# Pattern detection in server log messages with support for multiple log management systems

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Master 1 Internship with the CC-IN2P3

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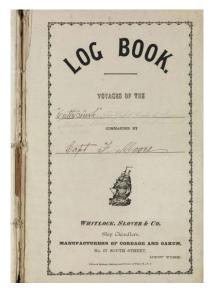
#### Plan

- Introduction
- Problem
- Goals
- Workflow
- Data Analysis
- SEQUENCE
  - How it works Scanner, Analyser, Parser
  - Extensions for this project
- Results
- Limitations
- Next steps Machine Learning/Anomaly detection
- Conclusion

#### Introduction

Why is logging important?

- Supplies **information** continuously on the **activities** of hardware, software and other equipment.
- Often first, if not only, alert of a problem.
- Critical for issue diagnosis, both real time and post mortem.
- State: Current and historical, normal and abnormal
- Audits and security





#### Problem

- approximately 100 million system logs every day.
- Uses a pattern database, for the analysis, whose patterns are created by hand.
- **Issues:** Scalability and maintenance.
- Events constantly changing.
- Approx 75-80% unknown.



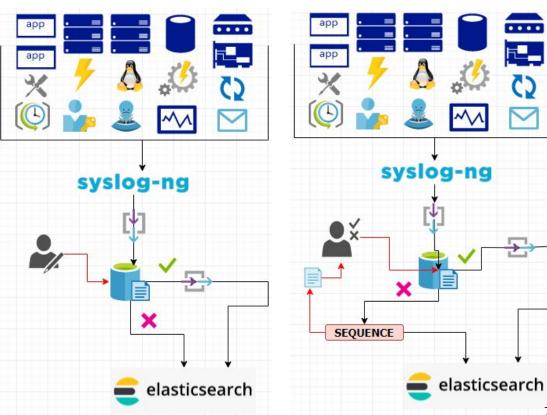
🕒 unknown 🔘 system 🕒 known 🔵 hardware

#### Goals

- Implement a pattern recognition algorithm into the message flow, to assist or automate the manual pattern creation.
- By adapting the SEQUENCE module implemented in Go language for the CC-IN2P3's data and workflow.
- Overall **goal** is to have **90%** or more of the messages **known** in production.
- Ideally return the modified software **back to the open source community** with support for **Syslog's patternDB** and **Logstash's Grok** pattern **parsers**.

### Workflow and other considerations

- Volume
- Variety
- Constant change
- No preprocessing •



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#### Data analysis

Log messages take many forms:

**Length**: 2-3 words to  $> \frac{1}{2}$  page of text.

**No strict rules** for construction, order or contents.

Not confined to one language.

Text and/or JSON format.

Elements that **do follow rules** that can be **extracted**.

Element	Data Type		
Date and Time stamps	DateTime		
MAC addresses	Hexidecimal		
IP addresses version 6	Hexidecimal		
IP addresses version 4	Text		
Port numbers	Integer		
Decimal numbers	Float		
Words, Brackets and Quotes	Text		
Line numbers and counts	Integer		
Punctuation and control characters	Text		
Email addresses	Text		
Urls with/without query strings	Text		
Host names and Protocols	Text		
Statuses, objects and actions	Text		
Uids and machine identifiers	Text/Integer		
Paths	Text		
Non-English characters	Text		
Durations	Text/Number		
Full SQL request queries	Text		
Key/value pairs in many formats	Text		



{"level":"debug","msg":"unregistering reader", "reader-id":"8c417dec-e854-469d-9744-c46f5bd14b2b", "time":"2019-04-09T09:08:23+02:00"}

warning: maildrop/10E66A7: error writing 1A648332: queue file write error

lcas\_userban.mod-plugin\_confirm\_authorization(): checking banned users in /etc/lcas/ban\_users.db

Callout to "LCMAPS" returned local user (service file): "ops008

134.158.172.113:46408 [18/Apr/2019:16:43:43.255] frontend puppetserver/ccpuppet03 1/0/183322 7479 cD 41/41/12/0 0/0

2019-04-18 16:54:36.148432 7faa47cac700 3 rocksdb: [/home/jenkins-build/build/workspace/ceph-build/ARCH/x86 64/AVAILABLE ARCH/x86 64/AVAILABLE DIST/centos7/DIST/ \*\* DB Stats \*\* Uptime(secs): 11768788.4 total, 2753.7 interval Cumulative writes: 538M writes, 2526M keys, 538M commit groups, 1.0 writes per commit group, ingest: 2846.58 GB, 0.25 MB/s Cumulative WAL: 538M writes, 265M syncs, 2.03 writes per sync, written: 2846.58 GB, 0.25 MB/s Cumulative stall: 00:00:0.000 H:M:S, 0.0 percent Interval writes: 156K writes, 741K keys, 156K commit groups, 1.0 writes per commit group, ingest: 978.22 MB, 0.36 MB/s Interval WAL: 156K writes, 76K syncs, 2.06 writes per sync, written: 0.96 MB, 0.36 MB/s Interval stall: 00:00:0.000 H:M:S, 0.0 percent\n\n\*\* Compaction Stats [default] \*\* Level Files Size Score Read(GB) Rn(GB) Rnpl(GB) Write(GB) Wnew(GB) Moved(GB) W-Amp Rd(MB/s) Wr(MB/s) Comp(sec) Comp(cnt) Avg(sec) KevIn KevDrop TO 4/0 21.48 MB 1.0 0.0 0.0 0.0 77.7 77.7 0.0 1.0 0.0 34.8 2286 11947 0.191 0 0 L1 4/0 192.52 MB 0.9 579.2 77.7 501.5 543.0 41.5 0.0 7.0 40.9 38.3 14512 2986 4.860 7841M 202M L2 303/0 2.45 GB 1.0 2319.9 41.5 2278.4 2278.5 0.1 0.0 54.9 57.3 56.3 41439 1477 28.056 5412M 765M Sum 316/0 3.00 GB 0.0 2899.1 119.2 2779.9 2899.2 119.3 0.0 37.3 51.0 51.0 58237 16410 3.549 13G 967M Int 0/0 0.00 KB 0.0 0.2 0.0 0.2 0.2 0.0 0.0 10.0 35.8 38.2 6 5 1.120 2950K 40K Uptime(secs): 11768788.4 total, 11768788.4 interval\nFlush(GB): cumulative 77.695, interval 0.021 AddFile(GB): cumulative 0.000, interval 0.000\nAddFile(Total Files): cumulative 0, interval 0 AddFile(L0 Files): cumulative 0, interval 0\nAddFile(Keys): cumulative 0, interval 0 Cumulative compaction: 2899.22 GB write, 0.25 MB/s write, 2899.10 GB read, 0.25 MB/s read, 58237.1 seconds Interval compaction: 0.21 GB write, 0.00 MB/s write, 0.20 GB read, 0.00 MB/s read, 5.6 seconds Stalls(count): 0 level0 slowdown, 0 level0 slowdown with compaction, 0 level0 numfiles, 0 level0 numfiles with compaction, 0 stop for pending compaction bytes,

\*\* File Read Latency Histogram By Level [default] \*\*

\*\* Compaction Stats [default] \*\* Level Files Size Score Read(GB) Rn(GB) Rnpl(GB) Write(GB) Wnew(GB) Moved(GB) W-Amp Rd(MB/s) Wr(MB/s) Comp(sec) Comp(cnt) Avg(sec) KeyIn KeyDrop

\*\* File Read Latency Histogram By Level [default] \*\*



Open source module using Go Lang written by Jian Zhen

Tested on a **small range** of **common** log messages, but not the **variety** seen at CC

It consists of:

- Scanner splits messages into pieces called tokens
- Analyser compares the sets of tokens to find patterns
- **Parser** tries to match new messages to already found patterns

#### Scanner

- Breaks the message into pieces: tokens
- Uses three separate processes, one for Hexidecimal values, one for Date/Time formats, one for everything else to find the tokens.
- **Reads** each log message only once, **character by character** and passes each character **simultaneously to the three processes.**
- Each process **stops** when it **finds a valid value** or can't continue as it **hits something invalid.**
- Fast: > 200,000 msg/s



#### Scanner cont.

#### • Token Types:

- Float
- Integer
- DateTime
- o IPv4, IPv6
- Urls (http/https)
- Literal
- MAC address

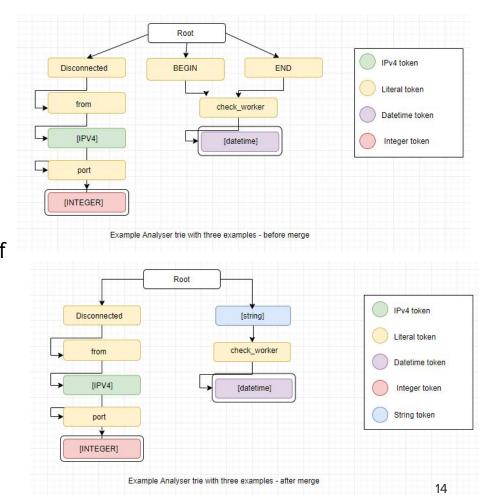
Disconnected from 134.158.106.8 port 41496 Disconnected from 127.0.0.1 port 49570 BEGIN check\_worker Thu Apr 18 16:51:12 CEST 2019 END check\_worker Thu Apr 18 16:51:29 CEST 2019

```
0: { Tag="funknown", Type="literal", Value="Disconnected"}
1: { Tag="funknown", Type="literal", Value="from"}
2: { Tag="funknown", Type="ipv4", Value="134.158.106.8"}
3: { Tag="funknown", Type="literal", Value="port"}
4: { Tag="funknown", Type="literal", Value="41496"}
0: { Tag="funknown", Type="literal", Value="BEGIN"}
1: { Tag="funknown", Type="literal", Value="Check_worker"}
2: { Tag="regextime", Type="time", Value="Thu Apr 18 16:51:12 CEST 2019"}
0: { Tag="funknown", Type="literal", Value="END"}
1: { Tag="funknown", Type="literal", Value="END"}
2: { Tag="funknown", Type="literal", Value="Check_worker"}
2: { Tag="funknown", Type="literal", Value="END"}
1: { Tag="funknown", Type="literal", Value="Check_worker"}
2: { Tag="regextime", Type="literal", Value="Check_worker"}
2: { Tag="regextime", Type="literal", Value="Check_worker"}
3: { Tag="regextime", Type="literal", Value="Check_worker"}
3:
```

## Analyser

1. Builds **a trie** from all the tokenised messages.

 Identifies and merges the tokens of the same type at the same level with the same parent and child node.



#### Analyser cont.

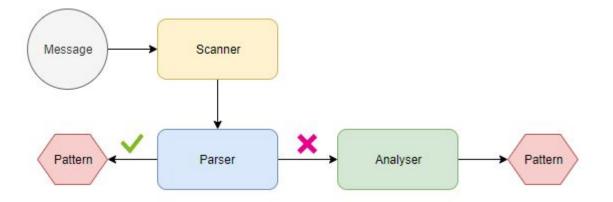
- 1. Looks for email addresses and hostnames and tags them.
- 2. Uses **prekeys** like 'from; and 'to' to tag source and destination variables.
- 3. Uses **keywords**, such as error, or file, to apply status or object tags for example.
- 4. Tries to find **IP and port combinations**.

Disconnected from 134.158.106.8 port 41496 Disconnected from 127.0.0.1 port 49570 BEGIN check\_worker Thu Apr 18 16:51:12 CEST 2019 END check\_worker Thu Apr 18 16:51:29 CEST 2019

> %action% from %srcip% port %srcport% %string% check\_worker %msgtime%

#### Parser

- Used for matching **new messages** to **existing known patterns.**
- If not matched **add to the analyser** for processing.



## Extension of SEQUENCE

To output for custom parsers and to deal with the volume of log messages at CC-IN2P3, we needed to:

- Added a database so SEQUENCE could run continuously.
- Added functionality to handle **multi-line messages**
- New approach to handle the volume of messages.
- Create a **pattern ID that is reproducible** always for the same pattern.
- **Preserve examples with the patterns** for testing with patternDB.
- **Translate** SEQUENCE patterns to **use with patternDB/Grok parsers.**

#### SEQUENCE pattern and Logstash Grok example

SEQUENCE: %action% from %srcip% port %srcport%

```
LOGSTASH:
```

```
filter {
   grok {
     match => {"message" => "%{DATA:action} from %{IP:srcip} port %{INT:srcport}"}
     add_tag => ["2908692bdd6cb4eca096eaa19afebd9e15650b4d", "pattern_id"]
   }
}
```

#### PatternDB output

- <rule id="2908692bdd6cb4eca096eaa19afebd9e15650b4d">

```
- <patterns>
```

```
-<pattern>@ESTRING:action: @from @IPvANY:srcip@ port @NUMBER:srcport@</pattern>
```

```
</patterns>
```

```
- <examples>
```

```
- <example>
```

```
<test_message program="sshd">Disconnected from 134.158.106.8 port 41496</test_message>
```

```
- <test_values>
```

```
<test_value name="action">Disconnected</test_value>
```

```
<test_value name="srcip">134.158.106.8</test_value>
```

```
<test_value name="srcport">41496</test_value>
```

```
</test_values>
```

```
</example>
```

```
- <example>
```

```
<test_message program="sshd">Disconnected from 127.0.0.1 port 49570</test_message>
```

```
- <test_values>
```

```
<test_value name="srcip">127.0.0.1</test_value>
```

```
<test_value name="srcport">49570</test_value>
```

```
<test_value name="action">Disconnected</test_value>
```

```
</test_values>
```

</example>

```
</examples>
```

```
- <values>
```

```
<value name="seq-matches">2</value>
<value name="seq-new">true</value>
<value name="seq-created">2019-06-18</value>
<value name="seq-last-match">2019-06-18</value>
</values>
</values>
```

- <rule id="cf7821e75182fb2738107d64ac1e9997eed01edf">

- <patterns>

<pattern>@ESTRING:status: @@ESTRING:method: @for @ESTRING:srcuser: @from @IPvANY:srcip@ port @NUMBER:srcport@ ssh2: RSA SHA256:@ESTRING:string:@</pattern> </patterns>

- <examples>

<example>

<test\_message program="sshd">Accepted publickey for root from 134.158.106.8 port 49084 ssh2: RSA SHA256:c3KQ+eoIEaK7zDNPuAXHrPuBHXep9LDX9+r2zqcdT9Q</test\_message> - <test values>

<test value name="method">publickey</test value>

<test\_value name="srcuser">root</test\_value>

<test\_value name="srcip">134.158.106.8</test\_value>

<test value name="srcport">49084</test value>

<test value name="string">c3KQ+eoIEaK7zDNPuAXHrPuBHXep9LDX9+r2zqcdT9Q</test value>

<test value name="status">Accepted</test value>

</test values>

</example>

- <example>

<test\_message program="sshd">Accepted publickey for root from 134.158.106.8 port 37484 ssh2: RSA SHA256:c3KQ+eoIEaK7zDNPuAXHrPuBHXep9LDX9+r2zqcdT9Q</test\_message> - <test\_values>

<test value name="method">publickev</test value>

<test value name="srcuser">root</test value>

<test\_value name="srcip">134.158.106.8</test\_value>

<test value name="srcport">37484</test value>

<test value name="string">c3KO+eoIEaK7zDNPuAXHrPuBHXep9LDX9+r2zgcdT9O</test value>

<test value name="status">Accepted</test value>

</test values>

</example>

- <example>

<test message program="sshd">Accepted publickey for root from 134.158.106.8 port 45368 ssh2: RSA SHA256:c3KO+eoIEaK7zDNPuAXHrPuBHXep9LDX9+r2zgcdT9O</test message> - <test values>

<test value name="string">c3KO+eoIEaK7zDNPuAXHrPuBHXep9LDX9+r2zgcdT9O</test value>

<test value name="status">Accepted</test value>

<test value name="method">publickey</test value>

<test value name="srcuser">root</test value>

<test value name="srcip">134.158.106.8</test value>

<test value name="srcport">45368</test value>

</test values>

</example>

</examples>

- <values>

<value name="seq-matches">105299</value>

<value name="seq-new">true</value>

<value name="seq-created">2019-06-18</value>

<value name="seq-last-match">2019-06-18</value>

</values> </rule>

1.12 -22 122 223 23 20 20 .....

- <rule id="f1f1a213a55e4a2c886acf6edfc55d9ab898d693" class="sequence">
  - <patterns>

```
<pattern>@ESTRING:string::@ @ESTRING:string1: @- rdac checker reports path is down</pattern>
```

- </patterns>
- <examples>
  - <example>

```
<test_message program="multipathd">nsd5602: sdb - rdac checker reports path is down</test_message>
```

- <test\_values>

```
<test_value name="string">nsd5602</test_value>
```

```
<test_value name="string1">sdb</test_value>
```

- </test\_values>
- </example>
- <example>

```
<test_message program="multipathd">nsd5603: sdm - rdac checker reports path is down</test_message>
```

- <test\_values>

```
<test_value name="string1">sdm</test_value>
```

```
<test_value name="string">nsd5603</test_value>
```

- </test\_values>
- </example>
- <example>
  - <test\_message program="multipathd">nsd5604: sdd rdac checker reports path is down</test\_message>
  - <test\_values>

```
<test_value name="string1">sdd</test_value>
```

```
<test_value name="string">nsd5604</test_value>
```

</test\_values>

```
</example>
```

</examples>

```
- <values>
```

```
<value name="seq-matches">2160746</value>
<value name="seq-new">true</value>
<value name="seq-created">2019-06-21</value>
<value name="seq-last-match">2019-07-01</value>
</values>
</rule>
```

#### Results

#### **SEQUENCE** Testing

File Name	Record Count	Time	No Patterns	
Coloss-with-service	92195	2.58s	396	
Coloss-json	967052	39.73s	1723	
Coloss-json-xl	13250853	15m22.36s	4034	

#### Syslog-ng PatternDB parser testing

File Name	Count	No Patterns Tested	% Matched	% Errored
Coloss-with-service	92195	64	81	19
Coloss-json-xl	13250853	131	87	13

#### **Results: Production**

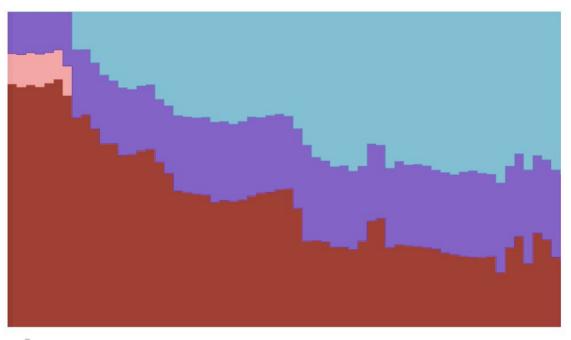
Known/unknown in production:

SEQUENCE **51.6%** 

Other known 27.4

Unknown 21%

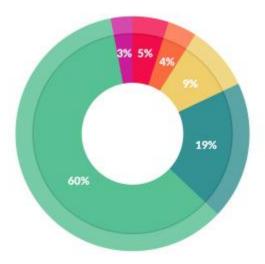
Runs every **15 mins**, takes **7 seconds** to process 100,000.



#### Limitations

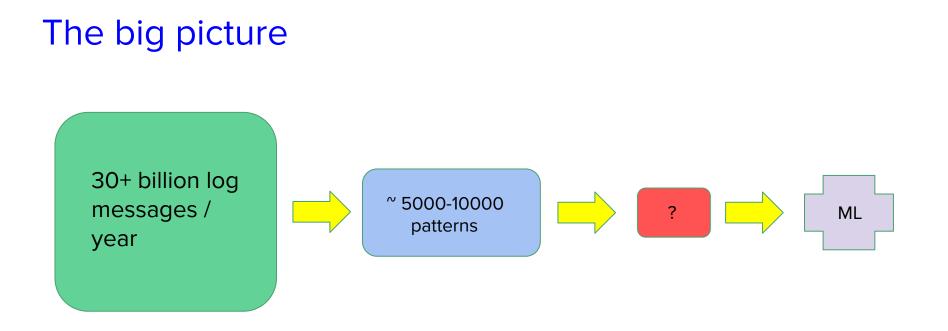
- Needs a **few examples** to find a good pattern.
- Some log messages have a pattern that matches a token type incorrectly.
- Keywords can cause more than one pattern for similar log messages.
- Struggles with some key/value pairs when the value is not delimited.
- Converting between **different parser types** will never be exact.

#### Machine Learning - Where does this fit?



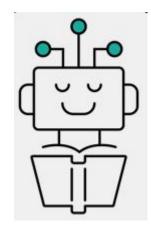
#### What data scientists spend the most time doing

- Building training sets: 3%
- Cleaning and organizing data: 60%
- Collecting data sets; 19%
- Mining data for patterns: 9%
- Refining algorithms: 4%
- Other: 5%



#### Possible next steps

- Machine Learning Anomaly Detection
  - Types of anomalies
    - Known events frequency
    - New/unseen events
    - Change in sequence of events
    - Change in event parameters
  - Considerations
    - Frequency of change/maintenance
    - Definition of 'normal'
    - Volume of messages



- Privacy?
- How do we communicate anomalies?
- Feedback dealing with false positives

#### Conclusion

- With close to **80% known** log messages, well on our way to the **goal of 90%**
- **Pattern discovery and creation** has made the **maintenance** of the patternDB more **manageable**.
- With the **extra meta-data** and **patternID**'s in **Elastic Search**, **easier to search** when diagnosing **issues** or looking for **information**,
- **First steps in preprocessing** the data for Machine Learning approaches like anomaly detection have been taken.
- **On track** for release back into **Open Source Community**.

#### Supervised vs unsupervised

Supervised learning is where you have input variables (x) and an output variable (Y) and you use an algorithm to learn the mapping function from the input to the output.

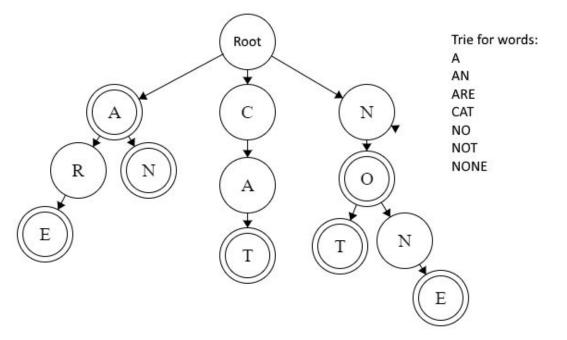
**Unsupervised learning** is where **you only have input data** (**X**) and no corresponding output variables. The goal for unsupervised learning is **to model the underlying structure or distribution** in the data in order to learn more about the data.

#### What approach to use? - Latest research

- Unsupervised, semi-supervised?
- Neural network using Long Short Term Memory (LSTM) for time series prediction to model frequency per pattern.
  - Compare predicted to actual significant difference = anomaly
- LSTM for learning log message sequences or,
- Learned finite state machines, markov models.

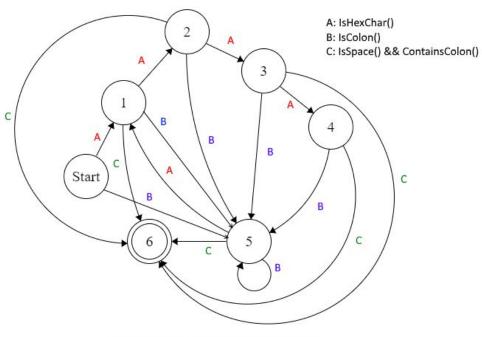
## **Analyser: Trie Introduction**

- data structure that
   specialises in working
   with strings
- allows for very fast
   search and retrieval of values.
- Most common use autocomplete



#### Finite State Machine - Hexidecimal

- 1. First character
- 2. Second character
- 3. Third character
- 4. Fourth character
- 5. Colon
- 6. Space



Finite State Machine for Hexidecimal tokens

### Date Time

Supports 49 different formats

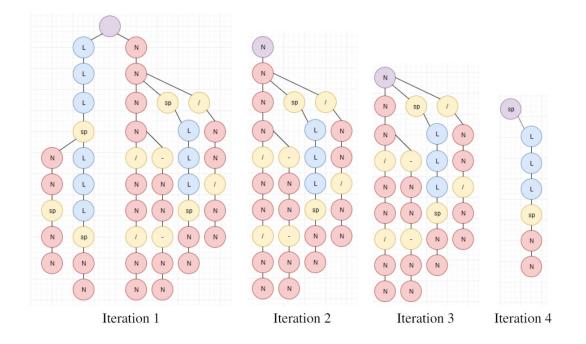


Figure 4: Trees passed to the time FSM with each iteration for example date 15 Jun 19.