# Open-Source Software for Appraisal and Processing of Email at Scale

#### **Christopher (Cal) Lee**

School of Information and Library Science

University of North Carolina at Chapel Hill

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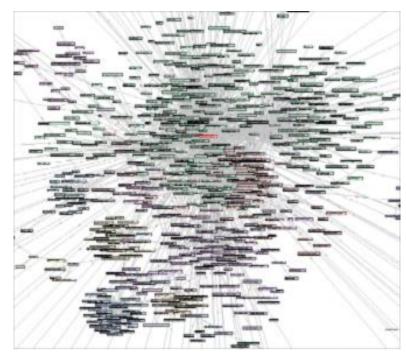


# Motivation - Selection/Appraisal

- Despite progress on various technologies to support data management and digital preservation, relatively little progress on software support for the core activities of selection and appraisal
- Selection/appraisal decisions are based on various patterns
- When patterns can be identified algorithmically, software can assist the process
- LAMs frequently want to take actions that reflect contextual relationships
- Timeline representations and visualizations can also provide useful, high-level views of materials

### **Motivation - Email**

- About 50 years of email creation
- Hundreds of billions of messages generated every day
- Most has little long-term retention value, but some absolutely does
- Despite presence of numerous other modalities, email still deeply embedded in activities, serving as massive source of evidence and information
- Often found in collections and acquisitions with other types of materials



http://hci.stanford.edu/~jheer/projects/enron/v1/

# Review, Appraisal, and Triage of Mail (RATOM)

- Funded by Andrew W. Mellon Foundation (2019-2020)
- Developing and repurposing software (including NLP and machine learning) for selection/appraisal in BitCurator environment with hooks and enhancements to TOMES output
- Support iterative processing information discovered at various points in the processing workflow can support further selection, redaction or description actions
- Mapping of timestamp, entity, sensitive features and other elements across the tools



**Ray Tomlinson**Implemented first email program on ARPANET.
Credited with invention of first email system.

### **Team Members**



**Cal Lee** Pl



**Antoine De Torcy**Software Engineer



**Camille Tyndall Watson** Co-Pl



Jamie Patrick-Burns
Investigator



**Eliscia Kinder** Project Manager



Kam Woods
Technical Lead (UNC)



Sangeeta Desai Technical Lead (NC DAR)



**Caktus Group** Additional Software Development

# Scope of the project

The RATOM project has several core development goals designed to serve the needs of collecting institutions tasked with preparing email collections for public access:

- Development of an integrated Python library to simplify parsing and processing PST, OST, and mbox email formats
- Development of utilities to support entity identification and export reports suitable for conducting automated and human-directed redaction actions at scale
- Development of an interface allowing processing archivists to browse email collections and mark messages as suitable for retention
- Development of utilities to apply machine learning techniques (by training on annotated message collections and/or unsupervised) to recognize candidate materials for retention

### RATOM tools - libratom

#### libratom (reusable library)

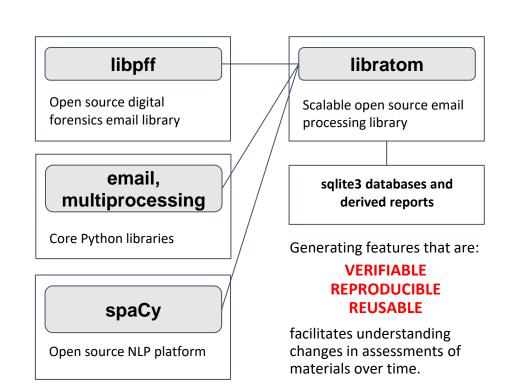
Python library to parse and analyze **PST**, **OST**, and **mbox** email formats

Wraps functions from **libpff**, Python **mailbox**, and **spaCy** (NLP)

Email message content, header, attachment extraction; entity identification and classification

Engineered to scale with core count and keep memory use flat per-core

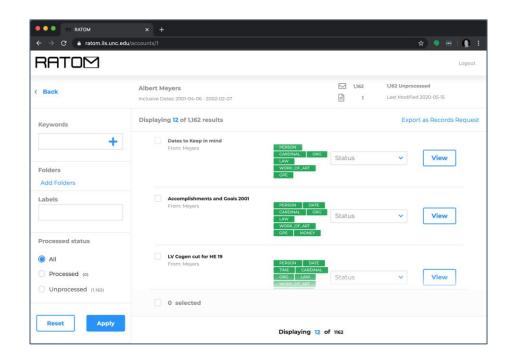
https://www.github.com/libratom/libratom



# RATOM tools - Iterative Processing Interface

Assist archivists in reviewing email materials for retention and/or release.

- Import of email accounts from PSTs and entity identification via libratom
- Creation of processing accounts associated with individual email users
- Interactive review and tagging of email messages within these accounts (e.g. "record", "nonrecord", "redact")
- Export of selected messages as EML for retention or release



https://github.com/StateArchivesOfNorthCarolina/ratom-deploy

```
883381 | the Department of Environmental Protection | ORG | david_delainey_000_1_2.pst | 2325380
883382 | Cellucci | PERSON | david delainey 000 1 2.pst | 2325380
883383 | the United States | GPE | david delainey 000 1 2.pst | 2325380
883384 | five | CARDINAL | david delainey 000 1 2.pst | 2325380
883385 | six | CARDINAL | david delainey 000 1 2.pst | 2325380
883386 | Jane Swift | PERSON | david delainey 000 1 2.pst | 2325380
883387 | the Department of Environmental Protection | ORG | david_delainey_000_1_2.pst | 2325380
883388 | six | CARDINAL | david delainey 000 1 2.pst | 2325380
883389 | the next few months | DATE | david delainey 000 1 2.pst | 2325380
883390 | 1.5 | CARDINAL | david delainey 000 1 2.pst | w2325380
883391 | 3 pounds | QUANTITY | david delainey 000 1 2.pst | 2325380
883392 | megawatt-hour | TIME | david delainey 000 1 2.pst | 2325380
883393 | five | CARDINAL | david delainey 000 1 2.pst | 2325380
883394 | Sithe Energies, Inc. | ORG | david delainey 000 1 2.pst | 2325380
```

Model: Spacy en\_core\_web\_sm, trained on OntoNotes 5, below stats for raw / no gold ref text:

MODEL	SPACY	TYPE	UAS	NER F	POS	WPS	SIZE
en_core_web_sm 2.0.0	2.x	neural	91.7	85.3	97.0	10.1k	35MB

With the current CLI, we can load different models (including user trained models) on demand for tasks / languages

### libratom commands

entities command now provides more structured and responsive feedback when progress is requested (progress bars for both file and message scans), performance improvements via tuning of job distribution

**model** command provides granular control of entity ident model(s) in use, including access to previously released models

**report** command generates a fast report, populating the sqlite3 schema without entities (but optionally including message text and headers)

**emldump** provides a mechanism for generating EML files using JSON structured message id lists produced by the web app (may also be used standalone)

```
kamwoods@blueash: ~/Research Q =
(venv) kamwoods@blueash > ~/Research > ratom -h
Usage: ratom [OPTIONS] COMMAND [ARGS]...
 Main command
Options:
 --version Show the version and exit.
 -h, --help Show this message and exit.
Commands:
           Generate .eml files from pst/mbox files.
 entities Extract named entities.
 model
           Manage spaCy models.
           Generate mailbox contents report.
venv) kamwoods@blueash > ~/Research
```

### libratom output

Many updates as of 0.4.x...

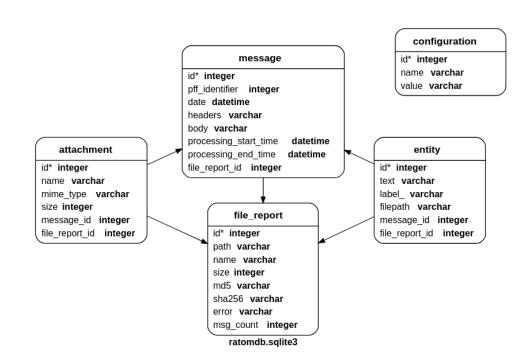
Datetime stamps now extracted from message header and stored in message table

Option to include all message text (stripped of markup and inline attachments) and headers in message table

MIME types for attachments now recorded in attachment table; types are verified vs IANA listed content types and subtypes

Various fixes and improvements (additional detail in configuration, character encoding checks, etc)

See this chart in detail in the README at https://github.com/libratom/libratom



### libratom processing the Enron (EDRM v1.3) corpus

EDRM v1.3 Enron Corpus: Approximately **54GB**, and includes **191 files**, containing **758,341 messages** 

**PST** internal directory structure and message count scan:

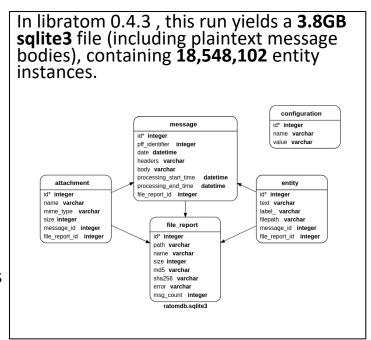
16-core Threadripper 2950X: 1 minute 32-core Threadripper 3970X: 30 seconds

Entity extraction from all 750K messages (spaCy

en core web sm model):

16-core Threadripper 2950X: **2 hrs** 32-core Threadripper 3970X: 1 hr 15m

Memory usage is bounded for the **spaCy** configuration and number of processes. For 32 processes, accessible memory is ~1.6GB/process, resident memory is ~500MB/process on average.



### libratom processing the Jeb Bush corpus

Jeb Bush corpus: Approximately **7.2GB**, includes **11 files** containing **251,509 messages** 

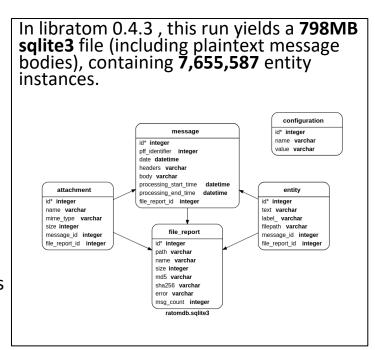
PST internal directory structure and message count scan:

16-core Threadripper 2950X: < 1 minute 32-core Threadripper 3970X: < 10 seconds

Entity extraction from all 252K messages (spaCy en core web sm model):

16-core Threadripper 2950X: ~48 minutes 32-core Threadripper 3970X: ~30 minutes

Memory usage is bounded for the **spaCy** configuration and number of processes. For 32 processes, accessible memory is **~1.6GB/process**, resident memory is **~500MB/process** on average.



# libratom processing the Gov. Kaine (Library of Virginia) sample corpus

Kaine sample corpus: Approximately **12.3GB**, includes **1 PST file** containing **79,538 messages** 

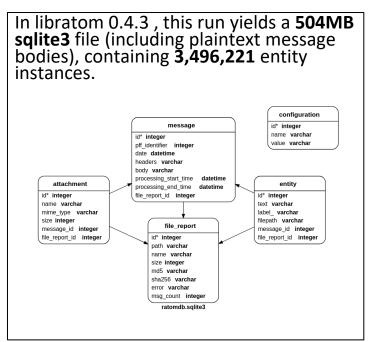
PST internal directory structure, message count, and attachment scan:

16-core Threadripper 2950X: < **10 seconds** 32-core Threadripper 3970X: < **5 seconds** 

Entity extraction from all 252K messages (spaCy en core web sm model):

16-core Threadripper 2950X: ~24 minutes 32-core Threadripper 3970X: ~15 minutes

Memory usage is bounded for the **spaCy** configuration and number of processes. For 32 processes, accessible memory is **~1.6GB/process**, resident memory is **~500MB/process** on average.



We can quickly examine various slices of the db output using a few simple SQL queries...

#### Example 1:

Entity groups by type, ordered by count.

```
sqlite> select count(*), label from entity group by label order by count(*)
DESC;
1288056 | PERSON
686345 ORG
444841 DATE
362347 CARDINAL
257576 GPE
173187 | TIME
60158 | MONEY
32378 | NORP
31311 ORDINAL
27683 FAC
27122 LOC
24327 PERCENT
23698 | WORK_OF_ART
22738 PRODUCT
14511 LAW
11194 | EVENT
8163 QUANTITY
586 LANGUAGE
salite>
```

#### Example 2:

Individual text elements identified by spaCy as "PERSON" that appear more than 10,000 times

Note that the NLP processor will return full names as entities; it just happens that this particular group of entities was mentioned independently by first or last name a large number of times in the collection.

Formatting quirks can also produce the behavior - additional introspection into the materials and experiments with models other than the default model would be needed to determine whether this baseline performance coulld be significantly improved.

(As an example - the "Gail" and "Jaspen" elements here almost certainly refer to the same person, "Gail Jaspen")

```
sqlite> select count(*), text from entity where label = 'PERSON' group by
text having count(*) > 10000 order by count(*) DESC;
41480|Gail
35978 | Marilyn
29690 Jaspen
19033 | Tavenner
18462|Bill
17264 Barbara
17077 | Mark
13858 | Wayne
13434|Brian
12723 Rubin
10933 | Craig
10767 | Leighty
10761 | Burns
10371|Heidi
salite>
```

#### Example 3:

Total number of attachments

Attachments by mime type, where there are more than 100 of that particular type

Note that given the total number of attachments this means there is a very long tail of additional attachment types...although some of these are variants of types that appear high in the list (Word, JPG, etc)

```
sqlite> select count(*) from attachment;
42783
sqlite> select count(*), mime type from attachment group by mime type having
count(*) > 100 order by count(*) DESC;
23396 application/msword
4621 application/octet-stream
4355 application/vnd.ms-excel
3483 application/pdf
1833 image/jpeg
1685 | image/gif
1363 application/vnd.ms-powerpoint
544 text/x-vcard
234 image/tiff
231 text/plain
179|text/html
136 application/msexcel
134 application/rtf
108 text/vcard
```

#### Example 4:

We can use this db to explore data that might be problematic for processing further down the line. For example:

All attachments with identical names that appear in the collection more than 100 times

```
sqlite> select count(*), name from attachment group by name having count(*) >
100;
303 | Document.pdf
108 | IQFormatFile.txt
161 | Scan001.PDF
729|image001.gif
1005 | image001.jpg
152|image002.gif
157 | image002.jpg
salite>
```

### Releases and Code Quality

#### Updates and improvements as of 0.4.x:

Releases have been generated in tandem with tags on GitHub main branch, tracking all minor and patch updates (currently 0.4.3).

All releases automatically pushed to PyPI.

Travis CI runs now performed using Python 3.6, 3.7, and 3.8

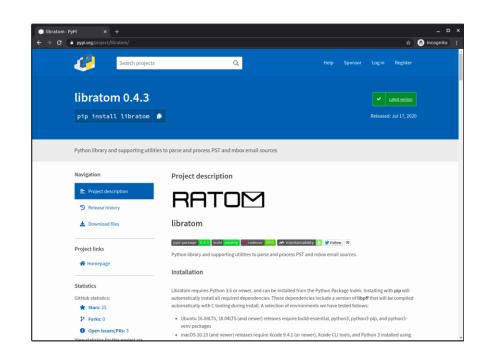
Codebase now tracked with codeclimate to assess maintainability

Code coverage tracked via codecov (currently 95.47%) - effectively all core code is exercised by the test suite

Routine dependency tree checks via dependabot

Code vulnerability/security checks via bandit

Many others...



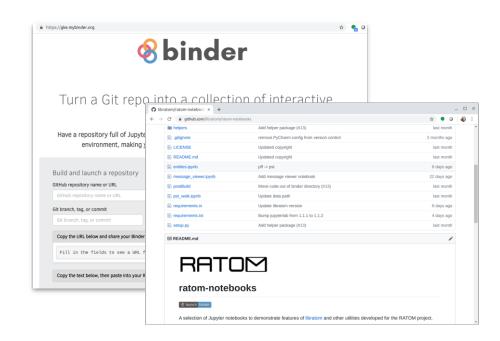
### Notebooks and examples

**Libratom** is still in development, but as we add and test new features we're making some of them available as Jupyter notebooks that you can try out.

These Jupyter notebooks can be run in any Jupyter Hub or Lab instance you choose, but for convenience we've configured them to run in a free hosted service - **mybinder** - in your web browser.

**Mybinder** can create a Jupyter Hub instance from any appropriately configured GitHub repository.

Click the "Launch Binder" badge in the ratomnotebooks repository to get started:



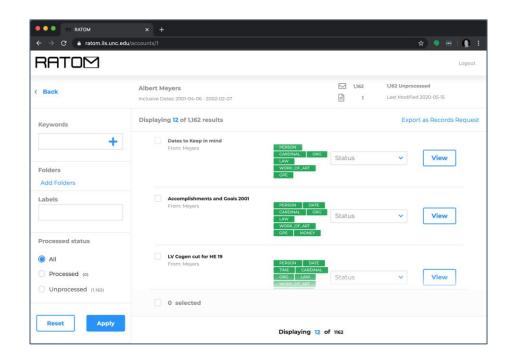
#### https://github.com/libratom/ratom-notebooks

Note that mybinder is a free hosted service. Depending on the current load, it make take a few minutes for the project to start. Be patient!

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Assist archivists in reviewing email materials for retention and/or release.

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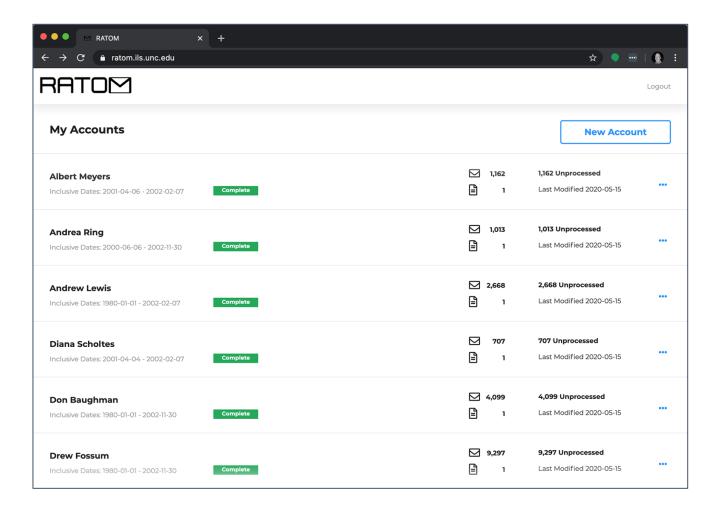


https://github.com/StateArchivesOfNorthCarolina/ratom-deploy

#### **Accounts View**

Accounts associated with imports of one or more imported PST files are displayed in the main interface.

Account processing indicates **Complete** when all entity identification and full-text indexing has finished.

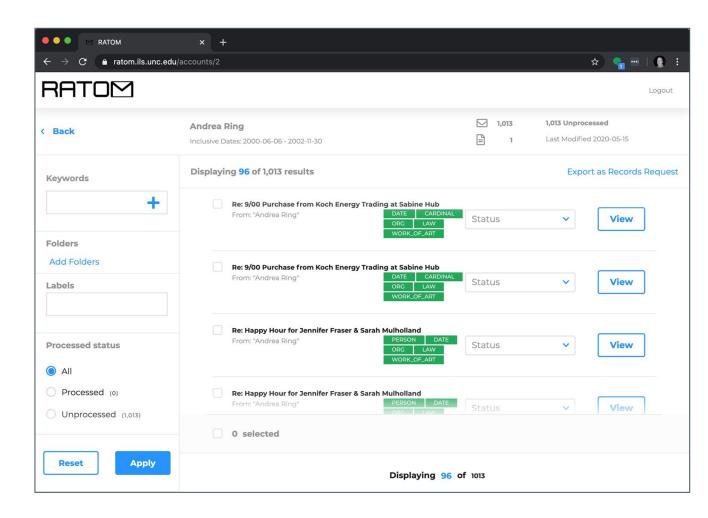


#### **Individual Account**

Selecting an account displays an infinite-scroll view of individual messages associated with that account.

Green tags indicate entity classes identified during processing.

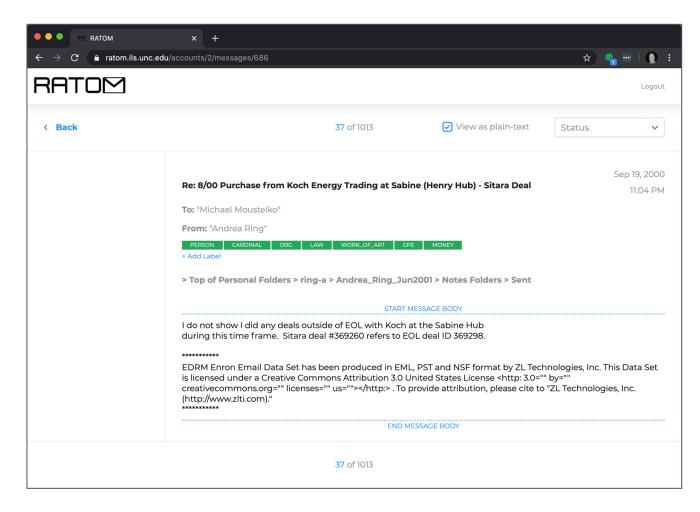
Status dropdown allows messages to be marked for retention or redaction (also appears in individual message view).



#### **Message View**

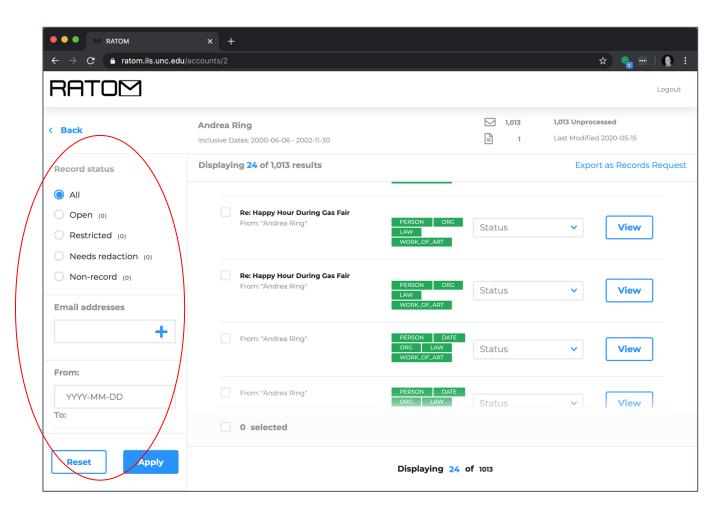
Messages are tagged during ingest using categories associated with entities identified in the body text.

(Note: this research dataset contains prior annotations, resulting in overtagging)



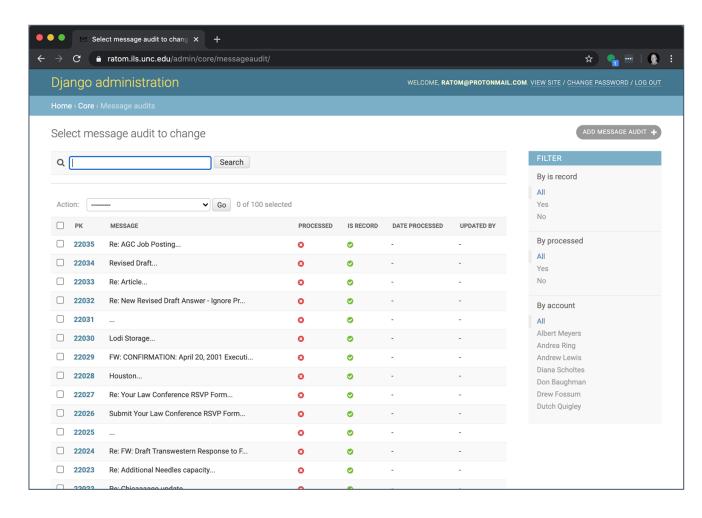
#### **Tagging and Search**

Selection by classification (e.g. record vs non-record) and date range.



#### **Audit History**

Audit histories for individual messages are retained, ensuring a clear record of initial processing actions and potential changes over time.





Project info, news, and blog posts:

https://ratom.web.unc.edu/

Core library:

https://github.com/libratom/libratom

Sample Jupyter notebooks:

https://github.com/libratom/ratomnotebooks



