



Developing Speech Corpora for Low-Resource Languages

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The data diversity in machine learning

Developing diverse, well-annotated speech corpora is essential for training modern machine learning models.

- ✓ However, most progress in AI has focused on a few *high-resource languages*.



The goal of the presentation

This presentation discusses the principles and methodologies of creating large-scale speech corpora, focusing on the *Lithuanian language* as a case study.

I am currently involved in three projects that focus on developing and annotating corpora

Project "Creation of the Great Lithuanian Speech Corpus (LIEPA-3)",

Project Nr. 02-023-K-0001, 12.03.2024 - 30.04.2026.

✓ ***10 000 hours of annotated Lithuanian speech for speech recognition***

Project "Creation of a Lithuanian Speech Corpus (for speech synthesis purposes to generate neural voices)", No. (1.57) 15600 INS 70, 16.10.2024 - 30.04.2026.

✓ ***200 hours of annotated Lithuanian speech for synthesis + 300 hours of annotated Lithuanian emotional speech***

Project "Research on propaganda and disinformation: automatic recognition by machine learning, impact and societal resilience", No. S-VIS-23-8, 2023-09-01-2026-06-30

✓ ***1 000 social network news articles, annotated for propaganda techniques and narratives***



Project LIEPA

“Services managed by Lithuanian language”

LIEPA *2013.02 – 2015.08*

LIEPA2 *2017.12 – 2020.12*

Key Outcomes:

- **Lithuanian Speech Recognition System:** A functional speech recognition system for Lithuanian.
- **Lithuanian Text-to-Speech System:** TTS system for converting Lithuanian text into speech.
- **Speech Corpora:** Annotated data resources for Lithuanian speech research and development.

100 hours of speech data **1000 hours** of speech data were collected. The corpus was designed to cover a wide range of phonetic variations in Lithuanian language, including all vowels, consonants, and diphthongs.



Project LIEPA-3

„Creation of the Great Lithuanian Language Corpus“

12 March 2024 to 30 April 2026

The project aims to develop an annotated Lithuanian language corpus of at least **10000 hours** for speech recognition, AI, and innovative technologies development



Vilniaus
universitetas



VYTAUTO
DIDŽIOJO
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MCMXXII



Recording Specifications

Recordings must be stored at: **44 kHz** sampling rate and **16-bit** mono format.

Originality and uniqueness

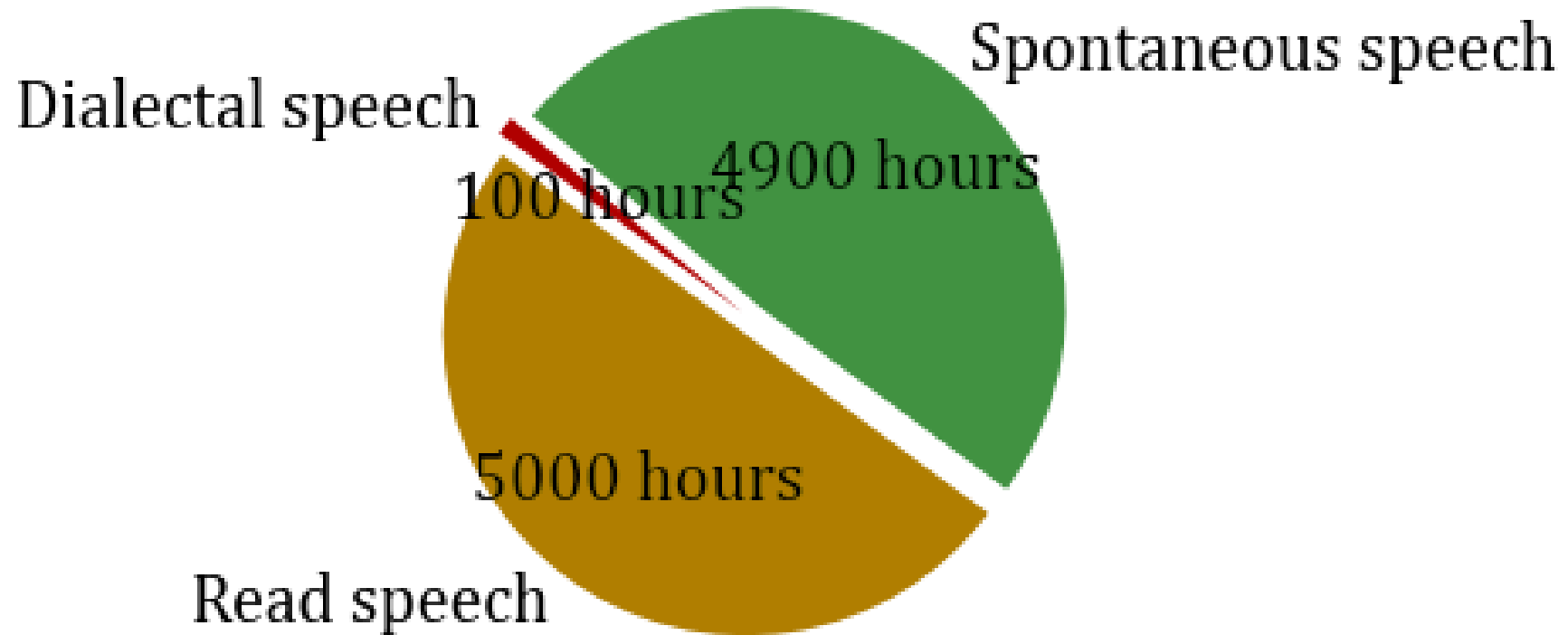
- ✓ The corpus must be a new linguistic resource.
- ✓ It cannot reuse or embed any previously created Lithuanian corpora.
- ✓ No duplicate recordings are allowed.

Dialectal speech

- ✓ The corpus must include dialectal speech samples representing the following dialects:
 - *Žemaitian*
 - *Dzūkian*
 - *Suvalkian*
 - *Aukštaitian*

Minimum: 100 hours in total.

Stylistic diversity



Recording Methodology in the LIEPA-3 Project

Read speech recordings come from two sources:

- ✓ direct in-person sessions in Lithuanian cities, and
- ✓ the <https://kurkgarsyna.lt/> online platform, where volunteers record speech remotely.

Spontaneous speech is collected from existing radio broadcasts under official collaboration agreements.

Dialectal speech samples are collected through in-person recordings.

Key features of corpus

- ✓ The corpus consists of Lithuanian speech recordings (***non-Lithuanian language content limited to 0.1%***).
- ✓ ***High-quality, noise-varied*** recordings.

Non-standard language

The corpus must include speakers using ***non-standard Lithuanian***, such as slang, or obscene language.

✓ Minimum: *100 hours of such recordings.*

Corpus composition must reflect

- ✓ Gender, age, and dialect region of speakers.
- ✓ ***Phonetic diversity***: balanced coverage of Lithuanian sounds and combinations.
- ✓ ***Topical diversity***: speech across various themes.

Age groups used in our corpus

Gender	Age groups
F (Female)	1: 0-12 years old
M (Male)	2: 13-17 years old
	3: 18-60 years old
	4: 60+ years old

Recording conditions

Must include recordings captured:

- ✓ With ***professional*** and ***consumer*** recording ***equipment***.
- ✓ In different ***acoustic environments***.
- ✓ Under various ***noise conditions***.

Annotation requirements

All recordings in the corpus must be annotated at

✓ ***The utterance level:***

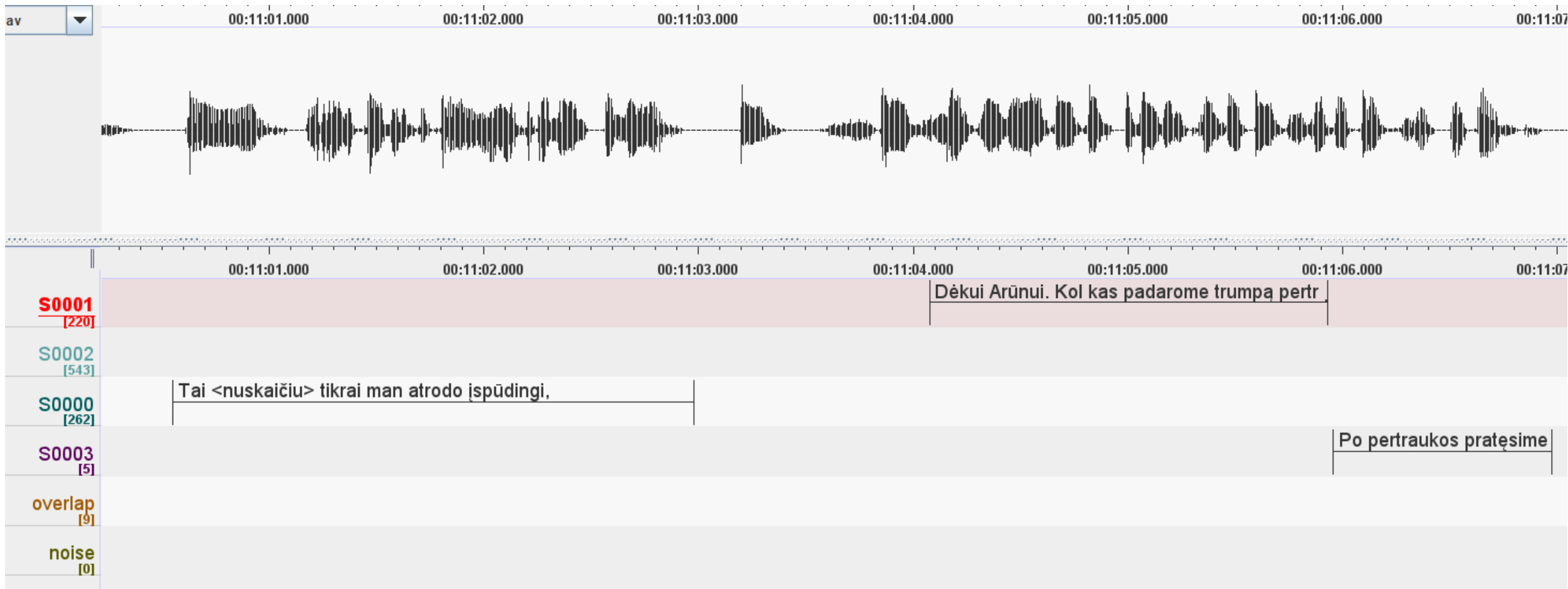
- The text of the spoken phrase.
- The start and end timestamps of that phrase.

At least 500 hours of recordings must be annotated at

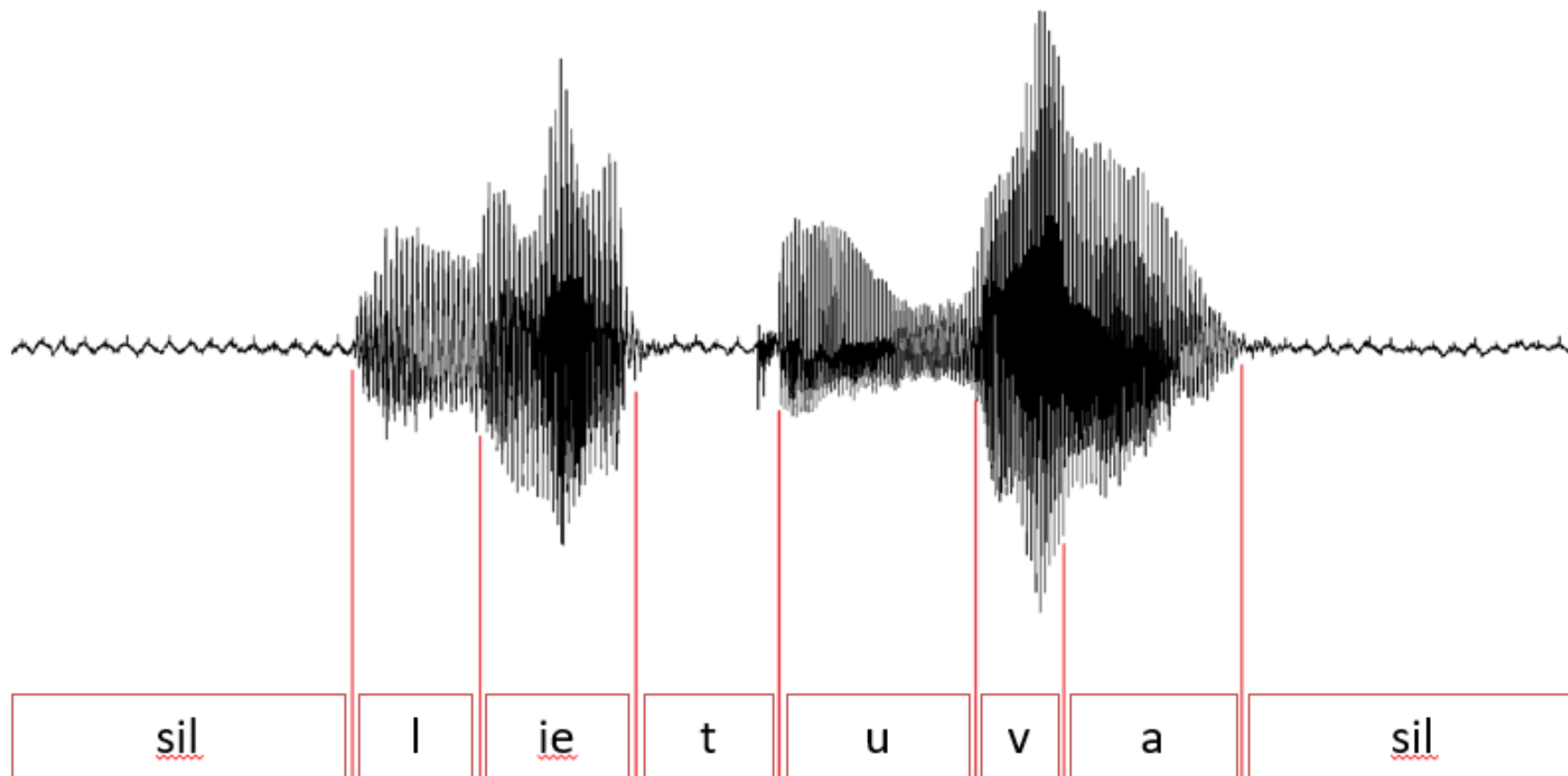
✓ ***The lexical unit level*** (individual words).

✓ ***The phoneme level*** (individual speech sounds).

The utterance level annotation



The phoneme level annotation



Corpus validation requirements

Purpose of Validation:

- ✓ To ensure that the ***annotations accurately match the audio recordings.***
- ✓ To demonstrate that the corpus ***is reliable and suitable for developing speech recognition technologies***

Annotation validation process

A formal annotation validation procedure must be performed.

- ✓ The level of mismatch between audio and annotation must not exceed: ***0.1% at the utterance level***, and ***0.05% at the lexical level***.

Demonstration of usability

Use 1% of the total corpus (minimum 100 hours) to build a demonstration speech recognition system.

✓ The system's **Word Error Rate** (WER) must be $\leq 20\%$.

$$\text{WER} = \frac{S + D + I}{N} \times 100\%$$

S – number of substitutions (incorrect words),

D – number of deletions (missing words),

I – number of insertions (extra words not in the reference)

N – total number of words in the reference (ground truth)

Open Access and Distribution

The corpus must be openly accessible under a Creative Commons license.

Must be published on at least two ***open-access platforms***, such as:

- Hugging Face
- CLARIN
- Lithuanian Open Data Portal

Access must be ***free of charge***.



Initial experiments with Liepa-3 dataset

What is Whisper?

An open-source neural model for Automatic Speech Recognition.

- ✓ Developed by OpenAI.
- ✓ Trained on $\approx$ **680,000 hours** of **multilingual** and **multitask** data.
- ✓ Based on the Transformer architecture.
- ✓ Freely available through Hugging Face and GitHub.

Multitask training data (680k hours)

English transcription



"Ask not what your country can do for ..."



Ask not what your country can do for ...

Any-to-English speech translation



"El rápido zorro marrón salta sobre ..."



The quick brown fox jumps over ...

Non-English transcription



"언덕 위에 올라 내려다보면 너무나 넓고 넓은 ..."



언덕 위에 올라 내려다보면 너무나 넓고 넓은 ...

No speech



(background music playing)



∅

A model was trained on:

- ✓ multilingual speech recognition
- ✓ speech translation
- ✓ spoken language identification
- ✓ voice activity detection

Radford, A., Kim, J. W., Xu, T., Brockman, G., McLeavey, C., & Sutskever, I. (2023, July). Robust speech recognition via large-scale weak supervision. In *Proceedings of the 40th International Conference on Machine Learning* (pp. 28492-28518).

How the dataset used for training is structured?

Speech and text in English	438,218 h	65 %
Speech and text in 96 non-English languages	117,113 h	17 %
Non-English speech paired with English transcripts	125,739 h	18 %
Non-speech training segments used for VAD	<i>Not quantified in the dataset statistics</i>	

Language	Task	Hours
Lithuanian	Multilingual ASR	67 h
Lithuanian	Translation (LT→EN)	99 h
Polish	Multilingual ASR	4,278 h
Polish	Translation (PL→EN)	2,200 h

All data came from large, noisy web sources.

Whisper ASR performance (Common Voice 9 dataset)

Model	WER (Lithuanian)	WER (Polish)
Tiny	103.5 %	45.3 %
Base	91.3 %	32.8 %
Small	72.8 %	16.9 %
Medium	49.4 %	10.1 %
Large V2	43.9 %	9.0 %
Whisper large-v2	35.2 %	7.6 %

Radford, A., Kim, J. W., Xu, T., Brockman, G., McLeavey, C., & Sutskever, I. (2023, July). Robust speech recognition via large-scale weak supervision. In *Proceedings of the 40th International Conference on Machine Learning* (pp. 28492-28518).

Fine-Tuning Whisper for Lithuanian ASR

Corpus	Hours
Liepa3	2 098 h
Liepa2	948 h
Total training	3 046 h (692,000 segments, avg 16 s)
Test set	20.2 h (42,000 files, segments, avg 16 s)

Training Configuration

	Whisper Small	WhisperMedium
GPU	RTX 4060 M (8 GB)	RTX A5000 (24 GB)
per-device batch size	8	8
grad accumulation	2	2
Effective batch	16	16
Optimizer	AdamW	AdamW
Epochs	4	4

Results (WER %)

Model	WER (%)
Whisper Small FT	11.98 %
Whisper Medium FT	9.89 %



***Thank you for
your attention!***

***I'll be happy to
answer any
questions***