BYOI: Build you own index

Build your own discovery index of scholary e-resources

5th German Vufind Meetup 2016, 2016–10–11, Freiburg, University Library Freiburg is.gd/iQYDZB Martin Czygan, David Aumüller, Leander Seige Leipzig University Library

- ▶ https://ub.uni-leipzig.de
- https://finc.info
- https://amsl.technology
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Welcome

During the next few hours, we will create an small aggregated index from scratch.

You can code along if you like. Code, data and slides are distributed in a VM (on a USB stick).

Why

At Leipzig University Library we built a version that serves as a successor to a commercial product.

Index includes data from *Crossref*, *DOAJ*, JSTOR, Elsevier, Genios, Thieme, DeGruyter among others.

Various coverages per library, from 55% to 70%. Potentially growable in breadth and depth.

Format

We will use a combination of

- slides to motivate concepts and
- live coding and experimentation

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We will not use a product, we will build it.

Goals

- a running VuFind 3 with a small aggregated index
- ▶ learn about a batch processing framework

First Steps

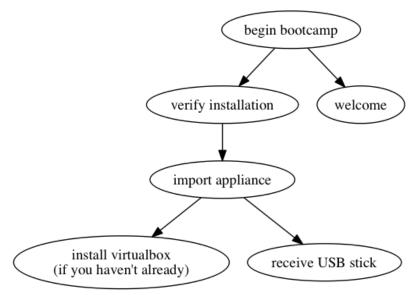


Figure 1: First steps

Prerequisites

► Virtualbox: https://www.virtualbox.org/wiki/Downloads

Import Appliance

On the USB-Stick you can find an OVA file that you can import into Virtualbox (or try to download it from https://goo.gl/J7hcYC). This VM contains:

- a VuFind 3 installation /usr/local/vufind
- raw metadata (around 3M records) ~/Bootcamp/input
- scripts and stubs for processing ~/Bootcamp/code
- ▶ these slides ~/Bootcamp/slides.pdf

Forwarded ports

```
Guest (VM) >> Host
80
           >> 8085
                     (HTTP, VuFind)
8080
           >> 8086
                     (SOLR)
8082
           >> 8087
                     (luigi)
           >> 2200 (SSH)
22
           >> 13306 (MySQL)
3306
SSH tip:
$ curl -sL https://git.io/vrxoC > vm.sh
 chmod +x vm.sh
$ ./vm.sh
```

Outline

Bootcamp play book:

- intro: problem setting (heterogenous data, batch processing)
- VM setup during intro

Then we will write some code:

- a basic pipeline with luigi python library
- combine various sources into a common format
- apply licensing information
- ▶ index into solr

Outline DAG

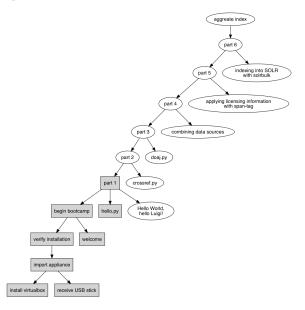


Figure 2: Tour

Intro: Problem setting

- batch processing, not small data (but not too big, either)
- regular processing required
- varying requirements
- multiple small steps to apply on all data
- iterative development

Intro: Rise of the DAG

- ► DAG = directed acyclic graph, partial ordering
- many things use DAGs, make, Excel, scheduling problems
- model tasks in a DAG, then run topological sort to determine order

Examples:

- http://goo.gl/FCpxiK (history is a DAG)
- http://i.stack.imgur.com/iVNcu.png (airflow)
- https://git.io/vw9rW (luigi)
- https://goo.gl/vMEezR (Azkaban)

Intro: Immutability

- immutability = data is not modified, after it is created
- immutable data has some advantages, e.g.
 - "human fault tolerance"
 - performance
- our use case: recompute everything from raw data
- tradeoff: more computation, but less to think about

Intro: Frameworks

- many libraries and frameworks for batch processing and scheduling, e.g. Oozie, Askaban, Airflow, luigi, . . .
- even more tools, when working with stream, Kafka, various queues, . . .
- luigi is nice, because it has only a few prerequisities

Intro: Luigi in one slide

```
import luigi
class MyTask(luigi.Task):
   param = luigi.Parameter(default='ABC')
   def requires(self):
        return SomeOtherTask()
   def run(self):
        with self.output().open('w') as output:
            output.write('%s business' % self.param)
    def output(self):
        return luigi.LocalTarget(path='output.txt')
if __name__ == '__main__':
   luigi.run()
```

Intro: luigi

- many integrations, e.g. MySQL, Postgres, elasticsearch, . . .
- ▶ support for Hadoop and HDFS, S3, Redshift, . . .
- ▶ 200+ contributors, 350+ ML
- hackable, extendable e.g. https://github.com/ubleipzig/gluish

Intro: Decomposing our goal

- clean and rearrange input data files
- convert data into a common (intermediate) format
- apply licensing information (from kbart)
- index into solr

Intro: Incremental Development

when we work with unknown data sources, we have to gradually move forward

Intro: Wrap up

- many approaches to data processing
- we will focus on one library only here
- concepts like immutability, recomputation and incremental development are more general

now back to the code

Test VuFind installation

We can SSH into the VM and start VuFind:

```
$ ./vm.sh
(vm) $ cd /usr/local/vufind
(vm) $ ./solr.sh start
Starting VuFind ...
...
```

- http://localhost:8085/vufind
- ▶ http://localhost:8085/vufind/Install/Home

Hello World

Test a Python script on guest. Go to the Bootcamp directory:

```
$ cd $HOME/Bootcamp
$ python hello.py
...
```

Note: Files follow PEP-8, so indent with space, here: 4.

Setup wrap-up

You can now edit Python files on your guest (or host) and run them inside the VM. You can start and stop VuFind inside the VM and access it through a browser on your host.

We are all set ot start exploring the data and to write some code.

Bootcamp outline

- parts 0 to 6: intro, crossref, doaj, combination, licensing, export
- each part is self contained, although we will reuse some artifacts

Bootcamp outline

- ▶ you can use scaffoldP_..., if you want to code along
- ▶ the partP_... files contain the target code

```
code/part{0-6}_....py
code/scaffold{0-6}_....py
```

► Hello World from luigi

```
$ cd code
```

\$ python part0_helloworld.py

Coding: Part 0 Recap

- simple things should be simple
- basic notion of a task
- command line integration

► An input and a task

\$ python part1_require.py

Coding: Part 1 Recap

- ▶ it is easy to start with static data
- business logic in python, can reuse any existing python library

- a first look at Crossref data
- harvest via API, the files contain batch responses
- custom format

Three things to do:

- find all relevant files (we will use just one for now)
- extract the records from the batch
- convert to an intermediate format

▶ Now on to the code

\$ python part2_crossref.py

Coding: Part 2 Recap

- used command line tools (fast, simple interface)
- chained three tasks together

Excursion: Normalization

- suggested and designed by system librarian
- internal name: intermediate schema https://github.com/ubleipzig/intermediateschema
- enough fields to accomodate various inputs
- can be extended carefully, if necessary
- tooling (licensing, export, quality checks) only for a single format

Excursion: Normalization

```
"finc.format": "ElectronicArticle",
"finc.mega collection": "DOAJ",
"finc.record id": "ai-28-00001...",
"finc.source id": "28",
"rft.atitle": "Importância da vitamina B12 na ...",
"rft.epage": "78",
"rft.issn": [
  "1806-5562",
  "1980-6108"
"rft.jtitle": "Scientia Medica",
```

- ▶ DOAJ index data
- a complete elasticsearch dump

This source is not batched and comes in a single file, so it is a bit simpler:

- ► locate file
- convert to intermediate schema

\$ python part3_require.py

Coding: Part 3 Recap

- ▶ it is easy to start with static data
- business logic in python, can reuse any existing python library

▶ after normalization, we can merge the two data sources

\$ python part4_combine.py

Coding: Part 4 Recap

- a list of dependencies
- python helps with modularization
- using the shell for performance and to reuse existing tools

- licensing turned out to be an important issue
- a complex topic
- we need to look at every record, so it is performance critical
- we use AMSL for ERM, and are on the way to a self-service interface
- AMSL has great APIs
- we convert collection information to an expression-tree-ish format – https://is.gd/Fxx0IU, https://is.gd/ZTqLqB

\$ python part5_licensing.py

- boolean expression trees allow us to specify complex licensing rules
- ▶ the result is a file, where each record is annotated with an ISIL
- at Leipzig University Library we currently do this for about 20 ISILs

Coding: Part 5 Recap

- dependencies as dictionary
- flexibility in modeling workflows
- ▶ again: use command line tools for performance critical parts

▶ a final conversion to a SOLR-importable format

\$ python part6_export.py

- slightly different from SOLRMARC style processing
- ▶ keep things (conversion, indexing) a bit separate
- standalone tool: solrbulk

Coding: Part 6 Recap

- flexibility in modeling workflows
- ▶ again: use command line tools for performance critical parts

- ▶ finally, we can index the data into SOLR
- ▶ make sure SOLR is running on your VM

```
$ solrbulk -host localhost -port 8080 \
    -w 2 -z -verbose -commit 100000 \
    -collection biblio \
    output/6/Export/output.ldj.gz
```

▶ might want to increase SOLR_HEAP (defaults to 512M)

- ▶ go to http://localhost:8085
- ▶ index should be slowly growing

Code recap

```
\_ Export()
  \_ ApplyLicensing()
  \_ CombinedIntermediateSchema()
     \_ DOAJIntermediateSchema()
     \_ DOAJInput()
  \_ CrossrefIntermediateSchema()
     \_ CrossrefItems()
     \_ CrossrefInput()
  \_ CreateConfiguration()
  \_ HoldingFile()
```

Code recap

```
$ python deps.py
\ Export()
   \_ ApplyLicensing()
       \ CombinedIntermediateSchema()
          \ DOAJIntermediateSchema()
             \_ DOAJInput()
          \ CrossrefIntermediateSchema()
             \ CrossrefItems()
                \ CrossrefInput()
       \ CreateConfiguration()
          \ HoldingFile()
```

Code recap

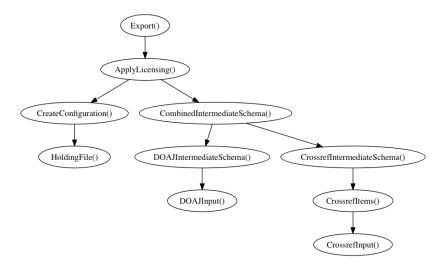


Figure 3: Deps

Follow up with workflow changes

https://git.io/vOZFQ

Production data points:

- sustained indexing rates between 2000-4000 docs/s
- a full reindex of about 100M docs currently takes about 10h with SOLR

Discussion

what we left out:

- more data sets
- larger data sets
- XML
- errors
- parameters
- collaboration and deployment

Discussion

- what are your experiences with batch systems?
- how do you manage large heterogeneous data?
- what could we add to the pipeline?

Q & A

Thanks for your attention.

For any questions, please get in touch during the conference or via e-mail:

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