Permutations avoiding 1324 and patterns in Łukasiewicz paths

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Abstract
The class \( Av(1324) \), of permutations avoiding the pattern 1324, is one of the simplest sets of combinatorial objects to define that has, thus far, failed to reveal its enumerative secrets. By considering certain large subsets of the class, which consist of permutations with a particularly regular structure, we prove that the growth rate of the class exceeds 9.81. This improves on a previous lower bound of 9.47. Central to our proof is an examination of the asymptotic distributions of certain substructures in the Hasse graphs of the permutations. In this context, we consider occurrences of patterns in Łukasiewicz paths and prove that in the limit they exhibit a concentrated Gaussian distribution.

1. Introduction
We identify a permutation with the sequence of its values. A permutation \( \sigma = \sigma_1 \ldots \sigma_n \) of \( \{1, \ldots, n\} \) is said to avoid a permutation (often referred to as a pattern) \( \pi = \pi_1 \ldots \pi_k \) of \( \{1, \ldots, k\} \) if there is no subsequence of \( \sigma \) that has the same relative order as \( \pi \). The class consisting of those permutations that avoid a permutation \( \pi \) is denoted by \( Av(\pi) \). Due to the celebrated proof of the Stanley–Wilf conjecture by Marcus & Tardos [13], it is known that \( Av(\pi) \) has a finite asymptotic growth rate
\[
\text{gr}(Av(\pi)) = \lim_{n \to \infty} \sqrt[n]{S_n(\pi)},
\]
where \( S_n(\pi) \) is the number of elements of \( Av(\pi) \) of length \( n \). The growth rate of \( Av(\pi) \) is also known as the Stanley–Wilf limit of \( \pi \).

Our interest is in \( Av(1324) \). This is the only class avoiding a pattern of length four that is yet to be enumerated exactly. Moreover, even the growth rate of the 1324-avoiders is currently unknown. In a recent paper, Conway & Guttmann [6] calculate the number of permutations avoiding 1324 up to length 36, building on earlier work by Johansson & Nakamura [11]. They then analyse the sequence of values and give an estimate for the growth rate of \( Av(1324) \) of 11.60 ± 0.01. However, rigorous bounds still differ from this value quite markedly.

The last few years have seen a steady reduction in upper bounds on the growth rate, based on a colouring scheme of Claesson, Jelínek & Steingrimsson [5] which yields a value of 16. Bóna [3] has now reduced this to 13.73718 by employing a refined counting argument.

As far as lower bounds go, Albert, Elder, Rechnitzer, Westcott & Zabrocki [1] have established that the growth rate is at least 9.47, by using the insertion encoding of 1324-avoiders to construct a sequence of finite automata that accept subclasses of \( Av(1324) \). The growth rate of a subclass is then determined from the transition matrix of the corresponding automaton. Our main result is an improvement on this lower bound:

\textbf{Theorem 1.1.} \( \text{gr}(Av(1324)) > 9.81 \).
Figure 1. The Hasse graph of a 1324-avoider, with an up-set and a down-set highlighted.

To each permutation $\sigma$, we associate a plane graph $H_\sigma$, which we call its Hasse graph. To create the Hasse graph for a permutation $\sigma = \sigma_1 \ldots \sigma_n$, let vertex $i$ be the point $(i, \sigma_i)$ in the Euclidean plane. Now, for each pair $i, j$ such that $i < j$, add an edge between vertices $i$ and $j$, if and only if $\sigma(i) < \sigma(j)$ and there is no vertex $k$ such that $i < k < j$ and $\sigma(i) < \sigma(k) < \sigma(j)$. See Figure 1 for an example. Note that the edges of $H_\sigma$ correspond to the edges of the Hasse diagram of the sub-poset, $P_\sigma$, of $\mathbb{N}^2$ consisting of the points $(i, \sigma_i)$. Hasse graphs of permutations were previously considered by Bousquet-Mélou & Butler [4] who determined the algebraic generating function of the class of forest-like permutations whose Hasse graphs are acyclic.

If a permutation avoids the pattern 1324, then its Hasse graph does not have the diamond graph $H_{1324} = \mathcal{D}$ as a minor. In particular, both up-sets and down-sets of the poset $P_\sigma$ are trees. In other words, the subgraph of $H_\sigma$ induced by a left-to-right minimum of $\sigma$ (a minimal element in the poset $P_\sigma$) and the points to its north-east is a tree, as is that induced by a right-to-left maximum of $\sigma$ (a maximal element in the poset $P_\sigma$) and the points to its south-west. See Figure 1 for an illustration.

Figure 2. The plot of a 1324-avoider of length 187 and its Hasse graph.

What does a typical 1324-avoider look like? Figures 2 and 4 contain illustrations of large 1324-avoiders.† As is noted by Flajolet & Sedgewick [8, p.169], the fact that a single example

†The data for Figure 4 was provided by Einar Steingrímsson from the investigations he describes in [14, Footnote 4].
can be used to illustrate the asymptotic structure of a large random combinatorial object can be attributed to concentration of distributions, of which we make much use below in determining our lower bound. Observe the cigar-shaped boundary regions consisting of numerous small subtrees, and also the relative scarcity of points in the interior, which tend to be partitioned into a few paths connecting the two boundaries. Many questions concerning the shape of a typical large 1324-avoider remain to be answered or even to be posed precisely. The recent investigations of Madras & Liu [12] and Atapour & Madras [2] provide a starting point.

We restrict our attention to 1324-avoiders whose Hasse graphs are spanned by a disjoint sequence of trees, rooted at alternate boundaries. In our investigation of how these trees can interact, we consider the asymptotic distribution of certain substructures of the Hasse graphs. In doing so, we exploit the fact that plane trees are in bijection with Łukasiewicz paths. A Łukasiewicz path of length $n$ is a sequence of integers $y_0, \ldots, y_n$ such that $y_0 = 0$, $y_i \geq 1$ for $i \geq 1$, and each step $s_i = y_i - y_{i-1} \leq 1$. Thus, at each step, a Łukasiewicz path may rise by at most one, but may fall by any amount as long as it doesn’t drop to zero or below.

![Figure 3. The plot of a Łukasiewicz path that contains three occurrences of the pattern 1, 0, 1, two of which overlap.](image)

In particular, we investigate the distribution of patterns in Łukasiewicz paths. A pattern $\omega$ of length $m$ in such a path is a sequence of steps $\omega_1, \ldots, \omega_m$ that occur contiguously in the path (i.e. there is some $k \geq 0$ such that $\omega_j = s_{k+j}$ for $1 \leq j \leq m$), with the restriction that the height $\sum_{j=1}^i \omega_j$ after the $i$th step is positive for $1 \leq i \leq m$. Note that multiple occurrences of a given pattern may overlap in a Łukasiewicz path. See Figure 3 for an illustration.

Under very general conditions, substructures of recursively defined combinatorial classes can be shown to be distributed normally in the limit. By generalising the correlation polynomial of Guibas & Odlyzko, and combining it with an application of the kernel method, we prove that patterns in Łukasiewicz paths also satisfy the conditions necessary for asymptotic normality:

**Theorem 1.2.** The number of occurrences of a fixed pattern in a Łukasiewicz path of length $n$ exhibits a Gaussian limit distribution with mean and standard deviation asymptotically linear in $n$.

In the next section, we introduce certain subsets of $Av(1324)$ for consideration, which consist of permutations having a particularly regular structure, and explore restrictions on their structure. We follow this in Section 3 by looking at a number of parameters that record the distribution of substructures in our permutations. Key to our result is the fact that these are asymptotically concentrated, and in this section we prove three of the four concentration results we need. Section 4 is reserved for the proof of Theorem 1.2, concerning the distribution of patterns in Łukasiewicz paths. This section may be read independently of the rest of the paper. To conclude, in Section 5, we use Theorem 1.2 to prove our final concentration result, and then pull everything together to calculate a lower bound for $gr(Av(1324))$, thus proving Theorem 1.1.
Figure 4. The plot of a $1324$-avoider of length 1000 and part of its Hasse graph.
2. **Permutations with a regular structure**

In this section, we present the structure and substructures of the permutations that are the subject of our investigation. Let $\mathcal{W}$ be the set of all permutations avoiding $1324$ whose Hasse graphs are spanned by a sequence of trees rooted alternately at the lower left and the upper right. See Figure 5 for an example.

![Figure 5](image-url)  

**Figure 5.** A permutation in $\mathcal{W}(3, 25, 19, 12)$ and the spanning of its Hasse graph by red and blue trees.

Trees rooted at a left-to-right minimum we colour red, and trees rooted at a right-to-left maximum we colour blue. We refer to these as *red trees* and *blue trees* respectively. As a mnemonic, note that Red trees grow towards the Right and bLue trees grow towards the Left.

Observe that the root of each non-initial blue tree is the uppermost point below the root of the previous red tree, and the root of each non-initial red tree is the leftmost point to the right of the root of the previous blue tree. Note that $\mathcal{W}$ does not contain every $1324$-avoider. For example, $2143 \not\in \mathcal{W}$.

We consider elements of $\mathcal{W}$ with a particularly regular structure. Each red tree has the same number of vertices. Similarly, each blue tree has the same number of vertices. Moreover, every blue tree also has the same root degree. Specifically, for any positive $t$, $k$, $\ell$ and $d$, let $\mathcal{W}(t, k, \ell, d)$ be the set of those permutations in $\mathcal{W}$ which satisfy the following four conditions.

1. Its Hasse graph is spanned by $t + 1$ red trees and $t$ blue trees.
2. Each red tree has $k$ vertices.
3. Each blue tree has $\ell$ vertices.
4. Each blue tree has root degree $d$.

See Figure 5 for an illustration of a permutation in $\mathcal{W}(3, 25, 19, 12)$.

To simplify our presentation, we use the term *blue subtree* to denote a principal subtree of a blue tree. (The principal subtrees of a rooted tree are the connected components resulting from deleting the root.) Thus each blue tree consists of a root vertex and a sequence of $d$ blue subtrees. We also refer to the roots of blue subtrees simply as *blue roots*.

Our goal is to determine a lower bound for the growth rate of the union of all the $\mathcal{W}(t, k, \ell, d)$. To achieve this, we focus on sets in which the number and sizes of the trees grow together along
with the root degree of blue trees. Specifically, we consider the parameterised sets

$$W_{\lambda, \delta}(k) = W(k, k, [\lambda k], [\delta \lambda k]),$$

for some \( \lambda > 0 \) and \( \delta \in (0, 1) \), consisting of \( k + 1 \) \( k \)-vertex red trees and \( k \) \([\lambda k]\)-vertex blue trees each having root degree \([\delta \lambda k]\). Thus, \( \lambda \) is the asymptotic ratio of the size of blue trees to red trees, and \( \delta \) is the limiting ratio of the root degree of each blue tree to its size. Note that, asymptotically, \( 1/\delta \) is the mean number of vertices in a blue subtree. Typically these subtrees are small.

Let \( g(\lambda, \delta) \) denote the upper growth rate of \( \bigcup_k W_{\lambda, \delta}(k) \):

$$g(\lambda, \delta) = \lim_{k \to \infty} \frac{|W_{\lambda, \delta}(k)|}{n(k, \lambda)},$$

where \( n(k, \lambda) = k(k + [\lambda k] + 1) \) is the length of each permutation in \( W_{\lambda, \delta}(k) \). In order to prove Theorem 1.1, we show that there is some \( \lambda \) and \( \delta \) for which \( g(\lambda, \delta) > 9.81 \).

![Figure 6. Valid and invalid horizontal interleavings, and valid and invalid vertical interleavings; occurrences of 1324 are shown with thicker edges.](image)

\( W(t, k, \ell, d) \) consists precisely of those permutations that can be built by starting with a \( k \)-vertex red tree and repeating the following two steps exactly \( t \) times (see Figure 5).

1. Place an \( \ell \)-vertex blue tree with root degree \( d \) below the previous red tree (with its root to the right of the red tree), horizontally interleaving its non-root vertices with the non-root vertices of the previous red tree in any way that avoids creating a 1324.

2. Place a \( k \)-vertex red tree to the right of the previous blue tree (with its root below the blue tree), vertically interleaving its non-root vertices with the non-root vertices of the previous blue tree without creating a 1324.

See Figure 6 for illustrations of valid interleavings of the non-root vertices of red and blue trees, and also of invalid interleavings containing occurrences of 1324. The configurations that have to be avoided when interleaving are shown schematically in Figure 7.

![Figure 7. Possible causes of a 1324 when interleaving horizontally and vertically.](image)
We simply call a valid interleaving of the non-root vertices of a red tree with those of a blue tree an *interleaving* of the trees. Note that the choice of interleaving at each step is completely independent of the interleaving at any previous or subsequent step. The only requirement is that no \textbf{1324} is created by any of the interleavings.

The key to our result is thus an analysis of how vertices of red and blue trees may be interleaved without forming a \textbf{1324}. The remainder of the paper consists of this analysis.

In what follows, we work exclusively with interleavings of red and blue trees in elements of \( W_{\lambda,\delta}(k) \), for some given \( \lambda > 0 \) and \( \delta \in (0,1) \). Thus, we assume, without restatement, that a red tree has \( k \) vertices, and that a blue tree has \( \ell = \lceil \lambda k \rceil \) vertices and is composed of \( d = \lceil \delta \lambda k \rceil \) blue subtrees.

We now consider how to avoid creating a \textbf{1324}. Without loss of generality, we limit our discussion to the horizontal case.

One way to guarantee that no \textbf{1324} is created when interleaving trees is to ensure that no blue subtree is split by a red vertex, since the pattern \textbf{1324} is avoided in any interleaving in which no red vertex occurs between two blue vertices of the same blue subtree. See Figure 8 for an illustration.

Let \( W^0_{\lambda,\delta}(k) \) be the subset of \( W_{\lambda,\delta}(k) \) in which red vertices are interleaved with blue subtrees in this manner in each interleaving.

\( W^0_{\lambda,\delta}(k) \) is easy to enumerate since trees and interleavings can be chosen independently. Indeed,

\[
| W^0_{\lambda,\delta}(k) | = R_k^{k+1} \times B_k^k \times P_k^{2k},
\]

where \( R_k \) is the number of distinct red trees, \( B_k \) is the number of distinct blue trees and \( P_k \) is the number of distinct ways of interleaving red vertices with blue subtrees.

We have \( R_k = \frac{1}{2} \binom{2k-1}{k-1} \), \( B_k = \frac{d}{\ell-1} \binom{2\ell-3-d}{\ell-2} \) (see [8, Example III.8]), and \( P_k = \binom{k-1+d}{d} \). Hence, by applying Stirling’s approximation we obtain the following expression for the growth rate of \( W^0_{\lambda,\delta}(k) \):

\[
g_0(\lambda, \delta) = \lim_{k \to \infty} | W^0_{\lambda,\delta}(k) |^{1/n(k, \lambda)} = E(\lambda, \delta)^{1/(1+\lambda)},
\]

\( (2.1) \)
where
\[ E(\lambda, \delta) = 4 \frac{(2 - \delta)^{2(\lambda)}(1 + \delta \lambda)^{2(1 + \delta \lambda)}}{(1 - \delta)^{(1 - \delta)\lambda}}. \] (2.2)

It is now elementary to determine the maximum value of this growth rate. For fixed \( \lambda \), \( E(\lambda, \delta) \) is maximal when \( \delta \) has the value
\[ \delta_\lambda = \frac{2\lambda - 1 + \sqrt{1 + 4\lambda + 8\lambda^2}}{2\lambda(2 + \lambda)}. \]

Thence, numerically maximising \( g_0(\lambda, \delta) \) by setting \( \lambda \approx 0.61840 \) (with \( \delta_\lambda \approx 0.86238 \)) yields a preliminary lower bound for \( gr(Av(1324)) \) of 9.40399. It is rather a surprise that such a simple construction exhibits a growth rate as large as this.

Form this analysis, we see that we have complete freedom in choosing the positions of the blue roots (roots of blue subtrees) relative to the vertices of the red tree. In the light of this, we divide the process of interleaving into two stages.

1. Freely interleave the blue roots with the red vertices.
2. Select positions for the non-root vertices of each blue subtree, while avoiding the creation of a 1324.

We call the outcome of the first stage a pre-interleaving. A pre-interleaving is thus a sequence consisting of \( k - 1 \) red vertices and \( d = \lceil \delta \lambda k \rceil \) blue vertices (the blue roots); the non-root vertices of the blue subtrees play no role in a pre-interleaving.

Note that in the second stage, each blue subtree can be considered independently since no 1324 can contain vertices from more than one blue subtree. We now consider where the non-root vertices may be positioned.

![Figure 9](image-url)

**Figure 9.** An interleaving of a blue subtree \( T \) with its two-component red forest.

Our first observation is as follows: Suppose \( v \) is the nearest red vertex to the right of the root \( u \) of some blue subtree \( T \). Now let \( x \) be the parent of \( v \) in the red tree. Then no vertex of \( T \) can be positioned to the left of \( x \), since otherwise a 1324 would be created in which \( xuv \) would be the 324. Thus, vertices of \( T \) can only be interleaved with those red vertices positioned between \( u \) and \( x \). We call the graph induced by this set of red vertices (which may be empty) a red forest. See Figure 9 for an illustration.

Our second (elementary) observation is as follows: Suppose \( u \) is the root of some blue subtree \( T \), and \( y \) is the next blue root to the left of \( u \). Then all the non-root vertices of \( T \) must occur to the right of \( y \) (else \( T \) would not be a tree). Note that \( y \) may occur either to the left of \( x \) or to its right. See Figure 10 for illustrations of both of these situations.

These two observations provide two independent constraints on the set of red vertices with which the non-root vertices of a blue subtree may be interleaved, the first determined by the structure of the red tree and the second by the pre-interleaving. This set consists of those
vertices of the red forest situated to the right of both $x$ and $y$. These red vertices induce a subgraph of the red forest which we call its red fringe. In the examples in Figure 10, the red fringes consist of those vertices in the shaded regions. The key fact that motivates the rest of our analysis is that vertices of a blue subtree may only be interleaved with vertices of its red fringe.

The size of a red fringe depends on both the size of the corresponding red forest and also on the location of the next blue root to the left. Let us call the number of red vertices positioned between a blue root $u$ and the next blue root to its left ($y$) the gap size of $u$; the gap size may be zero. The number of vertices in the red fringe is thus the smaller of the gap size and the number of vertices in the red forest.

If we combine this fact with results concerning the limiting distributions of blue subtrees and red fringes, then we can establish a lower bound for $g(\lambda, \delta)$. This is the focus of the next section.

3. Concentration of distributions

To determine our lower bound, we depend critically on the fact that the asymptotic distributions of substructures of permutations in $\mathcal{W}_{\lambda, \delta}(k)$ are concentrated. In this section we introduce certain parameters counting these substructures, show how their concentration enables us to bound $g(\lambda, \delta)$ from below, and prove three of the four concentration results we require.

It is frequently the case that distributions of parameters counting the proportion of particular substructures in combinatorial classes have a convergent mean and a variance that vanishes asymptotically. As a direct consequence of Chebyshev’s inequality, such distributions have the following concentration property:

**Proposition 3.1** [8, Proposition III.3]. If $\xi_n$ is a sequence of random variables with means $\mu_n = E[\xi_n]$ and variances $\nu_n = V[\xi_n]$ satisfying the conditions

$$\lim_{n \to \infty} \mu_n = \mu, \quad \lim_{n \to \infty} \nu_n = 0,$$

for some constant $\mu$, then $\xi_n$ is concentrated at $\mu$ in the sense that, for any $\varepsilon > 0$, given sufficiently large $n$,

$$\mathbb{P}\left[|\xi_n - \mu| \leq \varepsilon\right] > 1 - \varepsilon.$$
In practice this often means that we can work on the assumption that the value of any such parameter is entirely concentrated at its limiting mean. This is the case for the parameters in which we are interested.

We also make use of the following result concerning multiple concentrated parameters.

**Proposition 3.2.** If $\xi_n$ and $\xi'_n$ are two sequences of random variables on the same sample space concentrated at $\mu$ and $\mu'$ respectively, then they are jointly concentrated in the sense that, for any $\varepsilon > 0$, given sufficiently large $n$,

$$
\mathbb{P}[|\xi_n - \mu| \leq \varepsilon \text{ and } |\xi'_n - \mu'| \leq \varepsilon] > 1 - \varepsilon.
$$

**Proof.** For any $\eta > 0$ and sufficiently large $n$, the probability that $\xi_n$ differs from $\mu$ by less than $\eta$ exceeds $1 - \eta$, and similarly for $\xi'_n$ with $\mu'$. Hence the probability that both are simultaneously $\eta$-close to their asymptotic means is at least $1 - 2\eta$. Let $\eta = \varepsilon/2$.  

We now introduce the parameters we need:

**Blue subtrees** $\beta_k$: For each plane tree $T$, let $\beta_k(T)$ be the random variable that records the proportion of blue subtrees in a blue tree that are isomorphic to $T$.

**Gap sizes** $\gamma_k$: For each $j \geq 0$, let $\gamma_k(j)$ be the random variable that records the proportion of blue roots in a pre-interleaving that have gap size $j$. Also, let $\gamma_k(>j)$ record the proportion of blue roots in a pre-interleaving whose gap size exceeds $j$.

**Red forests** $\rho_k$: For each plane forest $F$, let $\rho_k(F)$ be the random variable that records the proportion of positions in a red tree whose red forest is isomorphic to $F$. Also, let $\rho_k(F^+)$ record the proportion of positions in a red tree whose red forest has at least $|F|$ vertices, and for which the graph induced by the rightmost $|F|$ vertices of the forest is isomorphic to $F$.

Below, we prove that each of these parameters is concentrated, and calculate their asymptotic means. First we describe how the parameters are combined.

Our first combined parameter counts red fringes. Given the combination of a red tree and a pre-interleaving of its vertices with a sequence of blue roots, let $\varphi_k(F)$ be the random variable that records the proportion of blue roots whose red fringe is isomorphic to $F$. Now, occurrences of blue roots with a given gap size $j$ are spread almost uniformly across the positions in a red tree, non-uniformity only occurring for the $j$ leftmost positions. This is also the case for the distribution of occurrences of blue roots whose gap size is at least $j$. Hence, by the definition of a red fringe at the end of Section 2, given any $\varepsilon > 0$, if $k$ is large enough, $\varphi_k(F)$ differs from

$$
\gamma_k(|F|)\rho_k(F^+) + \gamma_k(>|F|)\rho_k(F)
$$

by less than $\varepsilon$.

Our second combined parameter concerns pairs consisting of a blue subtree and a red fringe. Given a red tree, a blue tree and a pre-interleaving of their red vertices and blue roots, let $\psi_k(T,F)$ be the random variable that records the proportion of blue subtrees that are isomorphic to $T$ and have a red fringe that is isomorphic to $F$. We call such a blue subtree a $(T,F)$-subtree. Given that occurrences of a given blue subtree are distributed uniformly across the blue roots, we have

$$
\psi_k(T,F) = \beta_k(T)\varphi_k(F).
$$

Since, as we show below, $\beta_k$, $\gamma_k$ and $\rho_k$ are concentrated, it follows that $\psi_k$ is also concentrated. Let $\mu(T,F)$ denote the limiting mean of $\psi_k(T,F)$ as $k$ tends to infinity.

Finally, given a blue subtree $T$ and a red fringe $F$, let $Q(T,F)$ denote the number of distinct ways of interleaving the non-root vertices of $T$ and the vertices of $F$ without creating a $1324$. See Figure 11 for an example.
With all the relevant parameters defined, we are now in a position to present a lower bound on the value of \( g(\lambda, \delta) \).

**Proposition 3.3.** Let \( S \) be any finite set of pairs \((T, F)\) composed of a plane tree \( T \) and a plane forest \( F \). Then

\[
g(\lambda, \delta) \geq E(\lambda, \delta)^{1/(1+\lambda)} \times \prod_{(T, F) \in S} Q(T, F)^{2\delta \mu(T, F)/(1+\lambda)},
\]

where \( E(\lambda, \delta) \) is as defined in (2.2) on page 8.

**Proof.** Consider a red tree and a blue tree together with a pre-interleaving of their red vertices and blue roots. By Propositions 3.1 and 3.2, for any \( \epsilon > 0 \), if \( k \) is large enough, then with probability exceeding \( 1 - \epsilon \), it is the case that

\[
\left| \psi_k(T, F) - \mu(T, F) \right| \leq \epsilon
\]

for every \((T, F) \in S\). So the proportion of pre-interleaved pairs of trees with at least \( \lceil \delta \lambda k \rceil (\mu(T, F) - \epsilon) \) occurrences of \((T, F)\)-subtrees for every \((T, F) \in S\) exceeds \( 1 - \epsilon \).

Elements of \( W_{\lambda, \delta}(k) \) are constructed by independently choosing trees and interleavings. Thus, the size of \( W_{\lambda, \delta}(k) \) is bounded below by

\[
|W_{\lambda, \delta}(k)| \geq |W_{\lambda, \delta}^{\theta}(k)| \times \left( \prod_{(T, F) \in S} (1 - \epsilon)Q(T, F)^{[\delta \lambda k]([\mu(T, F) - \epsilon])} \right)^{2k}.
\]

Recall that

\[
g(\lambda, \delta) = \lim_{k \to \infty} \left| W_{\lambda, \delta}(k) \right|^{1/n(k, \lambda)},
\]

where \( n(k, \lambda) = k + [\lambda k] + 1 \) is the length of each permutation in \( W_{\lambda, \delta}(k) \). The desired result follows after expanding and taking the limit, making use of (2.1). \( \square \)

To determine the asymptotic mean and variance of our parameters, we utilise bivariate generating functions. The following standard result enables us to obtain the required moments directly as long as we can extract coefficients. We use \( [z^n]f(z) \) to denote the coefficient of \( z^n \) in the series expansion of \( f(z) \); we also use \( f_x \) for \( \frac{\partial f}{\partial x} \) and \( f_{xx} \) for \( \frac{\partial^2 f}{\partial x^2} \).

**Proposition 3.4** [8, Proposition III.2]. Suppose \( A(z, x) \) is the bivariate generating function for some combinatorial class, in which \( z \) marks size and \( x \) marks the value of a parameter \( \xi \). Then the mean and variance of \( \xi \) for elements of size \( n \) are given by

\[
E_n[\xi] = \frac{[z^n]A_x(z, 1)}{[z^n]A(z, 1)} \quad \text{and} \quad V_n[\xi] = \frac{[z^n]A_{xx}(z, 1)}{[z^n]A(z, 1)} + E_n[\xi] - E_n(\xi)^2,
\]

respectively.
The proofs of our first three concentration results each follow a similar pattern: establish the generating function; extract the coefficients; apply Proposition 3.4; take limits using Stirling’s approximation; finally apply Proposition 3.1.

First, we consider blue subtrees. Recall that the random variable $\beta_k(T)$ records the proportion of principal subtrees in a $[\lambda k]$-vertex plane tree with root degree $[\delta \lambda k]$ that are isomorphic to $T$.

**Proposition 3.5.** Let $i = |T|$. $\beta_k(T)$ is concentrated at

$$\mu_\beta(T) = \frac{(1-\delta)^{i-1}}{(2-\delta)^{2i-1}}.$$ 

**Proof.** Let $T(z) = \frac{1}{2}(1 - \sqrt{1 - 4z})$ be the generating function for plane trees. Then the bivariate generating function for plane trees with root degree $d$, in which $z$ marks vertices and $u$ marks principal subtrees isomorphic to $T$, is given by

$$B(z, u) = z(T(z) + (u-1)z^i)^d.$$ 

Extracting coefficients yields

$$[z^\ell]B(z, 1) = \frac{d}{\ell-1}\left(\frac{2\ell-d-3}{\ell-2}\right),$$

$$[z^\ell]B_u(z, 1) = \frac{d(d-1)}{\ell-2i-1}\left(\frac{2\ell-2i-d-2}{\ell-i-2}\right),$$

$$[z^\ell]B_{uu}(z, 1) = \frac{d(d-1)(d-2)}{\ell-2i-1}\left(\frac{2\ell-4i-d-1}{\ell-2i-2}\right).$$

Hence, with $\ell = [\lambda k]$ and $d = [\delta \lambda k]$, applying Proposition 3.4 and taking limits gives

$$\lim_{k \to \infty} E[\beta_k(T)] = \frac{(1-\delta)^{i-1}}{(2-\delta)^{2i-1}} \quad \text{and} \quad \lim_{k \to \infty} k \mathbb{V}[\beta_k(T)] = v_\beta(T),$$

where $v_\beta(T)$ is some rational function in $\delta$. So, $\beta_k(T)$ satisfies the conditions for Proposition 3.1 and is thus concentrated at $\mu_\beta(T)$ as required. \(\square\)

Secondly, we consider gap size. Recall that, given a pre-interleaving of the non-root vertices of a $k$-vertex red tree and $[\delta \lambda k]$ blue roots, the random variable $\gamma_k(j)$ records the proportion of blue roots that have gap size $j$. Similarly, $\gamma_k(>j)$ records the proportion that have gap size exceeding $j$.

**Proposition 3.6.** $\gamma_k(j)$ is concentrated at

$$\mu_\gamma(j) = \frac{\delta \lambda}{(1 + \delta \lambda)^j+1}.$$ 

Also, $\gamma_k(>j)$ is concentrated at

$$\mu_\gamma(>j) = \frac{1}{(1 + \delta \lambda)^j+1}.$$ 

**Proof.** The bivariate generating function for pre-interleavings containing $d$ blue roots, in which $z$ marks red vertices and $v$ marks gaps of size $j$, is given by

$$G(z, v) = \frac{z}{1-z} \left(\frac{1}{1-z} + (v-1)z^j\right)^d.$$
Extracting coefficients yields

\[
\begin{align*}
[z^k]G(z, 1) &= \binom{k+d-1}{d}, \\
[z^k]G_v(z, 1) &= \frac{d(k-j+d-2)}{d-1}, \\
[z^k]G_{vv}(z, 1) &= d(d-1)(k-2j+d-3). \end{align*}
\]

Hence, with \(d = \lceil \delta \lambda k \rceil\), applying Proposition 3.4 and taking limits gives

\[
\lim_{k \to \infty} E[\gamma_k(j)] = \delta \lambda (1 + \delta \lambda)^{j+1} \quad \text{and} \quad \lim_{k \to \infty} kV[\gamma_k] = v_\gamma(j),
\]

where \(v_\gamma(j)\) is some rational function in \(\delta\) and \(\lambda\). So, \(\gamma_k(j)\) satisfies the conditions for Proposition 3.1 and is thus concentrated at \(\mu_\gamma(j)\) as required.

Thirdly, we consider red forests. Recall that the random variable \(\rho_k(F)\) records the proportion of positions in a \(k\)-vertex red tree whose red forest is isomorphic to \(F\).

**Proposition 3.7.** Let \(m = |F|\). \(\rho_k(F)\) is concentrated at

\[
\mu_\rho(F) = \frac{1}{2^{2m+1}}.
\]

**Proof.** If \(F\) has \(h\) components, then an occurrence of \(F\) in a red tree comprises the leftmost \(h\) subtrees of some vertex \(x\) that has at least one additional child vertex to the right. See Figure 10 for an illustration. Hence, if \(R\) is the class of red trees augmented by marking occurrences of \(F\) with \(w\), then \(R\) satisfies the structural equation

\[
R = z(\text{Seq}[R] + (w-1)z^m\text{Seq}^+[R]).
\]

So the corresponding bivariate generating function, \(R(z, w)\), satisfies the functional equation

\[
R(z, w) = \frac{z(1 + (w-1)z^mR(z, w))}{1 - R(z, w)},
\]

and hence

\[
R(z, w) = \frac{1}{2} \left(1 + (1 - w)z^{m+1} - \sqrt{(1 + (1 - w)z^{m+1})^2 - 4z}\right).
\]

Extracting coefficients then yields

\[
\begin{align*}
[z^k]R(z, 1) &= \frac{1}{k} \binom{2k-2}{k-1}, \\
[z^k]R_w(z, 1) &= \binom{2k-2m-3}{k-m-1}, \\
[z^k]R_{ww}(z, 1) &= (k - 2m - 2) \binom{2k-4m-4}{k-2m-2}. \end{align*}
\]

Hence, applying Proposition 3.4 and taking limits gives

\[
\lim_{k \to \infty} E[\rho_k(F)] = \frac{1}{2^{2m+1}} \quad \text{and} \quad \lim_{k \to \infty} kV[\rho_k(F)] = v_\rho(F),
\]

where \(v_\rho(F)\) depends only on \(|F|\). So, \(\rho_k(F)\) satisfies the conditions for Proposition 3.1 and is thus concentrated at \(\mu_\rho(F)\) as required. \(\square\)
Our fourth and final concentration result concerns red fringes. Recall that the random variable \( \rho_k(F^+) \) records the proportion of positions in a red tree whose red forest has at least \(|F|\) vertices, and for which the graph induced by the rightmost \(|F|\) vertices of the forest is isomorphic to \(F\).

We would like to determine the bivariate generating function for red trees in which occurrences of the red fringe \(F\) are marked. This is considerably less straightforward than was the case for the other parameters. Primarily, this is because distinct occurrences of \(F\) may overlap. See the left of Figure 12 for an illustration. To achieve our goal, it is convenient to rephrase our problem in terms of Lukasiewicz paths.

Recall from Section 1 that a Lukasiewicz path of length \(n\) is a sequence of integers \(y_0, \ldots, y_n\) such that \(y_0 = 0\), \(y_i \geq 1\) for \(i \geq 1\), and each step \(s_i = y_i - y_{i-1} \leq 1\). It is easy to see that Lukasiewicz paths are in bijection with red trees: visit the vertices of the tree from right to left and let the height of the path be equal to the number of components in the forest induced by the vertices visited so far. Thus, for each leaf vertex, the path contains an up-step, and for each internal vertex with \(r\) children, the path contains a \((1-r)\)-step. See Figure 12 for an illustration.

Recall also that a pattern \(\omega\) of length \(m\) in a Lukasiewicz path is a sequence of contiguous steps \(\omega_1, \ldots, \omega_m\) in the path such that \(\sum_{j=1}^{i} \omega_j > 0\) for \(1 \leq i \leq m\). We do not consider sequences of steps for which the height drops to zero or below. Thus, a pattern in a Lukasiewicz path corresponds to an occurrence of a red fringe in a red tree. Again, see Figure 12, where this is illustrated.

4. Patterns in Lukasiewicz paths

The asymptotic distribution of patterns in words has been investigated before. For an exposition, see [8, Examples I.12, III.26 and IX.13]. The approach taken there makes use of the correlation polynomial of a pattern, introduced by Guibas & Odlyzko in [10] to analyse pattern-matching in strings, and also employs the cluster method of Goulden & Jackson [9]. We refine this approach for use with patterns in Lukasiewicz paths by utilising a generalisation of the correlation polynomial and combining it with an application of the kernel method.

It is readily seen that the bivariate generating function, \(L(z,y)\), for Lukasiewicz paths, in which \(z\) marks length and \(y\) marks height, satisfies the functional equation

\[
L(z,y) = zy + \frac{zy}{1-y} \left( L(z,1) - yL(z,y) \right).
\]  

(4.1)
Given a pattern \( \omega = \omega_1, \ldots, \omega_m \), let us use \( h_i(\omega) = \sum_{j=1}^{i} \omega_j \) to denote the height after the \( i \)th step of \( \omega \), and let us call \( h_m(\omega) \) the final height of \( \omega \).

The correlation polynomial of Guibas & Odlyzko is univariate. For our purposes, we define the bivariate autocorrelation polynomial, \( \hat{a}_\omega(z,y) \), for a pattern \( \omega = \omega_1, \ldots, \omega_m \) in a Lukasiewicz path as follows:

\[
\hat{a}_\omega(z,y) = \sum_{i=1}^{m-1} c_i z^i y^{h_i(\omega)},
\]

where

\[
c_i = \begin{cases} 
1, & \text{if } \omega_{i+1}, \ldots, \omega_m = \omega_1, \ldots, \omega_{m-i}; \\
0, & \text{otherwise.}
\end{cases}
\]

Thus, \( c_i \) records whether \( \omega \) matches itself when shifted (left or right) by \( i \), the variable \( z \) marks the shift, and \( y \) marks the height. For example, \( \hat{a}_{1.0.1.1}(z,y) = z^2 y^2 + z^4 y^4 \).

Given a fixed pattern \( \omega \) of length \( m \) and final height \( h \), we want to determine the trivariate generating function, \( L_\omega(z,y,u) \), for Lukasiewicz paths, where \( u \) marks the number of occurrences of the pattern \( \omega \) in a path. In order to achieve this, we first consider the class of Lukasiewicz paths augmented by distinguishing an arbitrary selection of occurrences of \( \omega \). Let \( M_\omega(z,y,v) \) be the corresponding generating function, in which \( v \) marks distinguished occurrences of the pattern in a path. By the standard inclusion-exclusion principle (see \([8, \text{p.208}]\)), we know that

\[
L_\omega(z,y,u) = M_\omega(z,y,u-1). \quad (4.2)
\]

In order to construct a functional equation for \( M_\omega \), we consider subpaths each consisting of a maximal collection of overlapping distinguished occurrences of \( \omega \). These collections are called clusters. It is readily seen that the generating function for clusters is

\[
C_\omega(z,y,v) = \frac{z^m y^h v}{1 - v \hat{a}_\omega(z,y)}, \quad (4.3)
\]

where \( v \) is used to mark distinguished occurrences of \( \omega \) in a cluster.

Furthermore, we have

\[
M_\omega(z,y,v) = zy + \frac{zy}{1-y} \left( M_\omega(z,1,v) - yM_\omega(z,y,v) \right) + M_\omega(z,y,v)C_\omega(z,y,v), \quad (4.4)
\]

since a path grows either by adding an arbitrary step, as in \((4.1)\), or else by adding a cluster.\(^1\)

Combining equations \((4.2), (4.3)\) and \((4.4)\) and rearranging gives us the following functional equation for \( L_\omega(z,y,u) \):

\[
L_\omega(z,y,u) = \frac{zy(1 + (1-u)\hat{a}_\omega(z,y))(1 - y + L_\omega(z,1,u))}{z^m y^h(1-y)(1-u) + (1-y + y^2)(1 + (1-u)\hat{a}_\omega(z,y))}.
\]

This equation is susceptible to the kernel method, so \( L_\omega(z,1,u) = y_0(z,u) - 1 \), where \( y_0 \) is the appropriate root for \( y \) of the denominator. Rearranging, we obtain the following polynomial functional equation for \( L = L(z,u) = L_\omega(z,1,u) \), the bivariate generating function for Lukasiewicz paths in which \( u \) marks occurrences of \( \omega \):

\[
L = z(1+L)^2 - (1-u) \left( z^m L(1+L)^h + (L - z(1+L)^2)\hat{a}_\omega(z,1 + L) \right). \quad (4.5)
\]

The fact that \( L \) satisfies this equation enables us to demonstrate that patterns in Lukasiewicz paths are concentrated, and moreover are distributed normally in the limit. The following proposition gives very general conditions for this to be the case for some parameter.

\(^1\)This equation excludes distinguished occurrences of \( \omega \) that begin with the first step of the path; this simplifies the algebra somewhat while having no effect on the asymptotics.
Proposition 4.1 ([8, Proposition IX.17 with Theorem IX.12]; see also [7, Theorem 1]). Let $F(z,u)$ be a bivariate function, analytic at $(0,0)$ and with non-negative Taylor coefficients, and let $\xi_n$ be the sequence of random variables with probability generating functions

\[ \frac{[z^n]F(z,u)}{[z^n]F(z,1)}. \]

Assume that $F(z,u)$ is a solution for $y$ of the equation

\[ y = \Phi(z,u,y), \]

where $\Phi$ is a polynomial of degree at least two in $y$, $\Phi(z,1,y)$ has non-negative Taylor coefficients and is analytic in some domain $|z| < R$ and $|y| < S$, $\Phi(0,1,0) = 0$, $\Phi_y(0,1,0) \neq 1$, $\Phi_{yy}(z,1,y) \neq 0$, and there exist positive $z_0 < R$ and $y_0 < S$ satisfying the pair of equations

\[ \Phi(z_0,1,y_0) = y_0, \quad \Phi_y(z_0,1,y_0) = 0. \]

Then, as long as its asymptotic variance is non-zero, $\xi_n$ converges in law to a Gaussian distribution with mean and standard deviation asymptotically linear in $n$.

All that remains is to check that $L$ satisfies the relevant requirements.

Theorem (Theorem 1.2). The number of occurrences of a fixed pattern in a Lukasiewicz path of length $n$ exhibits a Gaussian limit distribution with mean and standard deviation asymptotically linear in $n$.

Proof. From (4.5), it can easily be seen that $L(z,u)$ satisfies the conditions of Proposition 4.1, with $\Phi(z,1,y) = z(1+y)^2$, $z_0 = \frac{1}{4}$ and $y_0 = 1$. \hfill \qed

5. Summing up

Since patterns in Lukasiewicz paths are in bijection with red fringes in red trees, $L(z,u)$ is also the bivariate generating function for red trees in which $u$ marks occurrences of the red fringe $F$ corresponding to the pattern $\omega$, with $m = |F|$ and $h$ the number of components of $F$. Thus, we know that $\rho_k(F^+)$ is concentrated. It remains for us to determine the limiting mean.

Proposition 5.1. Let $m = |F|$ and $h$ be the number of components of $F$. $\rho_k(F^+)$ is concentrated at

\[ \mu_{\rho}(F^+) = \frac{1}{2^{2m-h}}. \]

Proof. Let $F(z) = \frac{1}{2z}(1 - \sqrt{1 - 4z})$ be the generating function for plane forests. Solving (4.5) with $u = 1$ gives $L(z,1) = F(z) - 1$ (as expected).

Similarly, differentiating (4.5) with respect to $u$, setting $u = 1$, and solving the resulting equation gives

\[ L_u(z,1) = \frac{z^m F(z)^h (1 - (1 - 2z) F(z))}{1 - 4z}. \]

Then, extracting coefficients yields

\[ [z^k]L(z,1) = \frac{1}{k+1} \binom{2k}{k}, \quad [z^k]L_u(z,1) = \binom{2k-2m+h}{k-m-1}. \]
Hence, applying Proposition 3.4 and taking limits,
\[
\lim_{k \to \infty} \mathbb{E}[\rho_k(F^+)] = \frac{1}{2^{2m-n}}.
\]
Concentration follows from Theorem 1.2.

We are finally in a position to compute a lower bound for the growth rate of the class of permutations avoiding 1324, proving our main theorem.

**Theorem (Theorem 1.1).** \( \text{gr}(\text{Av}(1324)) > 9.81. \)

**Proof.** We calculate the contribution to the growth rate from pairs consisting of a tree and a forest of bounded size. From Proposition 3.3, we know that, for each \( N > 0 \), the growth rate is at least
\[
g_N(\lambda, \delta) = E(\lambda, \delta)^{1/(1+\lambda)} \times \prod_{|T|+|F| \leq N} Q(T, F)^{2\delta \mu(T, F)/(1+\lambda)},
\]
where
\[
\mu(T, F) = \mu_\beta(T)(\mu_\gamma(|F|)\mu_\rho(F^+) + \mu_\gamma(|F|)\mu_\rho(F)),
\]
as follows from (3.1) and (3.2) and Propositions 3.5, 3.6, 3.7 and 5.1.

Using Mathematica to evaluate \( Q(T, F) \) and \( \mu(T, F) \) and then to apply numerical maximisation over values of \( \lambda \) and \( \delta \) yields
\[
g_{14}(\lambda, \delta) > 9.81056
\]
with \( \lambda \approx 0.69706 \) and \( \delta \approx 0.75887 \).

The determination of this value requires the processing of more than 1.6 million pairs consisting of a tree and a forest. Larger values of \( N \) would require more sophisticated programming techniques. However, increasing \( N \) is unlikely to lead to a significantly improved lower bound; although the rate of convergence at \( N = 14 \) is still quite slow, numerical analysis of the computational data suggests that \( \lim_{N \to \infty} \max_{\lambda, \delta} g_N(\lambda, \delta) \) is probably not far from 9.82.

We conclude with the observation that in the construction that gives our bound, the mean number of vertices in a blue subtree, \( 1/\delta \), is less than 1.32. We noted earlier that the cigar-shaped boundary regions of a typical 1324-avoider contain numerous small subtrees (although it is not immediately obvious how one should identify such a boundary tree). Is it the case that the mean size of these subtrees is asymptotically bounded? Perhaps, on the contrary, their average size grows unboundedly (but very slowly), and understanding how (and the rate at which) this occurs would lead to an improved lower bound. In the meantime, the following question might be somewhat easier to answer:

**Question.** Asymptotically, what proportion of the points in a typical 1324-avoider are left-to-right minima or right-to-left maxima?

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**References**


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