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Enumeration of self-avoiding walks on the square lattice

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Abstract

We describe a new algorithm for the enumeration of self-avoiding walks on the square lattice. Using up to 128 processors on a HP Alpha server cluster we have enumerated the number of self-avoiding walks on the square lattice to length 71. Series for the metric properties of mean-square end-to-end distance, mean-square radius of gyration and mean-square distance of monomers from the end points have been derived to length 59. An analysis of the resulting series yields accurate estimates of the critical exponents γ and ν confirming predictions of their exact values. Likewise we obtain accurate amplitude estimates yielding precise values for certain universal amplitude combinations. Finally we report on an analysis giving compelling evidence that the leading non-analytic correction-to-scaling exponent $\Delta_1 = 3/2$.

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1. Introduction

The self-avoiding walk (SAW) on regular lattices is one of the most important and classic combinatorial problems in statistical mechanics [24]. SAWs are often considered in the context of lattice models of polymers. The fundamental problem is the calculation (up to translation) of the number of SAWs, c_n , with *n* steps. As most interesting combinatorial problems, SAWs have exponential growth, $c_n \sim A\mu^n n^{\gamma-1}$, where μ is the connective constant, $\gamma = 43/32$ is a (known) critical exponent [25, 26], and *A* is a critical amplitude. So one major problem is the calculation, or at least an accurate estimation of critical amplitudes, such as *A*, in order to test predictions for various universal amplitude combinations for two-dimensional SAWs [2–4]. This requires, in addition to the calculation of c_n , the calculation of metric properties such as the end-to-end distance and the radius of gyration. Furthermore the enumeration of SAWs have traditionally served as a benchmark for both computer performance and algorithm design.

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An *n-step self-avoiding walk* ω on a regular lattice is a sequence of *distinct* vertices $\omega_0, \omega_1, \ldots, \omega_n$ such that each vertex is the nearest neighbour of its predecessor. SAWs are considered distinct up to translations of the starting point ω_0 . We shall use the symbol Ω_n to mean the set of all SAWs of length *n*.

In addition we also consider self-avoiding polygons (SAPs). A SAP can be viewed as a SAW whose end-points ω_0 and ω_n are the nearest neighbours and which therefore can be connected to form a closed loop by the addition of a single step. Note that there are 2(n + 1)SAWs which give rise to a given (n + 1) step SAP. Each vertex of the SAP can be used as ω_0 , and we could walk clockwise or counter-clockwise around the perimeter of the SAP.

The enumeration of SAWs and SAPs has a long and glorious history, which for the square lattice has recently been reviewed in [12]. Suffice to say that early calculations were based on various direct counting algorithms of exponential complexity, with computing time T(n)growing asymptotically as λ^n , where $\lambda = \mu \sim 2.638$, the connective constant for SAWs. Enting [8] was the first to produce a major breakthrough by applying transfer matrix (TM) methods to the enumeration of SAPs on finite lattices. This so-called finite lattice method (FLM) led to a very significant reduction in complexity to $3^{n/4}$, so $\lambda = \sqrt[4]{3} = 1.316...$ More recently we [18] refined the algorithm using the method of pruning and reduced the complexity to 1.2^n . The extension of the FLM to SAW enumeration had to wait until 1993 when Conway et al [5] implemented an algorithm with complexity $3^{n/4}$. The algorithm is difficult to implement and requires large amounts of physical memory. However, the algorithm cannot be used to calculate metric properties. In this paper we pursue a different FLM algorithm based on the same ideas used to improve the SAP algorithm. It appears that this pruning algorithm has a computational complexity of 1.334ⁿ very close to the CEG algorithm. So the CEG will ultimately beat the pruning algorithm for large enough n. For small n the pruning algorithm actually uses significantly less memory as we shall show in section 2.4, and it can in addition be used to calculate metric properties. To our knowledge this is the first time TM methods has been used to calculate metric properties of SAWs.

The quantities we consider in this paper are

• The number of SAWs of length *n*, believed to have the asymptotic behaviour

$$c_n = A\mu^n n^{\gamma - 1} [1 + o(1)] \tag{1.1a}$$

where μ is the connective constant and γ is a critical exponent. We shall also study the associated generating function

$$C(u) = \sum_{n=0}^{\infty} c_n u^n = A \Gamma(\gamma) (1 - u\mu)^{-\gamma} [1 + o(1)]$$
(1.1b)

so the generating function has a singularity at $u = u_c = 1/\mu$.

• The number of SAPs of length *n*, believed to grow asymptotically as

$$p_n = B\mu^n n^{\alpha-3} [1 + o(1)] \tag{1.2a}$$

where α is another critical exponent. In this case the generating function behaves as

$$\mathcal{P}(u) = \sum_{n=0}^{\infty} p_n u^n = B\Gamma(\alpha - 2)((1 - u\mu)^{2-\alpha} [1 + o(1)].$$
(1.2b)

• The mean-square end-to-end distance of *n* step SAWs

$$\langle R_e^2 \rangle_n = \frac{1}{c_n} \sum_{\Omega_n} (\omega_0 - \omega_n)^2 = C n^{2\nu} [1 + o(1)]$$
 (1.3*a*)

where ν is a new critical exponent. We also look at the generating function

$$\mathcal{R}_{e}(u) = \sum_{n} c_{n} \langle R_{e}^{2} \rangle_{n} u^{n} = A C \Gamma(\gamma + 2\nu) (1 - u\mu)^{-(\gamma + 2\nu)} [1 + o(1)].$$
(1.3b)

• The mean-square radius of gyration of *n* step SAWs

$$\langle R_g^2 \rangle_n = \frac{1}{c_n} \sum_{\Omega_n} \left[\frac{1}{2(n+1)^2} \sum_{i,j=0}^n (\omega_i - \omega_j)^2 \right] = Dn^{2\nu} [1 + o(1)] \quad (1.4a)$$

with the associated generating function

$$\mathcal{R}_g(u) = \sum_n (n+1)^2 c_n \langle R_g^2 \rangle_n u^n = A D \Gamma(\gamma + 2\nu + 2)(1 - u\mu)^{-(\gamma + 2\nu + 2)} [1 + o(1)] \quad (1.4b)$$

where the factors under the sum ensure that the coefficients are integer valued.

• The mean-square distance of a monomer from the end-points of *n* step SAWs

$$\left\langle R_m^2 \right\rangle_n = \frac{1}{c_n} \sum_{\Omega_n} \left[\frac{1}{2(n+1)} \sum_{i=0}^n \left[(\omega_0 - \omega_j)^2 + (\omega_n - \omega_j)^2 \right] \right] = E n^{2\nu} [1 + o(1)] \quad (1.5a)$$

with the associated generating function

$$\mathcal{R}_m(u) = \sum_n (n+1)c_n \langle R_m^2 \rangle_n u^n = AE\Gamma(\gamma + 2\nu + 1)(1 - u\mu)^{-(\gamma + 2\nu + 1)}[1 + o(1)].$$
(1.5b)

The critical exponents are believed to be universal in that they only depend on the dimension of the underlying lattice. μ on the other hand is non-universal. For SAWs in two dimensions the critical exponents $\gamma = 43/32$, $\alpha = 1/2$ and $\nu = 3/4$ have been predicted exactly, though non-rigorously, using Coulomb-gas arguments [25, 26].

While the amplitudes are non-universal, there are many universal amplitude ratios. Any ratio of the metric amplitudes, e.g., D/C and E/C, is expected to be universal [4]. Many other universal amplitude combinations in particular involving SAPs can be found in [3, 29]. Of particular interest is the linear combination [4, 2] (which we shall call the CSCPS relation)

$$F \equiv \left(2 + \frac{y_t}{y_h}\right) \frac{D}{C} - 2\frac{E}{C} + \frac{1}{2}$$
(1.6)

where $y_t = 1/\nu$ and $y_h = 1 + \gamma/(2\nu)$ are the thermal and magnetic renormalization-group eigenvalues, respectively, of the *n*-vector model at n = 0. In two dimensions ($y_t = 4/3$ and $y_h = 91/48$, hence $2 + y_t/y_h = 246/91$) Cardy and Saleur [4] (as corrected by Caracciolo *et al* [2]) have predicted, using conformal field theory, that F = 0. This conclusion has been confirmed by the previous high-precision Monte Carlo work [2] as well as by series extrapolations [13].

Privman and Redner [28] proved that the combination $BC/\sigma a_0$ is universal. σ is an integer constant such that p_n is non-zero when n is divisible by σ . So $\sigma = 1$ for the triangular lattice and 2 for the square and honeycomb lattices. a_0 is the area per lattice site and $a_0 = 1$ for the square lattice, $a_0 = 3\sqrt{3}/4$ for the honeycomb lattice and $a_0 = \sqrt{3}/2$ for the triangular lattice.

The asymptotic form (1.1a) for c_n only explicitly gives the leading contribution. In general one would expect corrections to scaling so

$$c_n = A\mu^n n^{\gamma-1} \left[1 + \frac{a_1}{n} + \frac{a_2}{n^2} + \dots + \frac{b_0}{n^{\Delta_1}} + \frac{b_1}{n^{\Delta_1+1}} + \frac{b_2}{n^{\Delta_1+2}} + \dots \right].$$
(1.7)

In addition to 'analytic' corrections to scaling of the form a_k/n^k , there are 'non-analytic' corrections to scaling of the form b_k/n^{Δ_1+k} , where the correction-to-scaling exponent Δ_1 is

not an integer. In fact one would expect a whole sequence of correction-to-scaling exponents $\Delta_1 < \Delta_2 \ldots$, which are both universal and also independent of the observable, that is the same for c_n , p_n and so on. Much effort has been devoted to determining the leading non-analytic correction-to-scaling exponent Δ_1 for two-dimensional SAWs and SAPs. At least two different theoretical predictions have been made for the exact value of this exponent: $\Delta_1 = 3/2$ based on Coulomb-gas arguments [25, 26], and $\Delta_1 = 11/16$ based on conformal-invariance methods [30].

In a recent paper [1] we studied the amplitudes and the correction-to-scaling exponents for two-dimensional SAWs, using a combination of series-extrapolation and Monte Carlo methods. We enumerated all self-avoiding walks up to 59 steps on the square lattice, and up to 40 steps on the triangular lattice, measuring the metric properties mentioned above, and then carried out a detailed and careful analysis of the data in order to accurately estimate the amplitudes and correction-to-scaling exponent. In this paper we give a detailed account of the algorithm used to calculate the square lattice series analysed in [1], report on a further extension of the SAW counts up to 71 steps, analyse the series and confirm to great accuracy the predicted exact values of the critical exponents and finally we briefly summarize the results of the analysis from [1].

2. Enumeration of self-avoiding walks

The algorithm we use to enumerate SAWs on the square lattice builds on the pioneering work of Enting [8] who enumerated square lattice self-avoiding polygons using the finite lattice method. More specifically our algorithm is based in large part on the one devised by Conway *et al* [5] for the enumeration of SAWs. The basic idea of the finite lattice method is to calculate partial generating functions for various properties of a given model on finite pieces, say $W \times L$ rectangles of the square lattice, and then reconstruct a series expansion for the infinite lattice limit by combining the results from the finite pieces. The generating function for any finite piece is calculated using transfer matrix (TM) techniques.

2.1. Basic transfer matrix algorithm

The most efficient implementation of the TM algorithm generally involves bisecting the finite lattice with a boundary (this is just a line in the case of rectangles) and moving the boundary in such a way as to build up the lattice cell by cell. The sum over all contributing graphs is calculated as the boundary is moved through the lattice. Due to the symmetry of the square lattice we need only consider rectangles with $L \ge W$. SAWs in a given rectangle are enumerated by moving the intersection so as to add one vertex at a time, as shown in figure 1. For each configuration of occupied or empty edges along the intersection, we maintain a generating function for partial walks cutting the intersection in that particular pattern. If we draw a SAW and then cut it by a line, we observe that the partial SAW to the left of this line consists of a number of loops connecting two edges (we shall refer to these as loop ends) in the intersection, and pieces which are connected to only one edge (we call these free ends). The other end of the free piece is an end point of the SAW so there are at most two free ends. In applying the transfer matrix technique to the enumeration of SAWs, we regard them as sets of edges on the finite lattice with the properties:

- (1) A weight u is associated with an occupied edge. In some cases one gives different weights u and v to occupied horizontal and vertical edges, respectively.
- (2) All vertices are of degree 0, 1 or 2.

			13 1 1		
		г - -	$\frac{1}{1}^{2}$		
		2 			
		1			
		1			
	-			 	

Figure 1. A snapshot of the boundary line (dashed line) during the transfer matrix calculation on the square lattice. SAWs are enumerated by successive moves of the kink in the boundary line, as exemplified by the position given by the dotted line, so that one vertex at a time is added to the rectangle. To the left of the boundary line we have drawn an example of a partially completed SAW.

- (3) There are at most two vertices of degree 1 and the final graph has exactly two vertices of degree 1 (the end points of the SAW).
- (4) Apart from isolated sites, the final graph has a single connected component.
- (5) In some implementations, each graph must span the rectangle from left to right, while in other implementations each graph must span the rectangle from left to right and from bottom to top.

We are not allowed to form closed loops, so two loop ends can only be joined if they belong to different loops. To exclude loops which close on themselves, we need to label the occupied edges in such a way that we can easily determine whether or not two loop ends belong to the same loop. The most obvious choice would be to give each loop a unique label. However, on two-dimensional lattices there is a more compact scheme relying on the fact that two loops can never intertwine. Each end of a loop is assigned one of two labels depending on whether it is the lower end or the upper end of a loop. Each configuration along the boundary line can thus be represented by a set of edge states { σ_i }, where

$$\sigma_i = \begin{cases} 0 & \text{empty edge} \\ 1 & \text{lower loop end} \\ 2 & \text{upper loop end} \\ 3 & \text{free end.} \end{cases}$$
(2.1)

If we read from the bottom to the top, the configuration along the intersection of the partial SAW in figure 1 is {011212320}. It is easy to see that this encoding uniquely describes which loop ends are connected. In order to find the upper loop end, matching a given lower end, we start at the lower end and work upwards in the configuration counting the number of '1's and '2's we pass (the '1' of the initial lower end is *not* included in the count). We stop when the number of '2's exceeds the number of '1's. This '2' marks the matching upper end of the loop. It is worth noting that there are some restrictions on the possible configurations. Firstly, every lower loop end must have a corresponding upper end, and it is therefore clear that the

Table 1. The various local 'input' states and the 'output' states which arise as the boundary line is moved in order to include one more vertex of the lattice.

			0			1			2			3	
0	00	12	03	30	01	10	00	02	20	$\overline{00}$	03	30	ADD
1	01	10	00			00		NO	T allo	wed		00	
2	02	20	00			00			00			$\overline{00}$	
3	03	30	ADD			00			00			ADD	

total number of '1's is equal to the total number of '2's. Secondly, as we look through the configuration starting from the bottom the number of '1's is never smaller than the number of '2's. Ignoring the '0's the '1's and '2's can be viewed as perfectly balanced parenthesis. Those familiar with algebraic languages will immediately recognize that each configuration (for now treating free ends and empty edges in the same way) of labelled loop ends forms a Motzkin word [7].

2.1.1. Derivation of updating rules. The updating of a partial generating function depends on the states of the edges to the left and above the new vertex. When the kink is moved we insert the edges to the right and below the new vertex. The way to avoid situations leading to graphs with more than a single connected component is to forbid free ends from terminating (or joining) at the vertex being processed unless the boundary line intersects no other occupied edges. In table 1 we have listed the possible local 'input' states and the 'output' states which arise as the kink in the boundary is propagated by one step. The rows in this table are labelled by the state of the left edge, while the columns are labelled by the state of the top edge. Each panel in the table contains the possible states of the bottom and right edges (in that order). We shall refer to the configuration before the move as the 'source' and a configuration produced as a result of the move as a 'target'. In each move the source generating function is multiplied by u^k , where k is the number of new occupied edges (just the number of non-zero entries in the local output state), and is then added to the target generating function.

In the following we give the details of how some of these updating rules are derived.

- 00: The left and top edges are empty. We have four possible outputs. We can leave the bottom and right edges empty (00), insert a new partial loop (12), or add a new free end on the right (03) or bottom (30) edge. Adding a free end increases by one the number of degree-1 vertices, so this is only allowed provided the source has at most one free end. Throughout, this restriction is indicated by the use of boldface entries.
- 01, 10, 02, 20: The left or top edge is occupied by a loop end. We can continue this loop end along either the right or bottom edge. Note that we cannot occupy both new edges since this would lead to vertices of degree 3. We can also leave both edges empty. This creates a new degree-1 vertex, and we have to relabel the matching end of the discontinued loop as free. Relabelling is indicated by over-lining. The way to identify the matching loop end is described below (2.1) in the previous section.
 - 03, 30: The left or top edge is occupied by a free end. We can continue the free end along either the right or bottom edge. We can also leave both edges empty. This creates a separate component and is only allowed if the resulting graph is a valid SAW. That is, the source contains no other occupied edges (and if required both the bottom and top of the rectangle has been touched).

The partial generating function is added to the running total. We mark this possibility by the entry ADD.

- 11, 22: Two lower (upper) loop ends are joined and the output edges must be empty (otherwise we would create vertices of degree greater than 2). The matching upper (lower) loop end of the innermost loop is relabelled as the new lower (upper) end of the combined loop.
 - 12: A closed loop would be formed. This is not allowed.
 - 21: Upper and lower loop ends are joined and the output edges must be empty.
- 13, 31, 23, 32: A free end is joined to a lower (upper) loop end. The output edges are empty and the matching loop end is relabelled free.
 - 33: Two free ends are joined. This results is a separate component. If the resulting graph is a valid SAW we add it to the generating function.

2.2. The Conway–Enting–Guttmann (CEG) algorithm

The algorithm used by Conway *et al* [5] to enumerate SAWs is ingenious but also quite complicated and relies heavily on manipulations of various generating functions. Here we shall only give the briefest of outlines of the algorithm sufficient for the reader to appreciate the differences between this algorithm and the one we used for the enumerations reported in this paper.

The CEG algorithm leads to the enumeration of anisotropic SAWs, that is the number of SAWs $c_{m,n}$ with *m* steps parallel to the *y*-axis and *n*-steps parallel to the *x*-axis. Obviously, $c_{m,n} = c_{n,m}$. The major 'trick' of the algorithm [5] is the realization that any SAW can be constructed by combining *irreducible* components. An irreducible component has at least two steps along the *y*-axis in each position, e.g., any line parallel to the *x*-axis will intersect the component at least twice (or not at all if the line lies beyond the extent of the components). There are five different types of irreducible components. The irreducible components are obtained from enumerations of anisotropic SAWs in rectangles. The SAWs span the rectangle in the *x*-direction but not necessarily in the *y*-direction. For each rectangle four enumerations are done with different restrictions on the allowed positions of the end-points, e.g., the endpoints may be allowed to lie only on the top border, on the top and/or bottom border etc. The enumeration of SAWs in a rectangle is done using the basic transfer matrix algorithm and updating rules as described above. The only difference is that we have to take care when creating a new free end that it is allowed under the restrictions imposed on the end-points.

If one enumerates anisotropic SAWs in rectangles up to width W it is possible to generate the series correctly to order $N_{\text{max}} = 4W - 1$. Note that this requires the calculation of a two parameter generating function since variables u and v must be kept for horizontal and vertical steps, respectively. The generating functions can be truncated if $m + n > N_{\text{max}}$.

The CEG algorithm requires the calculation of the anisotropic generating function even though one may ultimately only be interested in the isotropic SAW counts. However anisotropic series can be very useful and most importantly can yield valuable insights into the analytic properties of the generating function. In recent papers [9, 11], a numerical procedure was given (using anisotropic series) that indicates whether or not a given statistical mechanics problem is solvable in terms of D-finite functions. A D-finite function can be expressed as the solution to a linear ordinary differential equation of finite order with polynomial coefficients.

2.3. The pruning algorithm

The use of *pruning* to obtain more efficient TM algorithms was used for SAPs in [18]. We required valid SAPs to span the rectangle in *both* directions and directly enumerate SAPs

of width exactly W and length L rather than of width $\leq W$ and length L as done in [8]. At first glance, this appears inefficient since we have to keep four distinct generating functions depending on which borders have been touched. However, for SAPs [18] it actually leads to an algorithm which is both exponentially faster and whose memory requirement is exponentially smaller. Experimentally it was found that the computational complexity was close to $2^{n/4}$, much better than the $3^{n/4}$ of the original approach. We have used similar techniques for the enumerations of SAWs carried out for this paper.

Pruning, details of which are given in [18] for the SAP case, allows us to discard most of the possible configurations for large W because they only contribute to SAWs of length greater than N_{max} , where N_{max} is the maximal length to which we choose to carry out our calculations. The value of N_{max} is limited by the available computational resources, be they CPU time or physical memory. Briefly pruning works as follows. Firstly, for each configuration we keep track of the current minimum number of steps N_{cur} already inserted to the left of the boundary line in order to build up that particular configuration. Secondly, we calculate the minimum number of additional steps N_{add} required to produce a valid SAW. There are three contributions, namely the number of steps required to connect the loops and free ends, the number of steps needed (if any) to extend at least W edges in the lengthwise direction (remember we only need rectangles with $L \ge W$). If the sum $N_{\text{cur}} + N_{\text{add}} > N_{\text{max}}$ we can discard the partial generating function for that configuration, and of course the configuration itself, because it would not make a contribution to the SAW count up to the perimeter lengths we are trying to obtain.

There are no principal differences between pruning SAWs and SAPs though the detailed implementation is more complicated for the SAW case. We found it necessary to explicitly write subroutines to handle the three distinct cases where the TM configuration contains zero, one and two free ends, respectively. But in all cases we essentially have to go through all the possible ways of completing a SAW in order to find the minimum number of steps required. This is a fairly straight forward task though quite time consuming.

Note that the pruning algorithm can be used to generate either isotropic or anisotropic series. That is, unlike the CEG algorithm, we need only to maintain isotropic generating functions if we are after isotropic counts for SAWs. But if we wish to do so, say in order to perform the 'solvability' check mentioned above [9, 11], we could calculate anisotropic generating functions (at the expense of greatly increased memory requirements).

Inspired by Knuth's algorithm for the enumeration of polyominoes [20], we implemented a couple of further enhancements to our SAW algorithm. The first improvement uses a further symmetry of the square lattice. When a column has been completed the configuration are symmetric under reflection. That is the generating functions for the configurations, such as {010122030} and {030112020}, are identical. This symmetry also extends to the touching of the upper/lower borders of the rectangle. The second improvement is superior memory management. A given boundary line configuration only contributes from order $n = N_{cur}$ + N_{add} , so we need only retain the first $N_{\text{max}} - n$ terms in the associated generating function. In our case the maximum memory consumption occur at W = 27. Here there are approximately 1.12 billion distinct configurations and a total of about 4 billion terms in the generating functions. So on average there is a little less than four terms per configuration. At smaller widths there are fewer configurations but more terms per configuration. At larger widths both the number of configurations and the number of terms per configuration decrease. The important thing to note is that as N_{max} is increased the maximal number of terms seems to approach a constant (with a value less than 4) times the maximal number of configurations.



Figure 2. Lin-log plot of the maximal number of TM configurations from the pruned algorithm with increasing n.

2.4. Computational complexity

The time T(n) required to obtain the number of walks of length *n* grows exponentially with $n, T(n) \propto \lambda^n$. For the CEG algorithm the complexity can be calculated exactly. Time (and memory) requirements are basically proportional to a polynomial (in *n*) times the maximal number of configurations, N_{Conf} , generated during a calculation. When the boundary line is straight and intersects W + 1 edges we can find the exact number of configurations. First look at the situation with no free ends. The configurations correspond to Motzkin paths [7] (just map 0 to a horizontal step, 1 to a north-east step, and 2 to a south-east step) and the number of such paths M_n with *n* steps is easily derived from the generating function

$$M(x) = \sum_{n} M_{n} x^{x} = [1 - x - (1 - 2x - 3x^{2})^{1/2}]/2x^{2}.$$
 (2.2)

The number of transfer matrix configurations $N_S(W)$ in the CEG algorithm is simply obtained by inserting 0, 1 or 2 free ends into a Motzkin path and eliminating the path corresponding to a configuration of all 0s, hence

$$N_{\rm S}(W) = M_{W+1} + (W+1)M_W + (W+1)WM_{W-1}/2 - 1.$$
(2.3)

When the boundary line has a kink (such as in figure 1) N_{Conf} is no longer given exactly by (2.3). However, it is obvious that $N_{\text{S}}(W+1) \leq N_{\text{Conf}} \leq N_{\text{S}}(W)$ so from (2.2) we see that asymptotically $N_{\text{Conf}}(W)$ grows like 3^W . Since a calculation using rectangles of widths $\leq W$ yields the number of SAW up to n = 4W it follows that the complexity of the algorithm is $T(n) \propto \lambda^n$ with $\lambda = \sqrt[4]{3} = 1.316...$

The pruned algorithm is too much difficult to analyse exactly. So all we can give is a numerical calculation of the growth in the number of configurations with *n*. That is obtained by just running the SAW algorithm and measuring the maximal number of configurations generated for different values of *n*. The resulting graph is shown in figure 2. The straight line, drawn as a guide to the eye, has slope $\lambda = \sqrt[4]{3}$ and thus corresponds to the exponential growth of the CEG algorithm. From this figure it is clear that the computational complexities of the two algorithms are almost identical. A closer look at the actual numbers does however reveal that the pruned algorithm appears to have a slightly higher value of λ_p . Indeed it appears that increasing *n* by 8 increases the number of configurations by close to a factor of

10 (rather than the 9 expected if $\lambda_p = \sqrt[4]{3}$). This would mean that for the pruned algorithm $\lambda_p \approx \sqrt[8]{10} = 1.3335...$

The observed value of λ_p means that the CEG algorithm is asymptotically superior to the pruning algorithm, so that for very large values of *n* it will be not only be faster but require less memory as well. However for small *n* the pruning algorithm is highly competitive and can in fact use significantly less memory. This is because the CEG algorithm uses a two parameter generating function so memory requirements are $\alpha n^2 \lambda^n$. For the pruning algorithm memory growth is $\alpha \lambda_p^n$, rather than what one may naively have thought $\alpha n \lambda_p^n$ (see comments at the end of the previous section). More concretely, we can mention that the calculation in [6] of SAWs up to n = 51 required 10 GB of memory. The pruning algorithm can do the same job using less than 150 MB of memory.

2.5. Parallelization

The transfer-matrix algorithms used in the calculations of the finite lattice contributions are eminently suited for parallel computations. The bulk of the calculations for this paper were performed on the facilities of the Australian Partnership for Advanced Computing (APAC). The APAC facility is a HP Alpha server cluster with 125 ES45s each with four 1 GHz chips for a total of 500 processors in the compute partition. Each server node has at least 4 Gb of memory. Nodes are interconnected by a low latency high bandwidth Quadrics network.

The most basic concern in any efficient parallel algorithm is to minimize the communication between processors and ensure that each processor does the same amount of work and uses the same amount of memory. In practice one naturally has to strike some compromise and accept a certain degree of variation across the processors.

One of the main ways of achieving a good parallel algorithm using data decomposition is to try to find an invariant under the operation of the updating rules. That is we seek to find some property of the configurations along the boundary line which does not alter in a single iteration. The algorithm for the enumeration of SAWs is quite complicated since not all possible configurations occur due to pruning, and an update at a given set of edges might change the state of an edge far removed, e.g., when two lower loop ends are joined we have to relabel one of the associated upper loop ends as a lower loop end in the new configuration. However, there is still an invariant since any edge not directly involved in the update cannot change their occupation status. This invariant allows us to parallelize the algorithm in such a way that we can do the calculation completely independently on each processor with just two redistributions of the data set each time an extra column is added to the lattice. We have already used this scheme for SAPs [16] and lattice animals [15] and refer the interested reader to these publications for further detail. Our parallelization scheme is also very similar to that used by Conway and Guttmann [6, 12].

2.6. Metric properties

In a recent paper [14] we demonstrated that one can use transfer matrix techniques to calculate the radius of gyration of SAPs. Below we show how this work can be extended to calculate the metric properties of SAWs.

2.6.1. *Radius of gyrations*. We define the radius of gyration according to the *vertices* of the SAW. Note that the number of vertices is one more than the number of steps. The radius of

gyration of n + 1 points at positions \mathbf{r}_i is

$$(n+1)^{2} \langle R_{g}^{2} \rangle_{n} = \sum_{i>j} (\mathbf{r}_{i} - \mathbf{r}_{j})^{2} = n \sum_{i} \left(x_{i}^{2} + y_{i}^{2} \right) - 2 \sum_{i>j} (x_{i}x_{j} + y_{i}y_{j}).$$
(2.4)

This last expression is suitable for a transfer matrix calculation. We actually calculate the coefficients of the generating function (1.4b), $(n + 1)^2 c_n \langle R_g^2 \rangle_n$. In order to do this we have to maintain five partial generating functions for each possible boundary configuration, namely

- C(u), the number of (partially completed) SAWs,
- $X_{\rho}^{2}(u)$, the sum over SAWs of the squared components of the distance vectors,
- $X_g(u)$, the sum of the *x*-component of the distance vectors,
- $Y_g(u)$, the sum of the y-component of the distance vectors,
- $XY_g(u)$, the sum of the 'cross' product of the components of the distance vectors, that is $\sum_{i>j} (x_i x_j + y_i y_j)$.

As the boundary line is moved to a new position each target configuration S might be generated from several sources S' in the previous boundary position. The partial generation functions are updated as follows, with (x, y) being the coordinates of the newly added vertex:

$$C(u, S) = \sum_{S'} u^{n'} C(u, S')$$

$$X_{g}^{2}(u, S) = \sum_{S'} u^{n'} [X_{g}^{2}(u, S') + \delta_{g}(x^{2} + y^{2})C(u, S')]$$

$$X_{g}(u, S) = \sum_{S'} u^{n'} [X_{g}(u, S) + \delta_{g}xC(u, S')]$$

$$Y_{g}(u, S) = \sum_{S'} u^{n'} [Y_{g}(u, S) + \delta_{g}yC(u, S')]$$

$$XY_{g}(u, S) = \sum_{S'} u^{n'} [XY_{g}(u, S') + \delta_{g}xX_{g}(u, S') + \delta_{g}yY_{g}(u, S')]$$
(2.5)

where *n'* is the number of steps added to the SAW and $\delta_g = 0$ if the new vertex is empty (has degree 0), $\delta_g = 1$ if the new vertex is occupied (has degree >0).

Finally, when valid SAWs are completed, the partial generating functions are added to running totals for each case, and the results for coefficients in the generating function for the radius of gyration is

$$(n+1)^2 c_n \langle R_g^2 \rangle_n = n \langle X_g^2 \rangle_n - 2 \langle X Y_g \rangle_n.$$
(2.6)

2.6.2. *End-to-end distance*. The updating rules for the end-to-end distance are very similar to those for the radius of gyration except that we 'count' only the degree-1 vertices. We again maintain five partial generating functions for each possible boundary configuration, namely

- C(u), the number of (partially completed) SAWs,
- $X_{e}^{2}(u)$, the sum over SAWs of the squared components of the end-point vectors,
- $X_e(u)$, the sum of the x-component of the end-point vectors,
- $Y_e(u)$, the sum of the y-component of the end-point vectors,
- $XY_e(u)$, the sum of the 'cross' product of the components of the end-point vectors.

The partial generation functions are updated as described above (2.5) except that the corresponding quantity $\delta_e = 0$ if the new vertex has degree 0 or 2, while $\delta_e = 1$ if the new vertex has degree 1.

The results for coefficients in the generating function for the end-to-end distance is

$$c_n \langle R_e^2 \rangle_n = \langle X_e^2 \rangle_n - 2 \langle XY_e \rangle_n.$$
(2.7)

2.6.3. *Mean-square monomer distance from end points*. In order to calculate the mean-square distance of a monomer from the end points we have to introduce an additional partial generating function

• *XY_m(u)*, the sum of the 'cross'-product of the components of the end-points and distance vectors.

This is updated as follows:

$$XY_{m}(u, S) = \sum_{S'} u^{n'} [XY_{m}(u, S') + \delta_{g} x X_{e}(u, S') + \delta_{g} y Y_{e}(u, S') + \delta_{e} x X_{g}(u, S') + \delta_{e} y Y_{g}(u, S')].$$
(2.8)

The results for the coefficients in the generating function for the mean-square monomer distance from end points is

$$c_n \langle R_e^2 \rangle_m = (n-1) \langle X_e^2 \rangle_n + 2 \langle X_g^2 \rangle_n - 2 \langle XY_m \rangle_n.$$
(2.9)

2.7. Further details

Finally a few remarks of a more technical nature. The number of contributing configurations becomes very sparse in the total set of possible states along the boundary line and as is standard in such cases one uses a hash-addressing scheme. Since the integer coefficients occurring in the series expansion become very large, the calculation was performed using modular arithmetic [19]. This involves performing the calculation modulo various integers p_i and then reconstructing the full integer coefficients at the end. The p_i are called moduli and must be chosen so they are mutually prime, e.g., none of the p_i have a common divisor. The Chinese remainder theorem ensures that any integer has a unique representation in terms of residues. If the largest absolute values occurring in the final expansion is m, then we have to use a number of moduli k such that $p_1 p_2 \cdots p_k/2 > m$. Since we are using a heavily loaded shared facility CPU time was more of an immediate limitation than memory. So we used the moduli $p_0 = 2^{62}$ and $p_1 = 2^{62} - 1$, which allowed us to represent p_n correctly just using these two moduli.

The calculation of the metric properties requires a lot more memory for the generating functions, and involves multiplication with quite large integers, so in this case we used prime numbers of the form $2^{30} - r_i$ for the moduli p_i . Up to 4 primes were needed to represent the coefficients correctly.

We were able to extend the series for the square lattice SAW generating function from 51 terms to 71 terms using at most 100 Gb of memory. The calculations requiring the most resource were at widths 24–29. These cases were done using 128 processors and took from 16 to 26 h each. We also calculated the metric properties of SAWs up to length 59, thus extending these series from length 32 obtained previously using direct enumeration. In total the calculations used about 50 000 CPU hours.

In table 2 we list the number of SAWs from length 52 to 71. The number of SAWs up to length 51 is tabulated in [12] and [1] (this paper also tabulates the metric properties and several other series). The numbers are also available from our home page.

3. Analysis of the series

To obtain the singularity structure of the generating functions we used the numerical method of differential approximants [10]. The functions have critical points at u_c with exponents as in

Table 2. The number, c_n , of embeddings of *n*-step SAWs on the square lattice. Only terms for n > 51 are listed

п	<i>C</i> _n	п	Cn
52	37 325 046 962 536 847 970 116	62	646 684 752 476 890 688 940 276 172
53	99 121 668 912 462 180 162 908	63	1 715 538 780 705 298 093 042 635 884
54	263 090 298 246 050 489 804 708	64	4 549 252 727 304 405 545 665 901 684
55	698 501 700 277 581 954 674 604	65	12 066 271 136 346 725 726 547 810 652
56	1 853 589 151 789 474 253 830 500	66	31 992 427 160 420 423 715 150 496 804
57	4 920 146 075 313 000 860 596 140	67	84 841 788 997 462 209 800 131 419 244
58	13 053 884 641 516 572 778 155 044	68	224 916 973 773 967 421 352 838 735 684
59	34 642 792 634 590 824 499 672 196	69	596 373 847 126 147 985 434 982 575 724
60	91 895 836 025 056 214 634 047 716	70	1 580 784 678 250 571 882 017 480 243 636
61	243828023293849420839513468	71	4 190 893 020 903 935 054 619 120 005 916

Table 3. Estimates for the critical point u_c and exponent γ obtained from second- and third-order differential approximants to the series for square lattice SAW generating function. *L* is the order of the inhomogeneous polynomial.

	Second-ord	er DA	Third-order DA			
L	<i>u</i> _c	γ	<i>u</i> _c	γ		
0	0.379 052 2679(60)	1.343 735(29)	0.379 052 2735(11)	1.343 7397(18)		
2	0.379 052 2729(11)	1.343 7388(23)	0.379 052 2752(11)	1.343 7427(22)		
4	0.379 052 2720(13)	1.343 7387(32)	0.379 052 2756(27)	1.343 7438(61)		
6	0.379 052 27192(81)	1.343 7369(24)	0.379 052 2751(27)	1.343 7429(61)		
8	0.379 052 2733(15)	1.343 7395(24)	0.379 052 2752(27)	1.343 7434(63)		
10	0.379 052 2739(30)	1.343 744(12)	0.379 052 2751(22)	1.343 7430(39)		
12	0.379 052 2740(19)	1.343 7404(34)	0.379 052 2755(63)	1.343 748(11)		
14	0.379 052 2738(13)	1.343 7398(22)	0.379 052 2738(25)	1.343 7406(37)		
16	0.379 052 2739(12)	1.343 7403(20)	0.379 052 2733(39)	1.343 7408(53)		
18	0.379 052 2734(14)	1.343 7398(25)	0.379 052 2753(19)	1.343 7433(41)		
20	0.3790522749(32)	1.343 7437(87)	0.3790522755(25)	1.343 7435(78)		

(1.1*b*)–(1.5*b*). Our main objective is to obtain accurate estimates for the connective constant μ and the critical exponents γ and ν . In particular we confirm to a very high degree of precision the conjectured exact values of the exponents.

Once the exact values of the exponents have been confirmed we turn our attention to the 'fine structure' of the asymptotic form of the coefficients. In particular we are interested in obtaining accurate estimates for the amplitudes A, C, D and E. We do this by fitting the coefficients to the form given by (1.1a)–(1.5a). In this case our main aim is to test the validity of the predictions for the amplitude combinations mentioned in the introduction.

3.1. The SAW generating function

In table 3 we list estimates for the critical point u_c and exponent γ of the series for the square lattice SAW generating function. The estimates were obtained by averaging values obtained from second- and third-order differential approximants. For each order *L* of the inhomogeneous polynomial we averaged over those approximants to the series which used at least the first 60 terms of the series. The error quoted for these estimates reflects the spread (basically one standard deviation) among the approximants. Note that these error bounds



Figure 3. Estimates for γ versus u_c (left panel) and γ versus the number of terms used by the differential approximant (right panel).

should *not* be viewed as a measure of the true error as they cannot include possible systematic sources of error. Based on these estimates we conclude that $u_c = 0.379052274(4)$ and $\gamma = 1.343745(15)$. The estimate for u_c is compatible with the much more accurate estimate $u_c = 0.37905227773(7)$ obtained from the analysis of the SAP generating function [16]. The analysis adds further support to the already convincing evidence that the critical exponent $\gamma = 43/32 = 1.34375$ exactly. However, we do observe that both the central estimates for both u_c and γ are systematically very slightly lower than the expected values.

When analysing series it is always problematic to get a reliable error estimate. So in trying to confirm, as we are here, the exact value of a critical exponent, it is often useful to plot the behaviour of the estimates against both u_c and the number of terms used by the differential approximants. In this way it is often possible to gauge more clearly whether or not the highorder approximants have settled down to the limiting value of the true exponent. In figure 3 we carry out such an analysis. Each point in the left panel corresponds to estimates for u_c and γ obtained from a third-order differential approximant. The right panel shows the estimates of γ but now plotted against the number of terms used by the differential approximants. The straight lines indicate the expected exact value of γ , and the highly accurate estimates of u_c are obtained from the analysis of the SAP series. From the plot in the right panel we can see that the estimates of γ still exhibit a certain upwards drift as the number of terms in the approximants increases. So the estimates of γ have not yet settled at their limiting value, but there can be no doubt that the predicted exact value of γ is fully consistent with the estimates. In the left panel we observe that the (u_c, γ) -estimates fall in a narrow range. Note that this range does not include the intersection point between the exact γ and the precise u_c estimates. This is probably a reflection of the lack of 'convergence' to the true limiting values. This view is further supported by repeating the plot of figure 3, but only using those approximants using a number of terms at a prescribed interval, which we choose as 51–55, 56–60, 61–65 and 65–71. This corresponds to looking at the plots one would have obtained had the series only been known up to the lengths 55, 60, 65 and 71, respectively. These plots show that as more terms are included the (u_c, γ) -estimates move closer and closer to the expected intersection point. This drift is again a clear indication that the estimates have not yet settled at the true limiting values.

Table 4. Estimates for the critical point u_c and critical exponents obtained from second and third order differential approximants to the series for the end-to-end distance, radius of gyration, and monomer distance from the end point.

		$\mathcal{R}_{e}($	<i>(u)</i>			
	Second-ord	der DA	Third-orde	er DA		
L	<i>u</i> _c	$\gamma + 2\nu$	<i>u</i> _c	$\gamma + 2\nu$		
0	0.379 052 003(90)	2.843 24(62)	0.379 052 101(69)	2.843 33(19)		
2	0.379 051 985(57)	2.843 01(96)	0.379 052 116(58)	2.843 36(11)		
4	0.379 052 046(81)	2.843 45(18)	0.379 052 113(75)	2.843 36(10)		
6	0.379 052 034(80)	2.843 29(39)	0.379 052 119(86)	2.843 34(19)		
8	0.379 052 054(69)	2.8430(19)	0.379 052 115(75)	2.843 37(33)		
10	0.379 052 035(67)	2.843 29(23)	0.379 052 138(66)	2.843 38(11)		
		\mathcal{R}_{g} ((u)			
L	Second-ord	der DA	Third-orde	DA		
3	<i>u</i> _c	$\gamma + 2\nu + 2$	<i>u_c</i>	$\gamma + 2\nu + 2$		
0	0.379 052 2317(19)	4.843 6019(22)	0.379 052 2289(11)	4.843 5986(13)		
2	0.379 052 2317(26)	4.843 6017(29)	0.379 052 2295(10)	4.843 5992(11)		
4	0.379 052 2324(41)	4.843 6024(45)	0.379 052 2289(22)	4.843 5986(23)		
6	0.379 052 2290(94)	4.843 598(11)	0.379 052 2284(23)	4.843 5980(26)		
8	0.379 052 225(15)	4.843 595(16)	0.379 052 2294(43)	4.843 5992(48)		
10	0.379 052 2282(21)	4.843 5978(24)	0.379 052 2301(20)	4.843 6000(24)		
		\mathcal{R}_m	(<i>u</i>)			
	Second-order DA		Third-order DA			
L	<i>u</i> _c	$\gamma + 2\nu + 1$	<i>u</i> _c	$\gamma + 2\nu + 1$		
0	0.379 052 045(58)	3.84321(11)	0.379 052 131(39)	3.843 345(61)		
2	0.379 052 056(37)	3.843 256(63)	0.379 052 118(57)	3.843 327(93)		
4	0.379 052 044(70)	3.843 23(10)	0.379 052 107(68)	3.843 32(10)		
6	0.379 052 050(73)	3.843 22(10)	0.379 052 088(51)	3.843 281(96)		
8	0.379 052 081(99)	3.843 29(17)	0.379 052 081(52) 3.843 274			
10	0.379 052 069(95) 3.843 26(17)		0.379 052 135(55) 3.843 370			

3.2. The metric properties

We now turn our attention to the metric properties. The generating functions are expected to have a singularity at u_c where the end-to-end distance (1.3b) has exponent $\gamma + 2v = 91/32 =$ 2.84375, the radius of gyration (1.4b) has exponent $\gamma + 2v + 2 = 155/32 = 4.84375$, and the mean square monomer distance from the end points (1.5b) has exponent $\gamma + 2v + 1 =$ 123/32 = 3.84375. In table 4 we list the estimates obtained from a differential approximant analysis of these series. In summary we see that applying differential approximants to the metric series yields for the end-to-end distance $u_c = 0.37905205(15)$ and 2v = 2.8434(4), the radius of gyration yields $u_c = 0.379052230(5)$ and 2v + 2 = 4.84360(2), and the monomer distance yields $u_c = 0.3790521(1)$ and 2v + 1 = 3.84335(15). We immediately note that the exponent estimates are systematically lower that the expected exact values. Only the end-to-end distance is marginally consistent with the expected value, while there is a considerable discrepancy between the radius of gyration estimate and the expected value (similar though less pronounced for the monomer distance). However, we also note that the u_c estimates are quite far from the SAP estimates (in which we have considerable confidence) $u_c = 0.379\,052\,277\,73(7)$. So obviously the metric series are not that well behaved and might have large corrections to scaling.

As for the SAW generating function, we find it useful to plot the estimates for the critical exponents versus u_c and the number of terms. This we have done in figure 4. Clearly the estimates from the end-to-end distance have not yet converged and it is quite possible that the exponent estimates will eventually converge to the expected value (see top left panel). Also in the top right panel it is quite possible that the estimates will approach the point given by the intersection of the exact exponent value and the precise u_c value. The behaviour of the estimates for radius of gyration and monomer distance series are far more unsettling. In particular, the exponent estimates from the radius of gyration series appear well converged to a value 4.843 60 with a narrow spread which clearly does not include the expected exact value 4.84375, and in the plot of the exponent versus u_c (middle left panel) the estimates are quite far from the expected intersection. Similar remarks hold for the monomer distance (bottom panels) though convergence and discrepancy with expected values is less pronounced. Furthermore, the behaviour of the radius of gyration series is very different to the other series. In particular we note that in the plots of the exponents versus the number of terms (left panels) the estimates from the end-to-end and monomer distance seems to increase monotonically towards the expected value, while the estimates from the radius of gyration starts out above the expected value, then cross the expected value before apparently settling below the expected value. This behaviour is quite puzzling. Let us just note that if we look at the similar plots for the SAW generating function (figure 3) it is clear that convergence has not been achieved at n = 59 and not yet even at n = 71. It would be therefore surprising if we should not see a further drift in the exponent estimates for the metric properties with longer series.

Fortunately, we have the possibility of analysing other series involving the metric properties. We can look directly at the generating function with coefficients $\langle R_e^2 \rangle_{\rm r}$ and so on. While this has the advantage that we know that these series have a critical point at $u_c = 1$ it turns out that the estimates of the critical exponents, $-2\nu - 1$ in all cases, behave in exactly the same manner as the series for the original generating functions. A second, and as we shall see much more successful approach, is to take the original series and divide them by the SAW generating function. That is we study the series $\mathcal{R}_e(u)/(u * \mathcal{C}(u)) \propto (u - u_c)^{-2\nu}$, and so on, where again $u_c = 1/\mu$. We would not go through all the details here. Suffice to say that applying differential approximants to the resulting series yields for the end-to-end distance series $u_c = 0.37905229(2)$ and 2v = 1.50002(3), the radius of gyration series yields $u_c = 0.3790526(4)$ and $2\nu + 2 = 3.5006(4)$, and the monomer distance series yields $u_c = 0.3790524(2)$ and $2\nu + 1 = 2.5002(4)$. This clearly confirms that $\nu = 3/4$, exactly. We note that the error estimates of the modified series is quite different to the original series. The original end-to-end series has the largest error estimate of the three, while the modified end-to-end series has the smallest error estimate. The opposite happens for the radius of gyration series. The quite different behaviour of these series, as compared to the original ones, is probably even more clearly illustrated in figure 5 where we have plotted the exponent estimates versus the number of terms for the modified end-to-end and radius of gyration series. Clearly, the estimates from both of these series appear not yet to have settled at their limiting values, but it would seem that they are converging towards the expected exponent values. Note the very different behaviour of the original and modified radius of gyration series. So obviously dividing by the SAW generating function has a dramatic effect on the metric series. We can only guess that this procedure leads to modifications in the correction-to-scaling behaviour thus altering dramatically the convergence properties of the series.



Figure 4. Plots of estimates for the critical exponents versus the critical point (left panels) and critical exponents versus number of terms (right panels) obtained from third-order differential approximants to the generating functions $\mathcal{R}_e(u)$ (top panels), $\mathcal{R}_g(u)$ (middle panels) and $\mathcal{R}_m(u)$ (bottom panels).



Figure 5. Plots of estimates of the critical exponents versus the number of terms obtained from the third-order differential approximants to the functions $\mathcal{R}_e(u)/(u\mathcal{C}(u))$ (left panel) and $\mathcal{R}_g(u)/(u\mathcal{C}(u))$ (right panel).

3.3. Non-physical singularities

The generating functions have singularities on the negative axis at $u_{-} = -1/\mu = -x_c$. The exponents at u_{-} are compatible with simple exact values. For the SAW generating function the exponent is 1/2, for the end-to-end generating function (1.3b) the exponent is also 1/2, for the radius of gyration generating function (1.3b) the exponent is -3, while for the monomer distance generating function (1.5b) the exponent is -2.

3.4. Correction-to-scaling exponent

The correction-to-scaling exponent for SAWs is exhaustively studied in a recent paper [1] using series analysis and Monte Carlo simulations. In particular we performed a very careful and detailed analysis of the of the 59 step series for the square lattice SAW counts and metric properties and a 40 step series for the triangular lattice. In that study of the SAW correction-to-scaling exponents, a consistent picture emerged. We presented a compelling evidence that the first non-analytic correction term in the generating function for SAWs and SAPs is $\Delta_1 = 3/2$, as predicted by Nienhuis [25, 26]. We found no evidence for the presence of an exponent $\Delta_1 = 11/16$ in SAWs and SAPs on the square and triangular lattices, as proposed by Saleur [30].

Our method of analysis, both here and in [1], is based on direct fitting to the expected asymptotic form. Obviously (1.1*a*) only gives the leading term in the asymptotic expansion. We have to add in both analytic and non-analytic corrections to scaling. Furthermore, we have to take account of the presence of the singularity at $u_{-} = -1/\mu$. We thus expect c_n to have an asymptotic expansion of the form

$$c_n \sim \mu^n n^{\gamma - 1} \left[a_0 + \sum_{i=1}^k \frac{a_i}{n^{\Delta_i}} \right] + (-\mu)^n n^{\alpha - 2} \left[b_0 + \sum_{i=1}^m \frac{b_i}{n^{\Gamma_i}} \right]$$
(3.1)

where α is the critical exponent occurring in the polygon generating function.

We estimate the coefficients a_i and b_i , by inserting the estimated value of μ , the exact values of γ and α , and assumed values of Δ_i and Γ_i . The coefficients can then be fitted

to the assumed asymptotic form by solving a system of linear equations. By steadily increasing the number of series coefficients, many estimates for the $\{a_i\}$ and $\{b_i\}$ are found. Provided the different estimates are consistent over series of different lengths, we assume that they provide an acceptably accurate estimate of the actual asymptotic coefficients.

A noteworthy feature of the method is that, if a blatantly false exponent is given as input (for example, specifying $\Delta_1 = 1/2$ for the two-dimensional SAW), the sequence of amplitude estimates for the term corresponding to that exponent will converge rapidly to zero, giving a very strong signal that the exponent in question is absent. It was this feature which was used to rule out $\Delta_1 = 11/16$.

Note that one would expect a whole sequence $\Delta_j > \Delta_1$ of non-analytic corrections to scaling, as well as so-called mixing terms involving the exponents γ and α (see [1] for details). The first expected mixing term would give a contribution $n^{-59/32}$ [1]. However, in practice this is indistinguishable from the n^{-2} term. Higher order corrections can likewise not be detected since they make contributions well beyond the range of reasonable extrapolation.

3.5. Amplitudes

In our paper [1] we also obtained accurate amplitude estimates. Here we shall therefore only briefly review these results then report on the slightly improved estimates for the amplitudes of the SAW counts based on the extended 71 term series.

Given the value for the non-analytic correction-to-scaling exponent, we more concretely choose to fit to the form used in [1]:

$$c_n \sim \mu^n n^{11/32} [a_0 + a_1/n + a_2/n^{3/2} + a_3/n^2 + a_4/n^{5/2} + \cdots] + (-1)^n \mu^n n^{-3/2} [b_0 + b_1/n + b_2/n^2 + b_3/n^3 + \cdots].$$
(3.2)

Fitting to this form we found [1], using the 59 term SAW counts, $a_0 = A = 1.1770425(5)$ as well as reasonably accurate estimates for a_1-a_3 and b_0-b_2 . For the metric properties we found C = 0.77124(8), D = 0.108227(58)(5) and E = 0.33913(14). A similar analysis of the triangular lattice data yielded A = 1.183966(2), C = 0.71176(66), D = 0.09987(5) and E = 0.3130(5).

The ratios D/C and E/C were also estimated by direct extrapolation of the relevant quotient sequence, using the following method [27]: given a sequence $\{a_n\}$ defined for $n \ge 1$, assumed to converge to a limit a_{∞} with corrections of the form $a_n \sim a_{\infty}(1 + b/n + \cdots)$, we first construct a new sequence $\{h_n\}$ defined by $h_n = \prod_{m=1}^n a_m$. The generating function $\sum h_n x^n \sim (1 - a_{\infty} x)^{-(1+b)}$. Estimates for a_{∞} and the parameter *b* can then be obtained from differential approximants. In this way, we obtained the estimates [1], D/C = 0.140299(6) and E/C = 0.439647(6) for the square lattice and D/C = 0.140296(6) and E/C = 0.439649(9) for the triangular lattice.

The amplitude estimates leads to a high precision confirmation of the CSCPS relation (1.6) F = 0.000024(25).

In table 5 we have listed the estimates of various universal amplitude combinations, obtained from the work reported in this paper and elsewhere. As can be seen the estimates for all lattices are in perfect agreement strongly confirming the universality of the various combinations.

Finally we turn to the estimation of the amplitude A using the new extended 71 term series. As in previous work [14, 16] we find it very useful to plot the amplitude estimates versus 1/n, where c_n is the last coefficient used by the fit. In figure 6 we plot the estimates for the leading amplitude A from various fits. The legend numbers (k, m) indicate the number of terms used in the fit by each part of the asymptotic expansion (3.1), using the exponents given



Figure 6. Plots of estimates for the leading amplitude A from various fits of c_n to the assumed asymptotic expansion (3.1).

Table 5. Estimates of universal amplitude combinations on some two-dimensional lattices.

Lattice	D/C	E/C	$BC/\sigma a_0$	F
Square [1, 16]	0.140 299(6)	0.439647(6)	0.21683(4)	-0.000 024(28)
Triangular [1, 17]	0.140296(6)	0.439 649(9)	0.2169(2)	-0.000 036(34)
Honeycomb [21]	0.1403(1)	0.4397(2)	0.2170(3)	-0.000 13(67)
Kagomé [22, 23]	0.140(1)	0.440(1)	0.2144(25)	-0.0015(47)

in the explicit form (3.2). From the left panel we see a consistent trend emerging. As the number of terms used in (k, k)-fits is increased we see that the estimates settle down and that fits using more terms display less curvature. We take this as a clear indication that the fitting procedure is robust and that the assumed asymptotic expansion is correct. The fits using 5, 6 and 7 terms each can clearly be extrapolated to a value A = 1.770423(1). In the right panel we plot amplitude estimates from (k, m)-fits with k = 6 and m = 6, 5 and 4. We do this merely to point out that clearly only (k, m)-fits with m close to k are reliable. The (6, 4)-fit displays a pronounced oscillatory behaviour.

4. Summary and conclusion

We have presented a new algorithm for the enumeration of self-avoiding walks. Numerical data show that it has computational complexity only slightly worse than the Conway–Enting–Guttmann algorithm [5]. This means that the CEG algorithm will be superior to enumerating very long SAWs. However, the new algorithm uses much less memory at shorter lengths and remains competitive at lengths attainable at present and in the foreseeable future. Furthermore the new algorithm can be used to calculate metric properties such as the end-to-end distance, radius of gyration and average distance of monomers from the end points. We have used the algorithm to extend the series for the number of SAWs on the square lattice up to 71 steps and calculated the metric properties of SAWs up to 59 steps.

The analysis of the series yielded estimates of the critical exponents γ and ν which confirmed to a high degree of accuracy the predicted exact values $\gamma = 43/32$ and $\nu = 3/4$.

We reported on results from a comprehensive analysis [1] of the series providing very firm and convincing evidence that the leading non-analytic correction to scaling is $\Delta_1 = 3/2$, as well as giving accurate estimates for the critical amplitudes. The amplitude estimates led to a high precision confirmation of the CSCPS relation (1.6) F = 0.

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