

# Unsilencing Colonial Archives via Automated Entity Recognition

[Dataset](#)

Related Publication (under review)

Colonial archives remain difficult to access due to significant persisting barriers such as biases to be found in historical findings aids, such as indexes of person names, which perpetrate silences by omitting to include mentions of historically marginalized persons. In order to overcome such limitations and pluralize the scope of existing finding aids, we propose using automated entity recognition for content based indexing. To this end, we contribute a fit-for-purpose annotation typology and apply it on the colonial archive of the Dutch East India Company (VOC). We release a corpus of nearly 70,000 annotations as a shared task, for which we provide strong baselines using state-of-the-art neural network models.

## Authorship

### PUBLISHER(S)

Nationaal Archief,  
University of Amsterdam,  
Emerald Publishing

### INDUSTRY SECTOR

Academic

### DATASET CURATORS

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### FUNDING

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CREATE, UvA funds

### FUNDING TYPE

Public Research Funding

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## Motivations

### DATASET PURPOSE(S)

Research Purposes  
Machine Learning  
Training, testing and validation

### KEY APPLICATIONS

Machine Learning  
Entity Recognition

### PROBLEM SPACE

This dataset was created for training entity recognition models to create more inclusive content based indexes on the collection of VOC testaments. See accompanying article (in peer review).

### PRIMARY MOTIVATIONS

Provide ground truth for  
training entity recognition  
models on colonial archives

### INTENDED AND/OR SUITABLE USE CASE(S)

ML Model Evaluation & ML Model Training for:

- Entity detection
- Attribute detection

## Uses of Dataset

### SAFETY OF USE

Research Use

### CONJUNCTIONAL USE

Safe to use with other datasets

### KNOWN CONJUNCTIONAL USES AND DATASETS

Similar archival records.

### METHOD

Entity Recognition

### SUMMARY

An entity recognition and  
classification model can be  
trained for historical Dutch.

### KNOWN CAVEATS

This dataset contains a proportionally low number of  
organizations because of incomplete annotations.

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## Dataset Snapshot

### PRIMARY DATA TYPES

### DATASET SNAPSHOT

### DESCRIPTION OF CONTENT

Sensitive data about people

Data about places, organizations and proper names

Total Entities	32,203
Total Attributes	36,226
Total Annotations	68,429
Training	70%
Validation	10%
Testing	20%
Total Tokens Annotated	79,837
Average tokens per label	2.7
Human Annotated Labels	All

This dataset is based on the digitized collection of the Dutch East India Company (VOC) Testaments under the custody of the Dutch National Archives. These testaments are mainly from the 18th century, drawn up in the Dutch Republic, VOC ships and in the settlements. The testaments have a fixed order in the text structure and the language is a variety of historic Dutch.

The dataset has 68,429 annotations spanning over 79,837 tokens across 2193 unique pages. 47% of the total annotations correspond to entities and 53% to attributes of those entities. Of the 32,203 entity annotations, 11,715 (36.3%) correspond to instances that represent persons with associated attributes of gender, legal status and notarial role, 4,510 (14%) correspond to instances of places, 1,080 (3.4%) correspond to organizations with attribute beneficiary and 14,898 (46.2%) correspond to proper names (of places, organizations and persons).

### PRIMARY DATA MODALITY

### KNOWN CORRELATIONS

### HOW TO INTERPRET DATAPPOINT

Labels or Annotations

Gender presentation numbers are skewed towards predominantly **man** and **unspecified**;  
Legal status numbers are skewed towards **unspecified**

**Each datapoint** refers to a central entity that can be a person, place, organization or proper name or their attributes such as gender, legal status and notarial role of a person.

Each entity is represented by a span of characters across single or multiple connected tokens or words.

## Datapoint Example

The shared annotation task was performed on the Brat annotation software. For each page of annotations of the testaments corresponding to a .txt file, an annotation file with .ann suffix was created. The general annotation structure is that each line of the .ann file contains one annotation, and each annotation is given an ID that appears first on the line, separated from the rest of the annotation by a single TAB character. The initial ID character 'T' corresponds to text bound annotations whereas 'A' corresponds to an attribute. Consider this example of an annotation from the sentence "Emancipatie van lijfeigenen, en ...":

T1	Person 1298 1310	lijfeigenen
A1	Gender T1	Group
A2	LegalStatus T1	Enslaved
A3	Role T1	Beneficiary

Here, the term 'lijfeigenen' [serf] with characters spanning 1298 to 1310 on that particular page is annotated as entity: Person with attributes A1, A2 and A3 corresponding to that Person's gender, legal status and notarial role.

The dataset is also provided in **machine-readable IOB format**.

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## Data Collection & Sources

### DATA COLLECTION METHODS

Annotations by paid students and professionals

### DATA SOURCE

Digitized collection of the VOC Testaments. The testaments consist of 51 extant bundles consisting of 10,000 wills mainly from the 18th century.

### DESCRIPTION OF DATA SOURCE

**HTR Quality:** The testaments were extracted via handwritten text recognition by the Dutch National Archives with a character error rate of 5.3 on a test set and 7.3 on a held out sample.

**Speech Situation:** The testaments were collected in the 17th and 18th centuries and information about which varieties of Dutch are represented is not available.

### DATASET TYPE

Static

Data was collected once from a single source

### COLLECTION METHODS

Annotations were created as a shared annotation task on the Brat annotation [software](#).

### DATA SELECTION CRITERIA

Pages were randomly sampled from 13 non consecutive and equally spaced (every 4th) bundle to capture as much variation in content and transcription quality.

### DATA PROCESSING

The data i.e., the collection of annotations were cleaned to remove:

- Incomplete annotations: where a span is labeled as an entity but at least one of the corresponding attributes' value was not chosen by the annotator.
- Duplicate pages: HTR errors sometimes result in duplicate pages, these were labeled by the annotators as duplicates and were excluded from the dataset.

## Labeling Process

### LABELING METHOD

Manual Annotations

### ENTITY TYPES

Entity	#	%
Person	11,715	36.4
Place	4,510	14
Organization	1,080	3.4
ProperName	14,898	46.2

### METHOD SUMMARY

Annotations were created by highlighting the relevant span of text and choosing its entity type and where applicable exactly one attribute value through a drop down menu.

To tag the same span as two entities, the span must be selected two times and labeled accordingly. For example: 'Adam Domingo' has been labeled twice as a *Person* and *ProperName*.

### ENTITY TYPE

Person

### ATTRIBUTE DISTRIBUTION

Gender	#	%
Man	4,303	36.7
Woman	1,232	10.5
Group	420	3.6
Unspecified	5,760	49.2

### DESCRIPTIONS & MOTIVATIONS

When the mention of a person is followed or preceded by trigger words which reveal their gender, the text is annotated as a *Person* with the appropriate value of the attribute *Gender*.

When a person is mentioned without a gender trigger word, their gender is marked as *Unspecified*. This approach restricts possible 'annotator bias' due to unfounded inferences. Persons are annotated by trigger words alone, in the absence of a proper name and in the case marginalized persons such as enslaved and formerly enslaved persons. This is because such persons are often mentioned without name and are of particular interest to our research.

**iNote** Non-binary is not included in set of gender attribute values given that we couldn't find any instances in the data source.

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## Labeling Process

### ENTITY TYPE

### ATTRIBUTE DISTRIBUTION

### DESCRIPTIONS & MOTIVATIONS

Person

Legal Status	#	%
Free(d)	154	1.3
Enslaved	885	7.6
Unspecified	10,676	91.1

The attribute legal status takes the value *Enslaved* when the trigger words clearly identify the individual(s) to be enslaved. The attribute value *Free(d)* is most often triggered by the word ‘vrije’ [free]. It refers to persons who were set free, children of the manumitted slaves and the groups of free indigenous. The attribute value *Free(d)* captures these three different senses of the word ‘vrije’, for which there is no clear way to clearly disambiguate among. When no trigger words are used or don’t indicate legal status, the legal status is annotated as *Unspecified*.

The motivation to include legal status as a semantic category is because enabling findability of marginalized groups in colonial archives is one of the primary goals of the project.

### ENTITY TYPE

### ATTRIBUTE DISTRIBUTION

### DESCRIPTIONS & MOTIVATIONS

Person

Role	#	%
Testator	1,289	11
Beneficiary	1,830	15.6
Notary	473	4
ActingNotary	801	6.8
Testator Beneficiary	278	2.4
Witness	1,107	9.4
Other	5,937	50.7

In the historic index—used until now— only the male testator was indexed, thus silencing women co-testators, beneficiaries such as enslaved persons, concubines, children, etc. The attribute *Role* was thus created to refer to roles specific to testaments in notarial archives, which may take exactly one of the following values listed in the adjacent table.

An instance of a role is the *Testator beneficiary* which is attributed to those people who are both testator and beneficiary in the context of the testament. For instance, when man and wife collectively write down their testaments, each of them is a testator and often both of them are also each-other’s beneficiaries.

### ENTITY TYPE

### ATTRIBUTE DISTRIBUTION

### DESCRIPTIONS & MOTIVATIONS

Place

No attributes

The entity *Place* is used to annotate places or locations. This entity is often called *Location* in other typologies such as CoNLL.

### ENTITY TYPE

### ATTRIBUTE DISTRIBUTION

### DESCRIPTIONS & MOTIVATIONS

Proper Name

No attributes

The entity *Proper name* refers to names (proper nouns) of the other entities in this typology: *Person*, *Place* and *Organization*. In this typology we separate the name of an entity from a generic reference to an entity type because marginalized persons in colonial archives are frequently mentioned without name.

### ENTITY TYPE

### ATTRIBUTE DISTRIBUTION

### DESCRIPTIONS & MOTIVATIONS

Organization

Beneficiary	#	%
Yes	162	15
No	918	85

This entity refers to organizations such as companies, orphanages, religious institutions and other branches of the church. Organizations have the attribute *Beneficiary* which can take the value *Yes* or *No* depending on whether the testator decides an organization to be their beneficiary.

# Unsilencing Dutch Colonial Archives

## Use in Machine Learning or AI Systems

### DATASET USE(S)

Training  
Testing  
Validation

### DATASET SPLIT(S)

We divide the corpus of annotations into three splits: training (70%), validation (10%), and test (20%). We randomly sample annotated pages into splits by applying stratified sampling over annotation typologies and annotators, to maintain the overall data distribution within every split.

### USAGE GUIDELINES OR POLICIES

CRF baseline is a strong option compared with neural network-based approaches. For further information, refer to the paper.

## License & Access

### LICENSE TYPE(S)

CC BY 4.0

### LICENSE BREAKDOWN

Annotations are licensed under CC BY 4.0 License.

[CC BY 4.0](#)

### LICENSE PERMISSIONS

**Share** — copy and distribute the material in any medium or format.

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### ACCESS TYPE(S)

Open Access

### ACCESS COST

N/A - Open Access

### ACCESS PREREQUISITE(S)

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### ACCESS SUPPORT

Dataset Website

### DATASET WEBSITE

[https://github.com/budh333/UnSilence\\_VOC](https://github.com/budh333/UnSilence_VOC)

### ACCESS DETAILS

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Research Paper

### RESEARCH PAPER

Paper currently under review

### CITATION GUIDELINE(S)

Mrinalini Luthra, Konstantin Todorov, Charles Jeurgens, Giovanni Colavizza, and Leon van Wissen. "Unsilencing Dutch Colonial Archives". Zenodo, August 3, 2022. <https://doi.org/10.5281/zenodo.6958431>.

## Versioning & Maintenance

### VERSION STATUS

Limited Maintenance

This data will not be updated, but any technical issues will be addressed

### DATASET STATUS

Version	1.1
Last Updated	16/08/2022
First Released	16/08/2022

### MAINTENANCE PLAN

- No refreshes planned
- Dataset may be updated to incorporate feedback

# Unsilencing Dutch Colonial Archives

## Description of Annotators & Curators

### CURATORS

Mrinalini Luthra is responsible for overseeing the project.

Charles Jeurgens is the archival expert, who provided context of the archival records and terms that occur within them.

Giovanni Colavizza is the computer science expert.

Konstantin Todorov is the machine learning expert who set up and trained the baseline models.

Leon van Wissen set up the infrastructure for the collaborative annotation task.

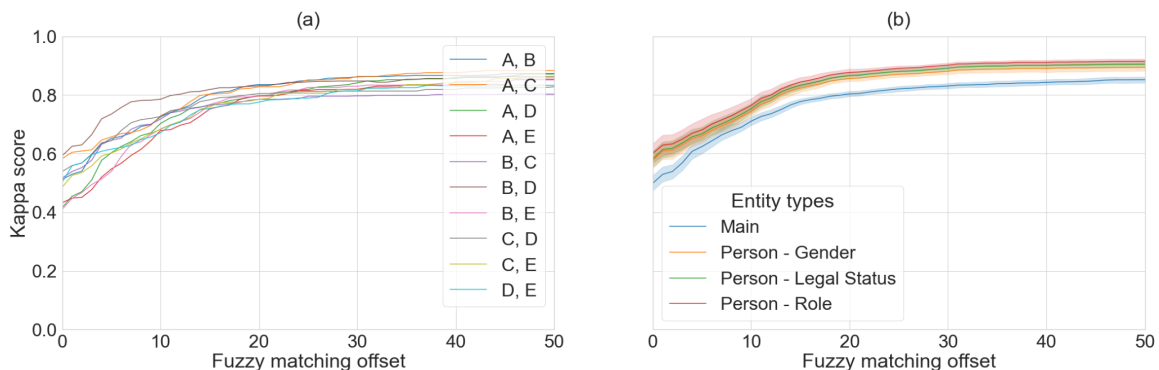
### ANNOTATORS

Annotators were recruited specifically for their expertise in 1) reading and understanding historical Dutch and 2) archival and historical knowledge. During the annotation process all annotators were trained to read and understand the original texts by the archival expert and were invited to compare the HTR texts with the scans of the original. This way of working proved instrumental in overcoming limitations of HTR quality.

† All annotators and curators were paid as per the Collective Labor Agreement of Dutch Universities

The group (annotators and curators) self-identified by gender as: women (45%) and men (55%) and ethnicity: afro caribbean (9%), south asian (9%) and white european (82%).

## INTER-ANNOTATOR AGREEMENT



Cohen's kappa score to evaluate the inter-annotator agreement. We measure it both exactly and using a *fuzzy matching offset*. This we define as the character offset that can exist between the same annotation given by two different annotators. Using an offset of 0 is equivalent to requiring an exact match, whereas an offset of 5 characters would entail considering two annotations to be the same if they overlap with a discrepancy of 5 characters at most. The inter-annotator agreement results between all pairs of annotators are shown in the first figure, while the average scores per entity are shown in the second). While with exact comparisons the kappa scores are only of moderate quality (0.5-0.6), with a modicum of fuzziness they converge to acceptable or strong values of 0.7-0.8 (at the 10 character offset mark).