

# Automated Subject Indexing and Classification using Annif

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HELDIG Summit, 23 October 2018

# Extrablad till ÅBO UNDERRÄTTELSE

N:o 2.  
Finlands oavhängighet.

Landsdagen omfattar regeringens proklamation om Finlands fullständiga oavhängighet och ansluter sig till huvudsaken i regeringens program för tryggandet av landets nya ställning. Beslutet fattades med 100 rösterna mot 58 vilka till följde ett av socialdemokraternas förordnad förslag.

**Veckans Nyheter**

Den 17:e maj 1918. I följande nummer av Åbo Underrättelser förmedlas de viktigaste nyheter som äro af intresse för den svenska befolkningen.

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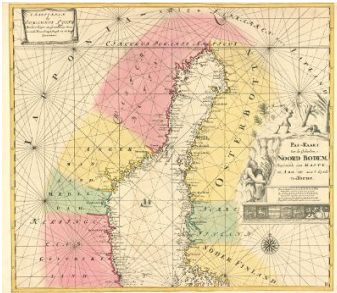
**Årsmötet i Helsingfors**

Årsmötet i Helsingfors öppnades den 17:e maj 1918. Närvarande vid mötet 100 röstberättigade medlemmar. Närvarande vid mötet 100 röstberättigade medlemmar.

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Pris 25 penn.



ÅBO 1918  
ÅBO 1918

Framework for Open Science and Research

ÅBO 2018

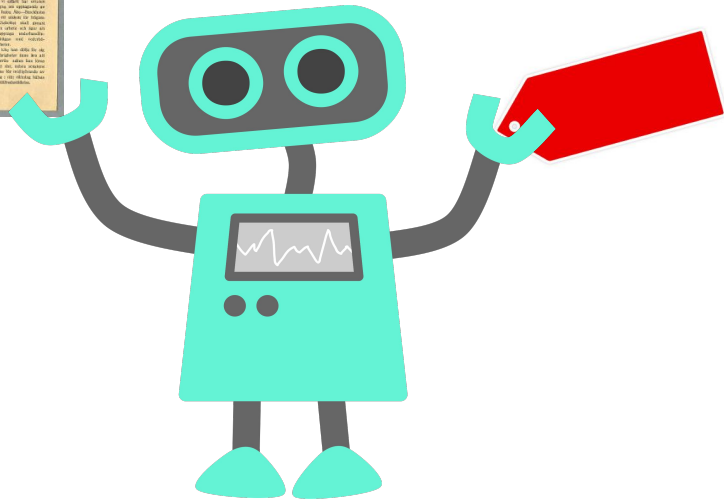
For example, which physical structures will store the digital data resources, who will be responsible using files for data management systems?

The Kartus EA method used in the higher education sector (described below) has been explained in this work.

Open Level	Open Level	Open Level	Open Level	Open Level
Level 1	Level 2	Level 3	Level 4	Level 5
Level 6	Level 7	Level 8	Level 9	Level 10
Level 11	Level 12	Level 13	Level 14	Level 15
Level 16	Level 17	Level 18	Level 19	Level 20
Level 21	Level 22	Level 23	Level 24	Level 25
Level 26	Level 27	Level 28	Level 29	Level 30
Level 31	Level 32	Level 33	Level 34	Level 35
Level 36	Level 37	Level 38	Level 39	Level 40
Level 41	Level 42	Level 43	Level 44	Level 45
Level 46	Level 47	Level 48	Level 49	Level 50
Level 51	Level 52	Level 53	Level 54	Level 55
Level 56	Level 57	Level 58	Level 59	Level 60
Level 61	Level 62	Level 63	Level 64	Level 65
Level 66	Level 67	Level 68	Level 69	Level 70
Level 71	Level 72	Level 73	Level 74	Level 75
Level 76	Level 77	Level 78	Level 79	Level 80
Level 81	Level 82	Level 83	Level 84	Level 85
Level 86	Level 87	Level 88	Level 89	Level 90
Level 91	Level 92	Level 93	Level 94	Level 95
Level 96	Level 97	Level 98	Level 99	Level 100

In accordance with the Kartus EA method, extensive work was conducted by the Kartus EA EA description model by using the physical structure for open science. As the object of the enterprise architecture description is the open science work for open science, the focus has been on the structure that describes the objectives of the digital open science, in accordance with the Kartus model. The following diagram roughly models the sub-description of the enterprise architecture (EAT) in science.





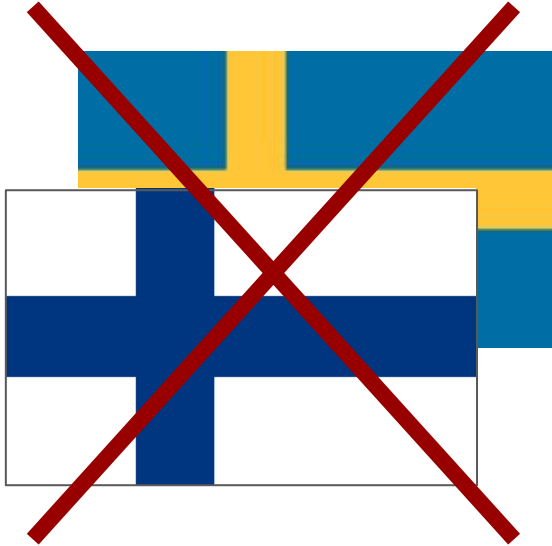


OPEN  
CALAIS



THOMSON REUTERS





~~YSA      YSO  
Allärs    KOKO~~

€ £ \$

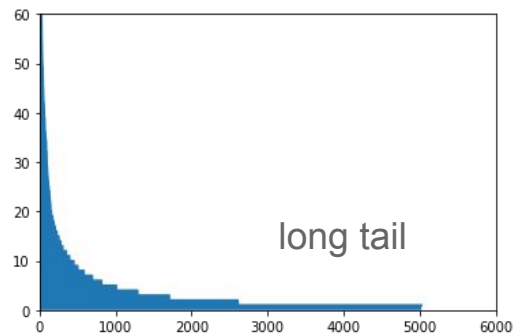
# Subject indexing is a hard problem

## for humans:

- **Subjectivity:** when two people index the same document, only  $\sim\frac{1}{3}$  of the subjects are the same
- **Many concepts:** tens of thousands of concepts to pick from
- **Vocabulary changes:** new concepts are added, existing ones are renamed and redefined

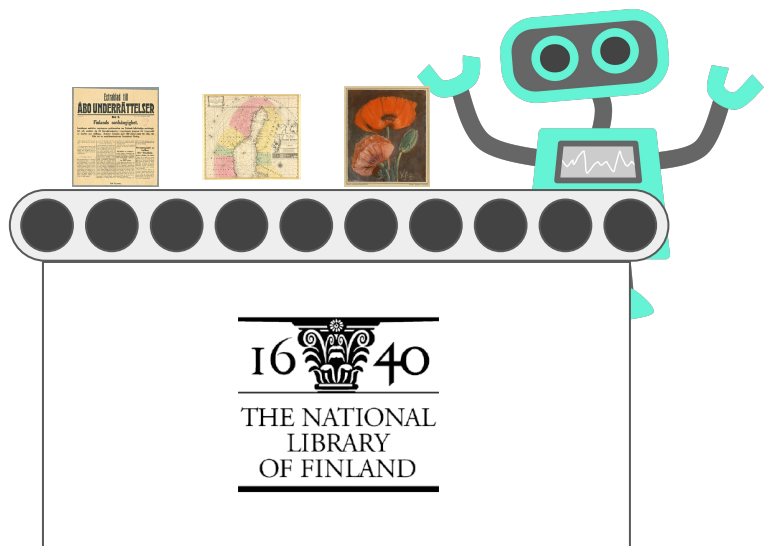
## for machines:

- **Long tail phenomenon:** even with large amounts of training data, most subjects are only used a small number of times
- **Many concepts:** requires complex models that are computationally intensive
- **Difficult to evaluate:** hard to tell “somewhat bad” answers from really wrong ones without human evaluation
- **Vocabulary changes:** models must be retrained



# Approach

Automating our own processes



vs.

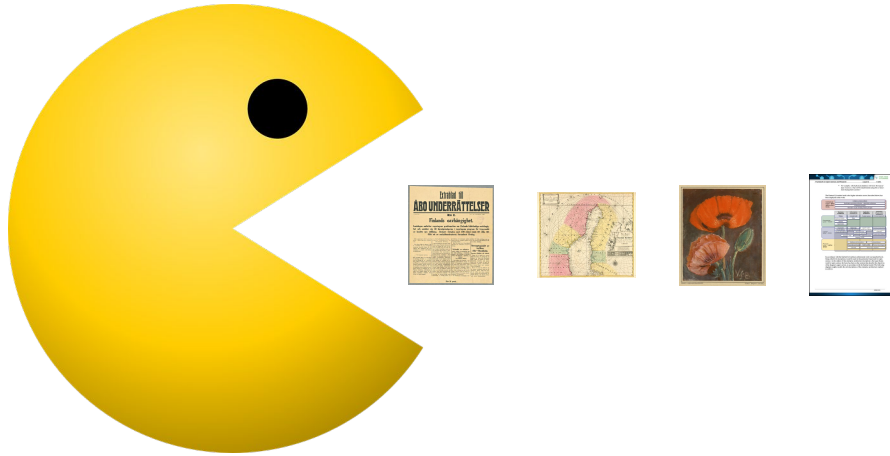
Creating generic tools for many contexts



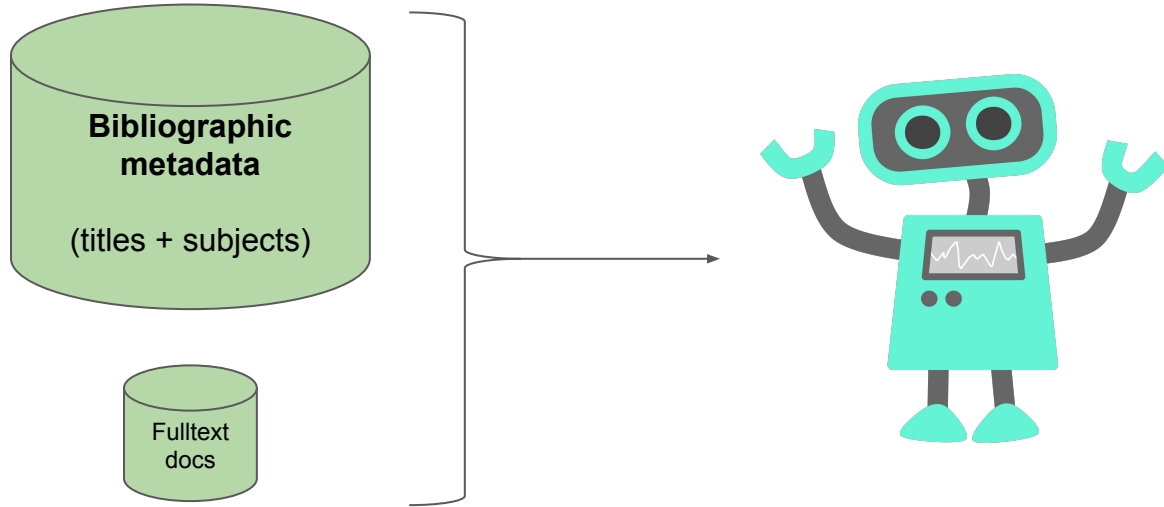


# Enter Annif

*Feed your subject indexing robot with bibliographic metadata!*



# Machine learning requires training data





The material of Finnish archives,  
libraries and museums with a  
single search



All fields



Advanced Search

## Finna Street

Uncover historical images of  
where you are!

 Show Images

## Search...

[Images](#)

[Unrestricted collections](#)

 [Archive collections](#)

 [Library collections](#)

 [Museum collections](#)

### Shortcuts:

[Other Finna websites](#)

[Organisations providing Finna content](#)

# Finna API

 **swagger**  [Explore](#)

## Finna.fi

### Record

[Show/Hide](#) | [List Operations](#) | [Expand Operations](#)

### Search

[Show/Hide](#) | [List Operations](#) | [Expand Operations](#)

[GET](#) /search [Search the index](#)

[ BASE URL: /api/v1 , API VERSION: 3.1.2 ]

[VALID](#) 

All Finna metadata is



!

TITLE            General Finnish upper ontology YSO  
                    YSO - General Finnish ontology

---

SUBJECT        **general concepts**

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DESCRIPTION    General Finnish Upper Ontology YSO is a trilingual ontology consisting mainly of general concepts. YSO has been founded on the basis of concepts in Finnish cultural sphere. As an indexing tool it is best applicable when indexed material is interdisciplinary and its themes vary to a great extent.

#### Resource counts by type

Type	Count
Concept	29031
• Individual concept	1890
• Hierarchical concept	101
• General concept	25940
Collection	241

---

} **~30 000 concepts** that can be used for subject indexing

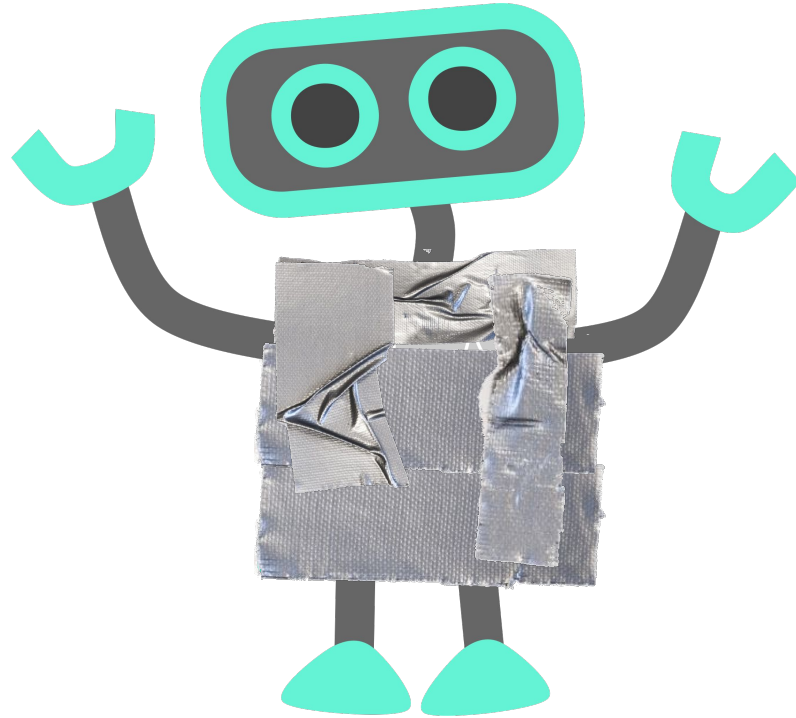
#### Term counts by language

Language	Preferred terms	Alternate terms	Hidden terms
English	28566	3245	11657
Finnish	29019	11491	14288
Swedish	28582	13072	11079

---



# Annif prototype (2017)



# Indexing Wikipedia by topics

Finnish Wikipedia has 410 000 articles (620 MB as raw text)

Automated subject indexing took 7 hours on a laptop, using the Annif prototype

1-3 topics per article (average ~2)

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## Examples: (random sample)

### Wikipedia article

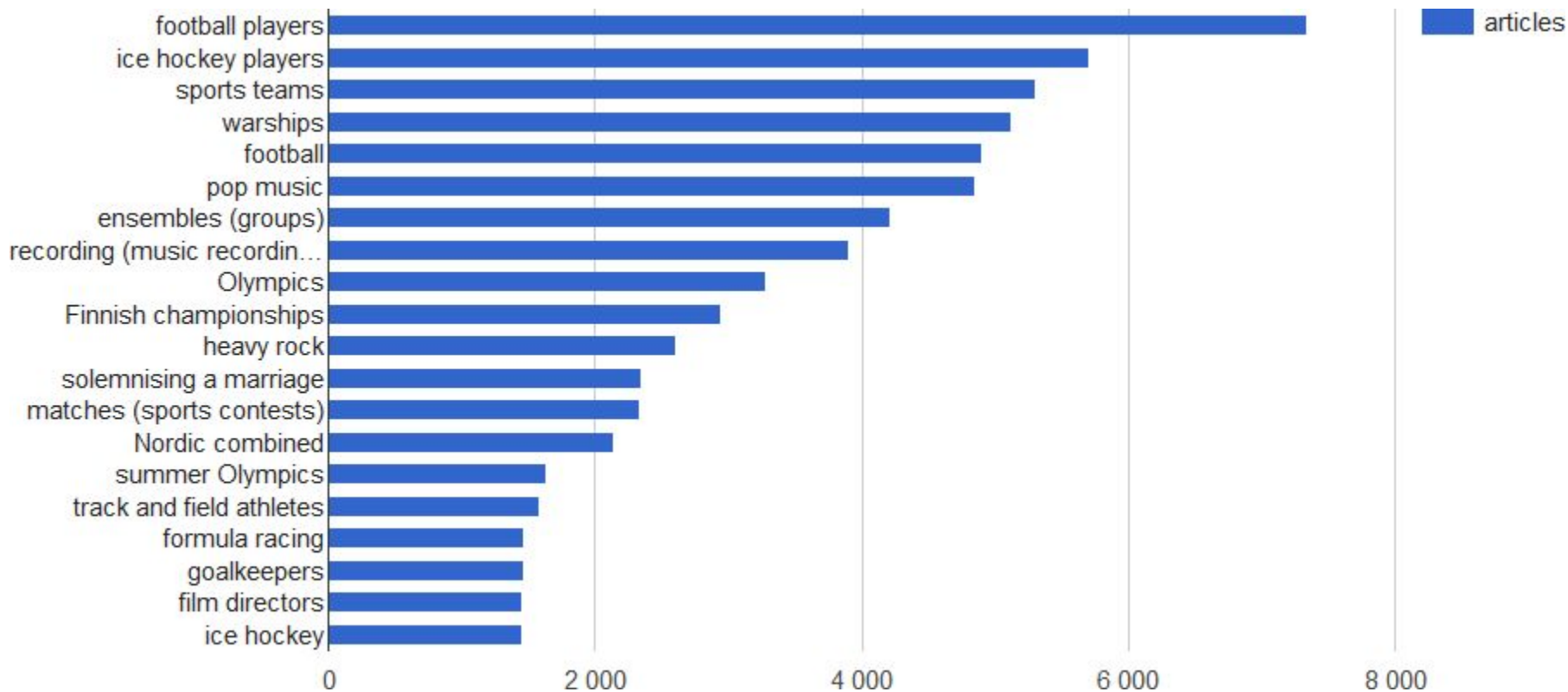
Ahvenuslammi (Urjala)  
Brasilian Grand Prix 2016  
Guy Topelius  
HMS Laforey  
Liigacup  
Pää Kii  
RT-21M Pioneer  
Runoja  
Sjur Røthe  
Veikko Lavi

### YSO topics

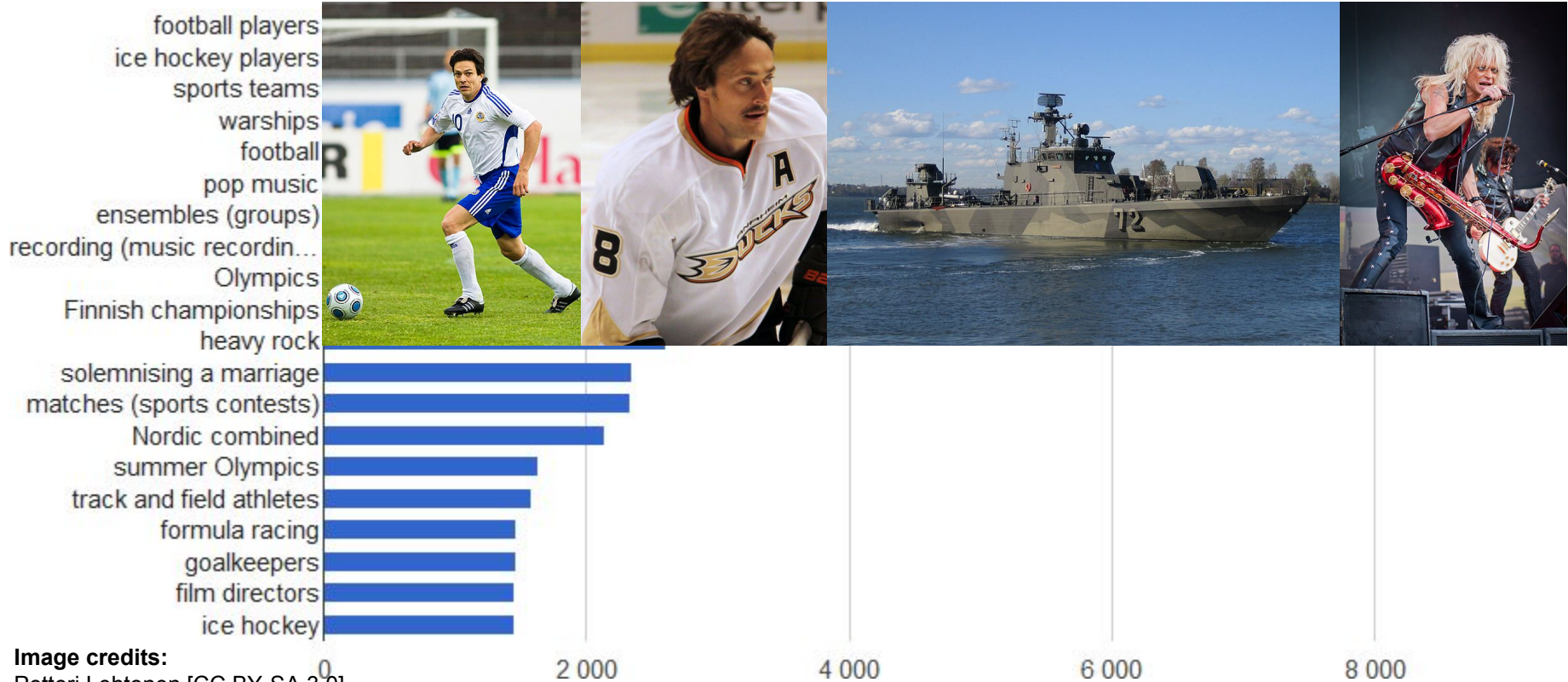
shores  
race drivers, formula racing, karting  
folk poetry researcher, saccharin  
warships  
football, football players  
ensembles (groups), pop music  
missiles  
pop music, recording (music recordings), compositions (music)  
skiers, skiing, Nordic combined  
lyricists, comic songs



# Most common topics in Finnish Wikipedia



# Most common topics in Finnish Wikipedia



## Image credits:

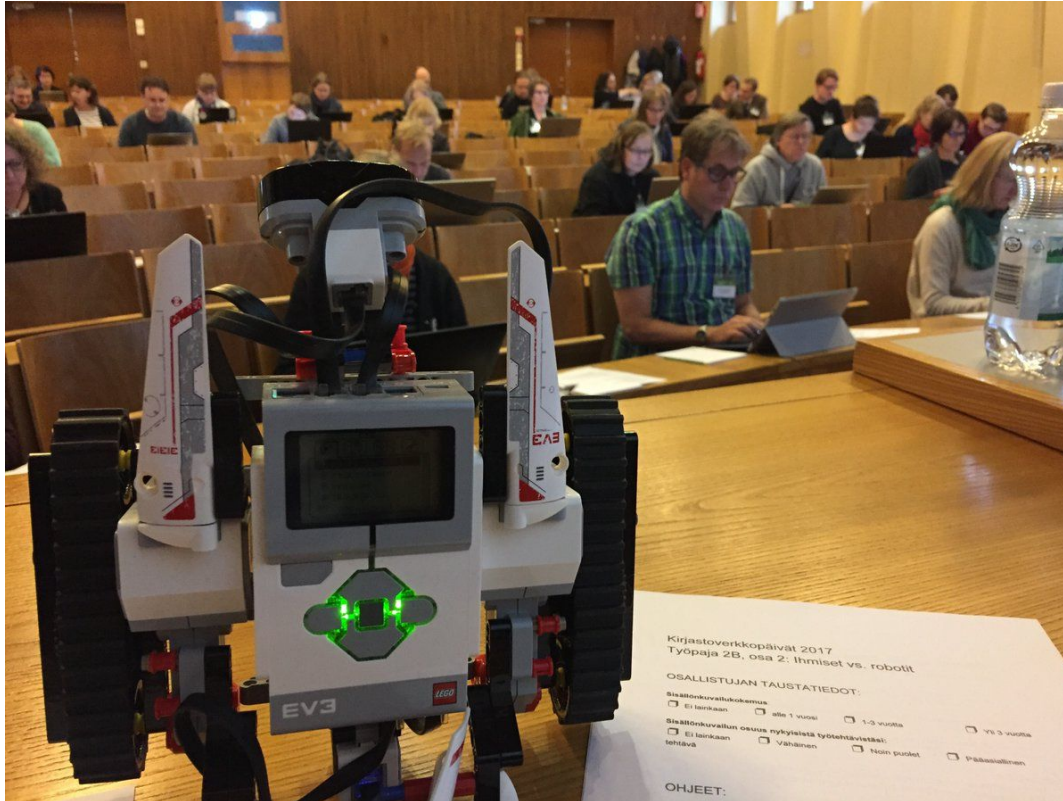
Petteri Lehtonen [CC BY-SA 3.0]

Hockeybroad/Cheryl Adams [CC BY-SA 3.0]

Tomisti [CC BY-SA 3.0]

Tuomas Vitikainen [CC BY-SA 3.0]

# People vs. Robots Workshop



20 documents  
40 librarians  
45 minutes

...

**225 indexing results**

- 11 per document
- 5.5 per person

# Average similarity of subject sets

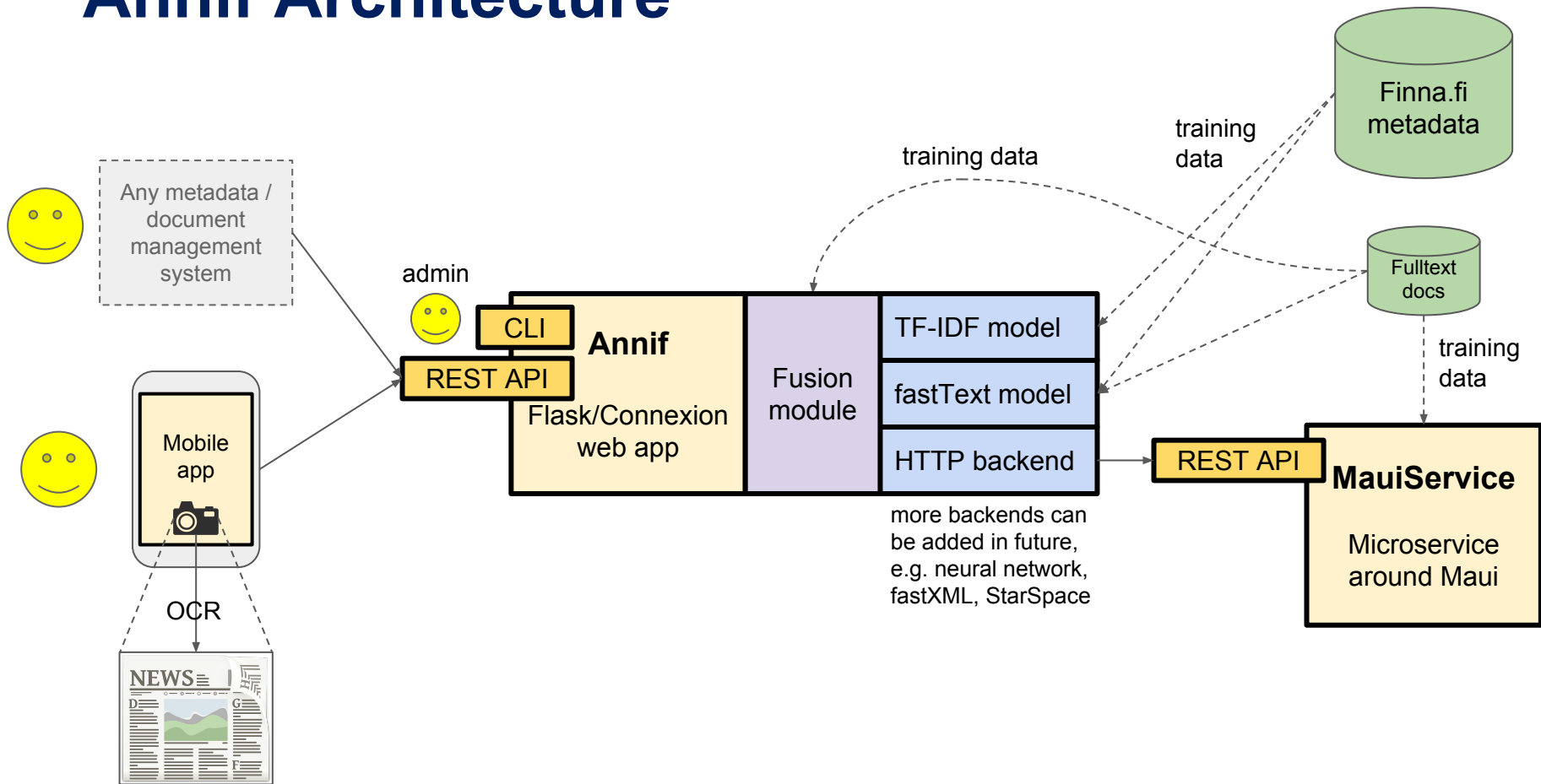
33.39 %

Using Rolling similarity, a.k.a. F1 score, to compare subject sets

# Annif prototype vs. new Annif

	Prototype (2017)	New Annif (2018→)
<i>architecture</i>	loose collection of scripts	Flask web application
<i>coding style</i>	quick and dirty	solid software engineering
<i>backends</i>	Elasticsearch index	TF-IDF, fastText, Maui ...
<i>language support</i>	Finnish, Swedish, English	any language supported by NLTK
<i>vocabulary support</i>	YSO, GACS ...	YSO, YKL, others coming
<i>REST API</i>	minimal	extended (e.g. list projects)
<i>user interface</i>	web form for testing	<a href="http://dev.annif.org">http://dev.annif.org</a>
<i>mobile app</i>	HTML/CSS/JS based	(native Android app?)
<i>open source license</i>	CC0	Apache License 2.0

# Annif Architecture



# Backends / Algorithms

- **TF-IDF similarity**

Baseline bag-of-words similarity measure. Implemented with the [Gensim](#) library.

- **[fastText](#)** by Facebook Research

Machine learning algorithm for text classification.

Uses word embeddings (similar to [word2vec](#)) and resembles a neural network architecture.

Promises to be good for e.g. library classifications (DDC, UDC, YKL...)

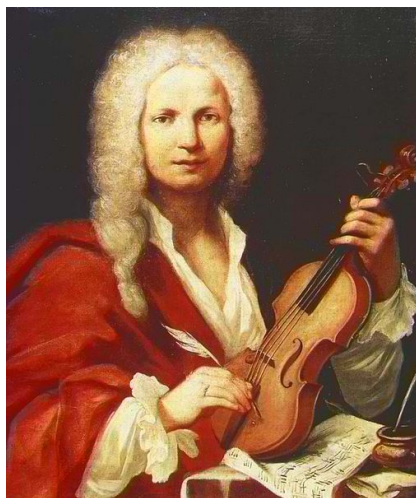
- **HTTP backend** for accessing MauiService REST API

[MauiService](#) is a microservice wrapper around the [Maui](#) automated indexing tool.

Based on traditional Natural Language Processing techniques - finds terms within text.

# Backend configuration

Backends may be used alone, or in combinations (ensembles)



**Current challenge:** Which fusion method works best for combining results from multiple backends?

[An experiment testing different fusion methods](#)



# Command line interface

## Load a vocabulary to be used by one or more models:

```
$ annif loadvoc yso-en yso-en.tsv
```

## Train a model:

```
$ annif train tfidf-en yso-finna-en.tsv.gz
```

## Analyze a document:

```
$ annif analyze tfidf-en <berries.txt>
<http://www.yso.fi/onto/yso/p772>    strawberry          0.39644203283656165
<http://www.yso.fi/onto/yso/p18109> wild strawberry     0.37539359094384245
<http://www.yso.fi/onto/yso/p25548> stolons            0.3261554545369906
<http://www.yso.fi/onto/yso/p6749>  berry cultivation  0.2394291077460799
<http://www.yso.fi/onto/yso/p10631> questionnaire survey 0.22714475653823335
<http://www.yso.fi/onto/yso/p6821>  farms             0.21725243067995587
<http://www.yso.fi/onto/yso/p3294>  customers         0.216395821347059
<http://www.yso.fi/onto/yso/p1834>  work motivation   0.21612376226244975
<http://www.yso.fi/onto/yso/p8531>  customership      0.21536113638508098
<http://www.yso.fi/onto/yso/p19047> corporate clients  0.21412270159920782
```

## Evaluate a model using several measures (e.g. recall, precision, F1 score, NDCG):

```
$ annif eval tfidf-en directory-with-gold-standard-docs/
```

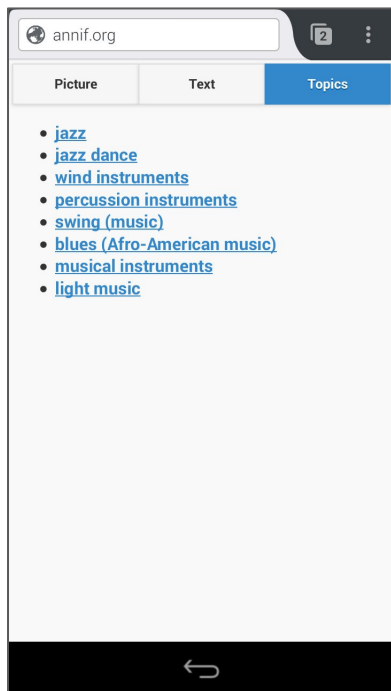
# REST API

## Main operations:

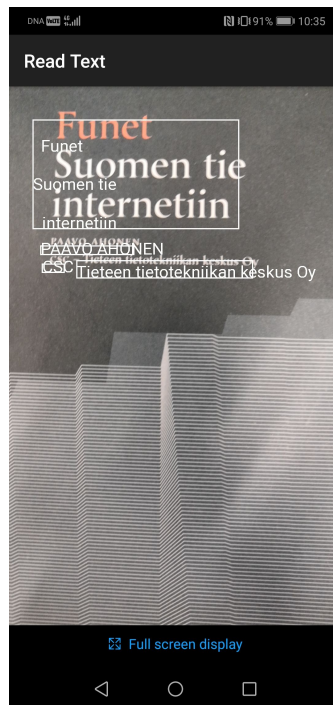
GET /projects/	list available projects
GET /projects/<project_id>	show information about a project
POST /projects/<project_id>/analyze	analyze text and return subjects
POST /projects/<project_id>/explain	analyze text and return subjects, with explanations indicating why they were chosen
POST /projects/<project_id>/train	train the model by giving a document and gold standard subjects

Defined using a Swagger / OpenAPI [specification](#)

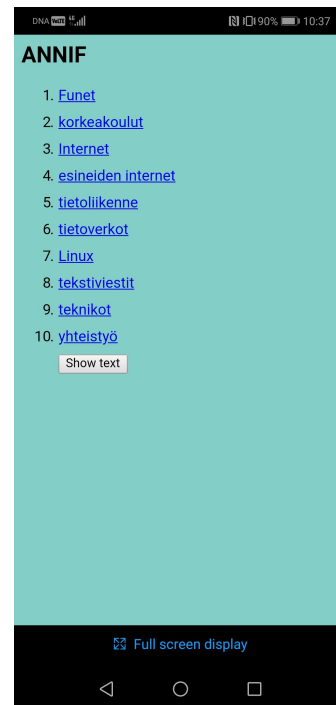
# Mobile apps



Prototype web app,  
ocr.space cloud OCR  
[m.annif.org](http://m.annif.org)



Prototype Android app with OCR on the device  
(by Okko Vainonen)



# Test corpora

Full text documents indexed with YSA/YSO for training and evaluation

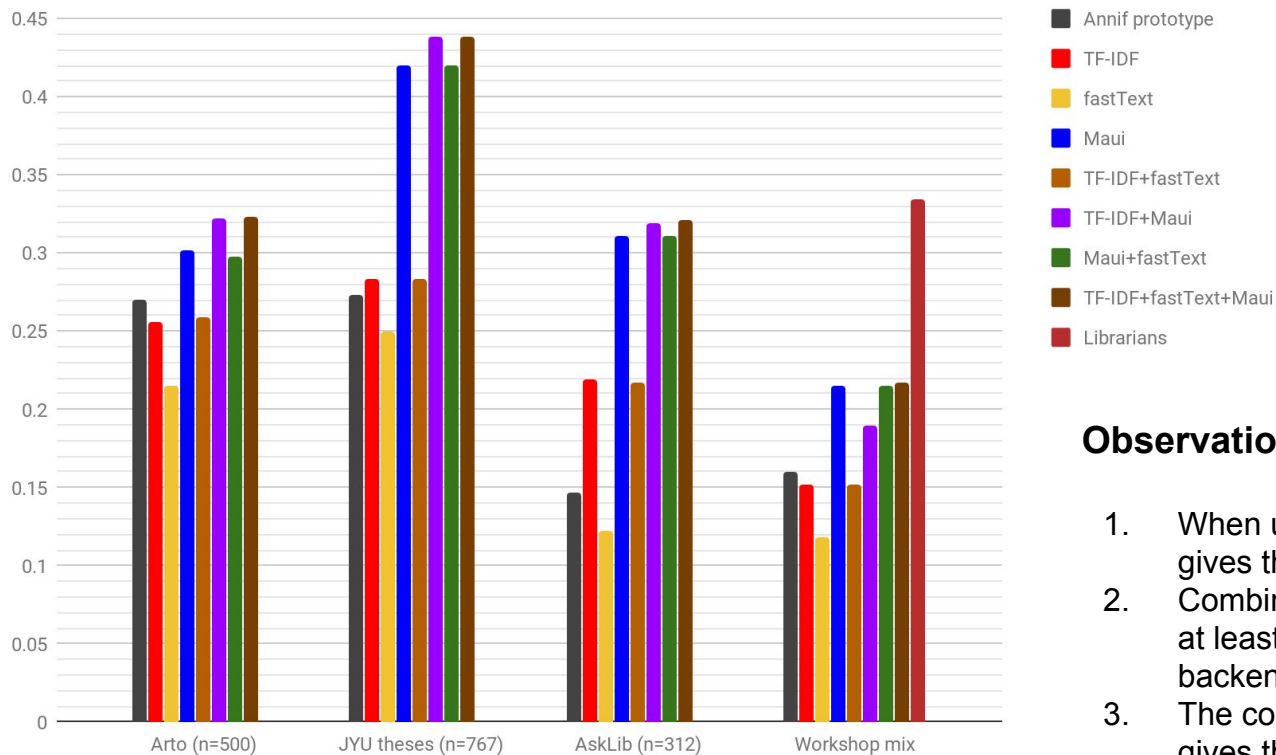
- Articles from Arto database (n=6287)  
Both scientific research papers and less formal publications. Many disciplines.
- Master's and Doctoral theses from Jyväskylä University (n=7400)  
Long, in-depth scientific documents. Many disciplines.
- Question/Answer pairs from an Ask a Librarian service (n=3150)  
Short, informal questions and answers about many different topics.

Available on GitHub: <https://github.com/NatLibFi/Annif-corpora>

(for the first two corpora, only links to PDFs are provided for copyright reasons)

# Evaluation of different backends

F-measure similarity scores against a gold standard



## Observations:

1. When using just one backend, Maui often gives the best results
2. Combinations (ensembles) usually give at least as good results as single backends
3. The combination of all three backends gives the best results

558 commits 10 branches 34 releases 3 contributors View license

Branch: master New pull request Create new file Upload files Find file Clone or download

osma Bump version: 0.33.0 → 0.34.0 Latest commit 4894a60 7 days ago

annif	break up AnnifProject.initialize() into smaller pieces (and rename of...	7 days ago
swagger	Handle errors in REST API. Part of #187	7 days ago
tests	More REST error handling tests	7 days ago
.codeclimate.yml	more comprehensive Code Climate configuration	a year ago
.codecov.yml	Codecov should ignore setup.py	6 months ago
.coveragerc	Generate Codecov reports	a year ago
.gitignore	Rename projects.cfg into projects.cfg.dist so deployments can use the...	5 months ago
.jgtn.yml	Add LGTM configuration excluding fasttext	26 days ago
.scrutinizer.yml	Try to fix pipenv/pip compatibility issue pypa/pipenv#2924 within Scr...	14 days ago
.travis.yml	use fasttextmirror package from official PyPI instead of fasttext fro...	26 days ago
LICENSE.txt	Switch to Apache license. Fixes #6	a year ago
Pipfile	Enable CORS requests to REST API using flask-cors. Fixes #190	7 days ago
README.md	add LGTM badge, drop Coveralls badge for now	26 days ago
autopep8.sh	refactor: separate merge_hits into a shared utility function	5 months ago
config.py	add tests for the initialize functionality	6 months ago
projects.cfg.dist	Add vocab settings to example configuration file, needed after #180	14 days ago
pytest.ini	add pep8 checks to pytest	7 months ago
setup.cfg	Bump version: 0.33.0 → 0.34.0	7 days ago
setup.py	Bump version: 0.33.0 → 0.34.0	7 days ago

README.md

# Annif

license Apache 2.0 build passing codecov 98% maintainability A Scrutinizer 9.95 codebeat A Better Code 9 / 10

code quality: python A+

# Annif on GitHub

Python 3.5+ code base

Apache License 2.0

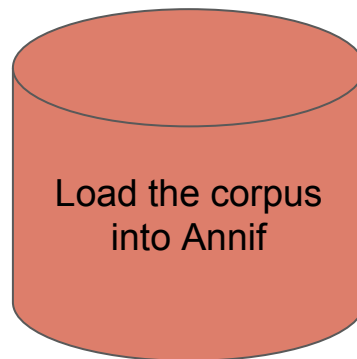
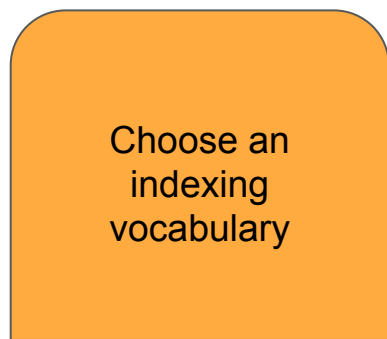
Fully unit tested (98% coverage)

PEP8 style guide compliant

Usage documentation in the wiki

<https://github.com/NatLibFi/Annif>

# Apply Annif on your own data!



# Lessons learned (so far)

1. Good quality training data is key for training and evaluation  
Don't expect good results if you don't have the data it takes
2. Gold standard subjects are useful, but human evaluation is necessary  
Subject indexing is inherently subjective; comparing to a single gold standard can be misleading
3. All algorithms have strong and weak points  
Combinations work better than any algorithm by itself
4. Surprising amount of interest also from non-library organizations  
Archives, media organizations, book distributors ... automation is better done together!



# Thank you!

## Questions?

[osma.suominen@helsinki.fi](mailto:osma.suominen@helsinki.fi) - [@OsmaSuominen](https://twitter.com/OsmaSuominen)

Website: <http://annif.org>

Code: <https://github.com/NatLibFi/Annif>

Test corpora: <https://github.com/NatLibFi/Annif-corpora>

These slides: <https://tinyurl.com/annif-heldig>