### UNESCO Radio Archives: Alfor Audio Metadata Enrichment

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### **Driving Question**

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How can AI enable better description of archival audio? WHO NH

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#### **Project Goals**

Understand relationship between a physical record and its digital surrogate

Evaluate whether diplomatic analyses apply to various genres of audio recordings

Analyze whether and how AI models improve performance during metadata creation, control, and enrichment.

Analyze risks, challenges, and potential biases

Size: ~16,000 Recordings (~1000 described)



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#### Metadata Enrichment Plan

Coverage_placename
Creator
Personality
Publisher
Contributor_organization
Contributor_person
Rights
Format_length
Program number
Associated Document
Language



#### Metadata Enrichment Plan

SpeakerID

LangID

Coverage\_placename Creator Personality Publisher Contributor\_organization Contributor\_person Rights Format\_length Program number Associated Document Language

Extractive Summarization

Transcript

ASR + Machine Translation

Title

Other\_lang\_title

Third\_lang\_title

Other\_lang\_description

Third\_lang\_description

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Description

File location

Source (script)

#### The story so far...

Initial focus group to examine ASR transcripts and Language ID Transcripts reasonable IF language ID predictions correct Started diplomatic analysis of transcripts (ongoing) Identify structure that can enable easy extraction of description Zero Shot Language ID & Speaker Embedding Experiments

#### Initial Language ID Lessons

Errors w/ certain language pairs

English predicted as being Welsh, Spanish predicted as Galician.

Ambiguity with Balkan languages. A11406.gl.txt – IDed as Galician 0.0-4.0: MÉXICO, PRÓXIMA 12.0-15.0: Aquí, la UNESCO en Paris. 15.0-20.0: Tenemos el gusto e el inmenso placer de recibir en nuestros estudios

A11903.gl.txt – IDed as Galician 0.0-3.0: Aquí, na Unesco, en París. 3.0-6.0: Temos o gozo de receber nos estudios 6.0-8.0: o professor Alain Jox

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#### **Diplomatics**

Consistent form across similar types of recordings.

Example: First 30 seconds of interviews almost always include who is being interviewed, and about what. 0.0-8.0: I have with me in the studio today Dr. George Stoddard, Dean of the School of Education of New York University.
8.0-21.0: He's one of the experts who've been convened to a UNESCO meeting to study the effect of mass media, that's films, press and radio, on juvenile delinquency. Dr. Stoddard.
22.0-36.0: Well, really the purpose of the meeting is somewhat broader than that. It's to study all the influences of the mass media, particularly the cinema and television, on the behavior of children.

0.0-7.0: Now here in the studio is Mr. Frederick Bellinger from the United States of America and for
8.18-15.18: the past year he's been in Egypt working for UNESCO Technical Assistance. Now I believe
15.22-22.06: Mr. Bellinger, you are a chemical engineer. What exactly were you doing in Egypt?
22.06-28.64: Under the UNESCO Technical Assistance Program I was asked to go to Egypt to assist the National
28.64-35.64: Research Council in definitely planning and starting a basic and applied research effort
36.24-39.76: in the industrial chemistry field.

#### **Diplomatics**

Consistent form across similar types of recordings.

Example: Last ~10 seconds "sign off" for reports and programs or thanking guests for interviews. 901.0-905.0: Sound of radio
910.0-916.0: Those noises from beyond the Earth mark the end of this, the fourth program in the series,
916.0-919.0: Signposts for the Atomic Age.
919.0-922.0: They are edited and introduced by Richie Calder.
922.0-927.0: The program was produced by Rex Keating in the studios of UNESCO, Paris.

210.26-217.3: This is Professor C. N. Vakil speaking from UNESCO headquarters in Paris and returning 217.3-220.18: you to UN Radio in New York.

285.84-289.04: That sounds very encouraging. Well thank you very much indeed Mr. Ballinger.

830.0-832.0: Thank you very much, Dr. Stallone

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#### Motivation

Right: Vittorino Veronese (UNESCO General Director 1958-1961)

Recordings in: French, Italian, Spanish, English, German

Photo Credit: https://unesdoc.unesco.org/ark:/48223/pf0000067034



### LangID & Weller Mb

How robust are Language ID models to L2 speakers?

Data: L2 English VoxPopuli [4] UNESCO Multilingual Speakers (L1 & L2)

Models: Whisper (large-v1,2,3) [3], MMS (l126) [2]

Benchmark off the shelf LID tools on L2 subsets



#### 

### Whisper v3 still best option



#### Speaker Processing in Brief

Embeddings represents the characteristics of a speaker in a fixed-size array

Speaker Embedding Extractor (i-,x-,r-vector etc.)



The same speaker *should* have similar embeddings

#### **Speaker Processing for Archives**

PLDA Classifier

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Challenges

Use Target

Speaker

Verification **Open SetSecurity** 

How

Speaker ID Closed Set Security

Speaker Diarization Open SetPre-Processing Transcripts

40307 Speaker Closed Set w/ Information Indexing For Archives "others" Retrieval

Clustering

**Direct Classification** 

ASR integrations [2] Cross talk [2], Domain [2]

Cluster / Direct / Domain Adaptation ? Multilinguality, Aging, Channel ?

Demographic [1], GenAl

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#### **Cross-lingual Embeddings**

How robust are embeddings for multilingual speakers?

Data: Single speaker recordings w/ multiple recordings in multiple languages Diarize and filter out recordings without a majority speaker (n\_speaker =75, n\_recordings= 363, n\_langs= 16)

Model: WeSpeaker Resnet34 [5]

Extract mean embeddings and compare monolingual vs. cross-lingual similarity

Monolingual: Mean: 0.71 Median: 0.76 Std: 0.19

Cross-lingual: Mean: 0.53 Median: 0.60 Std: 0.26



#### Cross-lingual Embedding Similarity

#### **Future Steps**

Diplomatics informed extractive summarization

Transcription and translation quality analysis

Domain adapted indexing of speakers

## Mahalo!

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