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# **Ropes: Theory and practice**

# Why and when to use Ropes for Java for string manipulations

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Systems that manipulate large quantities of string data are poorly served by the Java™ language's default String and StringBuilder classes. A *rope* data structure can be a better alternative. This article introduces Ropes for Java, a rope implementation for the Java platform; explores performance issues; and provides pointers for effective use of the library.

A *rope data structure* represents an immutable sequence of characters, much like a Java string. But ropes' highly efficient mutations make ropes — unlike <u>strings</u> and their mutable <u>stringBuffer</u> and <u>stringBuilder</u> cousins — ideal for applications that do heavy string manipulation, especially in multithreaded environments.

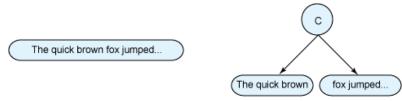
After briefly summarizing the rope data structure, this article introduces Ropes for Java, a rope implementation for the Java platform. Then it benchmarks string, stringBuffer, and the Ropes for Java Rope class on the Java platform; investigates some special performance issues; and concludes with a discussion about when (and when not) to use ropes in your applications.

# **Ropes: A brief overview**

A *rope* represents an immutable character sequence. A rope's *length* is simply the number of characters in the sequence. Most string representations store all their characters contiguously in memory. The defining characteristic of a rope is that it does away with this restriction, instead allowing fragments of the rope to reside noncontiguously and joining them using *concatenation nodes*. This design allows concatenation to run asymptotically faster than for Java strings. The string version of concatenation requires strings you want to join to be copied to a new location, which is an O(n) operation. The rope counterpart simply creates a new concatenation node, which is an O(1) operation. (If you're unfamiliar with big-O notation, see Related topics for a link to explanatory material.)

Figure 1 illustrates two types of string representations. In the one on the left, the characters are located in contiguous memory locations. Java strings rely on this representation. In the representation on the right, a disjointed string is combined using a concatenation node.

Figure 1. Two representations of a string



Rope implementations also often defer evaluation of large substring operations by introducing a *substring node*. Use of a substring node reduces the time for extracting a substring of length n from O(n) to  $O(\log n)$ , and often to O(1). It is important to note that Java <u>strings</u>, being immutable, also have a constant-time substring operation, but <u>stringBuilders</u> often do not.

A *flat rope* — a rope with no concatenation or substring nodes — has a *depth* of 1. The depth of concatenation and substring ropes is one more than the depth of the deepest node they enclose.

Ropes have two overheads that contiguous-character string representations do not. The first is that a superstructure of substring and concatenation nodes must be traversed to reach a specified character. Furthermore, this tree superstructure must be kept as balanced as possible to minimize traversal times, implying that ropes need occasional rebalancing to keep read performance good. Even when ropes are well balanced, obtaining the character at a specified position is an  $O(\log n)$  operation, where n is the number of concatenation and substring nodes the rope comprises. (For convenience, the rope's length can be substituted for n, because the length is always greater than the number of substring and concatenation nodes in the rope.)

Luckily, rope rebalancing is fast, and the determination of when to rebalance can be made automatically, for example by comparing the rope's length and depth. And, in most data-processing routines, sequential access to a rope's characters is what's required, in which case a rope iterator can provide amortized O(1) access speed.

Table 1 compares the expected run times of some common string operations on both ropes and Java Strings.

Table 1. Expected run times of common string operations

Operation	Rope	Java String
Concatenation	O(1)	O(n)
Substring	O(1)	O(1)
Retrieve character	O(log n)	O(1)
Retrieve all characters sequentially (cost per character)	O(1)	O(1)

# **Introducing Ropes for Java**

### Memory issues

Constant-time substring implementations in Java code can cause memory problems because the substring reference prevents the original string from being garbage collected. Both Rope and String suffer from this problem.

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Ropes for Java is a high-quality implementation of the rope data structure for the Java platform (see Related topics). It implements a wide range of optimizations to help provide excellent all-around performance and memory usage. This section explains how to integrate ropes into a Java application and compares rope performance with that of String and StringBuffer.

The Ropes for Java implementation exposes a single class to clients: Rope. Rope instances are created from arbitrary charsequences using the Rope.BUILDER factory builder.

Listing 1 shows how to create a rope.

### **Listing 1. Creating a rope**

```
Rope r = Rope.BUILDER.build("Hello World");
```

Rope exposes a standard battery of methods for manipulation, including methods to:

- · Append another character sequence
- Delete a subsequence
- Insert another character sequence into a rope

Keep in mind that, as with string, each of these mutations returns a new rope; the original is left unmodified. Listing 2 illustrates some of these operations.

### **Listing 2. Rope mutations**

```
Rope r = Rope.BUILDER.build("Hello World");
r = r.append("!"); // r is now "Hello World!"
r = r.delete(0,6); // r is now "World!"
```

## **Efficient rope Iteration**

Iterating over a rope requires some care. After examining the two code blocks in Listing 3, see if you can determine which one performs better.

# Listing 3. Two techniques for iterating over a rope

```
//Technique 1
final Rope r=some initialization code;
for (int j=0; j<r.length(); ++j)
    result+=r.charAt(j);

//Technique 2
final Rope r=some initialization code;
for (final char c: r)
    result+=c;</pre>
```

Recall that returning a single character at an arbitrary position within a rope is an  $O(\log n)$  operation. However, by using charAt for each character, the first code block in Listing 3 pays the  $O(\log n)$  lookup time n times. The second block, which uses an Iterator instead, should perform faster than the first. Table 2 summarizes the performance of iterating over a rope of

length 10,690,488 using both approaches. For comparison, Table 2 includes times for string and StringBuffer.

**Table 2. Complex rope iteration performance** 

Technique	Time (ns)
String	69,809,420
StringBuffer	251,652,393
Rope.charAt	79,441,772,895
Rope.iterator	1,910,836,551

The rope used to obtain the results in Table 2, which are in line with expectations, was created by performing a complex series of mutations to an initial string. However, if the rope is created directly from a character sequence, without any subsequent mutations, the performance numbers take a surprising twist. Table 3 compares the performance of both approaches, this time by iterating over all 182,029 characters of a rope initialized directly from a character array containing the Project Gutenberg edition of Charles Dickens' *A Christmas Carol*.

Table 3. Rope iteration performance

Technique	Time (ns)
String	602,162
StringBuffer	2,259,917
Rope.charAt	462,904
Rope.iterator	3,466,047

How can this performance reversal be explained in light of my earlier theoretical discussion? The rope's construction is a key factor: When a rope is constructed directly from an underlying characterSequence or character array, it has a simple structure consisting of a single flat rope. Because the rope contains no concatenation or substring nodes, character lookup consists of direct delegation to the underlying sequence's charat method (in the CharacterSequence case) or a direct array lookup (in the array case). The performance of Rope charat for a flat rope matches that of the underlying representation, which is most often O(1); hence the performance difference.

Astute readers might wonder why charAt is more than seven times faster than the iterator, given that both provide O(1) access time. This difference is due to the fact that in the Java language, Iterators must return Objects. Although the charAt implementation directly returns char primitives, the iterator implementation must box each char into a character object. Autoboxing might sooth the syntactic pain of primitive boxing, but it can't eliminate the performance penalties.

Finally, it is remarkable that the performance of Rope.charAt is better than the performance of String.charAt. The reason is that Rope represents lazy substrings using a dedicated class, allowing the implementation of charAt to remain simple for regular ropes. In contrast, the Java SE implementation of String uses the same class to represent regular strings and lazy substrings,

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which somewhat complicates the logic of charAt and thereby adds a small performance penalty during iteration over regular strings.

Rope iteration will resurface when I discuss optimizing the performance of regular-expression searches on ropes.

### **Optimizing output with Rope.write**

At some point, most rope instances must be written somewhere, usually to a stream. Unfortunately, writing an arbitrary object to a stream invokes the object's tostring method. This approach to serialization forces a string representation of the entire object to be created in memory before a single character can be written, which is a big performance drag for large objects. Ropes for Java was designed with large string objects in mind, so it provides a better approach.

To improve performance, Rope introduces a write method that accepts a writer and a range specification and writes the rope's specified range into the writer. This saves the time and memory cost of constructing a string from the rope, which, for large ropes, is enormous. Listing 4 shows both the standard and enhanced approaches to rope output.

### Listing 4. Two approaches to Rope output

```
out.write(rope);
rope.write(out);
```

Table 4 contains the benchmark results of writing a rope with length 10,690,488 and depth 65 to a stream backed by an in-memory buffer. Note that only time savings are shown, but the savings to temporarily allocated heap space are much larger.

Table 4. Rope output performance

Technique	Time (ns)
out.write	75,136,884
rope.write	13,591,023

### Benchmarking mutator performance

Rope mutations are, in theory, much faster than those of contiguous-character string representations. On the flip side, as you have seen, rope iteration can be slower as a result. In this section, you'll look at some benchmarks comparing the mutation performance of Ropes for Java with Strings and StringBuffers.

All the tests are initialized with the Project Gutenberg EBook of *A Christmas Carol* (182,029 bytes) and apply a successive series of mutations to it. In most cases, I varied the number of mutations from 20 to 480 to provide a picture of how well the data structures scale. (Figures 2, 3, and 4 refer to this as the *plan length*.) I performed each test seven times and used the median result.

#### **Insert benchmark**

In the insert benchmark, a substring was selected at random from the previous iteration's output and inserted at a random location in the string. Figure 2 shows the benchmark results.

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Insert Plan Length vs Median Time (ns) 20,000,000,000 18,000,000,000 16,000,000,000 14,000,000,000 12,000,000,000 Time (ns) 10,000,000,000 -StringBuffer 8,000,000,000 <u></u>
★ Rope 6,000,000,000 4,000,000,000 2,000,000,000 100 200 300 400 500 600 Insert Plan Length

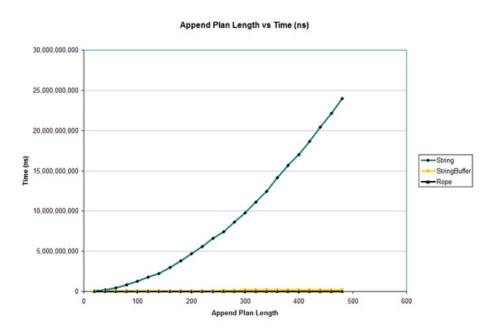
Figure 2. Insert benchmark results

For both strings and stringBuffers, the amount of time required to complete the benchmark rises exponentially as the plan length increases. Ropes, in contrast, perform extremely well.

### **Append benchmark**

The append benchmark consists of appending a random range of the input string to itself. This test is interesting because stringBuffers are optimized for performing fast appends. Figure 3 shows the benchmark results.

Figure 3. Append benchmark results

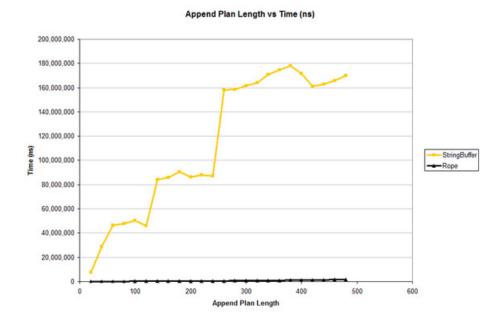


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Unfortunately, the view in Figure 3 is obscured by the atrocious performance of <u>string</u>. However, as expected, <u>stringBuffer</u> performance is excellent.

Figure 4 takes string out of the equation and rescales the chart to show more clearly how Rope and StringBuffer performance compare.

Figure 4. Append benchmark results, excluding String



### Rope immutability: A caveat

The Rope class is designed to be immutable, meaning that mutator functions never modify the original Rope instance, but, rather, return a modified copy. Immutability confers a number of benefits, the chief one being that Rope instances can be shared without precaution in a multithreaded environment, thereby greatly simplifying program logic.

Ropes for Java constructs ropes from an underlying character sequence. Because CharSequence is an interface, you gain a great deal of flexibility: a rope can be constructed from heterogeneous sources including files on disk, in-memory buffers, and documents stored on remote servers. However, unlike String, which is guaranteed to be immutable, CharSequence places no such restriction on its implementers. It is the application programmer's responsibility to ensure that the underlying character sequence used to build a rope is effectively immutable for the rope instance's lifespan.

Figure 4 shows a dramatic difference between Rope and StringBuffer that was not apparent in Figure 3. Rope barely manages to rise above the x-axis. However, what is really interesting is the graph of StringBuffer —it jumps and plateaus instead of rising smoothly. (As an exercise, try to explain why before reading the next paragraph.)

The reason has to do with how <u>StringBuffers</u> allocate space. Recall that they allocate extra space at the end of their internal array to allow for efficient appending. But after that space is exhausted, a brand new array must be allocated and all the data copied to it. The new array is generally sized as some factor of the current length. As the plan length increases, so does the length of the

resulting stringBuffer. And as resize thresholds are reached, the time taken jumps as an extra resize and copy occurs. Then performance plateaus (that is, time taken rises slowly) for the next several plan length increases. Because each plan item increases the total string length by roughly the same amount, the ratio of subsequent plateaus provides an indicator of the resize factor for the underlying array. Based on the results, an accurate estimate for this particular stringBuffer implementation is somewhere around 1.5.

### **Results summary**

So far, I've relied on graphs to illustrate the performance differences among Rope, String, and StringBuffer. Table 5 provides the timing results for all mutation benchmarks, using 480-item plan lengths.

Table 5. Median mutation times with a 480-item plan, plus delete results

Mutation/data structure	Time (ns)
Insert	
String	18,447,562,195
StringBuffer	6,540,357,890
Rope	31,571,989
Prepend	
String	22,730,410,698
StringBuffer	6,251,045,63
Rope	57,748,641
Append	
String	23,984,100,264
StringBuffer	169,927,944
Rope	1,532,799
Delete (delete 230 random ranges from initial text)	
String	162,563,010
StringBuffer	10,422,938
Rope	865,154

See Related topics for a link to the benchmark program. I encourage you to run it on your platform and verify the results I'm presenting here.

# Optimizing regular-expression performance

Regular expressions, introduced into the JDK in version 1.4, are a widely used feature in many applications. Therefore, it is critical that they perform well on ropes. In the Java language, regular expressions are represented as Patterns. To match a Pattern against regions of an arbitrary CharSequence, you use the pattern instance to construct a Matcher object, passing the CharSequence as a parameter.

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The ability to work with charsequences gives the Java regular-expression library excellent flexibility, but it also introduces one serious drawback. A charsequence provides only a single method for accessing its characters: charAt(x). As I mentioned in the overview section, random character access on a rope with many internal nodes is approximately  $O(\log n)$ , so traversal is  $O(n \log n)$ . To illustrate the problem that this causes, I benchmarked the time it takes to find all instances of the pattern "crachit\*" in a string with length 10,690,488. The rope used was constructed through the same series of mutations as the insert benchmark and has a depth of 65. Table 6 shows the results.

Table 6. Regular-expression search times

Technique	Time (ns)
String	75,286,078
StringBuffer	86,083,830
Rope	12,507,367,218
Rebalanced Rope	2,628,889,679

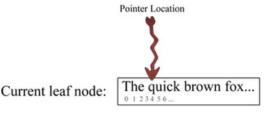
Clearly, the matching performance of a Rope is poor. (To be clear, this is true for a Rope with many internal nodes. For a flat rope, the performance is nearly identical to that of the underlying character representation.) Even explicitly rebalancing the Rope, which makes matching 3.5 times faster, doesn't put Ropes in the same league as either Strings or StringBuffers.

To improve this situation, Matchers should and can harness the much faster O(1) access times that Rope iterators provide. To grasp how this works you first need to understand the access pattern of a Matcher to its CharSequence.

The most common scenario for matching regular expressions is to start at some point in a character sequence and move forward until all matches have been found and the end of the sequence is reached. In this scenario, the matcher mainly moves in a forward direction, often by more than one character at a time. Occasionally, however, the matcher is forced to backtrack.

You can easily modify the Rope iterator to accommodate skipping forward by more than one character at a time. But moving backward is more complicated, because internally the iterator performs a depth-first, preorder traversal of the Rope, visiting each of its leaf nodes. The traversal stack doesn't have enough information to move to the previous leaf, but if the amount to move back does not take the iterator off of the current leaf it is possible to service the request. To illustrate, Figure 5 shows the state of a hypothetical iterator that can move back one, two, three, or four places, but no more, because that would require accessing the previously visited leaf.

Figure 5. Sample iterator state



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To enable this new capability, the Rope's charAt method can be modified so that, on the first invocation, an iterator is constructed at the specified position. Subsequent invocations move the iterator backward or forward by the required distance. If the iterator can't move backward the required distance, then the default charAt routine is executed to obtain the character value — a scenario that one hopes will occur rarely.

Because this optimization is not generally applicable and because it requires the introduction of new member variables, it is best not to add it directly to the Rope class. Instead, you can decorate the rope with this optimization on demand. To let you accomplish this, Ropes for Java includes a method on the Rope class that produces an optimized matcher for a specified pattern. Listing 5 illustrates this approach:

### Listing 5. Optimized regular-expression matching

```
Pattern p = ...
Matcher m = rope.matcher(p);
```

The call on the second line in Listing 5 decorates the rope to optimize regular-expression matching.

Table 7 provides benchmark results for this approach, with the previous results (from Table 6) included for clarity.

Table 7. Regular-expression search times with rope.matcher()

Technique	Time (ns)
String	75,286,078
StringBuffer	86,083,830
Rope	12,507,367,218
Rebalanced Rope	2,628,889,679
Rope.matcher	246,633,828

The optimization results in a significant 10.6x improvement over the rebalanced rope and brings the rope to within a factor of 3.2 times the string's performance.

# **Applications**

# When not to use ropes

```
Often, enterprise Java applications contain code similar to the following: String x = "<input type='text'
name='name' value='"
+ escapePcData(bean.getName())
+ "'>";
```

x is subsequently sent as part of an HTML page to the client's browser. Does it make sense to use ropes to compute the value of x, rather than a StringBuilder, which is what the compiler generates by default?

The answer is no, for a variety of reasons. To begin with, the amount of data being concatenated is, presumably, small, so using a rope is unlikely to improve performance, although it could improve robustness and scalability. (Consider how both solutions would behave if getName unexpectedly returned a 50 MB string.)

But for the sake of argument, imagine that many more chunks of data are being concatenated. Given that Rope's append performance is generally better than StringBuffer's, would a rope make sense then? Again, no. Whenever input data is being combined to produce formatted output, the cleanest and most efficient approach is to use a template engine such as StringTemplate or FreeMarker. Not only does this approach cleanly separate presentation markup from code, but templates are compiled once (often to JVM bytecode) and then can be reused, giving them excellent performance characteristics.

A second benefit of using templates exposes a fundamental flaw common to output-building routines like those in the code above, including one written using ropes. This benefit is that templates can be incrementally evaluated, and output can be written to a writer as soon as it is produced, rather than unnecessarily first being accumulated in memory. In a Java EE application, where the writer is actually a buffered connection to the client's browser, this approach to output rendering uses constant memory, O(1), versus the other solutions' O(n) memory usage. This is a *huge* improvement to application scalability and robustness, even though it's not apparent for smaller inputs or at lower application loads. (See Related topics for links to two articles on streaming architecture that further explain and quantify this approach.)

Now that you have a good understanding of rope performance, it's time to consider some traditional uses for ropes as well as a tempting, but likely, inappropriate use in Java EE applications.

Although ropes can serve as a general-purpose replacement for contiguous-memory representations of strings, only applications that extensively modify large strings will see a significant performance improvement. Perhaps it is not surprising, then, that one of the earliest applications of ropes was to represent documents in a text editor. Not only can text insertions and deletions be performed in near-constant time for extremely large documents, but ropes' immutability makes implementation of an undo stack trivial: simply store a reference to the previous rope with every change.

Another, more esoteric, application of ropes is to represent state in a virtual machine. For example, the ICFP 2007 Programming Contest involved implementing a virtual machine that modified its state with every cycle and ran for millions of cycles for some inputs (see Related topics). In one Java implementation, the virtual machine's speed was improved by three orders of magnitude, from ~50 cycles/sec to more than 50,000/sec, by changing the state representation to use a Rope instead of a specialized StringBuffer.

# **Directions for future investigation**

Although Ropes for Java is a new library, the underlying concepts are old, and the library appears to deliver on the performance promise of ropes. However, the project plans to improve some aspects of the library in future releases by:

- Providing high-performance implementations of other common string operations.
- Writing adapters to integrate ropes seamlessly into Scala (see Related topics) and other advanced languages for the Java platform.

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• Enhancing quality through further automated testing. Ropes for Java is tested using both manually written automated JUnit tests, as well as automatically generated automated tests using JUnit Factory. Incorporating Java Modeling Language (JML) annotations checked by ESC/Java2 (see Related topics) might further improve quality.

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# **Related topics**

- Ropes for Java: Visit the project Web site.
- Big-O notation: Computational complexity theory uses big-O notation to describe how inputdata size affects an algorithm's use of computational resources.
- The 10th ICFP Programming Contest: The author's entry in this contest demonstrated Ropes for Java's productivity.
- "Streaming Architecture" and "Streaming Presidents" (Amin Ahmad, Ahmadsoft.org): Detailed explanation and quantification of streaming architectures.
- The Scala programming language: Functional programming for the JVM.
- The busy Java developer's guide to Scala (Ted Neward, developerWorks): Learn Scala from the ground up.
- JUnit: and JUnitFactory: Ropes for Java is tested with JUnit and JUnitFactory.
- JML and ESC/Java2: ESC/Java2 is a programming tool that tries to find common run-time errors in JML-annotated Java programs.
- Ropes for Java: Download the latest release.
- PerformanceTest.java: The performance benchmark code used to test mutator performance for this article. You can download and run this code to obtain personalized results for your platform.
- Download IBM product evaluation versions and get your hands on application development tools and middleware products from DB2®, Lotus®, Rational®, Tivoli®, and WebSphere®.

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