Logisland event-mining@scale

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Schedule

- Introduction
- Core concepts
- Knowledge Paradigm
- API Design
- Quick start

Introduction

Logisland Big picture

- Multi-Purpose realtime analytics framework
- High scalability and Fault-tolerant.
- High throughput (billions messages / day).
- Easy to operate on Hadoop or on standalone containers
- Easily Extensible to build high level apps
- Open source, initiated by Hurence

Use cases

- Log aggregation : low latency log processing
- Stream processing: multiple stages of processing (enriching, ...)
- Complex Event processing: write custom business Rules to generate alerts, for fraud detection
- click stream tracking: capture user click stream data
- SIEM: security manager for intrusion detection
- **IoT**: generate alerts based on outliers and forecasting.

Challengers

- ELK is great to start with, but hard to centralize processing and lacks of real offline ML
- Splunk is fantastic but clients are not rich enough to afford it;)
- NIFI is a great tool but doesn't play well with distributed processing
- Metron, Eagle are security centric

Features

- out-of-the-box components (no code required)
- high level extensible framework
- raw data to structured records automatic conversion
- alert percolation or query matching
- event governance with Avro schema management
- online prediction with offline trained ML models

Features 2

- I/O to Elasticsearch, HBase, HDFS, RocksDB, ...
- telemetry sources (bro, pcap, netflow)
- live enrichement (geoip, custom lookups)
- SQL aggregations
- Time series sampling
- Outliers detection
- Network footprint clustering

CORE Concepts

event =

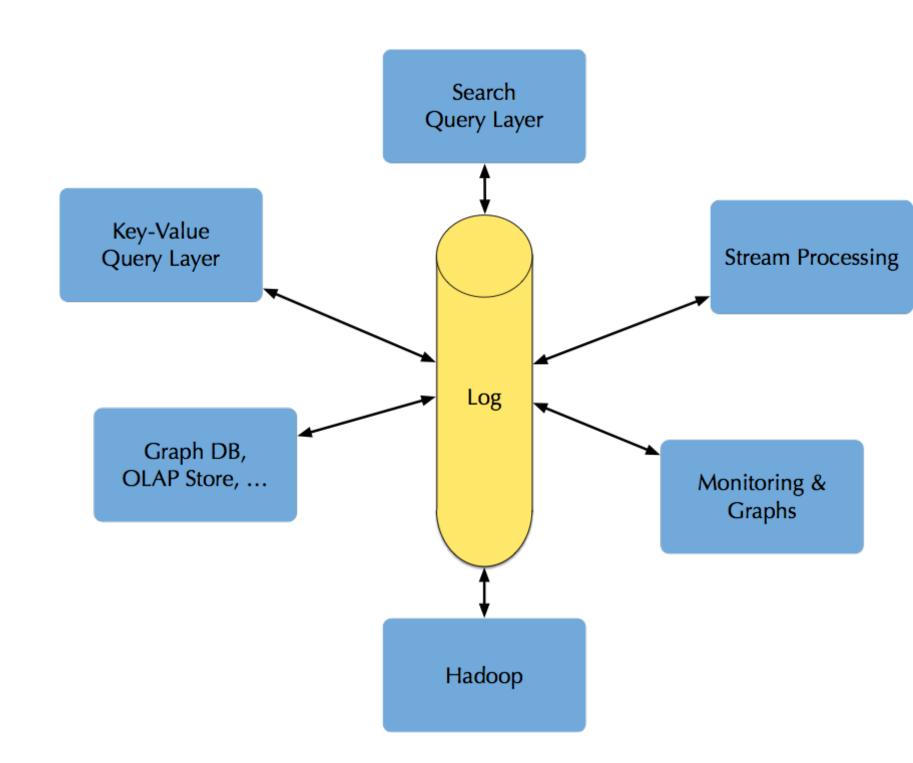
chronological change in the system state

centralized registry of

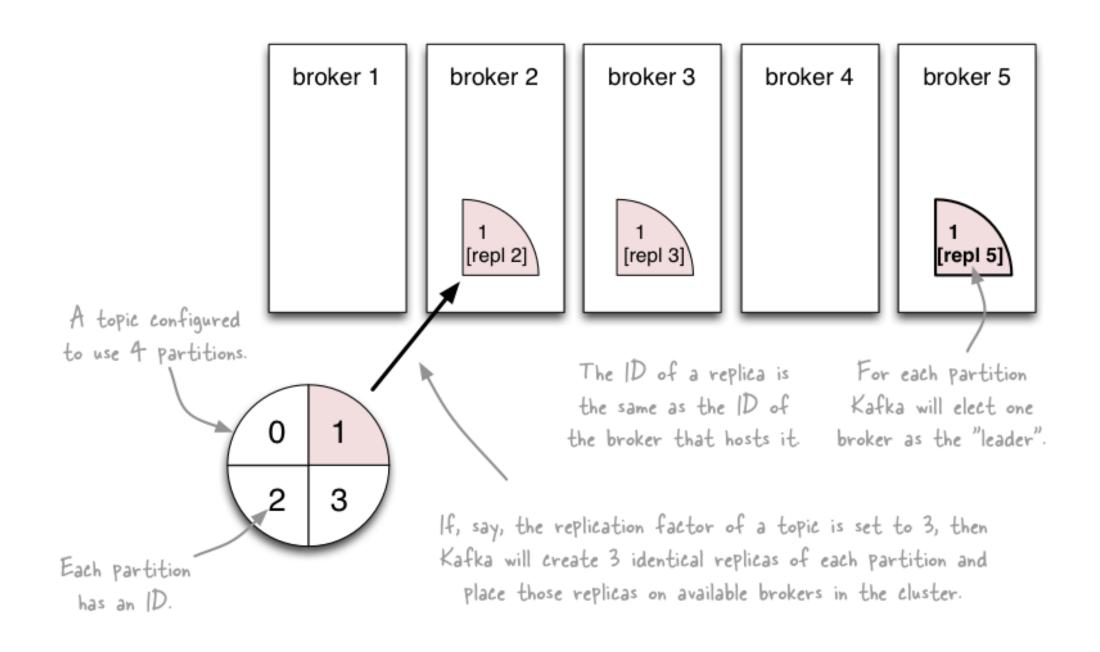
chronologically ordered events

Log centric architecture

- **async** event production and consumption.
- uncorrelated publishers and subscribers.
- acts as a **Messaging system**.
- replay the log from any point in time.
- realtime event availability.



How to handle distributed logs?

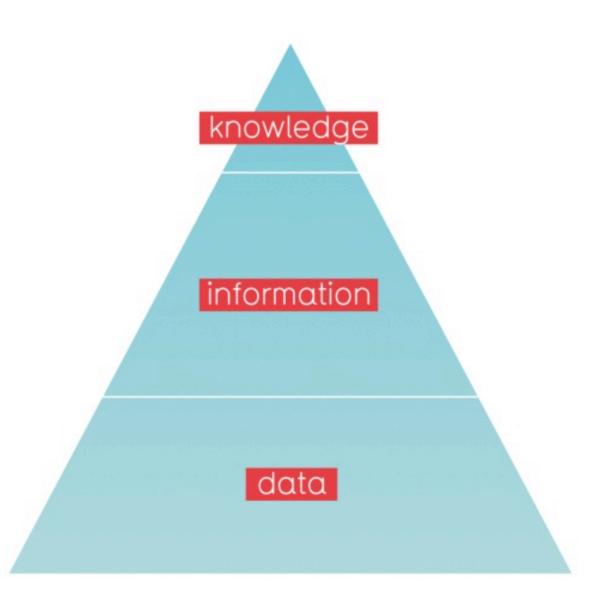


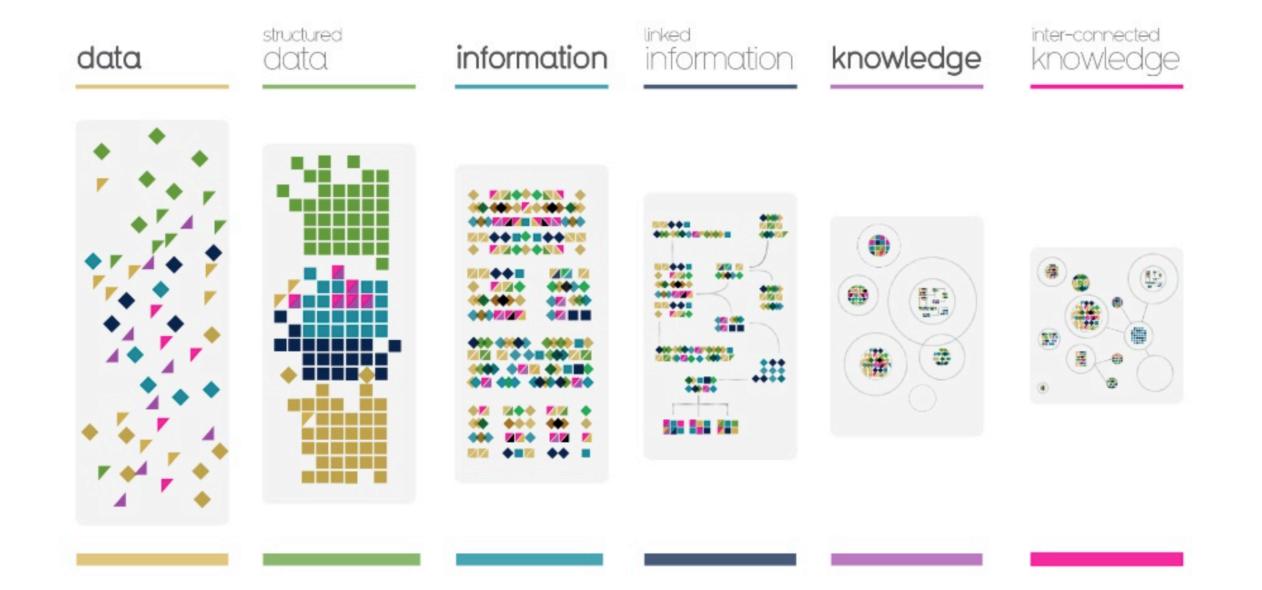
Logisland =

high level stream analytics solution to handle massive scale event processing

Pyramidal Knowledge Paradigm

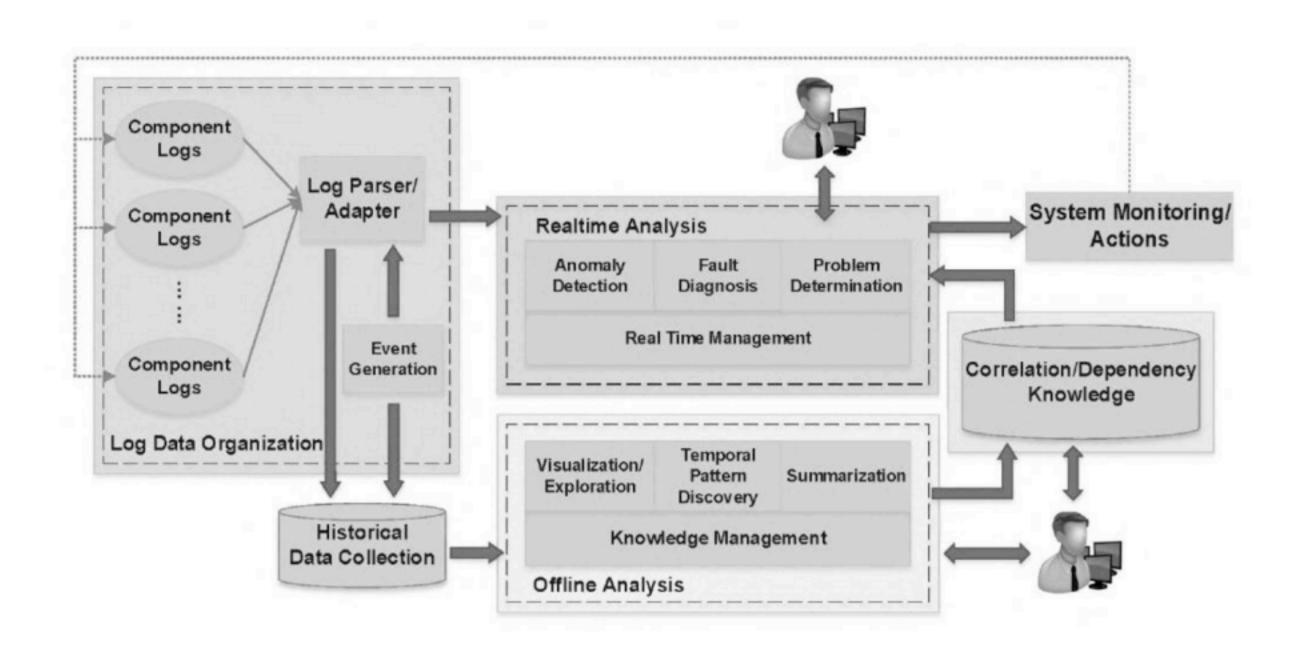
Logisland continuously transforms data into information & information into knowledge by using asynchronous processing on increasingly abstract and meaningful records.



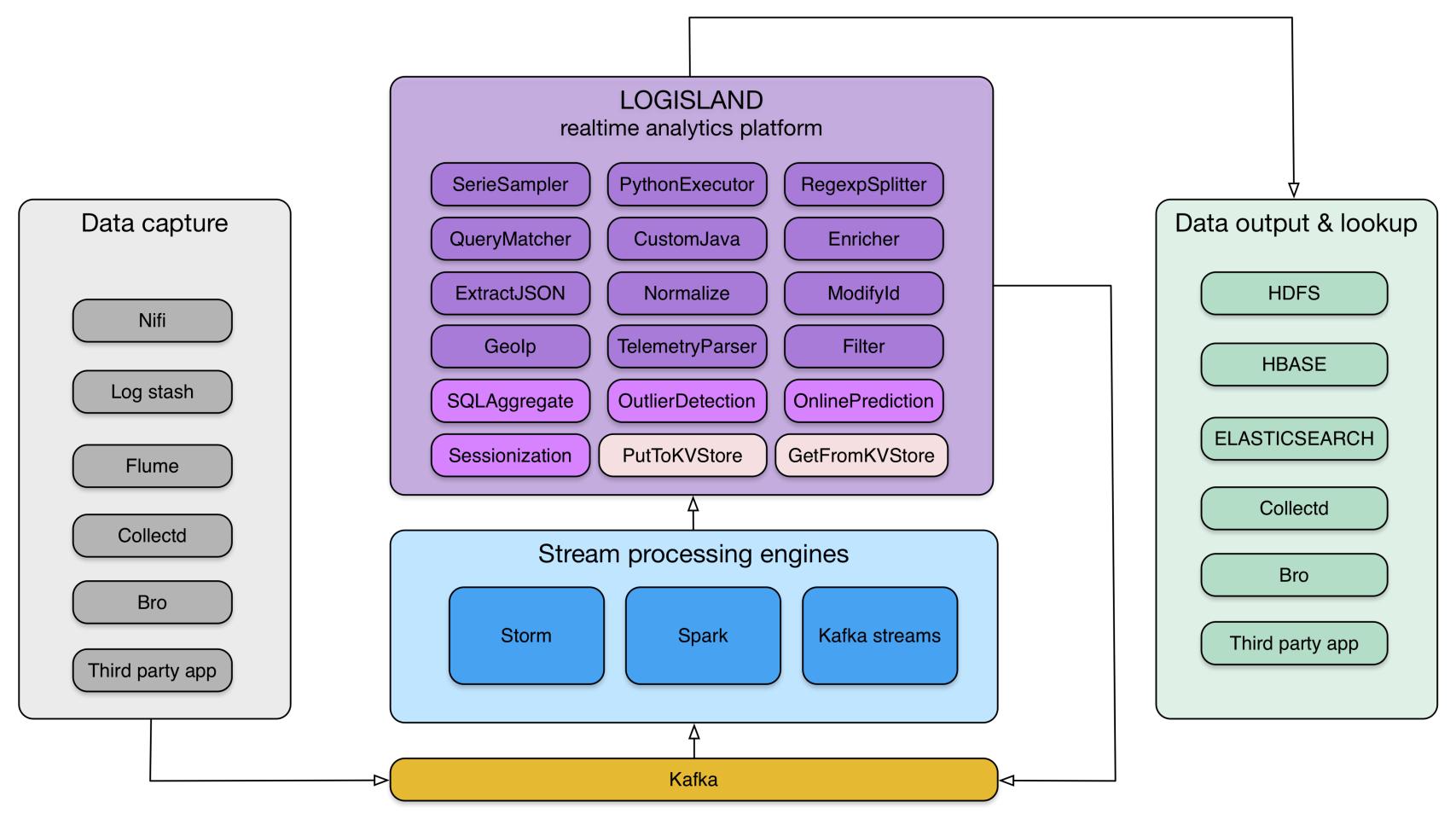


(credits: David McCandless, Information is Beautiful)

Data Driven Computing



Design



Record

The basic unit of processing is the Record.

A Record is a collection of Field, while a Field has a name, a type and a value.

```
String id = "firewall_record1";
String type = "cisco";
Record record = new Record(type).setId(id);
assertTrue(record.isEmpty());
assertEquals(record.size(), 0);
```

Field

A record holds a collection of fields.

```
record.setStringField("url_host", "origin-www.20minutes.fr")
      .setField("method", FieldType.STRING, "GET")
      .setField("response size", FieldType.INT, 452)
      .setField("is outside office hours", FieldType.BOOLEAN, false)
      .setField("tags",
                FieldType.ARRAY,
                Arrays.asList("spam", "filter", "mail"));
assertEquals( record.getField("method").asString(), "GET");
assertTrue( record.getField("response_size").asInteger() - 452 == 0);
record.removeField("is_outside_office hours");
assertFalse( record.hasField("is_outside_office_hours"));
```

Special Field

A Record also has some special fields (type, time and id).

```
// shortcut for id
assertEquals(record.getId(), id);
assertEquals(record.getField(FieldDictionary.RECORD ID).asString(),
             id);
// shortcut for time
assertEquals(record.getTime(),
  record.getField(FieldDictionary.RECORD TIME).asLong());
// shortcut for type
assertEquals(record.getType(), type);
```

Field typing and validation

Fields are strongly typed, you can validate them

```
Record record = new StandardRecord();
record.setField("request_size", FieldType.INT, 1399);
assertTrue(record.isValid());
record.setField("request size", FieldType.INT, "tom");
assertFalse(record.isValid());
record.setField("request_size", FieldType.DOUBLE, 45.5d);
assertTrue(record.isValid());
record.setField("request size", FieldType.STRING, 45L);
assertFalse(record.isValid());
```

Processor

Logisland is a component centric framework,

It's built over an abstraction layer to build **configurable components**.

A component can be Configurable and Configured.

The most common component you'll use is the Processor which takes a collection of Record and publish another collection of records

A configurable component that process Records

SplitText implementation

Define PropertyDescriptor to handle components config.

```
@Tags({"parser", "regex", "log", "record"})
@CapabilityDescription("This is a processor that is used ...")
@DynamicProperty(name = "alternative regex & mapping", ...)
public class SplitText extends AbstractProcessor {
    public static final PropertyDescriptor VALUE_REGEX =
     new PropertyDescriptor.Builder()
            .name("value.regex")
            .description("the regex to match for the message value")
            .required(true)
            .addValidator(StandardValidators.NON_EMPTY_VALIDATOR)
            .build();
```

SplitText config

Use the components with simple yaml blocs.

```
- processor: apache_parser
  component: com.hurence.logisland.processor.SplitText
  type: parser
  documentation: a parser for apache log REGEX
  configuration:
    record.type: apache_log
    value.regex: (\S+)\s+(\S+)\s+\[([\w:\/] ...
    value.fields: src_ip,identd,user,record_time,http_method, ...
```

Stream

a record Stream basically:

- is a configurable Component
- reads a distributed collection of Record from Kafka input topics
- transmits them to a chain of Processor
- write the output collection of Record to some Kafka output topics

Streaming paradigm

You can handle partionned data in 2 ways:

- **fully in parrallel**, eg. a thread by partition, like with KafkaRecordStreamParallelProcessing, when records have no link with each other
- by **joining partitions** like with KafkaRecordStreamSQLAggregator or KafkaRecordStreamHDFSBurner when you need to join related records (costly join and shuffling operations)

Sample Stream configuration

Define a processing pipeline

```
- stream: parsing_stream
 component: com.hurence.logisland.stream.spark.KafkaRecordStreamParallelProcessing
 type: stream
 documentation: a processor that links
 configuration:
   kafka.input.topics: logisland raw
   kafka.output.topics: logisland_events
   kafka.error.topics: logisland_errors
   kafka.input.topics.serializer: none
   kafka.output.topics.serializer: com.hurence.logisland.serializer.KryoSerializer
   kafka.error.topics.serializer: com.hurence.logisland.serializer.JsonSerializer
 processorConfigurations:
```

Engine

- The Engine manage a collection of Stream
- this is the abstraction of the execution model, mainly in Spark actually but plans are to integrate Beam to move on Storm and Kafka Streams
- you configure here your Spark job parameters

Sample engine configuration

Define a processing job

```
engine:
 component: com.hurence.logisland.engine.spark.KafkaStreamProcessingEngine
 type: engine
 documentation: Index some apache logs with logisland
 configuration:
    spark.app.name: IndexApacheLogsDemo
    spark.master: yarn-cluster
    spark.driver.memory: 1G
    spark.driver.cores: 1
    spark.executor.memory: 2G
    spark.executor.instances: 4
    spark.executor.cores: 2
    spark.yarn.queue: default
  streamConfigurations:
```

Transverse service injection: ControllerService

we often need to share access to external Services across the Processors, for example

- bulk buffers or client connections to external data
- a cache service that could cache K/V tuple across the worker node.

Sample ControllerService component

We need to provide an interface API for this service:

```
public interface CacheService<K,V> extends ControllerService {
   PropertyDescriptor CACHE_SIZE = new PropertyDescriptor.Builder()
            .name("cache.size")
            .description("The maximum number of element in the cache.")
            .required(false)
            .defaultValue("16384")
            .addValidator(StandardValidators.POSITIVE_INTEGER_VALIDATOR)
            .build();
   public V get(K k);
   public void set(K k, V v);
```

Inject service in Processor

You can then use this service in a custom processor:

```
public class TestProcessor extends AbstractProcessor {
    static final PropertyDescriptor CACHE_SERVICE = new PropertyDescriptor.Builder()
            .name("cache.service")
            .description("CacheService")
            .identifiesControllerService(CacheService.class)
            .required(true)
            .build();
   a0verride
   public boolean hasControllerService() {
        return true;
```

Define service in config

The injection is done through yaml config files by injecting the instance of lru_cache Service.

```
- controllerService: lru_cache
  component: com.hurence.logisland.service.elasticsearch.LRUKeyValueCacheService
  configuration:
    cache.size: 5000

streamConfigurations:
    stream: parsing_stream

processorConfigurations:
    processor: mock_processor
    component: com.hurence.logisland.processor.TestProcessorhing
    configuration:
        cache.service: lru_cache
```

controllerServiceConfigurations:

Ouick Quick Start

Getting started (Hadoop cluster)

Download the latest release from github

tar -xzf logisland-0.10.1-bin.tar.gz

Create a job configuration

vim conf/index-apache-logs.yml

Run the job

export SPARK_HOME=/usr/hdp/current/spark-client
bin/logisland.sh --conf conf/index-apache-logs.yml

Getting started (lightweight container)

Pull & run the image from Docker Repository

```
docker pull hurence/logisland
docker run -it --name logisland \
   -p 8080:8080 -p 5601:5601 -p 9200:9200 \
   -h sandbox hurence/logisland bash
```

Run the job

bin/logisland.sh --conf conf/index-apache-logs.yml

Next?

Roadmap

- Ambari Agent for job dynamic interaction (REST Api)
- visual Stream configuration / dashboards through Ambari views
- Auto-scaling to optimize cluster resources
- Density based automatic Usage profiling
- Pattern discovery through Deep Learning
- App store, per use-case knowledge bundles (cybersecurity, fraud, ...)

Resources

- source: https://github.com/Hurence/logisland/releases
- Docker: https://hub.docker.com/r/hurence/logisland/tags/
- Documentation : http://logisland.readthedocs.io/en/latest/ concepts.html
- support : https://gitter.im/logisland/logisland
- contact : bailet.thomas@gmail.com