

Storm Guide

Hortonworks Data Platform

(May 9, 2016)

Storm Guide: Hortonworks Data Platform

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Table of Contents

1. Using Apache Storm	1
1.1. Basic Storm Concepts	1
1.1.1. Spouts	2
1.1.2. Bolts	3
1.1.3. Stream Groupings	4
1.1.4. Topologies	4
1.1.5. Processing Reliability	5
1.1.6. Workers, Executors, and Tasks	6
1.1.7. Parallelism	6
2. Topology Development Guidelines	12
2.1. Debugging Topologies	12
2.2. Determining Topology Parallelism Units	14
3. Streaming Data to Hive	15
4. Ingesting Data with the Apache Kafka Spout Connector	18
4.1. KafkaSpout and OpaqueTridentKafkaSpout Examples	18
4.2. Storm-Kafka API Reference	19
4.3. KafkaSpout Limitations	21
4.4. Configuring Kafka for Use with the Storm-Kafka Connector	22
4.5. Configuring KafkaSpout to Connect to a Secure Kafka Cluster	22
4.6. Configuring KafkaSpout to Connect to HBase or Hive	23
5. Writing Data with Storm	24
5.1. Writing Data to HDFS with the Storm-HDFS Connector	24
5.2. Writing Data to HBase with the Storm-HBase Connector	27
5.3. Writing Data to Kafka using a Kafka Bolt	28
5.4. Configuring Connectors for a Secure Cluster	31
6. Packaging Storm Topologies	35
7. Deploying and Managing Apache Storm Topologies	37
8. Example: RollingTopWords Topology	39

List of Tables

1.1. Storm Concepts	1
1.2. Stream Groupings	4
1.3. Processing Guarantees	5
2.1. Storm Topology Guidelines	12
3.1. HiveMapper Arguments	15
3.2. HiveOptions Class Configuration Properties	16
4.1. KafkaConfig Parameters	19
4.2. SpoutConfig Parameters	20
4.3. TridentKafkaConfig Parameters	20
4.4. KafkaConfig Fields	20
4.5. SpoutConfig Parameters	21
4.6. TridentKafkaConfig Parameters	22
5.1. HdfsBolt Methods	24
5.2. SimpleHBaseMapper Methods	27
6.1. Topology Packing Errors	36
7.1. Topology Administrative Actions	37

1. Using Apache Storm

The exponential increase in data from real-time sources such as machine sensors creates a need for data processing systems that can ingest this data, process it, and respond in real time. A typical use case involves an automated system that responds to sensor data by sending email to support staff or placing an advertisement on a consumer's smart phone. Apache Storm enables such data-driven and automated activity by providing a realtime, scalable, and distributed solution for streaming data.

Apache Storm can be used with any programming language, and guarantees that data streams are processed without data loss.

Storm is datatype-agnostic; it processes data streams of any data type.

A complete introduction to the Storm API is beyond the scope of this documentation. However, the next section, [Basic Storm Concepts](#), provides a brief overview of the most essential concepts. See the [Apache Storm documentation](#) for a more thorough discussion of Apache Storm concepts.

Experienced Storm developers may want to skip to later sections for information about streaming data to Hive; ingesting data with the Apache Kafka spout; writing data to HDFS, HBase, and Kafka; and deploying Storm topologies.

The last section, RollingTopWords Topology, lists the source code for a sample application included with the `storm-starter.jar`.

1.1. Basic Storm Concepts

Writing Storm applications requires an understanding of the following basic concepts.

Table 1.1. Storm Concepts

Storm Concept	Description
Tuple	A named list of values of any data type. The native data structure used by Storm.
Stream	An unbounded sequence of tuples.
Spout	Generates a stream from a realtime data source.
Bolt	Contains data processing, persistence, and messaging alert logic. Can also emit tuples for downstream bolts.
Stream Grouping	Controls the routing of tuples to bolts for processing.
Topology	A group of spouts and bolts wired together into a workflow. A Storm application.
Processing Reliability	Storm guarantee about the delivery of tuples in a topology.
Parallelism	Attribute of distributed data processing that determines how many jobs are processed simultaneously for a topology. Topology developers adjust parallelism to tune their applications.
Workers	A Storm process. A worker may run one or more executors.
Executors	A Storm thread launched by a Storm worker. An executor may run one or more tasks.
Tasks	A Storm job from a spout or bolt.

Storm Concept	Description
Process Controller	Monitors and restarts failed Storm processes. Examples include supervisord, monit, and daemontools.
Master/Nimbus Node	The host in a multi-node Storm cluster that runs a process controller, such as supervisord, and the Storm nimbus, ui, and other related daemons. The process controller is responsible for restarting failed process controller daemons, such as supervisor, on slave nodes. The Storm nimbus daemon is responsible for monitoring the Storm cluster and assigning tasks to slave nodes for execution.
Slave Node	A host in a multi-node Storm cluster that runs a process controller daemon, such as supervisor, as well as the worker processes that run Storm topologies. The process controller daemon is responsible for restarting failed worker processes.

The following subsections describe several of these concepts in more detail.

1.1.1. Spouts

All spouts must implement the `backtype.storm.topology.IRichSpout` interface from the core-storm API. `BaseRichSpout` is the most basic implementation, but there are several others, including `ClojureSpout`, `DRPCSpout`, and `FeederSpout`. In addition, Hortonworks provides a Kafka spout to ingest data from a Kafka cluster. The following example, `RandomSentenceSpout`, is included with the `storm-starter` connector installed with Storm at `/usr/lib/storm/contrib/storm-starter`.

```
package storm.starter.spout;

import backtype.storm.spout.SpoutOutputCollector;
import backtype.storm.task.TopologyContext;
import backtype.storm.topology.OutputFieldsDeclarer;
import backtype.storm.topology.base.BaseRichSpout;
import backtype.storm.tuple.Fields;
import backtype.storm.tuple.Values;
import backtype.storm.utils.Utls;

import java.util.Map;
import java.util.Random;

public class RandomSentenceSpout extends BaseRichSpout {
    SpoutOutputCollector _collector;
    Random _rand;

    @Override
    public void open(Map conf, TopologyContext context, SpoutOutputCollector collector) {
        _collector = collector;
        _rand = new Random();
    }

    @Override
    public void nextTuple() {
        Utls.sleep(100);
        String[] sentences = new String[] { "the cow jumped over the moon", "an apple a day keeps the doctor away", "four score and seven years ago", "snow white and the seven dwarfs", "i am at two with nature" };
    }
}
```

```
String sentence = sentences[_rand.nextInt(sentences.length)];
_collector.emit(new Values(sentence));
}

@Override
public void ack(Object id) {
}

@Override
public void fail(Object id) {
}

@Override
public void declareOutputFields(OutputFieldsDeclarer declarer) {
    declarer.declare(new Fields("word"));
}
}
```

1.1.2. Bolts

All bolts must implement the `IRichBolt` interface. `BaseRichBolt` is the most basic implementation, but there are several others, including `BatchBoltExecutor`, `ClojureBolt`, and `JoinResult`. The following example, `TotalRankingsBolt.java`, is included with `storm-starter` and installed with Storm at `/usr/lib/storm/contrib/storm-starter`.

```
package storm.starter.bolt;

import backtype.storm.tuple.Tuple;
import org.apache.log4j.Logger;
import storm.starter.tools.Rankings;

/**
 * This bolt merges incoming {@link Rankings}.
 * <p/>
 * It can be used to merge intermediate rankings generated by {@link
 * IntermediateRankingsBolt} into a final,
 * consolidated ranking. To do so, configure this bolt with a globalGrouping
 * on {@link IntermediateRankingsBolt}.
 */
public final class TotalRankingsBolt extends AbstractRankerBolt {

    private static final long serialVersionUID = -8447525895532302198L;
    private static final Logger LOG = Logger.getLogger(TotalRankingsBolt.class);

    public TotalRankingsBolt() {
        super();
    }

    public TotalRankingsBolt(int topN) {
        super(topN);
    }

    public TotalRankingsBolt(int topN, int emitFrequencyInSeconds) {
        super(topN, emitFrequencyInSeconds);
    }

    @Override
```

```

void updateRankingsWithTuple(Tuple tuple) {
    Rankings rankingsToBeMerged = (Rankings) tuple.getValue(0);
    super.getRankings().updateWith(rankingsToBeMerged);
    super.getRankings().pruneZeroCounts();
}

@Override
Logger getLogger() {
    return LOG;
}
}

```

1.1.3. Stream Groupings

Stream grouping allows Storm developers to control how tuples are routed to bolts in a workflow. The following table describes the stream groupings available.

Table 1.2. Stream Groupings

Stream Grouping	Description
Shuffle	Sends tuples to bolts in random, round robin sequence. Use for atomic operations, such as math.
Fields	Sends tuples to a bolt based on one or more fields in the tuple. Use to segment an incoming stream and to count tuples of a specified type.
All	Sends a single copy of each tuple to all instances of a receiving bolt. Use to send a signal, such as clear cache or refresh state, to all bolts.
Custom	Customized processing sequence. Use to get maximum flexibility of topology processing based on factors such as data types, load, and seasonality.
Direct	Source decides which bolt receives a tuple.
Global	Sends tuples generated by all instances of a source to a single target instance. Use for global counting operations.

Storm developers specify the field grouping for each bolt using methods on the `TopologyBuilder.BoltGetter` inner class, as shown in the following excerpt from the `WordCountTopology.java` example included with `storm-starter`.

```

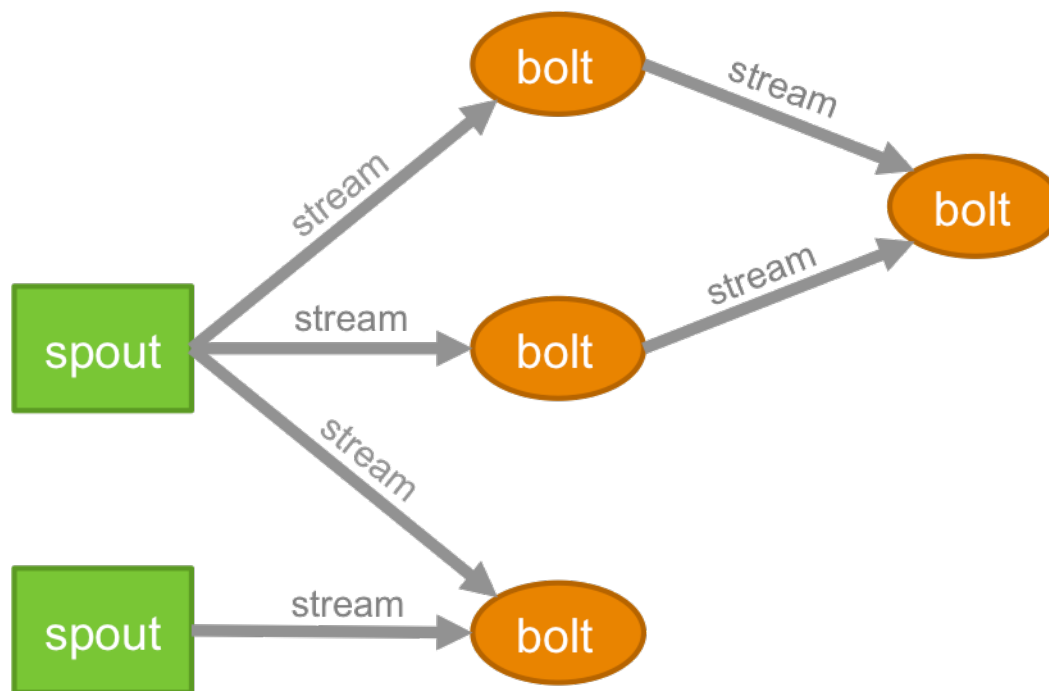
...
TopologyBuilder builder = new TopologyBuilder();
builder.setSpout("spout", new RandomSentenceSpout(), 5);
builder.setBolt("split", new SplitSentence(), 8).shuffleGrouping("spout");
builder.setBolt("count", new WordCount(), 12).fieldsGrouping("split", new
    Fields("word"));
...

```

The first bolt uses shuffle grouping to split random sentences generated with the `RandomSentenceSpout`. The second bolt uses fields grouping to segment and perform a count of individual words in the sentences.

1.1.4. Topologies

The following image depicts a Storm topology with a simple workflow.



Storm topology

The `TopologyBuilder` class is the starting point for quickly writing Storm topologies with the `storm-core` API. The class contains getter and setter methods for the spouts and bolts that comprise the streaming data workflow, as shown in the following sample code.

```

...
TopologyBuilder builder = new TopologyBuilder();
builder.setSpout("spout1", new BaseRichSpout());
builder.setSpout("spout2", new BaseRichSpout());
builder.setBolt("bolt1", new BaseBasicBolt());
builder.setBolt("bolt2", new BaseBasicBolt());
builder.setBolt("bolt3", new BaseBasicBolt());
...

```

1.1.5. Processing Reliability

Storm provides two types of guarantees when processing tuples for a Storm topology.

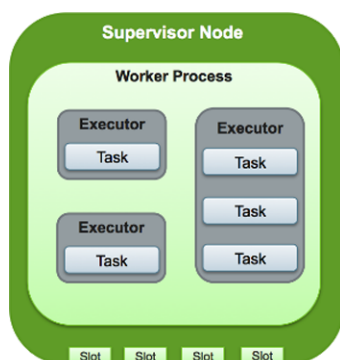
Table 1.3. Processing Guarantees

Guarantee	Description
At least once	Reliable; Tuples are processed at least once, but may be processed more than once. Use when subsecond latency is required and for unordered idempotent operations.
Exactly once	Reliable; Tuples are processed only once. Requires the use of a Trident spout and the Trident API.

1.1.6. Workers, Executors, and Tasks

Apache Storm processes, called workers, run on predefined ports on the machine that hosts Storm.

- Each worker process can run one or more executors, or threads, where each executor is a thread spawned by the worker process.
- Each executor runs one or more tasks from the same component, where a component is a spout or bolt from a topology.



1.1.7. Parallelism

Distributed applications take advantage of horizontally-scaled clusters by dividing computation tasks across nodes in a cluster. Storm offers this and additional finer-grained ways to increase the parallelism of a Storm topology:

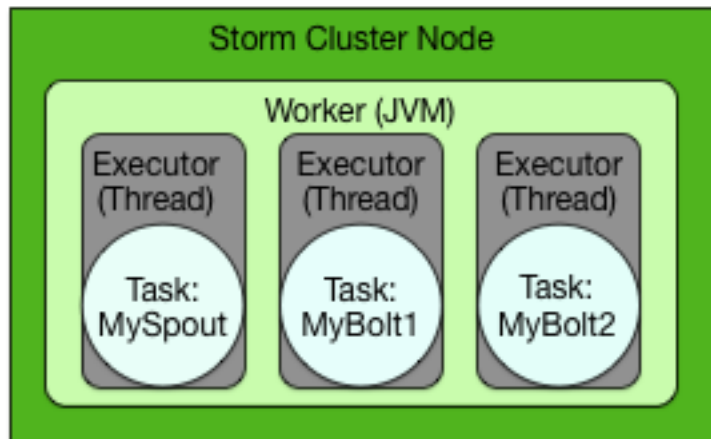
- Increase the number of workers
- Increase the number of executors
- Increase the number of tasks

By default, Storm uses a parallelism factor of 1. Assuming a single-node Storm cluster, a parallelism factor of 1 means that one worker, or JVM, is assigned to execute the topology, and each component in the topology is assigned to a single executor. The following diagram illustrates this scenario. The topology defines a data flow with three tasks, a spout and two bolts.



Note

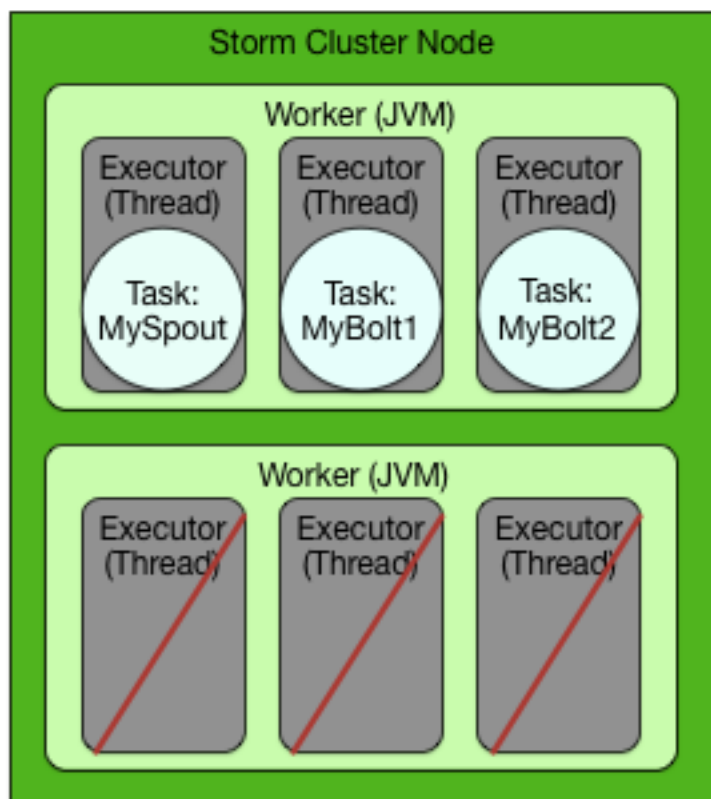
Hortonworks recommends that Storm developers store parallelism settings in a configuration file read by the topology at runtime rather than hard-coding the values passed to the Parallelism API. This topic describes and illustrates the use of the API, but developers can achieve the same effect by reading the parallelism values from a configuration file.



Increasing Parallelism with Workers

Storm developers can easily increase the number of workers assigned to execute a topology with the `Config.setNumWorkers()` method. This code assigns two workers to execute the topology, as the following figure illustrates.

```
...  
Config config = new Config();  
config.setNumWorkers(2);  
...
```



Adding new workers comes at a cost: additional overhead for a new JVM.

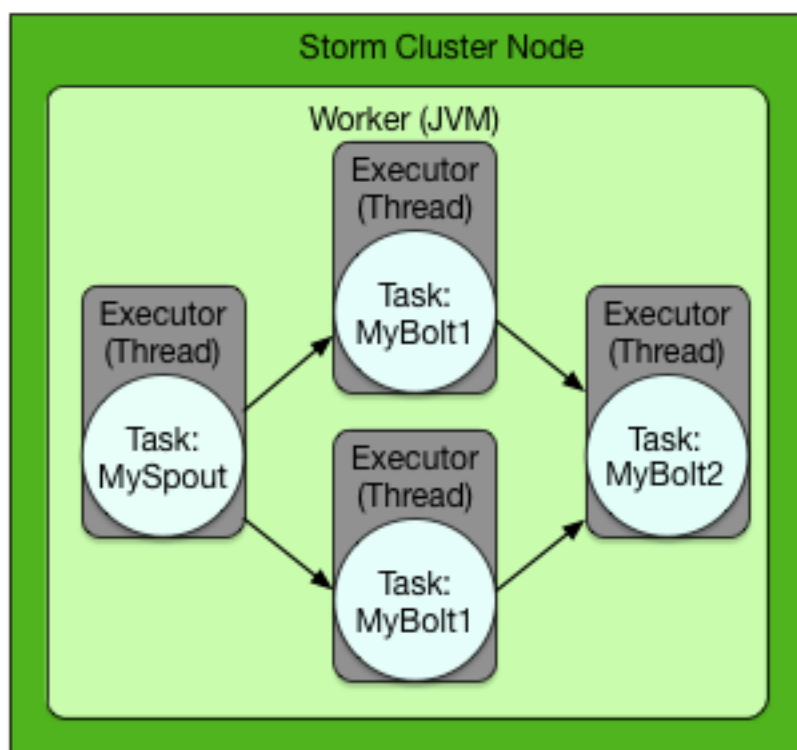
This example adds an additional worker without additional executors or tasks, but to take full advantage of this feature, Storm developers must add executors and tasks to the additional JVMs (described in the following examples).

Increasing Parallelism with Executors

The parallelism API enables Storm developers to specify the number of executors for each worker with a parallelism hint, an optional third parameter to the `setBolt()` method. The following code sample sets this parameter for the `MyBolt1` topology component.

```
...
Config config = new Config();
TopologyBuilder builder = new TopologyBuilder();
builder.setSpout(MY_SPOUT_ID, mySpout);
builder.setBolt(MY_BOLT1_ID, myBolt1, 2).shuffleGrouping(MY_SPOUT_ID);
builder.setBolt(MY_BOLT2_ID, myBolt2).shuffleGrouping(MY_SPOUT_ID);
...
```

This code sample assigns two executors to the single, default worker for the specified topology component, `MyBolt1`, as the following figure illustrates.



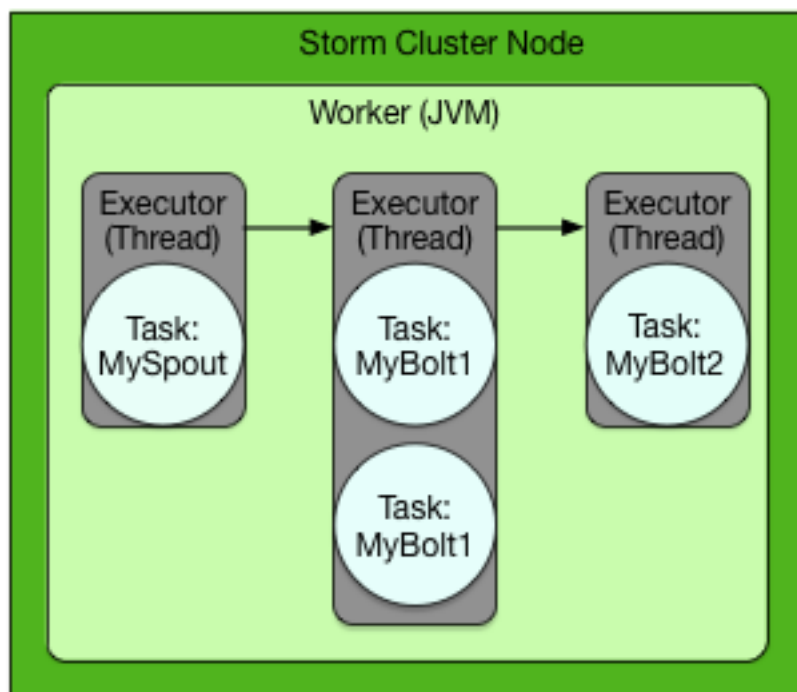
The number of executors is set at the level of individual topology components, so adding executors affects the code for the specified spouts and bolts. This differs from adding workers, which affects only the configuration of the topology.

Increasing Parallelism with Tasks

Finally, Storm developers can increase the number of tasks assigned to a single topology component, such as a spout or bolt. By default, Storm assigns a single task to each component, but developers can increase this number with the `setNumTasks()` method on the `BoltDeclarer` and `SpoutDeclarer` objects returned by the `setBolt()` and `setSpout()` methods.

```
...
Config config = new Config();
TopologyBuilder builder = new TopologyBuilder();
builder.setSpout(MY_SPOUT_ID, mySpout);
builder.setBolt(MY_BOLT1_ID, myBolt1).setNumTasks(2).
    shuffleGrouping(MY_SPOUT_ID);
builder.setBolt(MY_BOLT2_ID, myBolt2).shuffleGrouping(MY_BOLT1_ID);
...
```

This code sample assigns two tasks to execute `MyBolt1`, as the following figure illustrates. This added parallelism might be appropriate for a bolt containing a large amount of data processing logic. However, adding tasks is like adding executors because the code for the corresponding spouts or bolts also changes.



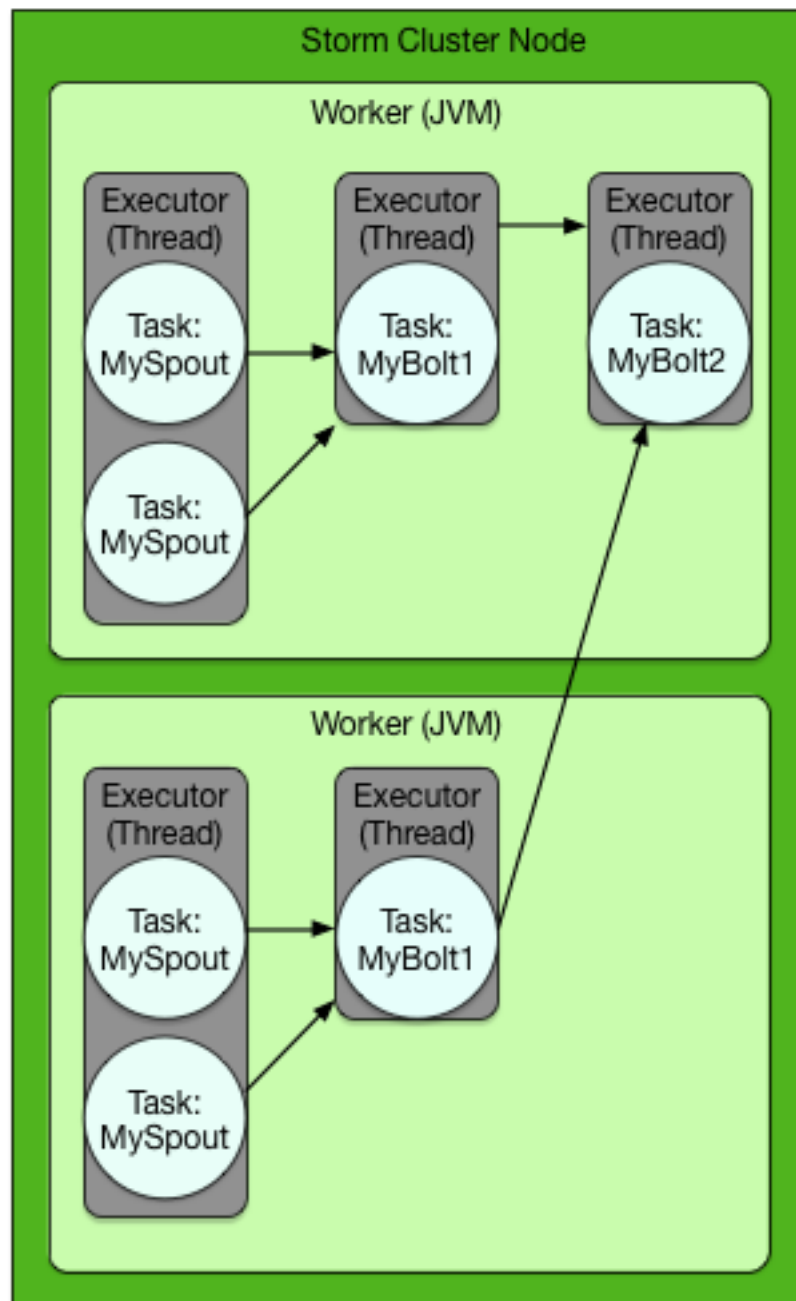
Putting it All Together

Storm developers can fine-tune the parallelism of their topologies by combining new workers, executors and tasks. The following code sample demonstrates all of the following:

- Split processing of the `MySpout` component between four tasks in two separate executors across two workers
- Split processing of the `MyBolt1` component between two executors across two workers

- Centralize processing of the MyBolt2 component in a single task in a single executor in a single worker on a single worker

```
...
Config config = new Config();
config.setNumWorkers(2);
TopologyBuilder builder = new TopologyBuilder();
builder.setSpout(MY_SPOUT_ID, mySpout, 2).setNumTasks(4);
builder.setBolt(MY_BOLT1_ID, myBolt1, 2).setNumTasks(2).
    shuffleGrouping(MY_SPOUT_ID);
builder.setBolt(MY_BOLT2_ID, myBolt2).shuffleGrouping(MY_SPOUT_ID);
...
```



The degree of parallelism depicted might be appropriate for the following topology requirements:

- High-volume streaming data input
- Moderate data processing logic
- Low-volume topology output

See the Storm javadocs at <http://storm.apache.org/releases/0.10.0/javadocs/index.html> for more information about the Storm API.

2. Topology Development Guidelines

Hortonworks recommends the following guidelines for all Storm topologies.



Note

These recommendations focus on guidelines for writing and debugging Storm topologies, rather than hardware tuning. Typically, most of the computation burden falls on the Supervisor and Worker nodes in a Storm cluster. The Nimbus node usually has a lighter load. For this reason, Hortonworks recommends that organizations save their hardware resources for the relatively burdened Supervisor and Worker nodes.

Table 2.1. Storm Topology Guidelines

Guideline	Description
Read topology configuration parameters from a file.	Rather than hard coding configuration information in your Storm application, read the configuration parameters, including parallelism hints for specific components, from a file inside the <code>main()</code> method of the topology. This speeds up the iterative process of debugging by eliminating the need to rewrite and recompile code for simple configuration changes.
Use a cache.	Use a cache to improve performance by eliminating unnecessary operations over the network, such as making frequent external service or lookup calls for reference data needed for processing.
Tighten code in the <code>execute()</code> method.	Every tuple is processed by the <code>execute()</code> method, so verify that the code in this method is as tight and efficient as possible.
Perform benchmark testing to determine latencies.	Perform benchmark testing of the critical points in the network flow of your topology. Knowing the capacity of your data "pipes" provides a reliable standard for judging the performance of your topology and its individual components.

2.1. Debugging Topologies

This topic describes best practices for debugging Storm topologies, including basic guidelines for configuring parallelism for individual topology components.

Debugging Storm topologies differs from debugging batch-oriented applications. Because Storm topologies operate on streaming data (rather than data at rest, as in HDFS) they are sensitive to data sources. When debugging Storm topologies, consider the following questions:

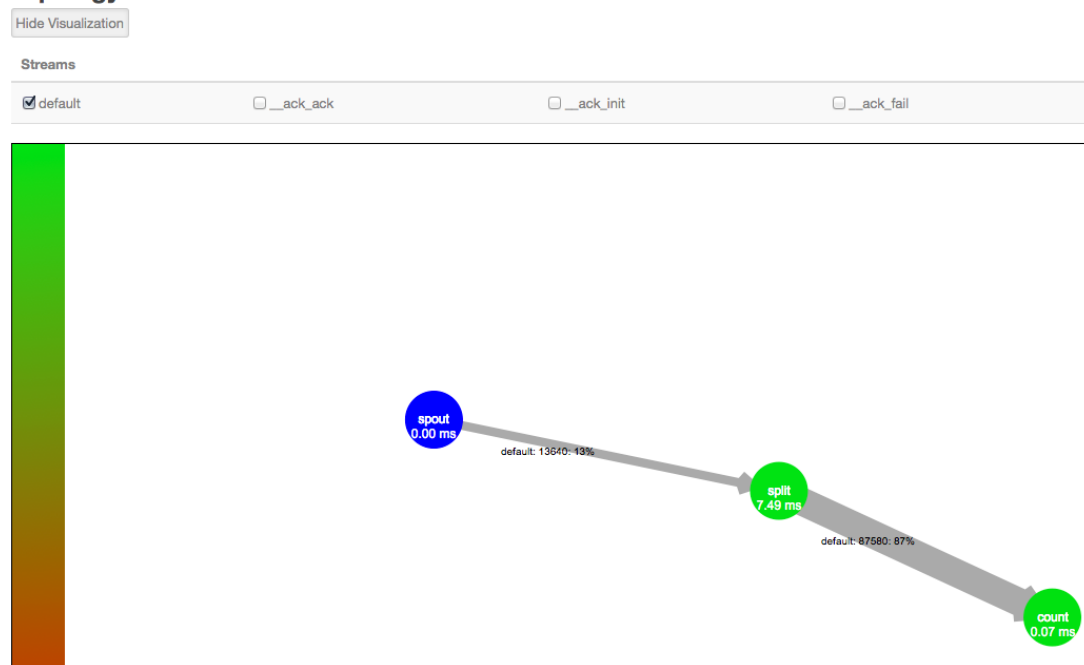
- What are my data sources?
- At what rate do these data sources deliver messages?
- What size are the messages?
- What is my slowest data sink?

The performance of a Storm topology degrades when it cannot ingest data fast enough to keep up with the data source. In addition, the velocity of incoming streaming data changes over time. When the data flow of the source exceeds what the topology can process, memory buffers fill up and the topology suffers frequent timeouts and must replay tuples to process them. (In contrast, MapReduce applications operate on data at rest in HDFS, with a constant data velocity. These applications suffer from poor latencies, but do not experience the buffer overflows and timeouts associated with streaming applications.)

Hortonworks recommends the following topology debugging technique to identify and overcome poor topology performance due to mismatched data flow rates between source and application.

1. Click Show Visualization in the Storm UI to display a visual representation of your topology and find the data bottleneck in your Storm application. Thicker lines between components denote larger data flows. A blue component represents the first component in the topology, such as the spout below from the WordCountTopology included with `storm-starter`. The color of the other topology components indicates whether the component is exceeding cluster capacity: red components denote a data bottleneck and green components indicate components operating within capacity.

Topology Visualization



Note

In addition to bolts in your topology, Storm uses its own bolts to perform background work when a topology component acknowledges that it either succeeded or failed to process a tuple. These names of these acker bolts are prefixed with an underscore in the visualization, but do not appear in the default view. You can display the component-specific data about successful acknowledgements by selecting the `__ack_ack` check box. Select the `__ack_fail` checkbox to display component-specific data about failed acknowledgements.

2. Verify that you have found the topology bottleneck by rewriting the `execute()` method of the target bolt or spout to perform no operations.

If the performance of the topology improves, you have found the bottleneck. Alternatively, turn off each topology component, one at a time, to find the component responsible for the processing bottleneck.

3. Increase the timeout value for the topology.

Edit the value of `topology.message.timeout.secs` in the `storm.yaml` configuration file. The default value is 30 seconds. This configuration parameter controls how long a tuple tree from the core-storm API or a batch from the Trident API has to complete processing before Storm times out and fails the operation.

4. Override the maximum number of tuples or batches waiting for processing before a spout temporarily stops emitting tuples to downstream bolts.

Edit the value of `topology.max.spout.pending` in the `storm.yaml` configuration file. The default is no limit. Hortonworks recommends that topologies using the core-storm API start with a value of 1000 and slowly decrease the value as necessary. Topologies using the Trident API should start with a much lower value, between 1 and 5.

5. Increase the parallelism for the target spout or bolt, as described in the next section.

2.2. Determining Topology Parallelism Units

Hortonworks recommends using the following calculation to determine the total number of parallelism units for a topology. Parallelism units are a useful conceptual tool for determining how to distribute processing tasks across a distributed application.

```
(number of worker nodes in cluster * number cores per worker node) - (number of acker tasks)
```

Acker tasks are topology components that acknowledge a successfully processed tuple. The following example assumes a Storm cluster with ten worker nodes, 16 CPU cores per worker node, and ten acker tasks in the topology. This Storm topology has 150 total parallelism units:

```
(10 * 16) - 10 = 150
```

Storm developers can mitigate the increased processing load associated with data persistence operations, such as writing to HDFS and generating reports, by distributing the most parallelism units to topology components that perform data persistence operations.

3. Streaming Data to Hive

Both the core-storm and Trident APIs support streaming data directly to Apache Hive using Hive transactions. Data committed in a transaction is immediately available to Hive queries from other Hive clients. Storm developers stream data to existing table partitions or configure the streaming Hive bolt to dynamically create desired table partitions. Use the following steps to perform this procedure:

1. Instantiate an implementation of the `HiveMapper` Interface.
2. Instantiate a `HiveOptions` class with the `HiveMapper` implementation.
3. Instantiate a `HiveBolt` with the `HiveOptions` class.



Note

Currently, data may be streamed only into bucketed tables using the ORC file format.

Core-storm API

```
DelimitedRecordHiveMapper mapper = new DelimitedRecordHiveMapper()
    .withColumnFields(new Fields(colNames));
HiveOptions hiveOptions = new
HiveOptions(metaStoreURI, dbName, tblName, mapper);
HiveBolt hiveBolt = new HiveBolt(hiveOptions);
```

Trident API

```
DelimitedRecordHiveMapper mapper = new DelimitedRecordHiveMapper()
    .withColumnFields(new Fields(colNames))
    .withTimeAsPartitionField("YYYY/MM/DD");

HiveOptions hiveOptions = new HiveOptions(metaStoreURI, dbName, tblName, mapper)
    .withTxnsPerBatch(10)
    .withBatchSize(1000)
    .withIdleTimeout(10);

StateFactory factory = new HiveStateFactory().withOptions(hiveOptions);
TridentState state = stream.partitionPersist(factory, hiveFields, new
    HiveUpdater(),
    new Fields());
```

The rest of this topic describes these steps in greater detail.

Instantiate an Implementation of `HiveMapper` Interface

The storm-hive streaming bolt uses the `HiveMapper` interface to map the names of tuple fields to the names of Hive table columns. Storm provides two implementations: `DelimitedRecordHiveMapper` and `JsonRecordHiveMapper`. Both implementations take the same arguments.

Table 3.1. `HiveMapper` Arguments

Argument	Data Type	Description
<code>withColumnFields</code>	<code>backtype.storm.tuple.Fields</code>	The name of the tuple fields that you want to map to table column names.

Argument	Data Type	Description
withPartitionFields	backtype.storm.tuple.Fields	The name of the tuple fields that you want to map to table partitions.
withTimeAsPartitionField	String	Requests that table partitions be created with names set to system time. Developers can specify any Java-supported date format, such as "YYYY/MM/DD".

The following sample code illustrates how to use `DelimitedRecordHiveMapper`:

```
...
DelimitedRecordHiveMapper mapper = new DelimitedRecordHiveMapper()
    .withColumnFields(new Fields(colNames))
    .withPartitionFields(new Fields(partNames));

DelimitedRecordHiveMapper mapper = new DelimitedRecordHiveMapper()
    .withColumnFields(new Fields(colNames))
    .withTimeAsPartitionField("YYYY/MM/DD");
...
```

Instantiate a HiveOptions Class with the HiveMapper Implementation

Use the `HiveOptions` class to configure the transactions used by Hive to ingest the streaming data, as illustrated in the following code sample.

```
...
HiveOptions hiveOptions = new HiveOptions(metaStoreURI, dbName, tblName, mapper)
    .withTxnsPerBatch(10)
    .withBatchSize(1000)
    .withIdleTimeout(10);
...
```

The following table describes all configuration properties for the `HiveOptions` class.

Table 3.2. HiveOptions Class Configuration Properties

HiveOptions Configuration Property	Data Type	Description
metaStoreURI	String	Hive Metastore URI. Storm developers can find this value in <code>hive-site.xml</code> .
dbName	String	Database name
tblName	String	Table name
mapper	Mapper	Two properties that start with "org.apache.storm.hive.bolt.": <code>mapper.DelimitedRecordHiveMapper</code> <code>mapper.JsonRecordHiveMapper</code>
withTxnsPerBatch	Integer	Configures the number of desired transactions per transaction batch. Data from all transactions in a single batch form a single compaction file. Storm developers use this property in conjunction with the <code>withBatchSize</code> property to control the size of compaction files. The default value is 100. Hive stores data in base files that cannot be updated by HDFS. Instead,

HiveOptions Configuration Property	Data Type	Description
		Hive creates a set of delta files for each transaction that alters a table or partition and stores them in a separate delta directory. Occasionally, Hive compacts, or merges, the base and delta files. Hive performs all compactions in the background without affecting concurrent reads and writes of other Hive clients. See Transactions for more information about Hive compactions.
withMaxOpenConnections	Integer	Specifies the maximum number of open connections. Each connection is to a single Hive table partition. The default value is 500. When Hive reaches this threshold, an idle connection is terminated for each new connection request. A connection is considered idle if no data is written to the table partition to which the connection is made.
withBatchSize	Integer	Specifies the maximum number of Storm tuples written to Hive in a single Hive transaction. The default value is 15000 tuples.
withCallTimeout	Integer	Specifies the interval in seconds between consecutive heartbeats sent to Hive. Hive uses heartbeats to prevent expiration of unused transactions. Set this value to 0 to disable heartbeats. The default value is 240.
withAutoCreatePartitions	Boolean	Indicates whether HiveBolt should automatically create the necessary Hive partitions needed to store streaming data. The default value is true.
withKerberosPrincipal	String	Kerberos user principal for accessing a secured Hive installation.
withKerberosKeytab	String	Kerberos keytab for accessing a secured Hive installation.

Instantiate the HiveBolt with the HiveOptions class

The next step is to instantiate the Hive streaming bolt. The core-storm and Trident APIs use different classes, as demonstrated in the following code samples:

Core-storm API

```
...
HiveBolt hiveBolt = new HiveBolt(hiveOptions);
...
```

Trident API

```
...
StateFactory factory = new HiveStateFactory().withOptions(hiveOptions);
TridentState state = stream.partitionPersist(factory, hiveFields, new
    HiveUpdater(),
    new Fields());
...
```

4. Ingesting Data with the Apache Kafka Spout Connector

Apache Kafka is a high-throughput, distributed messaging system. Apache Storm provides a Kafka spout to facilitate ingesting data from Kafka 0.8x brokers. Storm developers should include downstream bolts in their topologies to process data ingested with the Kafka spout.

The `storm-kafka` components include a core-storm spout, as well as a fully transactional Trident spout. Storm-kafka spouts provide the following key features:

- 'Exactly once' tuple processing with the Trident API
- Dynamic discovery of Kafka brokers and partitions

Hortonworks recommends that Storm developers use the Trident API. However, use the core-storm API if sub-second latency is critical for your Storm topology.

- The core-storm API represents a Kafka spout with the `KafkaSpout` class.
- The Trident API provides a `OpaqueTridentKafkaSpout` class to represent the spout.

To initialize `KafkaSpout` and `OpaqueTridentKafkaSpout`, Storm developers need an instance of a subclass of the `KafkaConfig` class, which represents configuration information needed to ingest data from a Kafka cluster.

- The `KafkaSpout` constructor requires the `SpoutConfig` subclass.
- The `OpaqueTridentKafkaSpout` requires the `TridentKafkaConfig` subclass.

In turn, the constructors for both `KafkaSpout` and `OpaqueTridentKafkaSpout` require an implementation of the `BrokerHosts` interface, which is used to map Kafka brokers to topic partitions. The storm-kafka component provides two implementations of `BrokerHosts`: `ZkHosts` and `StaticHosts`.

- Use the `ZkHosts` implementation to dynamically track broker-to-partition mapping.
- Use the `StaticHosts` implementation for static broker-to-partition mapping.

4.1. KafkaSpout and OpaqueTridentKafkaSpout Examples

The following code samples demonstrate the use of the `KafkaSpout` and `OpaqueTridentKafkaSpout` classes and related interfaces.

Core-storm API

```
BrokerHosts hosts = new ZkHosts(zkConnString);
SpoutConfig spoutConfig = new SpoutConfig(hosts,topicName,"/" + topicName,
    UUID.randomUUID().toString());
```

```
spoutConfig.scheme = new SchemeAsMultiScheme(new StringScheme());
KafkaSpout kafkaSpout = new KafkaSpout(spoutConfig);
```

Trident API

```
TridentTopology topology = new TridentTopology();
BrokerHosts zk = new ZkHosts("localhost");
TridentKafkaConfig spoutConf = new TridentKafkaConfig(zk, "test-topic");
spoutConf.scheme = new SchemeAsMultiScheme(new StringScheme());
OpaqueTridentKafkaSpout spout = new OpaqueTridentKafkaSpout(spoutConf);
```

```
zookeeper.connect=host1:2181,host2:2181,host3:2181
```

4.2. Storm-Kafka API Reference

Javadoc for the storm-kafka component is installed at `<$STORM_HOME>/contrib/storm-kafka/storm-kafka-0.9.3.2.2.6.0-<buildnumber>-javadoc.jar`. This section provides additional reference documentation for the primary classes and interfaces of the storm-kafka component.

BrokerHosts Interface

The storm-kafka component provides two implementations of the `BrokerHosts` interface: `ZkHosts` and `StaticHosts`. Use the `ZkHosts` implementation to dynamically track broker-to-partition mapping and the `StaticHosts` implementation when broker-to-partition mapping is static. The constructor for `StaticHosts` requires an instance of `GlobalPartitionInformation`:

```
Broker brokerForPartition0 = new Broker("localhost");//localhost:9092
Broker brokerForPartition1 = new Broker("localhost", 9092);//localhost:9092
  but we specified the port explicitly
Broker brokerForPartition2 = new Broker("localhost:9092");//localhost:9092
  specified as one string.
GlobalPartitionInformation partitionInfo = new GlobalPartitionInformation();
partitionInfo.add(0, brokerForPartition0)//mapping form partition 0 to
  brokerForPartition0
partitionInfo.add(1, brokerForPartition1)//mapping form partition 1 to
  brokerForPartition1
partitionInfo.add(2, brokerForPartition2)//mapping form partition 2 to
  brokerForPartition2
StaticHosts hosts = new StaticHosts(partitionInfo);
```

KafkaConfig Class

Instantiate an instance of `KafkaConfig` with one of the following constructors, each of which requires an implementation of the `BrokerHosts` interface:

```
public KafkaConfig(BrokerHosts hosts, String topic)
public KafkaConfig(BrokerHosts hosts, String topic, String clientId)
```

Table 4.1. KafkaConfig Parameters

KafkaConfig Parameter	Description
hosts	Any implementation of the <code>BrokerHosts</code> interface, currently either <code>ZkHosts</code> or <code>StaticHosts</code> .
topic	Name of the Kafka topic.

KafkaConfig Parameter	Description
clientId	Optional parameter used as part of the ZooKeeper path where the spout's current offset is stored.

Both `SpoutConfig` from the core-storm API and `TridentKafkaConfig` from the Trident API extend `KafkaConfig`. Instantiate these classes with the following constructors:

Core-Storm API

Constructor `public SpoutConfig(BrokerHosts hosts, String topic, String zkRoot, String id)`

Table 4.2. SpoutConfig Parameters

SpoutConfig Parameter	Description
hosts	Any implementation of the <code>BrokerHosts</code> interface, currently either <code>ZkHosts</code> or <code>StaticHosts</code> .
topic	Name of the Kafka topic.
zkroot	Root directory in Zookeeper where all topics and partition information is stored. By default, this is <code>/brokers</code> .
id	Unique identifier for this spout.

Trident API

Constructors `public TridentKafkaConfig(BrokerHosts hosts, String topic)` `public TridentKafkaConfig(BrokerHosts hosts, String topic, String id)`

Table 4.3. TridentKafkaConfig Parameters

TridentKafkaConfig	Description
hosts	Any implementation of the <code>BrokerHosts</code> interface, currently either <code>ZkHosts</code> or <code>StaticHosts</code> .
topic	Name of the Kafka topic.
clientId	Unique identifier for this spout.

`KafkaConfig` contains several fields used to configure the behavior of a Kafka spout in a Storm topology:

Table 4.4. KafkaConfig Fields

KafkaConfig Field	Description
fetchSizeBytes	Specifies the number of bytes to attempt to fetch in one request to a Kafka server. The default is 1MB.
socketTimeoutMs	Specifies the number of milliseconds to wait before a socket fails an operation with a timeout. The default value is 10 seconds.
bufferSizeBytes	Specifies the buffer size in bytes for network requests. The default is 1MB.
scheme	The interface that specifies how a <code>byte[]</code> from a Kafka topic is transformed into a Storm tuple. The default, <code>MultiScheme</code> , returns a tuple with the <code>byte[]</code> and no additional processing. The API provides the following

KafkaConfig Field	Description
	implementations: * storm.kafka.StringScheme * storm.kafka.KeyValueSchemeAsMultiScheme * storm.kafka.StringKeyValueScheme * storm.kafka.KeyValueSchemeAsMultiScheme
forceFromStart	Controls whether a Kafka spout fetches data from the beginning of a Kafka topic. The default is false.
startOffsetTime	Controls whether streaming for a topic starts from the beginning of the topic or whether only new messages are streamed. The following are valid values: * kafka.api.OffsetRequest.EarliestTime() - starts streaming from the beginning of the topic * kafka.api.OffsetRequest.LatestTime() - streams only new messages
maxOffsetBehind	Specifies how long a spout attempts to retry the processing of a failed tuple. If a failing tuple's offset is less than maxOffsetBehind, the spout stops retrying the tuple. The default is LONG.MAX_VALUE.
useStartOffsetTimeOfOffsetOutOfRange	Controls whether a spout streams messages from the beginning of a topic when the spout throws an exception for an out-of-range offset. The default value is true.
metricsTimeBucketSizeInSecs	Controls the time interval at which Storm reports spout-related metrics. The default is 60 seconds.

4.3. KafkaSpout Limitations

Limitations

The current version of the Kafka spout contains the following limitations:

- Does not support Kafka 0.7x brokers.
- Storm developers must include `${STORM_HOME}/lib/*` in the CLASSPATH environment variable from the command line when running `kafka-topology` in local mode. Otherwise, developers will likely receive a `java.lang.NoClassDefFoundError` exception:

```
java -cp "/usr/lib/storm/contrib/storm-kafka-example-0.9.1.2.1.1.0-320-jar-with-dependencies.jar:  
/usr/lib/storm/lib/*" org.apache.storm.kafka.TestKafkaTopology  
<zookeeper_host>
```

- Secure Hadoop clusters must comment out the following statement from `${STORM_HOME}/bin/kafka-server-start.sh`:

```
EXTRA_ARGS="-name kafkaServer -loggc"
```

- Core-storm API Constructor

```
public SpoutConfig(BrokerHosts hosts, String topic, String zkRoot, String  
id)
```

Table 4.5. SpoutConfig Parameters

SpoutConfig Parameter	Description
hosts	Any implementation of the BrokerHosts interface, currently either ZkHosts or StaticHosts.

SpoutConfig Parameter	Description
topic	Name of the Kafka topic.
zkroot	Root directory in Zookeeper where all topics and partition information is stored. By default, this is / brokers.
id	Unique identifier for this spout.

- Trident API Constructors

```
public TridentKafkaConfig(BrokerHosts hosts, String topic)
public TridentKafkaConfig(BrokerHosts hosts, String topic, String id)
```

Table 4.6. TridentKafkaConfig Parameters

TridentKafkaConfig	Description
hosts	Any implementation of the BrokerHosts interface, currently either ZkHosts or StaticHosts.
topic	Name of the Kafka topic.
clientid	Unique identifier for this spout.

4.4. Configuring Kafka for Use with the Storm-Kafka Connector

The storm-kafka connector requires configuration of the Apache Kafka installation.

Kafka administrators must add a `zookeeper.connect` property, with the hostnames and port numbers of the HDP Zookeeper nodes, to Kafka's `server.properties` file.

4.5. Configuring KafkaSpout to Connect to a Secure Kafka Cluster

To connect to a Kerberized Kafka topic:

1. **Code:** Add `spoutConfig.securityProtocol=PLAINTEXTSASL` to your Kafka Spout configuration.
2. **Configuration:** Add a `KafkaClient` section (excerpted from `/usr/hdp/current/kafka-broker/config/kafka_jaas.conf`) to `/usr/hdp/current/storm-supervisor/conf/storm_jaas.conf`:

```
KafkaClient {
    com.sun.security.auth.module.Krb5LoginModule required
    useKeyTab=true
    keyTab="/etc/security/keytabs/stormusr.service.keytab"
    storeKey=true
    useTicketCache=false
    serviceName="kafka"
    principal="stormusr/host.name@EXAMPLE.COM" ;
};
```

3. **Setup:** Add a Kafka ACL for the topic. For example:

```
bin/kafka-acls.sh --authorizer
kafka.security.auth.SimpleAclAuthorizer --authorizer-properties
zookeeper.connect=localhost:2181 --add --allow-principal
User:stormusr --allow-hosts * --operations Read --topic TEST
```

4.6. Configuring KafkaSpout to Connect to HBase or Hive

Before connecting to HBase or Hive, we recommend that you add the following exclusions to your POM file, for the curator framework:

```
<exclusion>
  <groupId>org.apache.curator</groupId>
  <artifactId>curator-client</artifactId>
</exclusion>

<exclusion>
  <groupId>org.apache.curator</groupId>
  <artifactId>curator-recipes</artifactId>
</exclusion>

<exclusion>
  <groupId>org.apache.curator</groupId>
  <artifactId>apache-curator</artifactId>
</exclusion>
```

5. Writing Data with Storm

Hortonworks provides a set of connectors that enable Storm developers to quickly write streaming data to a Hadoop cluster. These connectors are located at `/usr/lib/storm/contrib`. Each contains a `.jar` file containing the connector's packaged classes and dependencies, and another `.jar` file with javadoc reference documentation.

5.1. Writing Data to HDFS with the Storm-HDFS Connector

The `storm-hdfs` connector supports the following key features:

- HDFS 2.x
- High availability-enabled clusters
- Text and sequence files
- Configurable directory and file names
- Customizable synchronization, rotation policies, and rotation actions
- Tuple failure if HDFS update fails
- The Trident API
- Writing to a Kerberized Hadoop cluster (for more information, see [Configuring Connectors for a Secure Cluster](#))

The primary classes of the `storm-hdfs` connector are `HdfsBolt` and `SequenceFileBolt`, both located in the `org.apache.storm.hdfs.bolt` package. Use the `HdfsBolt` class to write text data to HDFS and the `SequenceFileBolt` class to write binary data.

For more information about the `HdfsBolt` class, refer to the Apache Storm [HdfsBolt API documentation](#).

Storm developers specify the following information when instantiating the bolt:

Table 5.1. HdfsBolt Methods

HdfsBolt Method	Description
<code>withFsUrl</code>	Specifies the target HDFS URL and port number.
<code>withRecordFormat</code>	Specifies the delimiter that indicates a boundary between data records. Storm developers can customize by writing their own implementation of the <code>org.apache.storm.hdfs.format.RecordFormat</code> interface. Use the provided <code>org.apache.storm.hdfs.format.DelimitedRecordFormat</code> class as a convenience class for writing delimited text data with delimiters such as tabs, comma-separated values, and pipes. The <code>storm-hdfs</code>

HdfsBolt Method	Description
	bolt uses the <code>RecordFormat</code> implementation to convert tuples to byte arrays, so this method can be used with both text and binary data.
<code>withRotationPolicy</code>	Specifies when to stop writing to a data file and begin writing to another. Storm developers can customize by writing their own implementation of the <code>org.apache.storm.hdfs.rotation.FileSizeRotationSizePolicy</code> interface.
<code>withSyncPolicy</code>	Specifies how frequently to flush buffered data to the HDFS filesystem. This action enables other Hive clients to read the synchronized data, even as the Storm client continues to write data. Storm developers can customize by writing their own implementation of the <code>org.apache.storm.hdfs.sync.SyncPolicy</code> interface.
<code>withFileNameFormat</code>	Specifies the name of the data file. Storm developers can customize by writing their own interface of the <code>org.apache.storm.hdfs.format.FileNameFormat</code> interface. The provided <code>org.apache.storm.hdfs.format.DefaultFileNameFormat</code> creates file names with the following naming format: {prefix}-{componentId}-{taskId}-{rotationNum}-{timestamp}-{extension}. Example: <code>MyBolt-5-7-1390579837830.txt</code> .

Example: Cluster Without High Availability ("HA")

The following example writes pipe-delimited files to the HDFS path `hdfs://localhost:8020/foo`. After every 1,000 tuples it will synchronize with the filesystem, making the data visible to other HDFS clients. It will rotate the files when they reach 5 MB in size.

Note that the `HdfsBolt` is instantiated with an HDFS URL and port number.

```

```java
// use "|" instead of "," for field delimiter
RecordFormat format = new DelimitedRecordFormat()
 .withFieldDelimiter("|");

// Synchronize the filesystem after every 1000 tuples
SyncPolicy syncPolicy = new CountSyncPolicy(1000);

// Rotate data files when they reach 5 MB
FileRotationPolicy rotationPolicy = new FileSizeRotationPolicy(5.0f, Units.
 MB);

// Use default, Storm-generated file names
FileNameFormat fileNameFormat = new DefaultFileNameFormat()
 .withPath("/foo/");

// Instantiate the HdfsBolt
HdfsBolt bolt = new HdfsBolt()
 .withFsUrl("hdfs://localhost:8020")
 .withFileNameFormat(fileNameFormat)
 .withRecordFormat(format)
 .withRotationPolicy(rotationPolicy)
 .withSyncPolicy(syncPolicy);

```

### Example: HA-Enabled Cluster

The following example shows how to modify the previous example for an HA-enabled cluster.

Here the `HdfsBolt` is instantiated with a nameservice ID, instead of using an HDFS URL and port number.

```
...
HdfsBolt bolt = new HdfsBolt()
 .withFsURL("hdfs://myNameserviceID")
 .withFileNameFormat(fileNameFormat)
 .withRecordFormat(format)
 .withRotationPolicy(rotationPolicy)
 .withSyncPolicy(syncPolicy);
...
```

To obtain the nameservice ID, check the `dfs.nameservices` property in your `hdfs-site.xml` file; `nnha` in the following example:

```
<property>
 <name>dfs.nameservices</name>
 <value>nnha</value>
</property>
```

### Trident API

The `storm-hdfs` connector supports the Trident API. Hortonworks recommends that Storm developers use the `trident` API unless your application requires sub-second latency.

The Trident API implements a `StateFactory` class with an API that resembles the methods from the `storm-code` API as shown in the following code sample:

```
...
Fields hdfsFields = new Fields("field1", "field2");

FileNameFormat fileNameFormat = new DefaultFileNameFormat()
 .withPrefix("trident")
 .withExtension(".txt")
 .withPath("/trident");

RecordFormat recordFormat = new DelimitedRecordFormat()
 .withFields(hdfsFields);

FileRotationPolicy rotationPolicy = new FileSizeRotationPolicy(5.0f,
 FileSizeRotationPolicy.Units.MB);

HdfsState.Options options = new HdfsState.HdfsFileOptions()
 .withFileNameFormat(fileNameFormat)
 .withRecordFormat(recordFormat)
 .withRotationPolicy(rotationPolicy)
 .withFsUrl("hdfs://localhost:8020");

StateFactory factory = new HdfsStateFactory().withOptions(options);

TridentState state = stream.partitionPersist(factory, hdfsFields, new
 HdfsUpdater(), new Fields());
```

See the javadoc for the Trident API, included with the `storm-hdfs` connector, for more information.

### Limitations

Directory and file names changes are limited to a prepackaged file name format based on a timestamp.

## 5.2. Writing Data to HBase with the Storm-HBase Connector

The `storm-hbase` connector supports the following key features:

- Apache HBase 0.96 and above
- Incrementing counter columns
- Tuples failure if an update to an HBase table fails
- Ability to group puts in a single batch
- Writing to Kerberized HBase clusters (for more information, see [Configuring Connectors for a Secure Cluster](#))

The `storm-hbase` connector enables Storm developers to collect several *PUTS* in a single operation and write to multiple HBase column families and counter columns. A PUT is an HBase operation that inserts data into a single HBase cell. Use the HBase client's write buffer to automatically batch: `hbase.client.write.buffer`. The primary interface in the `storm-hbase` connector is the `org.apache.storm.hbase.bolt.mapper.HBaseMapper` interface. However, the default implementation, `SimpleHBaseMapper`, writes a single column family. Storm developers can implement the `HBaseMapper` interface themselves or extend `SimpleHBaseMapper` if they want to change or override this behavior.

**Table 5.2. SimpleHBaseMapper Methods**

SimpleHBaseMapper Method	Description
<code>withRowKeyField</code>	Specifies the row key for the target HBase row. A row key uniquely identifies a row in HBase.
<code>withColumnFields</code>	Specifies the target HBase column.
<code>withCounterFields</code>	Specifies the target HBase counter.
<code>withColumnFamily</code>	Specifies the target HBase column family.

### Example

The following example specifies the 'word' tuple as the row key, adds an HBase column for the tuple 'word' field, adds an HBase counter column for the tuple 'count' field, and writes data to the 'cf' column family.

```
SimpleHBaseMapper mapper = new SimpleHBaseMapper()
 .withRowKeyField("word")
 .withColumnFields(new Fields("word"))
 .withCounterFields(new Fields("count"))
```

```
.withColumnFamily("cf");
```

The `storm-hbase` connector supports the following versions of HBase:

- 0.96
- 0.98

### Limitations

The current version of the `storm-hbase` connector has the following limitations:

- HBase table must be predefined
- Cannot dynamically add new HBase columns; can write to only one column family at a time
- Assumes that `hbase-site.xml` is in the `$CLASSPATH` environment variable
- Tuple field names must match HBase column names
- Does not support the Trident API
- Supports writes but not lookups

## 5.3. Writing Data to Kafka using a Kafka Bolt

Storm provides a Kafka Bolt for both the core-storm and Trident APIs that writes data to a Kafka cluster, also known as publishing to a topic using Kafka's terminology. Use the following procedure to add a Storm component to your topology that writes data to a Kafka cluster:

1. Instantiate a Kafka Bolt.
2. Configure the Kafka Bolt with a Tuple-to-Message mapper.
3. Configure the Kafka Bolt with a Kafka Topic Selector.
4. Configure the Kafka Bolt with the Kafka Producer properties.

The following code samples for each API illustrate the construction of a simple Kafka Bolt. The rest of this topic breaks the samples down to better describe each step.

### Core-storm API

```
TopologyBuilder builder = new TopologyBuilder();

Fields fields = new Fields("key", "message");
FixedBatchSpout spout = new FixedBatchSpout(fields, 4,
new Values("storm", "1"),
new Values("trident", "1"),
new Values("needs", "1"),
new Values("javadoc", "1")
);

spout.setCycle(true);
```



```

builder.setSpout("spout", spout, 5);
KafkaBolt bolt = new KafkaBolt()
 .withKafkaTopicSelector(new DefaultTopicSelector("test"))
 .withTridentTupleToKafkaMapper(new FieldNameBasedTupleToKafkaMapper());
builder.setBolt("forwardToKafka", bolt, 8).shuffleGrouping("spout");

Config conf = new Config();
//set producer properties.
Properties props = new Properties();
props.put("metadata.broker.list", "localhost:9092");
props.put("request.required.acks", "1");
props.put("serializer.class", "kafka.serializer.StringEncoder");
conf.put(TridentKafkaState.KAFKA_BROKER_PROPERTIES, props);

StormSubmitter.submitTopology("kafkaboltTest", conf, builder.
createTopology());

```

## Trident API

```

Fields fields = new Fields("word", "count");
FixedBatchSpout spout = new FixedBatchSpout(fields, 4,
 new Values("storm", "1"),
 new Values("trident", "1"),
 new Values("needs", "1"),
 new Values("javadoc", "1")
);

spout.setCycle(true);

TridentTopology topology = new TridentTopology();
Stream stream = topology.newStream("spout1", spout);

TridentKafkaStateFactory stateFactory = new TridentKafkaStateFactory()
 .withKafkaTopicSelector(new DefaultTopicSelector("test"))
 .withTridentTupleToKafkaMapper(new FieldNameBasedTupleToKafkaMapper("word",
"count"));
stream.partitionPersist(stateFactory, fields, new TridentKafkaUpdater(), new
Fields());

Config conf = new Config();
//set producer properties.
Properties props = new Properties();
props.put("metadata.broker.list", "localhost:9092");
props.put("request.required.acks", "1");
props.put("serializer.class", "kafka.serializer.StringEncoder");
conf.put(TridentKafkaState.KAFKA_BROKER_PROPERTIES, props);
StormSubmitter.submitTopology("kafkaTridentTest", conf, topology.build());

```

## Instantiate a KafkaBolt

The core-storm API uses the `storm.kafka.bolt.KafkaBolt` class to instantiate a Kafka Bolt. The Trident API uses a combination of the `storm.kafka.trident.TridentStateFactory` and `storm.kafka.trident.TridentKafkaStateFactory` classes.

## Core-storm API

```
KafkaBolt bolt = new KafkaBolt();
```

## Trident API

```
TridentTopology topology = new TridentTopology();
Stream stream = topology.newStream("spout");
TridentKafkaStateFactory stateFactory = new TridentKafkaStateFactory();
stream.partitionPersist(stateFactory, fields, new TridentKafkaUpdater(), new
Fields());
```

### Configure the KafkaBolt with a Tuple-to-Message Mapper

The KafkaBolt must map Storm tuples to Kafka messages. By default, KafkaBolt looks for fields named "key" and "message." Storm provides the `storm.kafka.trident.mapper.FieldNameBasedTupleToKafkaMapper` class to support this default behavior and provide backward compatibility. The class is used by both the core-storm and Trident APIs.

#### Core-storm API

```
KafkaBolt bolt = new KafkaBolt()
 .withTridentTupleToKafkaMapper(new FieldNameBasedTupleToKafkaMapper());
```

#### Trident API

```
TridentKafkaStateFactory stateFactory = new TridentKafkaStateFactory()
 .withTridentTupleToKafkaMapper(new FieldNameBasedTupleToKafkaMapper("word",
"count"));
```

Storm developers must specify the field names for the Storm tuple key and the Kafka message for any implementation of the `TridentKafkaState` in the Trident API. This interface does not provide a default constructor.

However, some Kafka bolts may require more than two fields. Storm developers may write their own implementation of the `TupleToKafkaMapper` and `TridentTupleToKafkaMapper` interfaces to customize the mapping of Storm tuples to Kafka messages. Both interfaces define 2 methods:

```
K getKeyFromTuple(Tuple/TridentTuple tuple);
```

```
V getMessageFromTuple(Tuple/TridentTuple tuple);
```

### Configure the Kafka Bolt with a Kafka Topic Selector



#### Note

To ignore a message, return NULL from the `getTopics()` method.

#### Core-storm API

```
KafkaBolt bolt = new KafkaBolt().withTupleToKafkaMapper(new
 FieldNameBasedTupleToKafkaMapper())
 .withTopicSelector(new DefaultTopicSelector());
```

#### Trident API

```
TridentKafkaStateFactory stateFactory = new TridentKafkaStateFactory()
 .withKafkaTopicSelector(new DefaultTopicSelector("test"))
 .withTridentTupleToKafkaMapper(new FieldNameBasedTupleToKafkaMapper("word",
"count"));
```

Storm developers can write their own implementation of the `KafkaTopicSelector` interface if they need to write to multiple Kafka topics:

```
public interface KafkaTopicSelector {
 String getTopics(Tuple/TridentTuple tuple);
}
```

## 5.4. Configuring Connectors for a Secure Cluster

If your topology uses Storm-HDFS, Storm-HBase, or Storm-Hive connectors, and if the corresponding components HDFS, HBase, and/or Hive are secured with Kerberos, then you will need to complete the following additional configuration steps.

### Storm-HDFS Connector Configuration

To use the `storm-hdfs` connector in topologies that run on secure clusters:

1. Provide your own Kerberos keytab and principal name to the connectors. The `Config` object that you pass into the topology must contain the storm keytab file and principal name.
2. Specify an `HdfsBolt` `configKey`, using the method `HdfsBolt.withConfigKey("somekey")`. The value map of this key should have the following two properties:

```
hdfs.keytab.file: "<path-to-keytab>"
```

```
hdfs.kerberos.principal: "<principal>@<host>"
```

where

`<path-to-keytab>` specifies the path to the keytab file on the supervisor hosts

`<principal>@<host>` specifies the user and domain; for example, `storm-admin@EXAMPLE.com`.

For example:

```
Config config = new Config();
config.put(HdfsSecurityUtil.STORM_KEYTAB_FILE_KEY, "$keytab");
config.put(HdfsSecurityUtil.STORM_USER_NAME_KEY, "$principal");

StormSubmitter.submitTopology("$topologyName", config, builder.
 createTopology());
```

On worker hosts the bolt/trident-state code will use the keytab file and principal to authenticate with the NameNode. Make sure that all workers have the keytab file, stored in the same location.



### Note

For more information about the `HdfsBolt` class, refer to the [Apache Storm HdfsBolt API documentation](#).

3. Distribute the keytab file that the Bolt is using in the `Config` object, to all supervisor nodes. This is the keytab that is being used to authenticate to HDFS, typically the Storm service keytab, `storm`. The user ID that the Storm worker is running under should have access to it.

On an Ambari-managed cluster this is `/etc/security/keytabs/storm.service.keytab` (the "path-to-keytab"), where the worker runs under `storm`.

4. If you set `supervisor.run.worker.as.user` to `true` (see [Running Workers as Users](#) in *Configuring Storm for Kerberos over Ambari*), make sure that the user that the workers are running under (typically the `storm` keytab) has read access on those keytabs. This is a manual step; an admin needs to go to each supervisor node and run `chmod` to give file system permissions to the users on these keytab files.



### Note

You do not need to create separate keytabs or principals; the general guideline is to create a principal and keytab for each group of users that requires the same access to these resources, and use that single keytab.

All of these connectors accept topology configurations. You can specify the keytab location on the host and the principal through which the connector will login to that system.

5. Configure the connector(s). Here is a sample configuration for the Storm-HDFS connector (see [Writing Data to HDFS with the Storm-HDFS Connector](#) for a more extensive example):

```
HdfsBolt bolt = new HdfsBolt()
 .withFsUrl("hdfs://localhost:8020")
 .withFileNameFormat(fileNameFormat)
 .withRecordFormat(format)
 .withRotationPolicy(rotationPolicy)
 .withSyncPolicy(syncPolicy);
.withConfigKey("hdfs.config");

Map<String, Object> map = new HashMap<String, Object>();
map.put("hdfs.keytab.file", "/etc/security/keytabs/storm.service.keytab");
map.put("hdfs.kerberos.principal", "storm@TEST.HORTONWORKS.COM");

Config config = new Config();
config.put("hdfs.config", map);

StormSubmitter.submitTopology("$topologyName", config, builder.
createTopology());
```



### Important

For the Storm-HDFS connector, you must package `hdfs-site.xml` and `core-site.xml` (from your cluster configuration) in the topology `.jar` file.

In addition, include any configuration files for HDP components used in your Storm topology, such as `hive-site.xml` and `hbase-site.xml`. This fulfills the requirement that all related configuration files appear in the CLASSPATH of your Storm topology at runtime.

## Storm-HBase Connector Configuration

To use the `storm-hbase` connector in topologies that run on secure clusters:

1. Provide your own Kerberos keytab and principal name to the connectors. The `Config` object that you pass into the topology must contain the storm keytab file and principal name.
2. Specify an `HBaseBolt configKey`, using the method `HBaseBolt.withConfigKey("somekey")`. The value map of this key should have the following two properties:

```
storm.keytab.file: "<path-to-keytab-file>"
```

```
storm.kerberos.principal: "<principal>@<host>"
```

For example:

```
Config config = new Config();
config.put(HBaseSecurityUtil.STORM_KEYTAB_FILE_KEY, "$keytab");
config.put(HBaseSecurityUtil.STORM_USER_NAME_KEY, "$principal");

StormSubmitter.submitTopology("$topologyName", config, builder.
createTopology());
```

On worker hosts the bolt/trident-state code will use the keytab file and principal to authenticate with the NameNode. Make sure that all workers have the keytab file, stored in the same location.



### Note

For more information about the `HBaseBolt` class, refer to the Apache Storm [HBaseBolt API documentation](#).

3. Distribute the keytab file that the Bolt is using in the `Config` object, to all supervisor nodes. This is the keytab that is being used to authenticate to HBase, typically the Storm service keytab, `storm`. The user ID that the Storm worker is running under should have access to it.

On an Ambari-managed cluster this is `/etc/security/keytabs/storm.service.keytab` (the "path-to-keytab"), where the worker runs under `storm`.

4. If you set `supervisor.run.worker.as.user` to `true` (see [Running Workers as Users](#) in *Configuring Storm for Kerberos over Ambari*), make sure that the user that the workers are running under (typically the `storm` keytab) has read access on those keytabs. This is a manual step; an admin needs to go to each supervisor node and run `chmod` to give file system permissions to the users on these keytab files.



### Note

You do not need to create separate keytabs or principals; the general guideline is to create a principal and keytab for each group of users that requires the same access to these resources, and use that single keytab.

All of these connectors accept topology configurations. You can specify the keytab location on the host and the principal through which the connector will login to that system.

## 5. Configure the connector(s). Here is a sample configuration for the Storm-HBase connector:

```
HBaseBolt hbase = new HBaseBolt("WordCount", mapper).withConfigKey("hbase.config");

Map<String, Object> mapHbase = new HashMap<String, Object>();
mapHbase.put("storm.keytab.file", "/etc/security/keytabs/storm.service.keytab");
mapHbase.put("storm.kerberos.principal", "storm@TEST.HORTONWORKS.COM");

Config config = new Config();
config.put("hbase.config", mapHbase);

StormSubmitter.submitTopology("$topologyName", config, builder.createTopology());
```



### Important

For the Storm-HBase connector, you must package `hdfs-site.xml`, `core-site.xml`, and `hbase-site.xml` (from your cluster configuration) in the topology .jar file.

In addition, include any other configuration files for HDP components used in your Storm topology, such as `hive-site.xml`. This fulfills the requirement that all related configuration files appear in the CLASSPATH of your Storm topology at runtime.

## Storm-Hive Connector Configuration

The Storm-Hive connector accepts configuration settings as part of the `HiveOptions` class. For more information about the `HiveBolt` and `HiveOptions` classes, see the Apache Storm [HiveOptions](#) and [HiveBolt](#) API documentation.

There are two required settings for accessing secure Hive:

- `withKerberosPrincipal`, the Kerberos principal for accessing Hive:

```
public HiveOptions withKerberosPrincipal(String kerberosPrincipal)
```

- `withKerberosKeytab`, the Kerberos keytab for accessing Hive:

```
public HiveOptions withKerberosKeytab(String kerberosKeytab)
```

## 6. Packaging Storm Topologies

Storm developers should verify that the following conditions are met when packaging their topology into a .jar file:

- Use the maven-shade-plugin, rather than the maven-assembly-plugin to package your Apache Storm topologies. The maven-shade-plugin provides the ability to merge JAR manifest entries, which are used by the Hadoop client to resolve URL schemes.
- Include a dependency for the Hadoop version used in the Hadoop cluster.
- Include both of the Hadoop configuration files, `hdfs-site.xml` and `core-site.xml`, in the .jar file. In addition, include any configuration files for HDP components used in your Storm topology, such as `hive-site.xml` and `hbase-site.xml`. This is the easiest way to meet the requirement that all required configuration files appear in the CLASSPATH of your Storm topology at runtime.

### Maven Shade Plugin

Use the maven-shade-plugin, rather than the maven-assembly-plugin to package your Apache Storm topologies. The maven-shade-plugin provides the ability to merge JAR manifest entries, which are used by the Hadoop client to resolve URL schemes.

Use the following Maven configuration file to package your topology:

```
<plugin>
 <groupId>org.apache.maven.plugins</groupId>
 <artifactId>maven-shade-plugin</artifactId>
 <version>1.4</version>
 <configuration>
 <createDependencyReducedPom>true</createDependencyReducedPom>
 </configuration>
 <executions>
 <execution>
 <phase>package</phase>
 <goals>
 <goal>shade</goal>
 </goals>
 <configuration>
 <transformers>
 <transformer implementation="org.apache.maven.
plugins.shade.resource.ServicesResourceTransformer"/>
 <transformer implementation="org.apache.maven.
plugins.shade.resource.ManifestResourceTransformer">
 <mainClass></mainClass>
 </transformer>
 </transformers>
 </configuration>
 </execution>
 </executions>
</plugin>
```

### Hadoop Dependency

Include a dependency for the Hadoop version used in the Hadoop cluster; for example:

```
<dependency>
 <groupId>org.apache.hadoop</groupId>
 <artifactId>hadoop-client</artifactId>
 <version>2.7.1.2.4.2.0-258</version>
 <exclusions>
 <exclusion>
 <groupId>org.slf4j</groupId>
 <artifactId>slf4j-log4j12</artifactId>
 </exclusion>
 </exclusions>
</dependency>
```

## Troubleshooting

The following table describes common packaging errors.

**Table 6.1. Topology Packing Errors**

Error	Description
com.google.protobuf. InvalidProtocolBufferException: Protocol message contained an invalid tag (zero)	Hadoop client version incompatibility
java.lang.RuntimeException: Error preparing HdfsBolt: No FileSystem for scheme: hdfs	The .jar manifest files have not properly merged in the topology.jar



## 7. Deploying and Managing Apache Storm Topologies

Use the command-line interface to deploy a Storm topology after packaging it in a jar. For example, use the following command to deploy WordCountTopology from the `storm-starter` jar:

```
storm jar storm-starter-<starter_version>-storm-<storm_version>
.jar storm.starter.WordCountTopology WordCount -c nimbus.host=sandbox.
hortonworks.com
```

Point a browser to the following URL to access the Storm UI and to manage deployed topologies.

`http://<storm-ui-server>:8080`



### Note

You may need to configure Apache Storm to use a different port if Ambari is also running on the same host with Apache Storm. Both applications use port 8080 by default.

Storm UI

Cluster Summary

Version	Nimbus uptime	Supervisors	Used slots	Free slots	Total slots	Executors	Tasks
0.9.1.2.1.1.0-187	12s	0	0	0	0	0	0

Topology summary

Name	Id	Status	Uptime	Num workers	Num executors	Num tasks
WordCount	WordCount-1-1395077335	ACTIVE	52m 22s	0	0	0

Supervisor summary

Id	Host	Uptime	Slots	Used slots
----	------	--------	-------	------------

Nimbus Configuration

Key	Value
dev.zookeeper.path	/tmp/dev-storm-zookeeper
drpc.childopts	-Xmx768m
drpc.invocations.port	3773

In the image above, no workers, executors, or tasks are running. However, the status of the topology remains active and the uptime continues to increase. Storm topologies, unlike traditional applications, remain active until an administrator deactivates or kills them. Storm administrators use the Storm user interface to perform the following administrative actions:

**Table 7.1. Topology Administrative Actions**

Topology Administrative Action	Description
Activate	Returns a topology to active status after it has been deactivated.
Deactivate	Sets the status of a topology to inactive. Topology uptime is not affected by deactivation.

Topology Administrative Action	Description
Rebalance	Dynamically increase or decrease the number of worker processes and/or executors. The administrator does not need to restart the cluster or the topology.
Kill	Stops the topology and removes it from Apache Storm. The topology no longer appears in the Storm UI, and the administrator must deploy the application again to activate it.

Click any topology in the Topology Summary section to launch the Topology Summary page. Administrators perform any of the topology actions in the table above by clicking the corresponding button, shown in the following image.

## Storm UI

### Topology summary

Name	Id	Status	Uptime	Num workers	Num executors	Num tasks
WordCount	WordCount-1-1395077335	ACTIVE	2h 28m 31s	0	0	0

### Topology actions

Activate | Deactivate | Rebalance | Kill

### Topology stats

Window	Emitted	Transferred	Complete latency (ms)	Acked	Failed
All time			0		

### Spouts (All time)

Id	Executors	Tasks	Emitted	Transferred	Complete latency (ms)	Acked	Failed	Last error
----	-----------	-------	---------	-------------	-----------------------	-------	--------	------------

### Bolts (All time)

Id	Executors	Tasks	Emitted	Transferred	Capacity (last 10m)	Execute latency (ms)	Executed	Process latency (ms)	Acked	Failed	Last error
----	-----------	-------	---------	-------------	---------------------	----------------------	----------	----------------------	-------	--------	------------

### Topology Configuration

Key	Value
dev.zookeeper.path	/tmp/dev-storm-zookeeper
drpc.childopts	-Xmx768m

The Executors field in the Spouts and Bolts sections show all running Storm threads, including the host and port. If a bolt is experiencing latency issues, Storm developers should look here to determine which executor has reached capacity. Click the port number to display the log file for the corresponding executor.

## 8. Example: RollingTopWords Topology

The `RollingTopWords.java` is included with `storm-starter`.

```
package storm.starter;

import backtype.storm.Config;
import backtype.storm.testing.TestWordSpout;
import backtype.storm.topology.TopologyBuilder;
import backtype.storm.tuple.Fields;
import storm.starter.bolt.IntermediateRankingsBolt;
import storm.starter.bolt.RollingCountBolt;
import storm.starter.bolt.TotalRankingsBolt;
import storm.starter.util.StormRunner;

/**
 * This topology does a continuous computation of the top N words that the
 * topology has seen in terms of cardinality.
 * The top N computation is done in a completely scalable way, and a similar
 * approach could be used to compute things
 * like trending topics or trending images on Twitter.
 */
public class RollingTopWords {

 private static final int DEFAULT_RUNTIME_IN_SECONDS = 60;
 private static final int TOP_N = 5;

 private final TopologyBuilder builder;
 private final String topologyName;
 private final Config topologyConfig;
 private final int runtimeInSeconds;

 public RollingTopWords() throws InterruptedException {
 builder = new TopologyBuilder();
 topologyName = "slidingWindowCounts";
 topologyConfig = createTopologyConfiguration();
 runtimeInSeconds = DEFAULT_RUNTIME_IN_SECONDS;

 wireTopology();
 }

 private static Config createTopologyConfiguration() {
 Config conf = new Config();
 conf.setDebug(true);
 return conf;
 }

 private void wireTopology() throws InterruptedException {
 String spoutId = "wordGenerator";
 String counterId = "counter";
 String intermediateRankerId = "intermediateRanker";
 String totalRankerId = "finalRanker";
 builder.setSpout(spoutId, new TestWordSpout(), 5);
 builder.setBolt(counterId, new RollingCountBolt(9, 3), 4).
 fieldsGrouping(spoutId, new Fields("word"));
 builder.setBolt(intermediateRankerId, new IntermediateRankingsBolt(TOP_N),
 4).fieldsGrouping(counterId, new Fields("obj"));
 builder.setBolt(totalRankerId, new TotalRankingsBolt(TOP_N)).
 globalGrouping(intermediateRankerId);
 }
}
```

```
}

public void run() throws InterruptedException {
 StormRunner.runTopologyLocally(builder.createTopology(), topologyName,
 topologyConfig, runtimeInSeconds);
}

public static void main(String[] args) throws Exception {
 new RollingTopWords().run();
}
}
```