Hybrid linked data approaches in traditional discovery environments using Share-VDE linked data

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https://guides.library.upenn.edu/linked-data

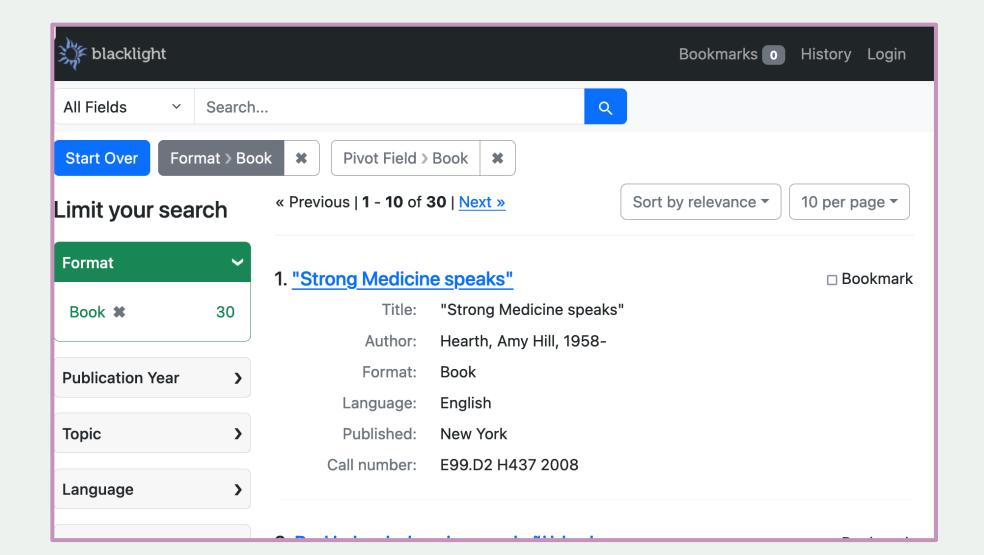


Linked data and more familiar formats

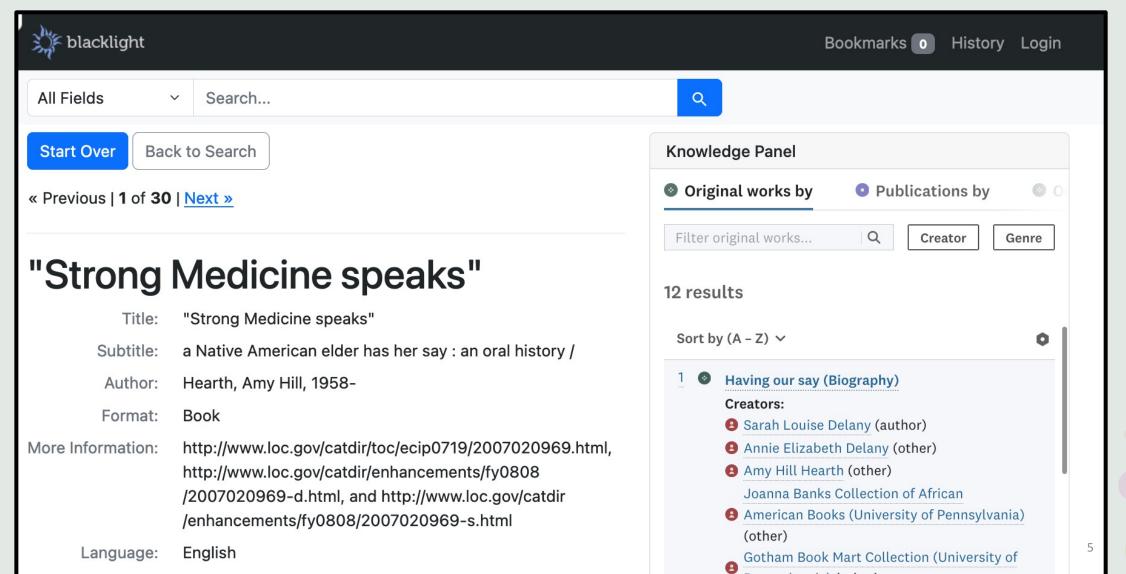
It is often possible to include linked data in more traditional representations, or to make connections between linked data and more familiar formats.

We expect these mixed-format or "hybrid" linked data environments to be the most common way in which linked data is used in production in the next few years

Traditional Discovery

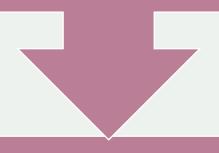


Hybrid Linked Data Discovery



SVDE Linked Data

The Share-VDE project (https://svde.org) is a collaborative discovery environment based on linked data. Explored in this talk are several lesser known and non-intuitive uses of Share-VDE linked data.



Deliverables from Share-VDE

Enriched MARC

RDF triples

SVDE.org

Share-VDE Data



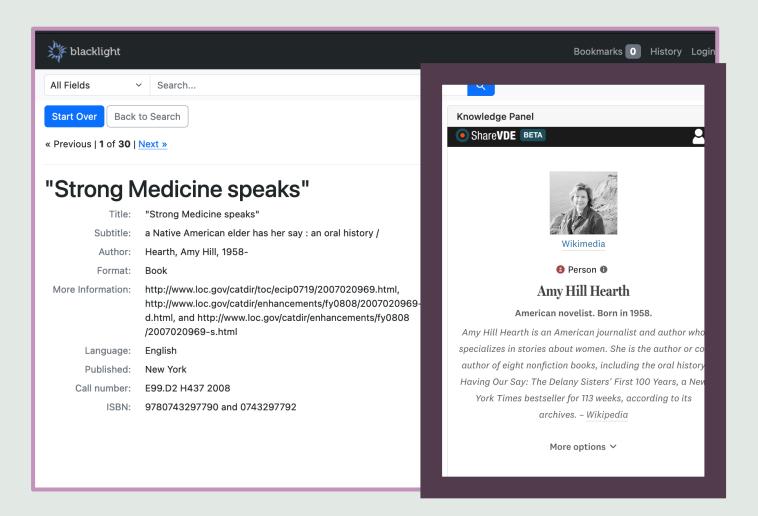




Knowledge Panel



SVDE Person page as Knowledge panel



SVDE Linked Data for Alma Automated Subject Assignment



Alma Sandbox Experiment

Alma/Annif Nightshift

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The project was inspired by the BookOps Nightshift project (https://github.com/BookOps-CAT/NightShift), a copy cataloging bot that used OCLC numbers to match brief records.

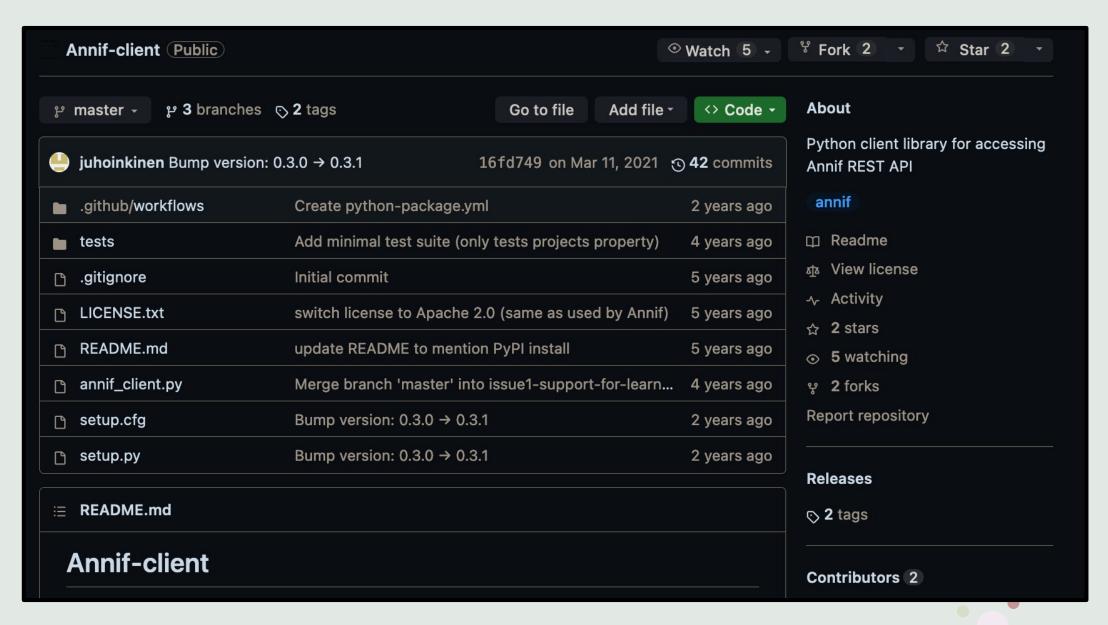
This variation may be useful to enhance brief records when OCLC matching isn't possible.

The data flow begins from Alma Brief record reports. The titles (and/or authors) are sent to a prepackaged machine learning service for FAST subject assignment using the Annif codebase (https://github.com/NatLibFi/Annif).

The machine learning model used for this project is a NN-ensemble (https://github.com/NatLibFi/Annif /wiki/Backend%3A-nn_ensemble) of Omikuji (https://github.com/NatLibFi/Annif/wiki/Backend%3A-Omikuji) and TF-IDF (https://github.com/NatLibFi/Annif/wiki/Backend%3A-TF-IDF) models.

The data in the customized Fast Annif API were collected from Penn Libraries and IvyPlus POD data (https://pod.stanford.edu/).

```
Get a row of the Alma report of titles to be processed
In [...
         #import the data as string data types
nosubjectsdf = pd.read_csv('/Users/jimhahn/Documents/GitHub/alma-nightshift/emptyk
         #what does the data look like?
nosubjectsdf.head()
Out[...
                Type /
                                             Inventory Receiving
                                                                             Library Temporary
                                                                                                     Creation
                                                                    Library
             Creator /
                             Title Barcode
                                               Number
                                                          Number
                                                                                Unit
                                                                                         Library
                                                                                                        Date
               Imprint
                           Premio
                            FIL de
              Book By
                         Literatura
                                                                                                  03/30/2023
           Cartarescu,
                                       NaN
                                                              NaN
                                                                     LIBRA
                                                                                NaN
                                                                                            NaN
                                                   NaN
                               en
                                                                                                     16:24:59
                Mircea
                          Lenguas
                (2022)
                        Romances
        1 rows × 44 columns
```



```
if name == ' main ':
      annif = AnnifClient()
      #select the title from the nosubjectsdf,
title = nosubjectsdf['Title'].iloc[0]
      # send it to Annif API
      url = 'http://jimhahn-dev.library.upenn.int:5000/v1/projects/nn-ensemble-Fast,
      payload = {'text': title}
      req = requests.post(url, data=payload)
      req.raise_for_status()
print(req.json()['results'])
      #print(req.json()['results'][0]['uri'])
      #save the response as a key value pair in pandas dataframe called subjectsdf
subjectsdf = pd.DataFrame(req.json()['results'])
      print(subjectsdf)
label notation
                                                           None
                                       Literary prizes
                                                           None 0.241649
         Premio FIL de Literatura en Lenguas Romances
                        Premio Nacional de Literatura
                                                                 0.177207
                                                           None
                                                                 0.173243
                               Contests in literature
                                                           None
    Spanish literature——Study and teaching (Second...
                                                                 0.172929
                                                           None
                                                           None 0.172634
                                        Premio Cremona
                                                           None 0.162656
                              Building, Brick——Awards
                                Drawing——Competitions
                                                           None 0.160081
           Revolutionary literature, Spanish American
                                                           None 0.159907
                                        Press releases
                                                           None 0.151161
```

Evaluate the subject recommendations looking for any have a confidence score above 0.5

```
# check if the subjects dataframe has any values above 0.5
# and if so, add the subject to a new column in the nosubjectsdf and the subject
if subjectsdf['score'].iloc[0] > 0.5:
    nosubjectsdf['subject'] = subjectsdf['uri'].iloc[0]
    nosubjectsdf['label'] = subjectsdf['label'].iloc[0]
else:
    print("No subject found")
```

Get the brief record from Alma for the title so that we can add the subject to the MARC record

```
# use for statement in https://api-na.hosted.exlibrisgroup.com/almaws/v1/bibs/:mm.
# to download the marc records
 #create a for loop to iterate through the mmsid list
# and download the marc record for each mmsid
 # and save it to a folder on the desktop
 # create a list of mmsids from the nosubjectsdf
 mmsids = nosubjectsdf['MMS_ID'].tolist()
print(mmsids)
 # create a for loop to iterate through the mmsid list
# and download the marc record for each mmsid
 # and save it to a folder on the desktop
 for mmsid in mmsids:
    url = 'https://api-na.hosted.exlibrisgroup.com/almaws/v1/bibs?mms_id='+ mmsid
      print(url)
      r = requests.get(url)
      # parse the response as xml
      root = ET.fromstring(r.content)
      #print(ET.tostring(root, pretty_print=True))
#we only want the record
      record = root.find('.//record')
      #print(record)
      #print(ET.tostring(record, pretty_print=True))
      #save the record as a file
filename = mmsid + '.xml'
      print(filename)
      with open(filename, 'wb') as f:
    f.write(ET.tostring(record, pretty_print=True))
            f.close()
11007006220170260111
```

Before we can add the subject to the MARC record we need to know the type of subject we are working with here

we need to check the type of subject
we can lookup FASTAll/lookup/nt to see the type of subject # If the subject is found in FASTChronological nt it is a Chronological subject # If the subject is found in FASTCorporate nt it is a Corporate subject # If the subject is found in FASTCOrporate.nt it is a Corporate subject
If the subject is found in FASTEvent.nt it is an Event subject
If the subject is found in FASTGeographic.nt it is a Geographic subject
If the subject is found in FASTNamedEvent.nt it is a Named Event subject
If the subject is found in FASTPersonal.nt it is a Personal subject
If the subject is found in FASTTopic.nt it is a Topic subject
If the subject is found in FASTUniformTitle.nt it is a Uniform Title subject #query the FASTAll/lookup/nt/ folder to see if the subject is there # if it is, then add the subject type to the nosubjectsdf # open the nt and read it # if the subject is found in the nt, then add the subject type to the nosubjectsd # create a list of subjects from the nosubjectsdf #make the FASTAll/lookup/FASTTopical-nt.csv into a dataframe fasttopicaldf = pd.read_csv('./FASTAll/lookup/FASTTopical-nt.csv', dtype=str) #make the FASTAll/lookup/FASTChronological-nt.csv into a dataframe
fastchronologicaldf = pd.read_csv('./FASTAll/lookup/FASTChronological-nt.csv', dty #make the FASTAll/lookup/FASTCorporate-nt.csv into a dataframe
fastcorporatedf = pd.read_csv('./FASTAll/lookup/FASTCorporate-nt.csv', dtype=str) #make the FASTAll/lookup/FASTEvent-nt.csv into a dataframe fasteventdf = pd.read_csv('./FASTAll/lookup/FASTEvent-nt.csv', dtype=str) #make the FASTAll/lookup/FASTForGenrem-nt.csv into a dataframe
fastformdf = pd.read_csv('./FASTAll/lookup/FASTFormGenre-nt.csv', dtype=str) #make the FASTAll/lookup/FASTGeographic-nt.csv into a dataframe fastgeographicdf = pd.read_csv('./FASTAll/lookup/FASTGeographic-nt.csv', dtype=st #make the FASTAll/lookup/FASTEvent-nt.csv into a dataframe fastnamedeventdf = pd.read csv('./FASTAll/lookup/FASTEvent-nt.csv', dtype=str) #make the FASTAll/lookup/FASTPersonal-nt.csv into a dataframe fastpersonaldf = pd.read csv('./FASTAll/lookup/FASTPersonal-nt.csv', dtype=str) #make the FASTAll/lookup/FASTTitle-nt.csv into a dataframe
fastuniformtitledf = pd.read_csv('./FASTAll/lookup/FASTTitle-nt.csv', dtype=str)

```
#we're going to be using pymarc to read the marc records and add a new subject fro
#convert the MARCXML at ./alma marc/ to MARC usiing pymarc
mmsids = nosubjectsdfl'MMS_ID'].tolist()
print(mmsids)
# create a for loop to iterate through the mmsid list
# and download the marc record for each mmsid
# and save it to a folder on the desktop
 for mmsid in mmsids:
       with open(mmsid + '.mrc', 'rb') as fh:
    reader = MARCReader(fh)
              for record in reader:
                    record add field

#evaluate The type of topic from the nosubjects, if topical use 650:

if nosubjectsdf['subject_type'].iloc[0] == 'Topical':
    record.add_field(
                                        tag = '650',
indicators = [' ', '7'],
subfields = [
                                               'a', nosubjectsdf['label'].iloc[0],
'2', 'fast',
'0', nosubjectsdf['subject'].iloc[0]
                    #evaluate the type of topic from the nosubjects, if chronological use
elif_nosubjectsdf['subject_type'].iloc[0] == 'Chronological':
                           record.add_field(
                                 Field(
                                        tag = '648',
indicators = [' ', '7'],
                                        subfields =
                                              'a', nosubjectsdf['label'].iloc[0],
'2', 'fast',
'0', nosubjectsdf['subject'].iloc[0]
                    #evaluate the type of topic from the nosubjects, if corporate use 610
elif nosubjectsdf['subject_type'].iloc[0] == 'Corporate':
    record.add_field(
                                 Field(
                                        tàg = '610',
indicators = [' ', '7'],
                                        subfields =
                                               'a', nosubjectsdf['label'].iloc[0],
'2', 'fast',
'0', nosubjectsdf['subject'].iloc[0]
                    #evaluate the type of topic from the nosubjects, if event use 611:
elif nosubjectsdf['subject_type'].iloc[0] == 'Event':
                           record add_field(
Field(
                                        tag = '611',
indicators = [' ', '7'],
                                        subfields = |
                                               'a', nosubjectsdf['label'].iloc[0],
'2', 'fast',
'0', nosubjectsdf['subject'].iloc[0]
                    tag = '655',
indicators = [' ', '7'],
                                        subfields =
                                               'a', nosubjectsdf['label'].iloc[0],
'2', 'fast',
                                                                     - 1051 - 11 - 111 - 11 - 101
```

 Adding the 650 with the FAST heading from Annif.



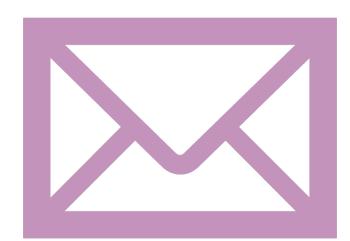
Concluding thoughts

- SVDE data has been useful for several nontraditional uses for Linked Data experimentation and for hybrid linked data discovery.
 - Subject Indexing
 - Knowledge Panels for author agents bringing in related works and biographical assertions



Thank you!

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Acknowledgement

This presentation contains information from FAST (Faceted Application of Subject Terminology) Data which is made available by OCLC Online Computer Library Center, Inc. under the ODC Attribution License.

