深入理解Spark 2.1 Core (七): 任务执行的原理与源码分析

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上篇博文《深入理解Spark 2.1 Core (六): 资源调度的实现与源码分析》中我们讲解了,AppClient和Executor是如何启动,如何为逻辑上与物理上的资源调度,以及分析了在Spark1.4之前逻辑上资源调度算法的bug。

这篇博文,我们就来讲讲Executor启动后,是如何在Executor上执行Task的,以及其后续处理。

执行Task

我们在《深入理解Spark 2.1 Core (三): 任务调度器的原理与源码分析》中提到了,任务调度完成后,CoarseGrainedSchedulerBackend.DriverEndpoint会调用launchTasks向CoarseGrainedExecutorBackend发送带着serializedTask的LaunchTask信号。接下来,我们就来讲讲CoarseGrainedExecutorBackend接收到LaunchTask信号后,是如何执行Task的。

调用栈如下:

- CoarseGrainedExecutorBackend.receive
 - Executor.launchTask
 - Executor.TaskRunner.run
 - · Executor.updateDependencies
 - Task.run
 - ShuffleMapTask.runTask
 - ResultTask.runTask

CoarseGrainedExecutorBackend.receive

```
1
       case LaunchTask(data) =>
 2
         if (executor == null) {
 3
           exitExecutor(1, "Received LaunchTask command but executor was null")
 4
         } else {
 5
           // 反序列话task描述
 6
           val taskDesc = ser.deserialize[TaskDescription](data.value)
 7
           logInfo("Got assigned task " + taskDesc.taskId)
 8
           // 调用executor.launchTask
 9
           executor.launchTask(this, taskId = taskDesc.taskId, attemptNumber = taskDesc.attemptNumber,
10
             taskDesc.name, taskDesc.serializedTask)
         }
11
```

Executor.launchTask

```
def launchTask(
 1
 2
         context: ExecutorBackend,
 3
         taskId: Long,
 4
         attemptNumber: Int,
 5
         taskName: String,
         serializedTask: ByteBuffer): Unit = {
 6
 7
         // 创建TaskRunner
       val tr = new TaskRunner(context, taskId = taskId, attemptNumber = attemptNumber, taskName,
 8
 9
         serializedTask)
10
         // 把taskID 以及 对应的 TaskRunner,
11
         // 加入到 ConcurrentHashMap[Long, TaskRunner]
       runningTasks.put(taskId, tr)
```

```
13 // 线程池 执行 TaskRunner
14 threadPool.execute(tr)
15 }
```

Executor.TaskRunner.run

```
1
       override def run(): Unit = {
 2
         val threadMXBean = ManagementFactory.getThreadMXBean
 3
         val taskMemoryManager = new TaskMemoryManager(env.memoryManager, taskId)
 4
         // 记录开始反序列化的时间
 5
         val deserializeStartTime = System.currentTimeMillis()
 6
         // 记录开始反序列化的时的Cpu时间
         val deserializeStartCpuTime = if (threadMXBean.isCurrentThreadCpuTimeSupported) {
 7
           threadMXBean.getCurrentThreadCpuTime
 8
 9
         } else 0L
         Thread.currentThread.setContextClassLoader(replClassLoader)
10
11
         val ser = env.closureSerializer.newInstance()
12
         logInfo(s"Running $taskName (TID $taskId)")
13
         execBackend.statusUpdate(taskId, TaskState.RUNNING, EMPTY_BYTE_BUFFER)
14
         var taskStart: Long = 0
15
         var taskStartCpu: Long = 0
         // 开始GC的时间
16
17
         startGCTime = computeTotalGcTime()
18
19
         trv {
          //反序列化任务信息
20
21
           val (taskFiles, taskJars, taskProps, taskBytes) =
22
             Task.deserializeWithDependencies(serializedTask)
23
24
          // 根据taskProps设置executor属性
25
           Executor.taskDeserializationProps.set(taskProps)
26
          // 根据taskFiles和taskJars,
27
          // 下载任务所需的File 和 加载所需的Jar包
28
           updateDependencies(taskFiles, taskJars)
29
          // 根据taskBytes生成task
30
           task = ser.deserialize[Task[Any]](taskBytes, Thread.currentThread.getContextClassLoader)
31
           //设置task属性
           task.localProperties = taskProps
32
33
           //设置task内存管理
34
           task.setTaskMemoryManager(taskMemoryManager)
35
           // 若在反序列话之前Task就被kill了,
36
37
           // 抛出异常
38
           if (killed) {
39
             throw new TaskKilledException
40
41
           logDebug("Task " + taskId + "'s epoch is " + task.epoch)
42
43
           //更新mapOutputTracker Epoch 为task epoch
           env.mapOutputTracker.updateEpoch(task.epoch)
44
45
46
           // 记录任务开始时间
           taskStart = System.currentTimeMillis()
47
           // 记录任务开始时的cpu时间
48
           taskStartCpu = if (threadMXBean.isCurrentThreadCpuTimeSupported) {
49
             threadMXBean.getCurrentThreadCpuTime
50
51
           } else 0L
52
           var threwException = true
           val value = try {
53
54
           // 运行Task
             val res = task.run(
55
56
               taskAttemptId = taskId,
57
               attemptNumber = attemptNumber,
               metricsSystem = env.metricsSystem)
```

```
threwException = false
 59
 60
              res
 61
            } finally {
 62
              val releasedLocks = env.blockManager.releaseAllLocksForTask(taskId)
 63
              val freedMemory = taskMemoryManager.cleanUpAllAllocatedMemory()
 64
 65
              if (freedMemory > 0 && !threwException) {
                val errMsg = s"Managed memory leak detected; size = $freedMemory bytes, TID = $taskId"
 66
                if (conf.getBoolean("spark.unsafe.exceptionOnMemoryLeak", false)) {
 67
 68
                  throw new SparkException(errMsg)
 69
                } else {
 70
                  logWarning(errMsg)
 71
 72
              }
 73
 74
              if (releasedLocks.nonEmpty && !threwException) {
 75
                val errMsg =
                  s"${releasedLocks.size} block locks were not released by TID = $taskId:\n" +
 76
 77
                    releasedLocks.mkString("[", ", ", "]")
 78
                if (conf.getBoolean("spark.storage.exceptionOnPinLeak", false)) {
 79
                  throw new SparkException(errMsg)
 80
                } else {
 81
                  logWarning(errMsg)
 82
                }
 83
              }
 84
            // 记录任务结束时间
 85
 86
            val taskFinish = System.currentTimeMillis()
            // 记录任务结束时的cpu时间
 87
            val taskFinishCpu = if (threadMXBean.isCurrentThreadCpuTimeSupported) {
 88
              thread {\tt MXBean.getCurrentThreadCpuTime}
 89
 90
            } else 0L
 91
            // 若task在运行中被kill了
 92
 93
            // 则抛出异常
            if (task.killed) {
 94
              throw new TaskKilledException
 95
 96
 97
 98
 99
            val resultSer = env.serializer.newInstance()
100
            // 结果记录序列化开始的系统时间
101
            val beforeSerialization = System.currentTimeMillis()
102
            // 序列化结果
103
            val valueBytes = resultSer.serialize(value)
104
            // 结果记录序列化完成的系统时间
105
            val afterSerialization = System.currentTimeMillis()
106
            // 反序列话发生在两个地方:
107
108
            // 1. 在该函数下反序列化Task信息以及Task实例。
109
            // 2. 在任务启动后, Task.run 反序列化 RDD 和 函数
110
111
            // 计算task的反序列化费时
112
            task.metrics.setExecutorDeserializeTime(
              (taskStart - deserializeStartTime) + task.executorDeserializeTime)
113
114
            // 计算task的反序列化cpu费时
115
            task.metrics.setExecutorDeserializeCpuTime(
              (taskStartCpu - deserializeStartCpuTime) + task.executorDeserializeCpuTime)
116
117
            // 计算task运行费时
118
            task.metrics.setExecutorRunTime((taskFinish - taskStart) - task.executorDeserializeTime)
119
            // 计算task运行cpu费时
120
            task.metrics.setExecutorCpuTime(
121
              (taskFinishCpu - taskStartCpu) - task.executorDeserializeCpuTime)
             // 计算GC时间
122
```

```
123
            task.metrics.setJvmGCTime(computeTotalGcTime() - startGCTime)
124
             //计算结果序列化时间
125
126
            task.metrics.setResultSerializationTime(afterSerialization - beforeSerialization)
127
128
129
            val accumUpdates = task.collectAccumulatorUpdates()
            // 这里代码存在缺陷:
130
            // value相当于被序列化了两次
131
132
            val directResult = new DirectTaskResult(valueBytes, accumUpdates)
133
            val serializedDirectResult = ser.serialize(directResult)
            // 得到结果的大小
134
135
            val resultSize = serializedDirectResult.limit
136
           // 对于计算结果,会根据结果的大小有不同的策略:
137
138
           // 1.生成结果在(正无穷,1GB):
139
           // 超过1GB的部分结果直接丢弃,
           // 可以通过spark.driver.maxResultSize实现
140
141
           // 默认为1G
142
           // 2.生成结果大小在$[1GB,128MB - 200KB]
           // 会把该结果以taskId为编号存入BlockManager中,
143
144
           // 然后把该编号通过Netty发送给Driver,
145
           // 该阈值是Netty框架传输的最大值
           // spark.akka.frameSize (默认为128MB) 和Netty的预留空间reservedSizeBytes (200KB) 的差值
146
           // 3.生成结果大小在 (128MB - 200KB,0):
147
148
           // 直接通过Netty发送到Driver
            val serializedResult: ByteBuffer = {
149
150
              if (maxResultSize > 0 && resultSize > maxResultSize) {
                logWarning(s"Finished $taskName (TID $taskId). Result is larger than maxResultSize " +
151
                  s"(${Utils.bytesToString(resultSize)}) > ${Utils.bytesToString(maxResultSize)}), " +
152
153
                  s"dropping it.")
                ser.serialize(new IndirectTaskResult[Any](TaskResultBlockId(taskId), resultSize))
154
              } else if (resultSize > maxDirectResultSize) {
155
156
                val blockId = TaskResultBlockId(taskId)
157
                env.blockManager.putBytes(
158
                  blockId.
                  new ChunkedByteBuffer(serializedDirectResult.duplicate()),
159
                  StorageLevel.MEMORY_AND_DISK_SER)
160
                logInfo(
161
                  s"Finished $taskName (TID $taskId). $resultSize bytes result sent via BlockManager)")
162
                ser.serialize(new IndirectTaskResult[Any](blockId, resultSize))
163
164
165
                logInfo(s"Finished $taskName (TID $taskId). $resultSize bytes result sent to driver")
                serializedDirectResult
166
167
              }
168
            }
169
            // 更新execBackend 状态
170
            execBackend.statusUpdate(taskId, TaskState.FINISHED, serializedResult)
171
172
          } catch {
173
            case ffe: FetchFailedException =>
              val reason = ffe.toTaskFailedReason
174
175
              setTaskFinishedAndClearInterruptStatus()
176
              execBackend.statusUpdate(taskId, TaskState.FAILED, ser.serialize(reason))
177
178
            case _: TaskKilledException =>
179
              logInfo(s"Executor killed $taskName (TID $taskId)")
180
              setTaskFinishedAndClearInterruptStatus()
181
              execBackend.statusUpdate(taskId, TaskState.KILLED, ser.serialize(TaskKilled))
182
183
            case _: InterruptedException if task.killed =>
184
              logInfo(s"Executor interrupted and killed $taskName (TID $taskId)")
185
              setTaskFinishedAndClearInterruptStatus()
186
              execBackend.statusUpdate(taskId, TaskState.KILLED, ser.serialize(TaskKilled))
```

```
187
188
            case CausedBy(cDE: CommitDeniedException) =>
               val reason = cDE.toTaskFailedReason
189
190
               setTaskFinishedAndClearInterruptStatus()
191
               execBackend.statusUpdate(taskId, TaskState.FAILED, ser.serialize(reason))
192
193
             case t: Throwable =>
194
               logError(s"Exception in $taskName (TID $taskId)", t)
195
196
               val accums: Seq[AccumulatorV2[_, _]] =
197
                 if (task != null) {
                   task.metrics.setExecutorRunTime(System.currentTimeMillis() - taskStart)
198
199
                   task.metrics.setJvmGCTime(computeTotalGcTime() - startGCTime)
200
                   task.collectAccumulatorUpdates(taskFailed = true)
201
                 } else {
202
                   Seq.empty
203
                 }
204
205
               val accUpdates = accums.map(acc => acc.toInfo(Some(acc.value), None))
206
207
               val serializedTaskEndReason = {
208
                 try {
209
                   ser.serialize(new ExceptionFailure(t, accUpdates).withAccums(accums))
                 } catch {
210
211
                   case _: NotSerializableException =>
212
                     ser.serialize(new ExceptionFailure(t, accUpdates, false).withAccums(accums))
213
                 }
214
               }
215
               setTaskFinishedAndClearInterruptStatus()
               execBackend.statusUpdate(taskId, TaskState.FAILED, serializedTaskEndReason)
216
217
               if (Utils.isFatalError(t)) {
218
                 SparkUncaughtExceptionHandler.uncaughtException(t)
219
220
               }
221
          } finally {
222
          // 任务结束后移除
223
224
             runningTasks.remove(taskId)
225
          }
226
        }
```

Executor.updateDependencies

接下来,我们来看看更新executor的依赖,即下载任务所需的File和加载所需的Jar包:

```
1
      private def updateDependencies(newFiles: HashMap[String, Long], newJars: HashMap[String, Long]) {
 2
        lazy val hadoopConf = SparkHadoopUtil.get.newConfiguration(conf)
 3
        synchronized {
 4
          // 下载任务所需的File
 5
          for ((name, timestamp) <- newFiles if currentFiles.getOrElse(name, -1L) < timestamp) {</pre>
 6
            logInfo("Fetching " + name + " with timestamp " + timestamp)
 7
            Utils.fetchFile(name, new File(SparkFiles.getRootDirectory()), conf,
 8
              env.securityManager, hadoopConf, timestamp, useCache = !isLocal)
 9
            currentFiles(name) = timestamp
10
          }
11
          // 加载所需的Jar包
12
          for ((name, timestamp) <- newJars) {</pre>
13
            val localName = name.split("/").last
            val currentTimeStamp = currentJars.get(name)
14
15
              .orElse(currentJars.get(localName))
16
              .getOrElse(-1L)
17
            if (currentTimeStamp < timestamp) {</pre>
              logInfo("Fetching " + name + " with timestamp " + timestamp)
18
              Utils.fetchFile(name, new File(SparkFiles.getRootDirectory()), conf,
```

```
20
                env.securityManager, hadoopConf, timestamp, useCache = !isLocal)
21
              currentJars(name) = timestamp
              // 把它加入到 class loader
22
23
              val url = new File(SparkFiles.getRootDirectory(), localName).toURI.toURL
24
              if (!urlClassLoader.getURLs().contains(url)) {
                logInfo("Adding " + url + " to class loader")
25
26
                urlClassLoader.addURL(url)
27
              }
28
           }
29
         }
30
        }
      }
31
```

Task.run

接下来,我们来看看这篇博文最核心的部分——task运行:

```
final def run(
 1
 2
          taskAttemptId: Long,
 3
          attemptNumber: Int,
 4
          metricsSystem: MetricsSystem): T = {
 5
        SparkEnv.get.blockManager.registerTask(taskAttemptId)
 6
        //创建TaskContextImpl
 7
        context = new TaskContextImpl(
 8
          stageId,
 9
          partitionId,
10
          taskAttemptId,
11
          attemptNumber,
12
          taskMemoryManager,
          localProperties,
13
14
          metricsSystem,
15
          metrics)
          //在TaskContext中设置TaskContextImpl
16
17
        TaskContext.setTaskContext(context)
        taskThread = Thread.currentThread()
18
19
20
        if (_killed) {
21
          kill(interruptThread = false)
22
        }
23
24
        new CallerContext("TASK", appId, appAttemptId, jobId, Option(stageId), Option(stageAttemptId),
25
          Option(taskAttemptId), Option(attemptNumber)).setCurrentContext()
26
27
        try {
28
        // 调用runTask
29
          runTask(context)
30
        } catch {
31
          case e: Throwable =>
32
            try {
33
              context.markTaskFailed(e)
34
            } catch {
              case t: Throwable =>
35
36
                e.addSuppressed(t)
37
            }
38
            throw e
39
        } finally {
40
          // 标记Task完成
41
          context.markTaskCompleted()
42
          try {
43
            Utils.tryLogNonFatalError {
              // 释放内存
44
45
              SparkEnv.get.blockManager.memoryStore.releaseUnrollMemoryForThisTask(MemoryMode.ON\_HEAP)
46
              SparkEnv.get.blockManager.memoryStore.releaseUnrollMemoryForThisTask(MemoryMode.OFF_HEAP)
              val memoryManager = SparkEnv.get.memoryManager
```

Task有两个子类,一个是非最后的Stage的Task,ShuffleMapTask;一个是最后的Stage的Task,ResultTask。它们都覆盖了Task的runTask方法,接下来我们就分别来讲下它们的runTask方法。

ShuffleMapTask.runTask

根据每个Stage的partition数量来生成ShuffleMapTask,ShuffleMapTask会根据下游的Partition数量和Shuffle的策略来生成一系列文件。

```
1
     override def runTask(context: TaskContext): MapStatus = {
 2
 3
       val threadMXBean = ManagementFactory.getThreadMXBean
 4
       // 记录反序列化开始时间
       val deserializeStartTime = System.currentTimeMillis()
       // 记录反序列化开始时的Cpu时间
 6
       val deserializeStartCpuTime = if (threadMXBean.isCurrentThreadCpuTimeSupported) {
 8
         threadMXBean.getCurrentThreadCpuTime
       } else 0L
 9
       val ser = SparkEnv.get.closureSerializer.newInstance()
10
       // 反序列化rdd 及其 依赖
11
       val (rdd, dep) = ser.deserialize[(RDD[_], ShuffleDependency[_, _, _])](
12
         ByteBuffer.wrap(taskBinary.value), Thread.currentThread.getContextClassLoader)
13
14
       _executorDeserializeTime = System.currentTimeMillis() - deserializeStartTime
15
16
       // 计算 反序列化Cpu费时
17
       _executorDeserializeCpuTime = if (threadMXBean.isCurrentThreadCpuTimeSupported) {
         threadMXBean.getCurrentThreadCpuTime - deserializeStartCpuTime
18
       } else 0L
19
20
21
       var writer: ShuffleWriter[Any, Any] = null
22
23
       //获取shuffleManager
         val manager = SparkEnv.get.shuffleManager
24
25
26
         writer = manager.getWriter[Any, Any](dep.shuffleHandle, partitionId, context)
27
        // 调用writer.write 开始计算RDD,
28
        // 这部分 我们会在后续博文讲解
29
         writer.write(rdd.iterator(partition, context).asInstanceOf[Iterator[_ <: Product2[Any, Any]]])</pre>
30
         // 停止计算, 并返回结果
31
         writer.stop(success = true).get
32
       } catch {
33
         case e: Exception =>
34
           try {
35
             if (writer != null) {
36
               writer.stop(success = false)
37
             }
38
           } catch {
39
             case e: Exception =>
               log.debug("Could not stop writer", e)
40
41
42
           throw e
43
       }
44
     }
```

ResultTask.runTask

```
override def runTask(context: TaskContext): U = {
   1
   2
                        val threadMXBean = ManagementFactory.getThreadMXBean
   3
                        // 记录反序列化开始时间
   4
                       val deserializeStartTime = System.currentTimeMillis()
   5
                       // 记录反序列化开始时的Cpu时间
                       val deserializeStartCpuTime = if (threadMXBean.isCurrentThreadCpuTimeSupported) {
   6
   7
                              threadMXBean.getCurrentThreadCpuTime
   8
                        } else 0L
   9
                        val ser = SparkEnv.get.closureSerializer.newInstance()
                        // 反序列化rdd 及其 作用于RDD的结果函数
10
                       val (rdd, func) = ser.deserialize[(RDD[T], (TaskContext, Iterator[T]) => U)](
11
                              ByteBuffer.wrap(taskBinary.value), Thread.currentThread.getContextClassLoader)
12
13
                              // 计算 反序列化费时
                        _executorDeserializeTime = System.currentTimeMillis() - deserializeStartTime
14
15
                        // 计算 反序列化Cpu费时
                        _executorDeserializeCpuTime = if (threadMXBean.isCurrentThreadCpuTimeSupported) {
16
17
                              thread {\tt MXBean.getCurrentThreadCpuTime} \ - \ describing {\tt constantCpuTime} \ - 
18
                        // 这部分 我们会在后续博文讲解
19
20
                        func(context, rdd.iterator(partition, context))
21
```

后续处理

计量统计

对各个费时的统计、上章已经讲解。

回收内存

这在上章Task.run也已经讲解。

处理执行结果

Executor.TaskRunner.run的execBackend.statusUpdate,在《深入理解Spark 2.1 Core (四): 运算结果处理和容错的原理与源码分析 》中我们已经讲解过。