

BigDebug: Interactive Debugger for Big Data Analytics in Apache Spark

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Developing Big Data Analytics

- Data scientists gather insights from massive quantities of data using data intensive scalable computing systems



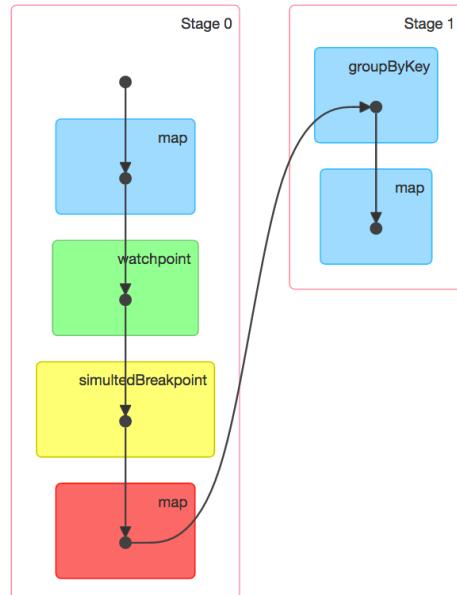
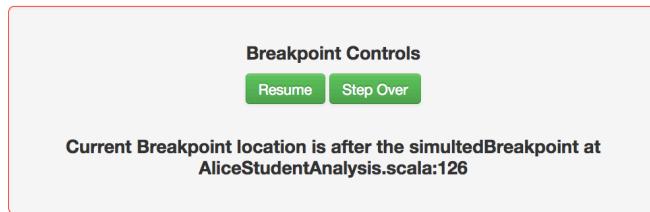
- Debugging of big data analytic workflows on the cloud is time consuming and error-prone

Traditional Debugging in Big Data Analytics

Enabling interactive debugging requires us to **re-think the features of traditional debugger** such as GDB

- Pausing the entire computation on the cloud could reduce throughput
- It is clearly infeasible for a user to inspect billion of records through a regular watchpoint
- Even launching remote JVM debuggers to individual worker nodes cannot scale for big data computing

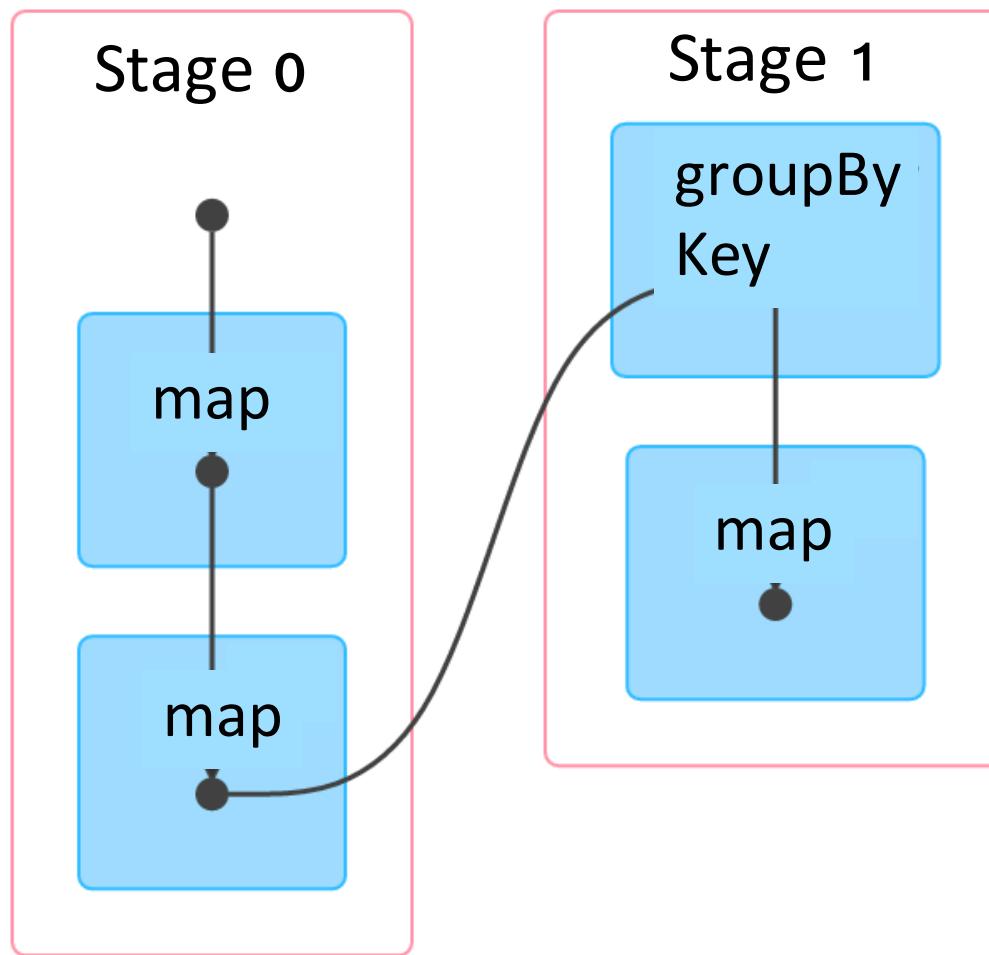
BigDebug : Interactive Debugger for Apache Spark [ICSE 2016]



AliceStudentAnalysis.scala

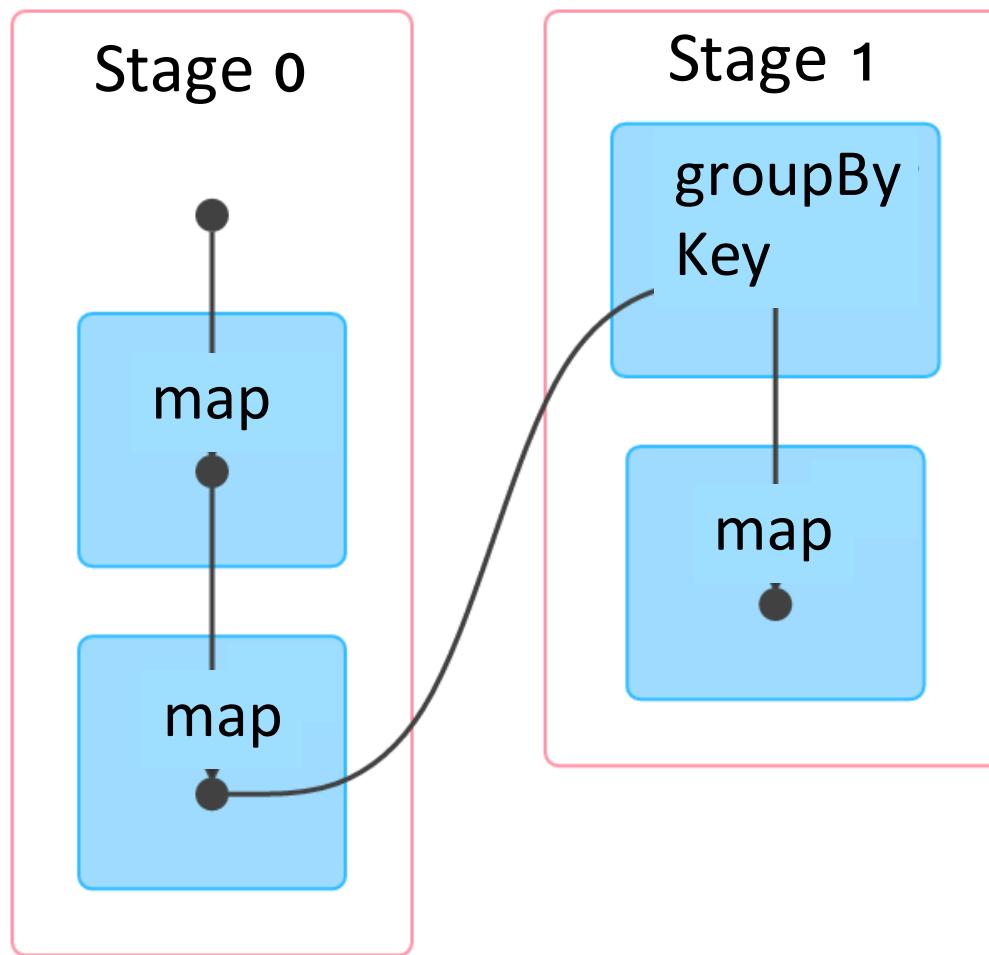
```
110
111 object AliceStudentAnalysisWP {
112
113     val COLLEGEYEAR = List("Sophomore" , "Freshman" , "Junior", "Senior")
114     def main(args: Array[String]): Unit = {
115
116         //set up spark configuration
117         val sparkConf = new SparkConf()
118         val bdconf = new BigDebugConfiguration
119         bdconf.setFilePath("/home/ali/work/temp/git/dsbigdebug/spark-lineage/examples/src/main"
120         //set up spark context
121         val ctx = new SparkContext(sparkConf)
122         ctx.setBigDebugConfiguration(bdconf)
123         //spark program starts here
124         val records = ctx.textFile("/home/ali/Desktop/myfile.txt", 1)
125             .watchpoint(s=> !COLLEGEYEAR.contains(s.split(" ")(2)))
126             .simultedBreakpoint()
127 >         val grade_age_pair = records.map(line => {
128             val list = line.split(" ")
129             (list(2), list(3).toInt)
130         })
131         val average_age_by_grade = grade_age_pair.groupByKey
132             .map(pair => {
133                 val itr = pair._2.toIterator
134                 var moving_average = 0
135                 var num = 1
136                 while (itr.hasNext) {
137                     moving_average = moving_average + itr.next()
138                     num = num + 1
139                 }
140                 (pair._1, moving_average/num)
141             })
142         val out = average_age_by_grade.collect()
143         out.foreach(println)
144 }
```

Feature 1: Simulated Breakpoint



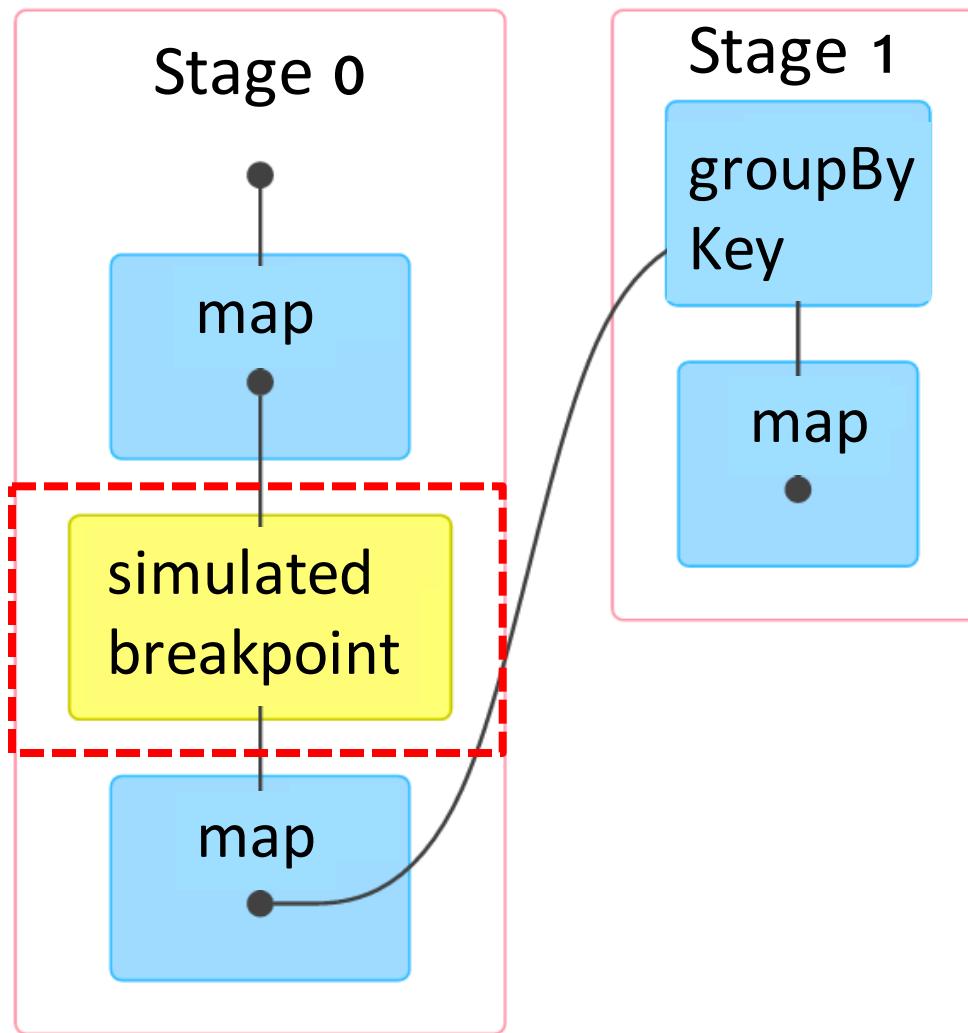
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125     val records = ctx.textFile("/home/ali/Desktop/grades.txt")
126     .simulatedBreakpoint()
127 >     val grade_age_pair = records.map(line => {
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Feature 1: Simulated Breakpoint



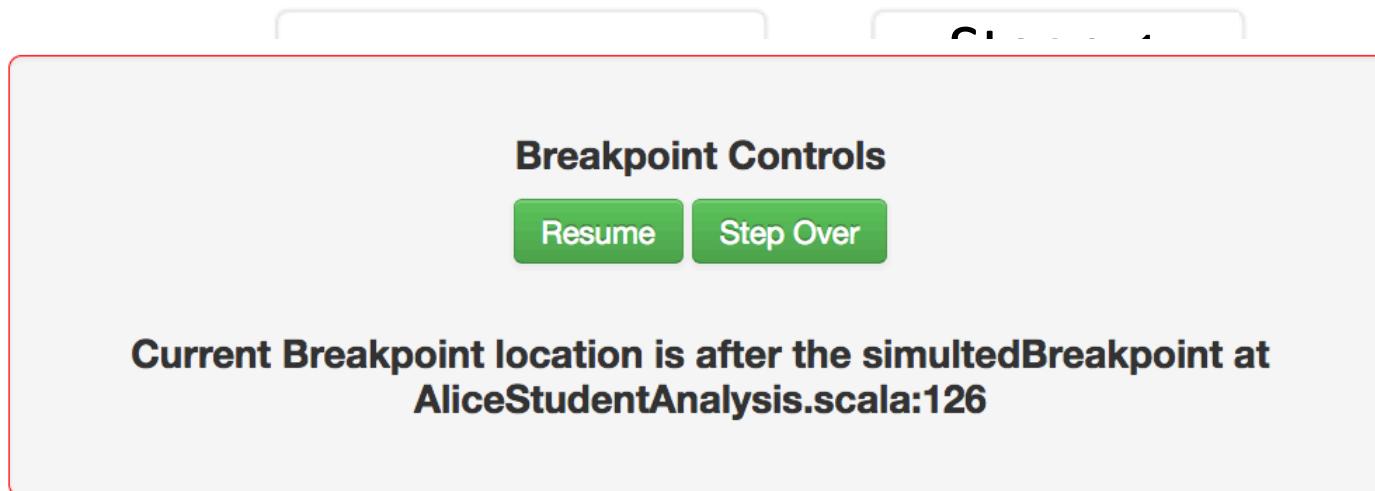
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125 val records = ctx.textFile("/home/ali/Desktop/grades.txt")  
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Feature 1: Simulated Breakpoint



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Feature 1: Simulated Breakpoint



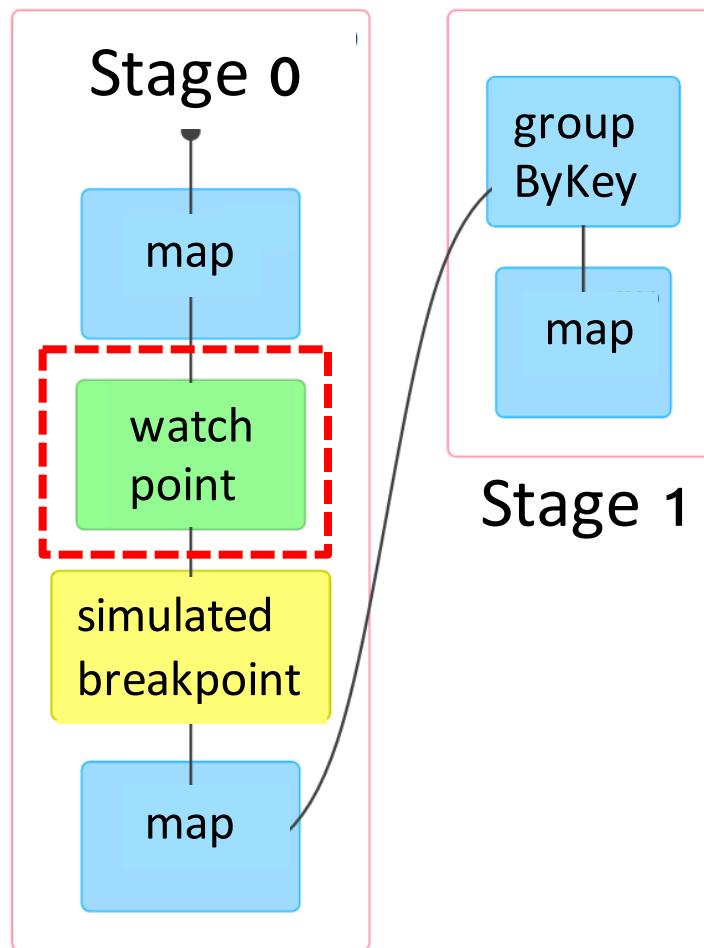
Realtime Code Fix

```
1 def function(value:  
2 /**Write input types. For Example : (String, Int) */  
3 ):  
4 /**Write output types. For Example : (String, Int) */  
5 = {  
6 /**Write code here**/  
7 }
```

Patch the Code

Simulated breakpoint enables user to inspect intermediate program state without pausing the computation

Feature 2: On Demand Guarded Watchpoint



```
124 val records = ctx.textFile("/home/ali/Desktop/SparkCourse/SparkSQL/SparkSQL  
125 .watchpoint(s=> !COLLEGEYEAR.contains(s.substring(0, 2)))  
126 .simulatedBreakpoint()
```

Feature 2: On Demand Guarded Watchpoint

Captured Data Records

1 Timothy 2 21

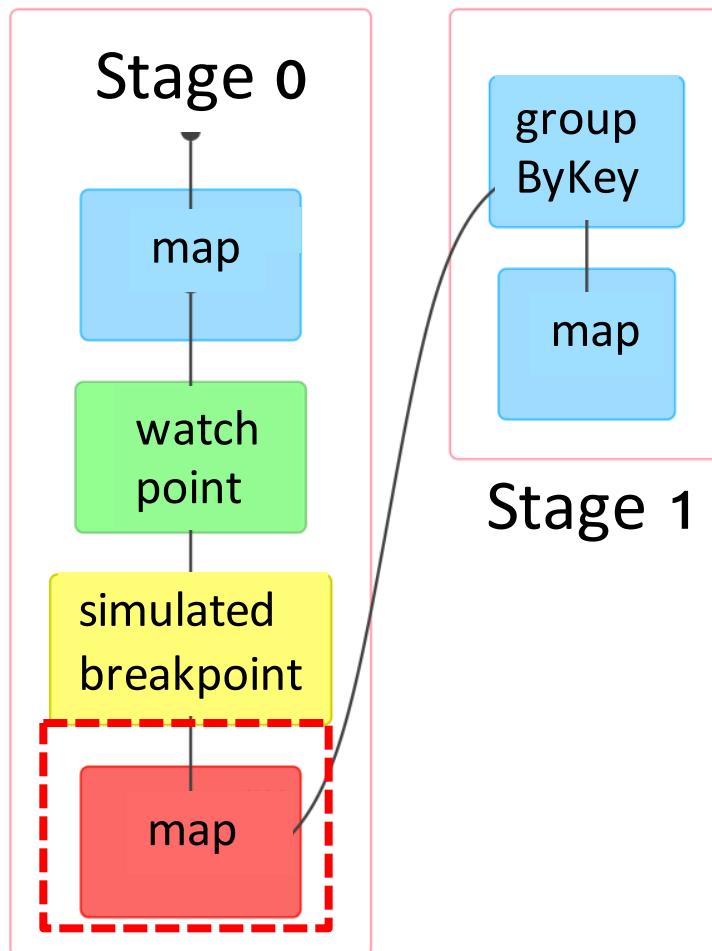
265 Alan 1 24

```
1 def guard(value:  
2   /**Write input types for this watchpoint  
3   guard below. For Example : (String, Int) */  
4   ): Boolean = {  
5     /**Write your guard here**/  
6   }
```

Submit New Guard

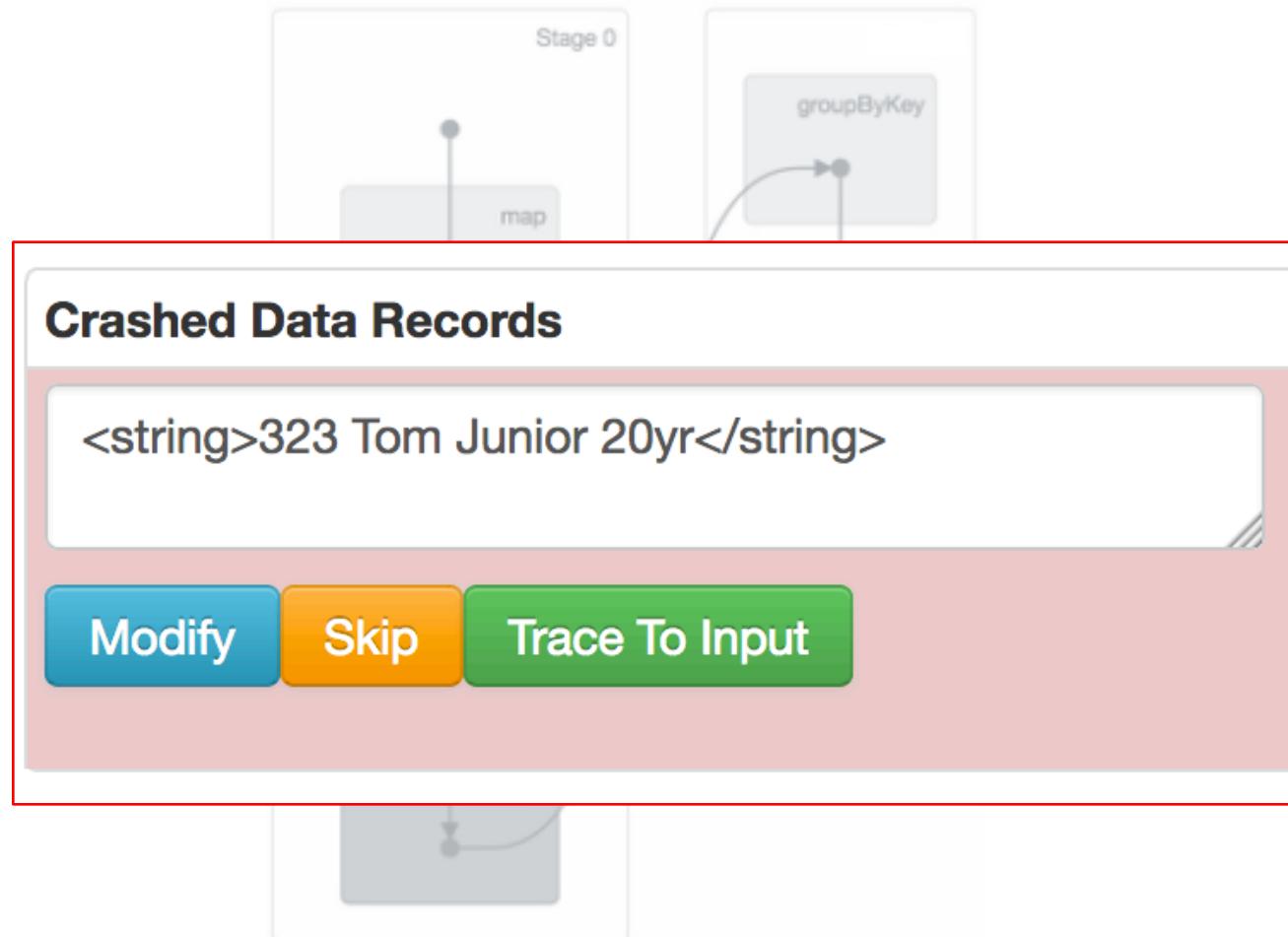
A user can inspect intermediate data using a guard and also update it on the fly

Feature 3: Crash Culprit Identification and Remediation



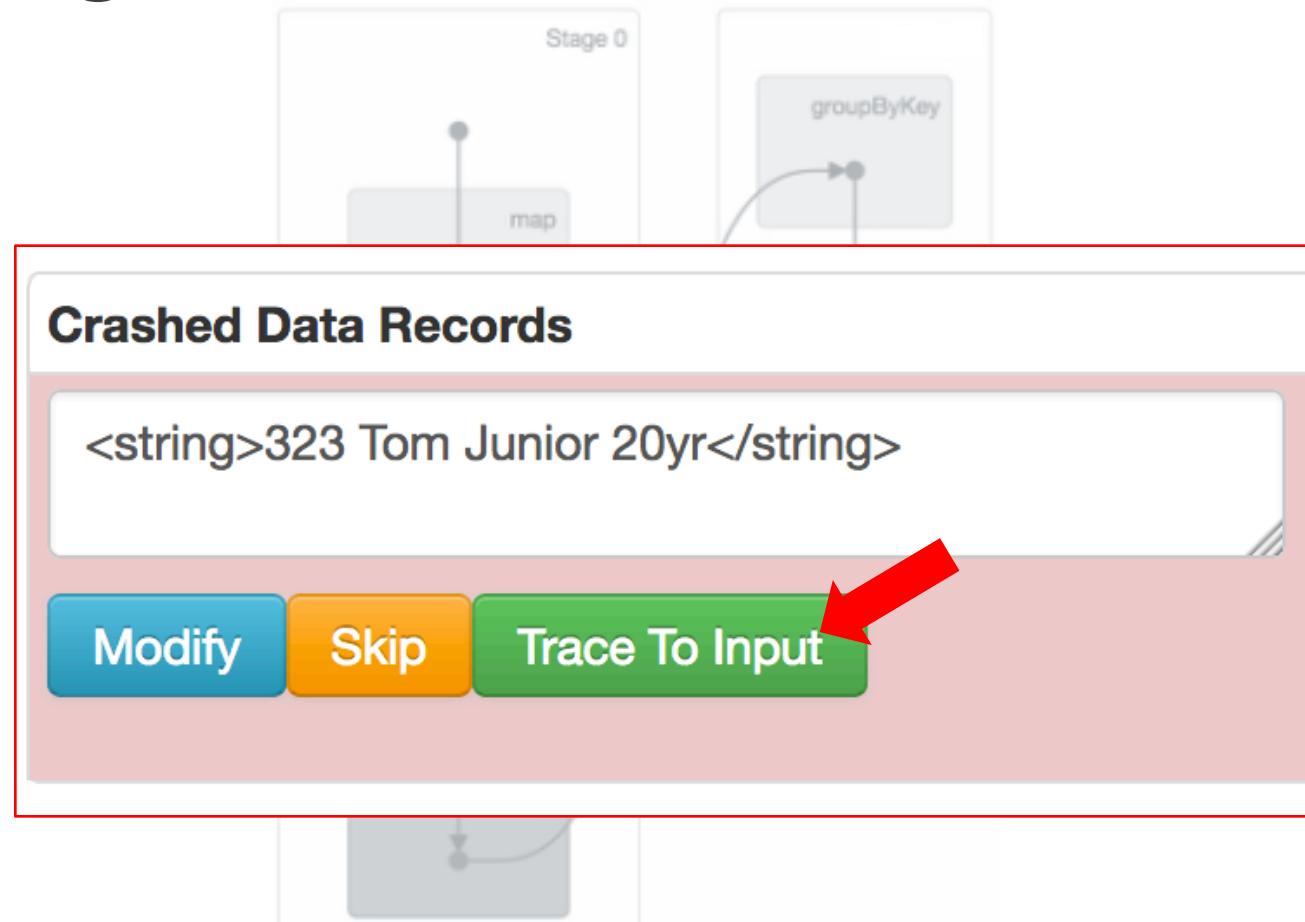
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127     val grade_age_pair = records.map(line => {
```

Feature 3: Crash Culprit Identification and Remediation



A user can use BigDebug to identify the crashing records and remediate from the failure

Feature 4: Forward and Backward Tracing [VLDB '15]

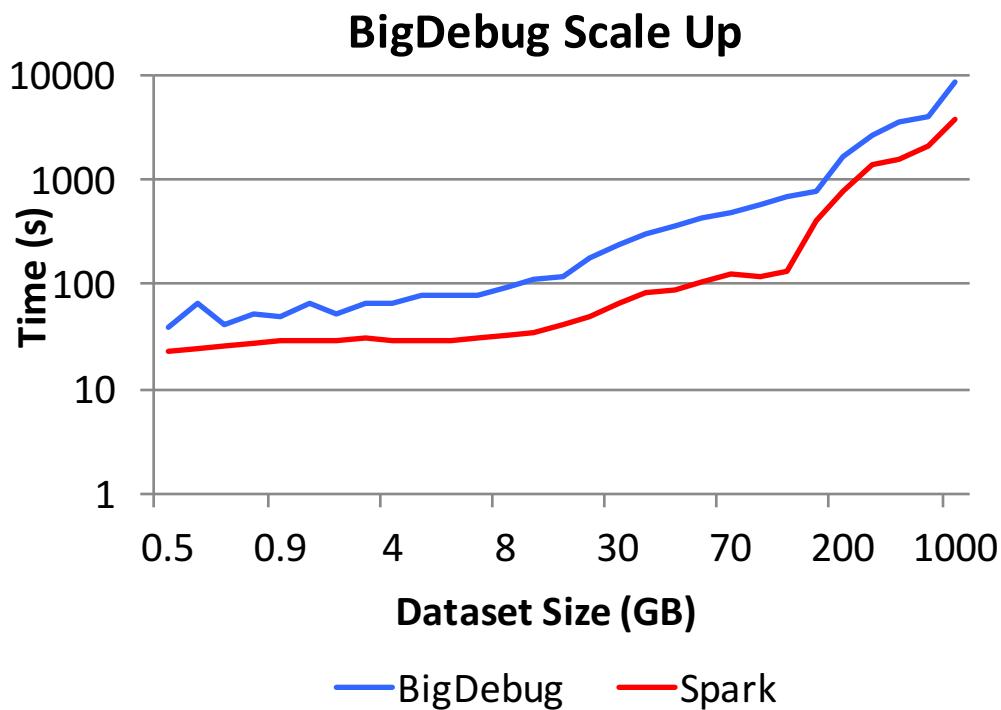


Data provenance enables users to identify crash inducing inputs records

125
126
127

```
.simultedBreakpoint()  
val grade_age_pair = records.map(line => {
```

Performance Evaluation [ICSE '16]



With maximum instrumentation, BigDebug takes 2.4X the time of baseline Spark while the average case is at 1.34X

Time Saving

Arthur [Dave et al. 2013]

The first run
crashes



The second run instruments all
records leading to a crash



The third run removes
the crash.

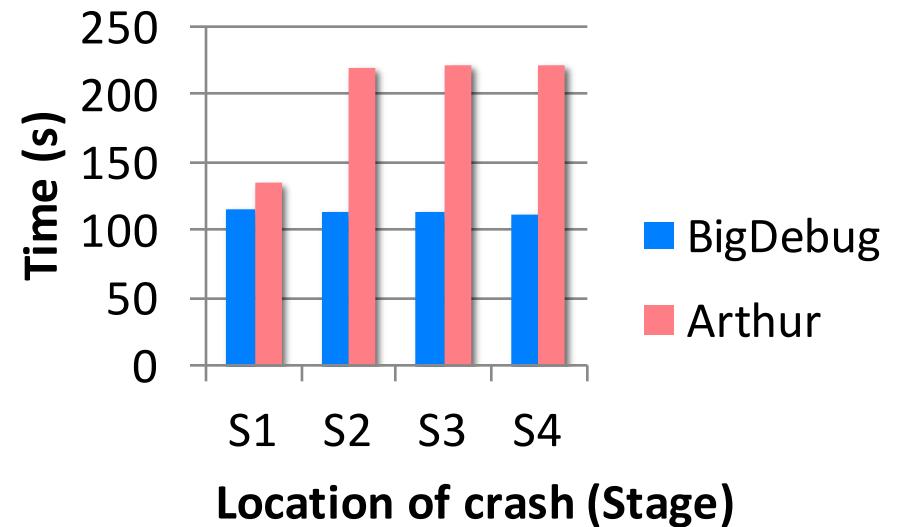


BigDebug

A single run can detect and remove the
crash and resumes the job



BigDebug finds a crash inducing record with 100% accuracy and
saves upto 100% time saving through runtime crash remediation



In Summary

- BigDebug provides primitives to enable interactive debugging on the cloud without sacrificing the performance
- Using data provenance a user can understand how errors propagate through data processing steps
- On average, BigDebug poses 34% overhead and saves 100% of time in case of crash
- BigDebug is publicly available at

<https://sites.google.com/site/sparkbigdebug/>