

Fast RSK correspondence by doubling search

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Abstract

The Robinson–Schensted–Knuth (RSK) correspondence is a fundamental concept in combinatorics and representation theory. It is defined as a certain bijection between permutations and pairs of Young tableaux of a given order. We consider the RSK correspondence as an algorithmic problem, along with the closely related k -chain problem. We give a simple, direct description of the symmetric RSK algorithm, which is implied by the k -chain algorithm of Felsner and Wernisch. We also show how the doubling search of Bentley and Yao can be used as a subroutine by the symmetric RSK algorithm, replacing the default binary search. Surprisingly, such a straightforward replacement improves the asymptotic running time for the RSK correspondence that has been best known since 1998. A similar improvement also holds for the average running time of RSK on uniformly random permutations.

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1 Introduction

The Robinson–Schensted–Knuth (RSK) correspondence is a fundamental concept in combinatorics and representation theory; for the background on the combinatorial aspects of RSK, see e.g. [18, 16]. It is defined as a certain bijection between pairs of standard Young tableaux and permutations of a given order, and represents a far-reaching generalisation of the longest increasing subsequence problem in a permutation. A common definition of RSK correspondence is algorithmic, via Robinson–Schensted tableau insertions or, alternatively, via the Viennot geometric construction.

The combinatorial properties of RSK are well-studied. In this paper, we consider the RSK correspondence as an algorithmic problem, along with the closely related k -chain problem. In particular, we are interested in both the worst-case and the average asymptotic running time of algorithms for these problems. This aspect of the RSK correspondence seems to have been studied relatively less thoroughly than its combinatorial aspects.

In the rest of this paper, we describe the standard RSK algorithm by Robinson and Schensted (which by itself is the most common definition of the RSK correspondence). We then give a simple, direct description of the symmetric RSK algorithm, which is implied by the k -chain algorithm of Felsner and Wernisch [9]. Further, we recall the doubling search algorithm of Bentley and Yao [1], and show how it can be used as a subroutine by the symmetric RSK algorithm, replacing the default binary search. Surprisingly, such a straightforward replacement improves the asymptotic worst-case running time for the RSK correspondence from $O(n^{3/2} \log n)$, which has been best-known since [9], to $O(n^{3/2})$. A similar improvement also holds for the average running time of RSK on uniformly random permutations.

2 The RSK correspondence

Planar dominance We will use the standard terminology related to partial orders: *downset*, *principal downset*, *chain*, *antichain*. We will also consider two specific (strict) partial orders between points on the Euclidean plane.



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23:2 Fast RSK correspondence by doubling search

44 ► **Definition 1.** Point (x, y) is dominated (respectively, anti-dominated) by point (x', y') , if
45 $x \leq x'$ and $y \leq y'$ (respectively, $y \geq y'$). Whenever $x \neq x'$, $y \neq y'$, we denote dominance by
46 $(x, y) \ll (x', y')$, and anti-dominance by $(x, y) \leq (x', y')$

47 When considering a point set P as a partial order, we will assume the dominance order, unless
48 indicated otherwise. We will also always assume that all x -coordinates in P are distinct, and
49 so are all the y -coordinates.

50 **Young tableaux** Let \mathbb{N}_+ denote the set of all positive integers. Given $n \in \mathbb{N}_+$, let $\mathbb{N}_n =$
51 $\{1, \dots, n\} \subset \mathbb{N}_+$.

52 ► **Definition 2.** A Young diagram of order n is a subset of \mathbb{N}_+^2 of cardinality n , that is a
53 downset in the dominance partial order. A Young tableau¹ of order n is an order-preserving
54 bijection from a Young diagram of order n (called the tableau's shape) to a subset of \mathbb{R} of
55 cardinality n .

56 We use the so-called French notation for visual representation of Young diagrams and tableaux.
57 The elements of a diagram are represented by cells of an integer grid, arranged in left-aligned
58 rows and bottom-aligned columns. Column indices increase from left to right, and row indices
59 from below upwards. The value of each cell of a tableau is written within that cell; these
60 values increase from left to right in rows, and from below upwards in columns.

61 ► **Example 3.** Figure 1 (middle, right) gives several examples of Young tableaux.

62 **Canonical antichain partitioning** The theory of Young tableaux is intimately connected
63 with the combinatorics of permutations. We take a symmetric view of this connection,
64 due to Viennot [20, 21]. A permutation $\pi : \mathbb{N}_n \rightarrow \mathbb{N}_n$ is identified with the point set
65 $P_\pi = \{(x, \pi(x)) \mid x \in \mathbb{N}_n\}$.

66 ► **Definition 4.** The height of an element in a finite partial order is the maximum cardinality
67 of a chain in the principal downset generated by that element. A canonical antichain is formed
68 by all the elements of a given height. The partitioning of a partial order P into disjoint
69 canonical antichains is called the canonical antichain partition (CAP), denoted $\text{cap}(P)$.

70 Canonical antichains are also sometimes called *layers of minima* (*maxima*) [4, 3], *Pareto*
71 *fronts* [5], or *terraces* [13]. The canonical antichain partition is also sometimes called the
72 partial order's *greedy cover* [11], *patience sorting* [2], or *non-dominated sorting* [5].

73 ► **Example 5.** Figure 1 (top-left) shows a point set P of cardinality 10, and the partitioning
74 $\text{cap}(P)$ into five antichains.

75 We recall the following classical theorem, relied upon by [9].

76 ► **Theorem 6** (see e.g. [9]). *Partitioning $\text{cap}(P)$ has the minimum possible number of*
77 *antichains among all antichain partitionings of P . This number is also equal to the maximum*
78 *cardinality of a chain in P .*

79 **Proof.** See e.g. [11]; the connection between antichain partitions and chains is via Dilworth's
80 theorem. ◀

¹ Young tableaux as defined here are often called “standard”, to distinguish them from more general types of tableaux; we omit this qualifier, since it is the only type of Young tableaux we are dealing with.

81 The problem of finding the cardinality of $\text{cap}(P)$ is equivalent to the problem of finding the
 82 length of a longest increasing subsequence (LIS) in a corresponding permutation. The LIS
 83 problem has a long history, going back to Erdős and Szekeres [8] and Robinson [14]. Based
 84 on their ideas, a classical LIS algorithm running in time $O(n \log n)$ was made explicit by
 85 Knuth [12], Fredman [10] and Dijkstra [6].

86 ► **Definition 7.** Let $A = \{(x_1, y_1) \leq (x_2, y_2) \leq \dots \leq (x_r, y_r)\}$ be an antichain of cardinality
 87 r . Values x_1 and y_r will be called respectively the head and the tail of A . The skeleton of A
 88 is the antichain $\text{sk}(A) = \{(x_2, y_1) \leq (x_3, y_2) \leq \dots \leq (x_r, y_{r-1})\}$ of cardinality $r - 1$. Given
 89 a point set P , the skeleton of P is the point set $\text{sk}(P) = \bigcup_{A \in \text{cap}(P)} \text{sk}(A)$.

90 **The RSK correspondence** The Robinson–Schensted–Knuth (RSK) correspondence, dis-
 91 covered independently by Robinson [14] and Schensted [17] (see also Romik [16]), is a bijection
 92 between permutations of a given order and pairs of Young tableaux of the same order and
 93 identical shape. The two tableaux in the pair will be called the *head* and the *tail* tableaux
 94 (such a terminology is chosen for its symmetry and consistency with the rest of our exposition,
 95 whereas the traditional terminology calls them the *recording* and the *insertion* tableaux).

96 ► **Definition 8.** Let P be a set of points. The RSK image² of P is a pair of Young tableaux
 97 $\text{rsk}(P) = (H, T)$, defined recursively as follows. The initial row in H (respectively, T) is
 98 formed by the heads (respectively, the tails) of the antichains in $\text{cap}(P)$. The remaining rows
 99 of H, T are formed as $\text{rsk}(\text{sk}(P))$.

100 ► **Example 9.** Figure 1 (top) shows the construction of the initial rows in tableaux H (top
 101 middle) and T (top right) from a set of points $P \subseteq \mathbb{N}_{10}^2$ (top left). Figure 1 (middle, bottom)
 102 shows the recursive construction of the remaining rows in the tableaux H, T .

103 ► **Definition 10.** Let P be a set of points. The transpose of P is the point set obtained by
 104 exchanging the x - and y -coordinates of each point: $P^* = \{(y, x) \mid (x, y) \in P\}$. The x -reversal
 105 (respectively, y -reversal) of P is the point set obtained by negating the first (respectively, the
 106 second) coordinate: $P^\mp = \{(-x, y) \mid (x, y) \in P\}$, $P^\pm = \{(x, -y) \mid (x, y) \in P\}$.

107 ► **Observation 11.** Let π be a permutation. We have $P_\pi^* = P_{\pi^{-1}}$.

108 ► **Proposition 12.** Let P be a set of points, $\text{rsk}(P) = (H, T)$. We have (i) $\text{rsk}(P^*) = (T, H)$,
 109 (ii) $\text{rsk}(P^\pm) = (H, T')$, (iii) $\text{rsk}(P^\mp) = (H', T)$, where H', T' are some Young diagrams.

110 **Proof.** Statement (i) is obvious by symmetry.

111 Let us establish statement (iii). Let (x_0, y_0) be the point with the least y -coordinate in
 112 $\text{sk}(P)$, so y_0 is the tail of the least-height antichain in $\text{sk}(P)$. Let A be an antichain in P
 113 such that $(x_0, y_0) \in \text{sk}(A)$. Then, there is a pair of points $(x, y_0) \leq (x_0, y)$ in A . The subset
 114 of P with y -coordinate less than y_0 must consist of a single chain including point (x_0, y)
 115 (otherwise, there would be two points with y -coordinate less than y_0 on some antichain B of
 116 P , and then $\text{sk}(B)$ would contain a point of $\text{sk}(P)$ with y -coordinate less than y_0). Consider
 117 the points corresponding to this chain in P^\mp , with x -coordinates negated. These points,
 118 including point $(-x_0, y)$, form a subset of the least height antichain in P^\mp . Point $(-x, y_0)$

² The terms “RSK correspondence”, “RSK image” as defined here are often called just “Robinson–Schensted”, reserving the name “RSK” for a more general type of combinatorial bijection. Since this is an algorithmic study, we use the term RSK throughout, in order to highlight the contribution of Donald Knuth to the development of RSK algorithms.

23:4 Fast RSK correspondence by doubling search

■ **Table 1** Standard RSK

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1: procedure RSK( $P$ )           ▷ given point set  $P$  sorted by  $x$ -coordinate, returns  $rsk(P)$ 
2:    $H_1 \leftarrow \emptyset; T_1 \leftarrow \emptyset$            ▷ initialise variables for initial rows of  $H, T$ 
3:    $S \leftarrow \emptyset$                                    ▷ initialise variable for  $sk(P)$ 
4:   while  $P \neq \emptyset$  do
5:      $(x, y) \leftarrow$  point in  $P$  with least  $x$ -coordinate
6:      $y' \leftarrow$  least value in  $T_1$  greater than  $y$ ;  $+\infty$  if none exists           ▷ binary search
7:     if  $y' = +\infty$  then
8:       append  $x$  to  $H_1$ ; append  $y$  to  $T_1$            ▷ start new antichain
9:     else
10:      replace  $y'$  by  $y$  in  $T_1$ ; append  $(x, y')$  to  $S$            ▷ extend antichain
11:      remove  $(x, y)$  from  $P$ 
12:       $(H_+, T_+) \leftarrow$  RSK( $S$ )           ▷ recursive call
13:       $H \leftarrow$  tableau with initial row  $H_1$  and remaining rows  $H_+$ 
14:       $T \leftarrow$  tableau with initial row  $T_1$  and remaining rows  $T_+$ 
15:      return  $(H, T)$            ▷  $rsk(P) = (H, T)$ 

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119 must belong to the second-least height antichain in P^\mp , and must be the \leq -minimal point
 120 on that antichain. Therefore, y_0 is the tail of the second-least height antichain in P^\mp .

121 We have established that the tail of the least height antichain in $sk(P)$ is equal to the
 122 tail of the second-least height antichain in P^\mp . Now (iii) follows by (i) and the recursive
 123 construction of Definition 8, and (ii) follows from (iii) by symmetry. ◀

124 **Multichains** The following definition and proposition are not essential for this paper.
 125 However, we include them for highlighting the connection with [9].

126 ▶ **Definition 13.** *Let P be a set of points. A k -chain is a subset of P that can be represented
 127 as a union of k chains.*

128 ▶ **Proposition 14** ([9]). *Let P be a set of points. The maximum cardinality of a k -chain in
 129 P is equal to the number of cells in the initial k rows of $rsk(P)$.*

130 3 RSK algorithms

131 **Standard RSK algorithm** The classical definition of the RSK correspondence leads directly
 132 to an algorithm for computing the RSK image of a given point set.

133 Given a point set P , the pair of tableaux $rsk(P) = (H, T)$ are constructed by rows. To
 134 obtain the initial rows of H, T , the points in P are scanned in order of increasing x -coordinate.
 135 For the subset P' of points seen so far, we maintain the partitioning $cap(P')$; in particular,
 136 the heads and the tails of antichains in that partitioning are kept in sorted order. We also
 137 maintain the skeleton $sk(P')$ in order of increasing x -coordinate. When the scan of P is
 138 complete, the heads (respectively, tails) of antichains in $cap(P)$ become the initial row of
 139 tableau H (respectively, T) in $rsk(P)$. To obtain the remaining rows of $rsk(P)$, we repeat
 140 the above procedure on point set $sk(P)$. Table 1 gives the algorithm's pseudocode.

141 ▶ **Example 15.** Figure 1 shows the execution of the standard RSK algorithm in three
 142 successive iterations: the point set at the beginning of each iteration and its CAP (left), and
 143 the state of the tableaux H and T at the end of the respective iteration (middle, right).

144 The computation of the initial row in the standard RSK algorithm (before the recursive
 145 call in line 13 of Table 1) is essentially identical to the classical algorithm for the LIS problem
 146 [12, 10, 6]. The processing of each of n points in set P is done by binary search, therefore
 147 the whole initial row is obtained in time $O(n \log n)$. In total, there are at most n rows in
 148 $rsk(P)$, therefore the overall time is $n \cdot O(n \log n) = O(n^2 \log n)$.

149 Romik [15] analysed the average-case running time of the standard RSK algorithm on
 150 a uniformly random permutation (in this case, the shape of diagrams H, T follows the
 151 *Plancherel* distribution). The average-case running time in this setting is $O(n^{3/2} \log n)$.

152 **Symmetric RSK algorithm** Felsner and Wernisch [9] proposed a more efficient, symmetric
 153 version of the RSK algorithm. Their algorithm was described in the language of k -chains.
 154 In particular, they gave two algorithms for computing maximum k -chains (and, by sym-
 155 metry, also k -antichains) of a planar point set: one with running time $O(kn \log n)$, another
 156 $O((n^2/k) \log n)$. By running both algorithms simultaneously, maximum k -chains can be
 157 obtained in time $O(n^{3/2} \log n)$ for all k .

158 Here, we give a simpler, more direct description of this algorithm as an extension of the
 159 standard RSK algorithm. The main idea of the symmetric RSK algorithm is to construct the
 160 pair of tableaux $rsk(P) = (H, T)$ simultaneously by rows and by columns. The successive
 161 rows of tableaux H, T are constructed as in the standard RSK algorithm. At the same time,
 162 the successive columns in tableau H (respectively, T) are obtained by running the standard
 163 RSK algorithm on point set P^\pm (respectively, P^\mp). The correctness of the symmetric RSK
 164 algorithm follows from Proposition 12.

165 There is clearly some redundancy in running the standard algorithm three times on
 166 point sets P, P^\pm, P^\mp . However, this constant-factor redundancy allows one to reduce the
 167 overall asymptotic running time. Notice that after the first iteration of each of the three
 168 runs, we have obtained the union of the initial row and initial column in each of H and T ;
 169 this union is called the initial *principal hook* of the respective tableau. Likewise, after the
 170 second iteration, we obtain the second principal hook of both H and T (i.e. the union of the
 171 second row and column, minus the initial principal hook). Crucially, while the number of
 172 both rows and columns in a Young tableau of order n can be as high as n , the number of its
 173 principal hooks is always at most $n^{1/2}$. Thus, the algorithm can be terminated after at most
 174 $\lceil n^{1/2} \rceil$ iterations made by each of the three simultaneous runs on P, P^\pm, P^\mp . The worst-case
 175 running time of the symmetric RSK algorithm is $n^{1/2} \cdot O(n \log n) = O(n^{3/2} \log n)$.

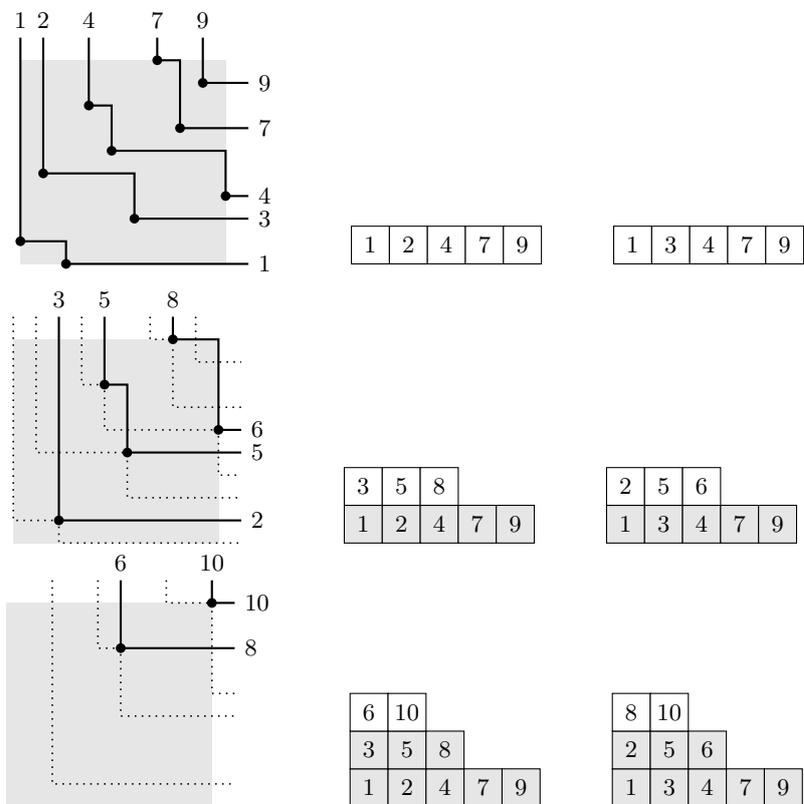
176 ► **Example 16.** Figure 2 shows the execution of the symmetric RSK algorithm on the same
 177 input point set as in Figure 1 (only the computation of tableau T on point sets P, P^\mp is
 178 shown explicitly, while the symmetric computation of tableau H on point sets P, P^\pm is
 179 omitted.) Compared to the three iterations of the standard algorithm in Figure 1, now only
 180 two iterations are required.

181 4 Speeding up RSK by doubling search

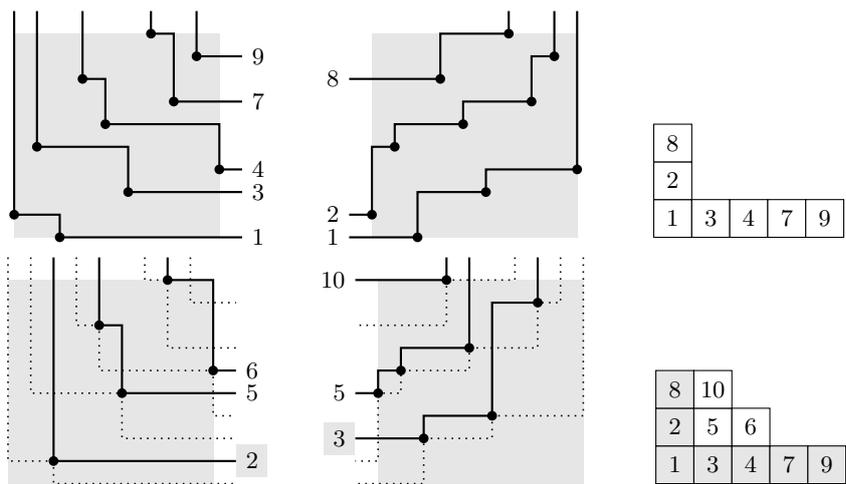
182 **Doubling search** The *doubling search* technique (also called *exponential search*) was intro-
 183 duced by Bentley and Yao [1], and represents a hybrid between linear and binary search.
 184 Doubling search is guaranteed to be as fast asymptotically as each of the binary and the
 185 linear search alone, even for a non-uniform distribution of the target index, skewed towards
 186 an end of the array being searched.

187 We describe doubling search with the starting point at the upper end of the array, in
 188 order to be consistent with its intended application as a subroutine for RSK. Given an array

23:6 Fast RSK correspondence by doubling search



■ Figure 1 The standard RSK algorithm: running on P , returning tableaux H, T



■ Figure 2 The symmetric RSK algorithm: shown running on P, P^\mp only, returning T

■ **Table 2** Doubling search

```

1: procedure DSEARCH( $a, q$ )      ▷ given sorted array  $a$  and  $q$ , returns index for  $q$  in  $a$ 
2:   if  $a_s < q$  then return 0
3:    $t \leftarrow 1$ 
4:   while  $a_{s-t} > q$  do  $t \leftarrow 2t$       ▷ doubling phase
5:   return index  $k$  in  $\{1, \dots, t\}$ , such that  $a_{s-k} < q < a_{s-k+1}$       ▷ binary search

```

189 a_i , $1 \leq i \leq s$, and a value q distinct from all a_i , we consider the problem of finding the
 190 greatest value in a less than q , that is index $k \geq 0$ such that $a_{s-k} < q < a_{s-k+1}$. We assume
 191 $a_i = -\infty$ for $i \leq 0$, and $a_{s+1} = +\infty$.

192 The search begins at the upper end of the array, comparing q against a_s . If $a_s < q$, we
 193 have found $k = 0$. Otherwise, the search continues in two phases. In the *doubling phase*, we
 194 compare q against a_{s-1} , a_{s-2} , a_{s-4} , a_{s-8} , \dots , until we find a subtrahend t that is the least
 195 power of 2 such that $a_{s-t} < q$. This phase takes $\lfloor \log k \rfloor + 1$ comparisons.

196 We now know that $k \in \{1, \dots, t\}$, and move on to the *binary search phase*. In this phase,
 197 we find the exact value of k in this range by binary search, taking at most $\lfloor \log t \rfloor \leq \lfloor \log k \rfloor + 1$
 198 comparisons. Overall, the doubling search algorithm takes at most $2\lfloor \log k \rfloor + 3$ comparisons.
 199 Table 2 shows the pseudocode for the doubling search algorithm.

200 **Symmetric RSK with doubling search** We now present a simple improvement to the
 201 symmetric RSK algorithm, replacing binary search within each iteration by doubling search.

202 Consider a specific value x for a point's x -coordinate, as the RSK algorithm iterates on
 203 P , $sk(P)$, $sk(sk(P))$, etc. These iterations form respectively row 1, 2, 3, \dots of diagrams
 204 H , T . Let l_r denote the length of row r before a point with coordinate x is processed
 205 for that row. Let b_r denote the insertion index in row r of a point with coordinate x
 206 (as per lines 8 or 10 of Table 1); b_r is undefined if no point with coordinate x is left in
 207 iteration r (that is, in the $r - 1$ -th skeleton of P). Let the *displacement interval* in row r
 208 be $\{b_r, b_r + 1, \dots, \min(l_r, b_{r-1} - 1)\}$, where $b_{-1} = +\infty$. We denote this interval's length by
 209 $d_r = \min(l_r + 1, b_{r-1}) - b_r$; in particular, $d_r = 0$ if $b_r = b_{r-1}$. We also define $d_r = 0$ if b_r is
 210 undefined due to no point with coordinate x being left in iteration r .

211 ► **Theorem 17.** *The symmetric RSK algorithm with doubling search solves the RSK corres-*
 212 *pondence problem in worst-case time $O(n^{3/2})$.*

213 **Proof.** Without loss of generality, assume that n is a perfect square (otherwise, the input
 214 can be extended by extra points with a suitably high y -value). Let $m = n^{1/2}$.

215 For a fixed x -coordinate, consider the displacement interval in a given row r . The rectangle
 216 of tableau cells below and including this interval consists of rd_r cells. All these rectangles
 217 for different values of r are pairwise disjoint. The symmetric RSK algorithm terminates
 218 after processing at most m rows. The total number of cells in the rectangles defined by the
 219 displacement intervals in these rows is obviously at most n :

$$220 \quad \sum_{r=1}^m rd_r \leq n$$

222 While searching for an insertion index of a point with coordinate x into row r , doubling
 223 search makes at most $\lfloor 2 \log d_r \rfloor + 3$ comparisons. For the total number of comparisons made
 224 for the given x -coordinate, we have by the arithmetic-geometric mean inequality and the

225 Stirling lower bound on the factorial (cancelling the rounding down of the logarithms, and
 226 omitting the constant factor 2 and the additive term $\sum_{r=1}^m 3 = O(m)$):

$$\begin{aligned}
 227 \quad \sum_{r=1}^m \log d_r &= \sum_{r=1}^m \log \frac{rd_r}{r} = \log \prod_{r=1}^m \frac{rd_r}{r} = \log \left(\frac{1}{m!} \prod_{r=1}^m rd_r \right) \leq \\
 228 \quad \log \left(\frac{1}{m!} \left(\frac{1}{m} \sum_{r=1}^m rd_r \right)^m \right) &\leq \log \frac{(n/m)^m}{m!} = \log \frac{m^m}{m!} \leq \log \frac{m^m}{(m/e)^m} = m \log e = O(m) \\
 229
 \end{aligned}$$

230 There are n different x -coordinates to consider, therefore the algorithm makes $n \cdot O(m) =$
 231 $O(n^{3/2})$ comparisons in total. \blacktriangleleft

232 5 Conclusion

233 We have given a simple, direct description of the symmetric RSK algorithm by Felsner and
 234 Wernisch [9]. We have shown how the default binary search in this algorithm can be replaced
 235 by doubling search, improving the asymptotic running time from $O(n^{3/2} \log n)$ to $O(n^{3/2})$.
 236 It should also be noted that the (worst-case) running time of our algorithm is lower than the
 237 average-case running time of the standard (or the symmetric) RSK algorithm on uniformly
 238 random permutations, as analysed by Romik [15]. Our result implies a similar improvement
 239 for the k -chain problem for arbitrary k .

240 A natural lower bound on the running time of RSK correspondence is provided by the
 241 LIS problem, which is a subproblem for RSK, and requires $\Omega(n \log n)$ comparisons in the
 242 comparison model [10]. Thus, there remains a substantial gap between the known upper and
 243 lower bounds for the asymptotic complexity of the RSK correspondence.

244 Apart from potential improvements in the algorithm or the lower bound, there is scope
 245 for future work in extending the algorithm for more general versions of the RSK correspond-
 246 ence, e.g. that between positive integer matrices and semistandard Young tableaux. An
 247 experimental confirmation of the efficiency of our algorithm also remains an endeavor for
 248 future work; this is a non-trivial task, since most existing experiments with RSK, e.g. those
 249 by Vasilyev and Duzhin [7, 19], concentrate on either Plancherel-random Young diagrams, or
 250 on Young diagrams with (near-)maximum dimensions; such a diagram shape seems to be far
 251 away from the worst-case shape suggested by the proof of Theorem 17.

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