \sim \exp(-7.14 \times 10^{-4} s) = \exp(-8.93 \times 10^{-5} n)$$

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In  $\mathbb{Z}^3$ , it has been estimated  $P(0_1) \sim \exp(-4.15 \times 10^{-6} n)$ .

	The model	Large forces	Hamiltonian polygons	Random sampling
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Probability of unknot  $(0_1)$  for all polygons in  $3 \times 1$  tube:



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Probability of unknot  $(0_1)$  for all polygons in  $3 \times 1$  tube:



By taking the log, we see that

$$P(0_1) \sim \exp(-1.40 \times 10^{-4} s)$$

	The model	Large forces	Hamiltonian polygons	Random sampling
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Probability of trefoil  $(3^\pm_1)$  for Hamiltonian polygons in  $3\times 1$  tube:



	The model	Large forces	Hamiltonian polygons	Random sampling
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Probability of trefoil  $(3^\pm_1)$  for Hamiltonian polygons in  $3\times 1$  tube:



By taking  $P(3_1^{\pm})/P(0_1)$ , we see that

 $P(3_1^{\pm}) \sim 6.98 \times 10^{-4} sP(0_1) \sim 8.73 \times 10^{-5} n \exp(-4.15 \times 10^{-6} n).$ 

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This relationship seems to hold for any prime knot type (with different constants).

Further analysis (in progress) appears to confirm the expectation that if knot type K is the connected sum of k prime knots, then

 $P(K) \sim const. \times n^k P(0_1).$ 

Similar results also hold for non-Hamiltonian polygons in  $\mathbb T.$  However, the transfer matrix is much bigger, so it is harder to get good estimates from the data.

	The model	Large forces	Hamiltonian polygons	Random sampling
Ongoing work				

- Determine how knotting probability behaves for larger tube sizes
- $\bullet\,$  The "knotted part" of a polygon in  $\mathbb T$  tends to be very small look at the distribution of its location/size
- Include the force f in the simulations, and examine how knotting probability etc. changes with force
- Examine writhe, twisting, etc. and how they affect knotting
- In cases where the transfer matrix is too big to be used, develop new method (Markov chain?) for sampling Hamiltonian polygons

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arXiv:1604.07465 – NRB, Jeremy Eng and Chris Soteros, Polygons in restricted geometries subjected to infinite forces. To appear in *Journal of Physics A: Mathematical and Theoretical.* 

More work in preparation.

Thank you!