

Speech Recognition with Python

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whoami



mpfmorawski/README.md

Welcome!

- My name is Maciej Morawski! 🙋
- I'm a Python Backend Developer Intern @ Schneider Electric, living in Warsaw, Poland
- I have graduated from the MSE in Robotics and Automatic Control @ WUT but...
- That's not the end of my university education! I'm still a Computer Science Student @ WUT!
- l'm interested in Robotics, Al and... all those amazing things you can do using Python (that's why I'm here with you today (3))
- Today is my first time speaking at such a large conference, but I hope to speak many more times!

Agenda

- 1. Speech-to-text intro
- 2. SpeechRecognition
- 3. AssemblyAl
- 4. OpenAl's Whisper
- 5. Transformers
- 6. Summary

Slides

Slides can be found here

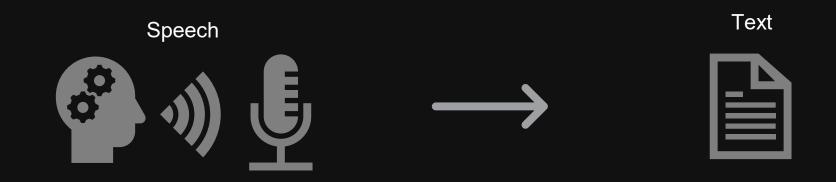




https://bit.ly/pyconpl-asr-slides



What is... speech recognition?



Speech Recognition



Automatic Speech Recognition (ASR)

Speech-To-Text (STT)

Today we will not focus on theory. We will answer the question...

What solutions can we use?

If your first thought was to make your own speech recognition engine

putting it simply...



Data Requirements

Expertise

Why it's bad idea to build your own speech recognition engine?

Time and Cost

Accuracy and Performance

Maintenance and Updates

And because...

there are plenty of solutions available on the market!

Speech recognition engines/APIs can be divided by their availability into two main groups:

tools that can only be used online



tools that can be used offline



But... speech recognition engines/APIs can also be divided by **service providers** into two main groups:

external solutions



internal solutions



Hint \mathbb{Q} : Which is much more important from the company's point of view \mathfrak{S}



External solutions



Advantages (+):

- Easy to use
- Scalability
- No need for infrastructure and maintenance
- Cost (when processing a small amount of audio data)

Disadvantages (-):

- Privacy concerns
- Dependence on internet connection
- Cost (when processing a large amount of audio data)

Internal solutions



Advantages (+):

- Data security
- Cost (when processing a large amount of audio data)

Disadvantages (-):

- Limited scalability
- Maintenance
- Cost (when processing a small amount of audio data)

What options are available to us as Python Developers 2?

Let's go straight to practice together!

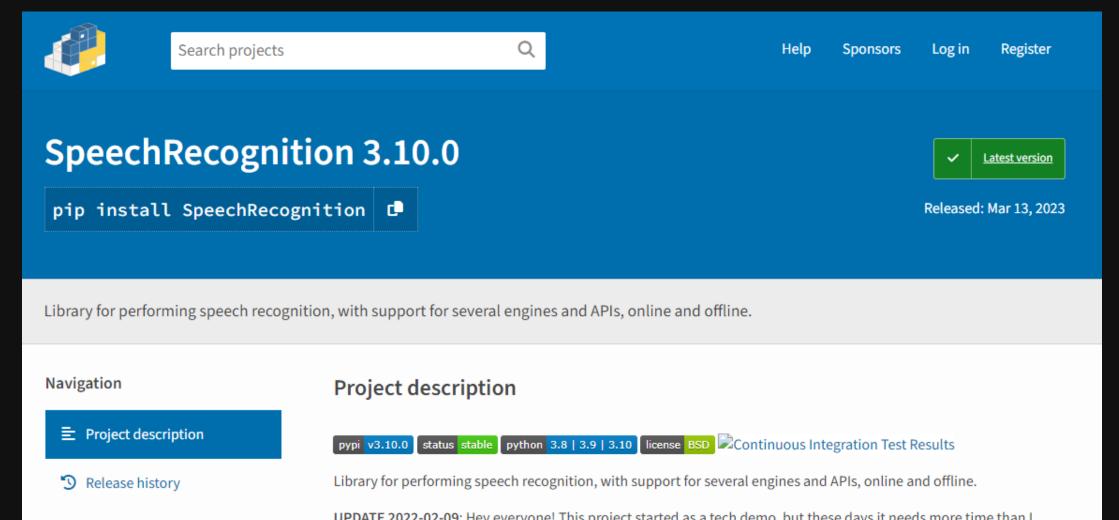




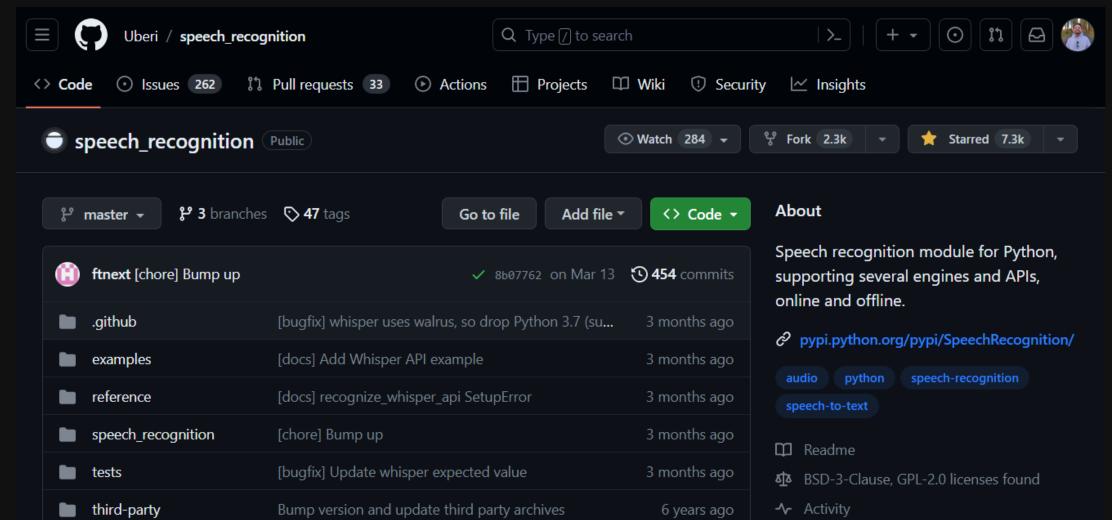
https://bit.ly/pyconpl-asr-notebook



Source: https://pypi.org/project/SpeechRecognition/



Source: https://github.com/Uberi/speech_recognition



Speech recognition engine/API support:

- CMU Sphinx (works offline)
- Google Speech Recognition
- Google Cloud Speech API
- Wit.ai
- <u>Microsoft Azure Speech</u>
- Houndify API
- IBM Speech to Text
- <u>Snowboy Hotword Detection</u> (works offline)
- Tensorflow
- Vosk API (works offline)
- OpenAl whisper (works offline)
- Whisper API



\$☆ added in the last release in March \$☆

pip install SpeechRecognition

```
import speech recognition as sr
r = sr.Recognizer()
  with sr.AudioFile (audio path) as source:
    audio = r.record(source)
r.recognize google (audio)
```

Google Speech Recognition (using the default API key)

```
import speech_recognition as sr

r = sr.Recognizer()
  with sr.AudioFile(audio_path) as source:
    audio = r.record(source)

try:
  print( r.recognize_google(audio) )
  except sr.UnknownValueError:
  print( "Could not understand audio" )
  except sr.RequestError as e:
  print( f"Could not request results service; {e}" )
```

Google Cloud Speech

```
import speech recognition as sr
r = sr.Recognizer()
 with sr.AudioFile(audio path) as source:
    audio = r.record(source)
GOOGLE CLOUD SPEECH CREDENTIALS = "CREDENTIALS"
try:
 print( r.recognize google cloud(audio, credentials json=GOOGLE CLOUD SPEECH CREDENTIALS) )
except sr.UnknownValueError:
 print( "Could not understand audio" )
except sr.RequestError as e:
 print( f"Could not request results from service; {e}" )
```

Wit.ai

```
import speech recognition as sr
r = sr.Recognizer()
 with sr.AudioFile(audio path) as source:
    audio = r.record(source)
WIT AI KEY = "KEY"
try:
 print( r.recognize wit(audio, key=WIT AI KEY) )
except sr.UnknownValueError:
 print( "Could not understand audio" )
except sr.RequestError as e:
 print( f"Could not request results from service; {e}" )
```

Houndify

```
import speech recognition as sr
r = sr.Recognizer()
 with sr.AudioFile(audio path) as source:
    audio = r.record(source)
HOUNDIFY CLIENT ID = "ID"
HOUNDIFY CLIENT KEY = "KEY"
try:
 print( r.recognize houndify(audio,
                              client id=HOUNDIFY CLIENT ID,
                              client key=HOUNDIFY CLIENT KEY)
except sr.UnknownValueError:
 print( "Could not understand audio" )
except sr.RequestError as e:
 print( f"Could not request results from service; {e}" )
```

IBM Speech to Text

```
import speech recognition as sr
r = sr.Recognizer()
 with sr.AudioFile(audio path) as source:
    audio = r.record(source)
IBM USERNAME = "USERNAME"
IBM PASSWORD = "PASSWORD"
try:
 print( r.recognize ibm(audio,
                         username=IBM USERNAME,
                         password=IBM PASSWORD) )
except sr.UnknownValueError:
 print( "Could not understand audio" )
except sr.RequestError as e:
 print( f"Could not request results from service; {e}" )
```

Whisper API



→ ♠ added in the last release in March ♠

```
import speech recognition as sr
r = sr.Recognizer()
 with sr.AudioFile(audio path) as source:
    audio = r.record(source)
OPENAI API KEY = "KEY"
try:
 print( r.recognize whisper api(audio, api key=OPENAI API KEY) )
except sr.UnknownValueError:
 print( "Could not understand audio" )
except sr.RequestError as e:
 print( f"Could not request results from service; {e}" )
```

Let's get back to Google Colab!





AssemblyAl

Source: https://www.assemblyai.com/



Why Assembly Al Models Use Cases Pricing Developers Playground

Your dashboard >

NEW • Introducing LeMUR, our new framework for applying powerful LLMs to transcribed speech >

Access powerful Al models to transcribe and understand speech

Our simple API exposes AI models for speech recognition, speaker detection, speech summarization, and more. We build on the latest state-of-the-art AI research to offer production-ready, scalable, and secure AI models through a simple API. Used by thousands of breakthrough startups and dozens of global enterprises for mission-critical workloads.

Try the API >

Contact sales >

Source: https://www.assemblyai.com/

```
Try the API >
                                               Contact sales >
           тs TypeScript
Python
                           ● PHP 🔏 Ruby
    import requests
    endpoint = "https://api.assemblyai.com/v2/transcript"
    json = {
      "audio_url": "https://storage.googleapis.com/bucket/b2c31290d9d8.wav"
    headers = {
      "Authorization": "c2a41970d9d811ec9d640242ac12",
11
      "Content-Type": "application/json"
12
13
    response = requests.post(endpoint, json=json, headers=headers)
    AssemblyAI nse)
```

import requests

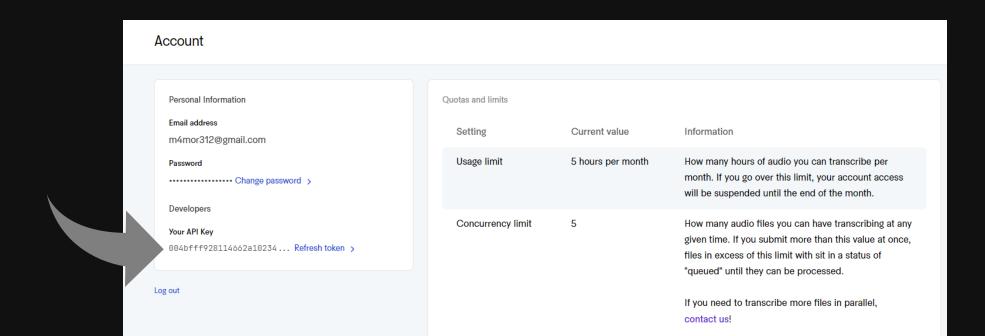
```
endpoint = "https://api.assemblyai.com/v2/transcript"
json = {
  "audio url": "https://storage.googleapis.com/bucket/b2c31290d9d8.wav"
headers = {
  "Authorization": "c2a41970d9d811ec9d640242ac12",
  "Content-Type": "application/json"
response = requests.post(endpoint, json=json, headers=headers)
parse (response)
```

Output

```
endpoint = "https://api.assemblyai.com/v2/transcript"
json =
  "audio url": "https://storage.googleapis.com/bucket/b2c31290d9d8.wav"
headers =
  "Authorization": "c2a41970d9d811ec9d640242ac12",
  "Content-Type": "application/json"
response = requests.post(endpoint, json=json, headers=headers)
parse (response)
```

Authorization

- 1. Go to: https://www.assemblyai.com/dashboard/signup
- 2. Sign up
- 3. Go to: https://www.assemblyai.com/app/account
- 4. Copy your API Key



```
ASSEMBLY_AI_API_KEY = "your_api_key"
```

Uploading audio

Output

{'upload_url': 'https://cdn.assemblyai.com/upload/random-letters-and-numbers'}

Making request for transcription

```
TRANSCRIPT ENDPOINT = "https://api.assemblyai.com/v2/transcript"
json = {
  "audio url": response.json()["upload url"]
headers = {
  "authorization": ASSEMBLY AI API KEY
response = requests.post(TRANSCRIPT ENDPOINT,
                         json=json,
                         headers=headers)
print(response.json())
```

Output

```
json full of data!
```

But if we take a closer look...

```
response.json()['text']=None
response.json()['status']='queued'
```

Polling

```
polling_endpoint = f"{TRANSCRIPT_ENDPOINT}/{response.json()['id']}"
while True:
  response = requests.get(polling endpoint, headers=headers).json()
  if response['status'] == 'completed':
    break
  elif response['status'] == 'error':
    raise RuntimeError(f"Transcription failed: {response['error']}")
  else:
    time.sleep(3)
print(response["text"])
```

Output

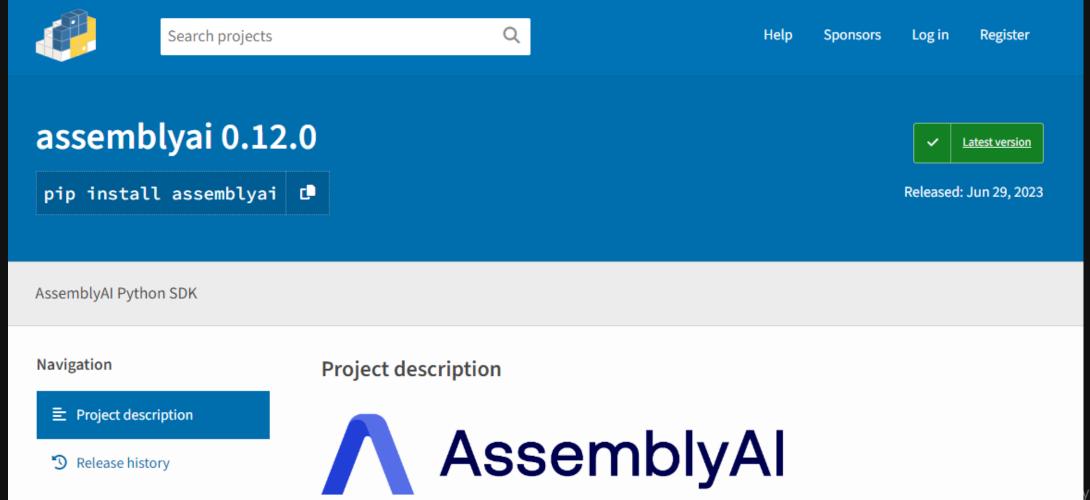
Our transcription!

Let's get back to Google Colab!



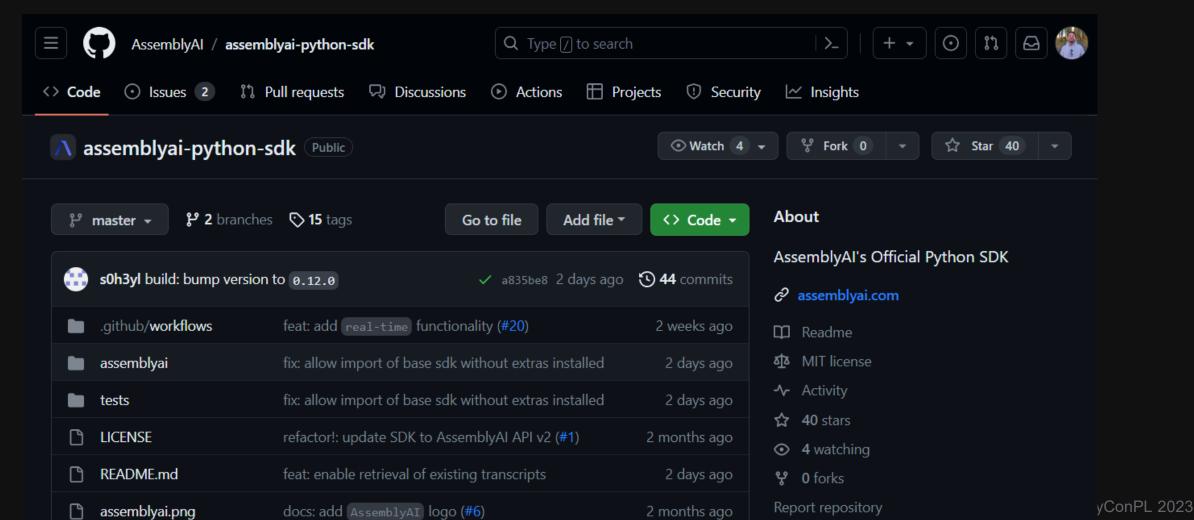
On the other hand, we can use...

Source: https://pypi.org/project/assemblyai/



ConPL 2023

Source: https://github.com/AssemblyAI/assemblyai-python-sdk

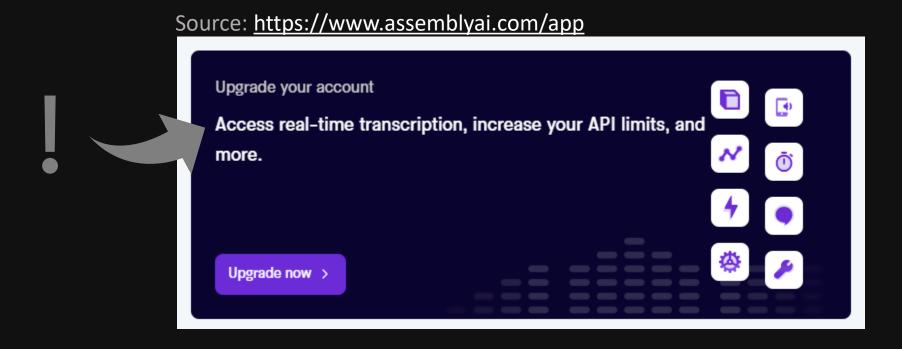


pip install assemblyai

```
import assemblyai as aai
aai.settings.api_key = "YOUR API KEY"

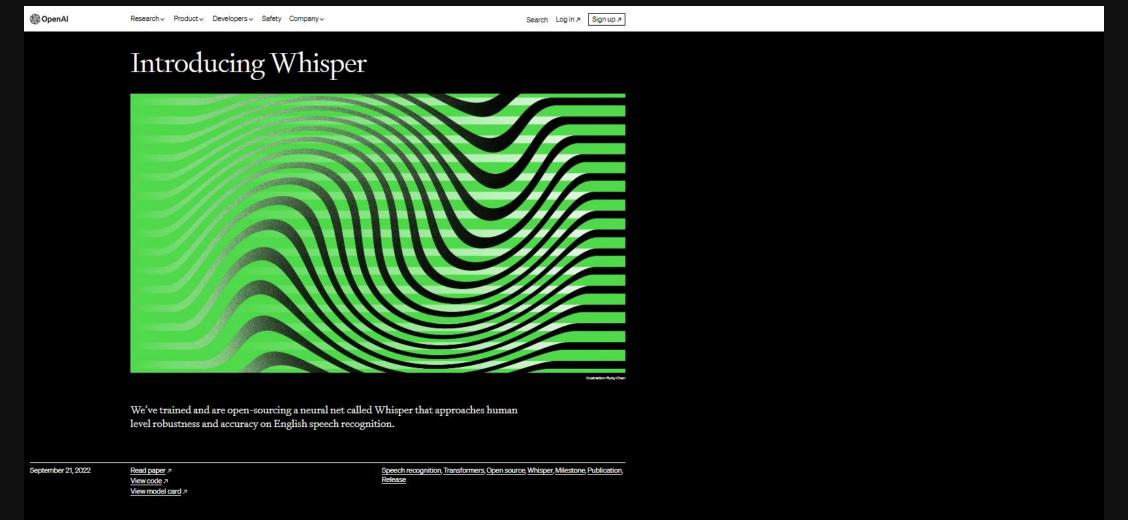
transcriber = aai.Transcriber()
transcript = transcriber.transcribe(audio_path)

print( transcript.text )
```

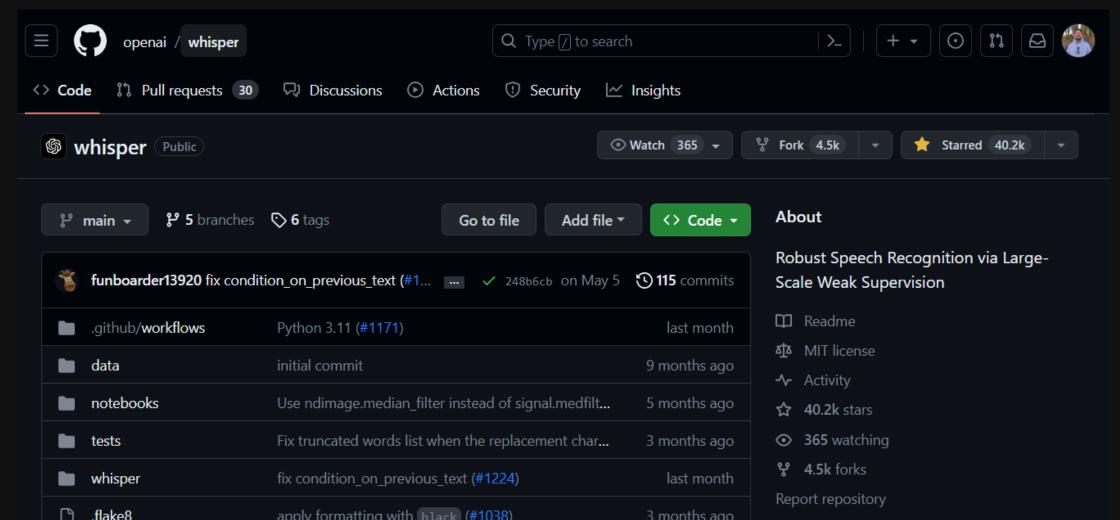




Source: https://openai.com/research/whisper



Source: https://openai.com/research/whisper



pip install -U openai-whisper

```
import whisper

model = whisper.load_model(model_name)

transcription = model.transcribe(audio_path)
print(transcription["text"])
```

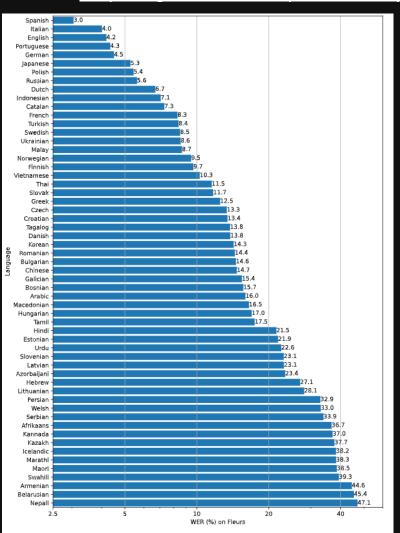
Source: https://github.com/openai/whisper

Available models and languages

There are five model sizes, four with English-only versions, offering speed and accuracy tradeoffs. Below are the names of the available models and their approximate memory requirements and relative speed.

Size	Parameters	English-only model	Multilingual model	Required VRAM	Relative speed
tiny	39 M	tiny.en	tiny	~1 GB	~32x
base	74 M	base.en	base	~1 GB	~16x
small	244 M	small.en	small	~2 GB	~6x
medium	769 M	medium.en	medium	~5 GB	~2x
large	1550 M	N/A	large	~10 GB	1x



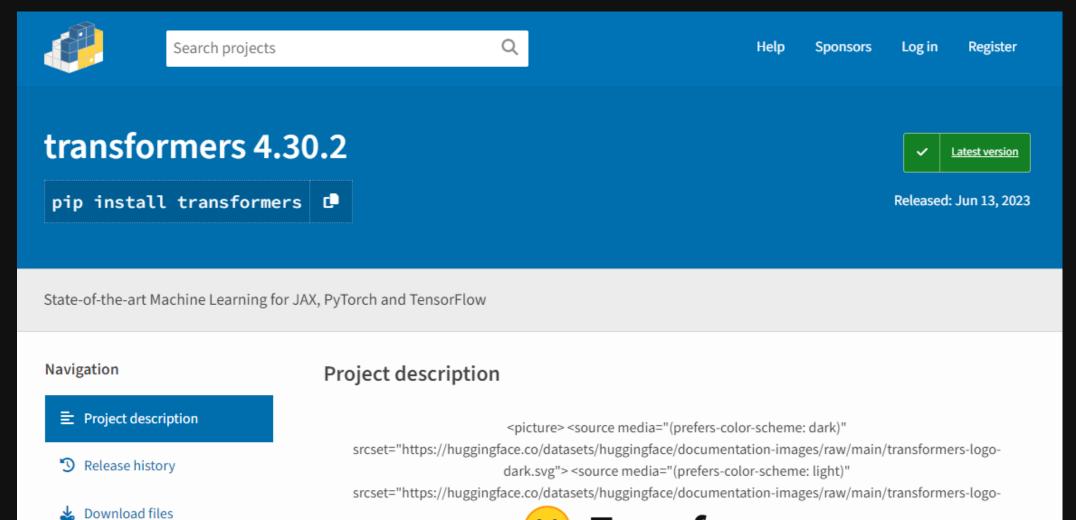


Let's get back to Google Colab!

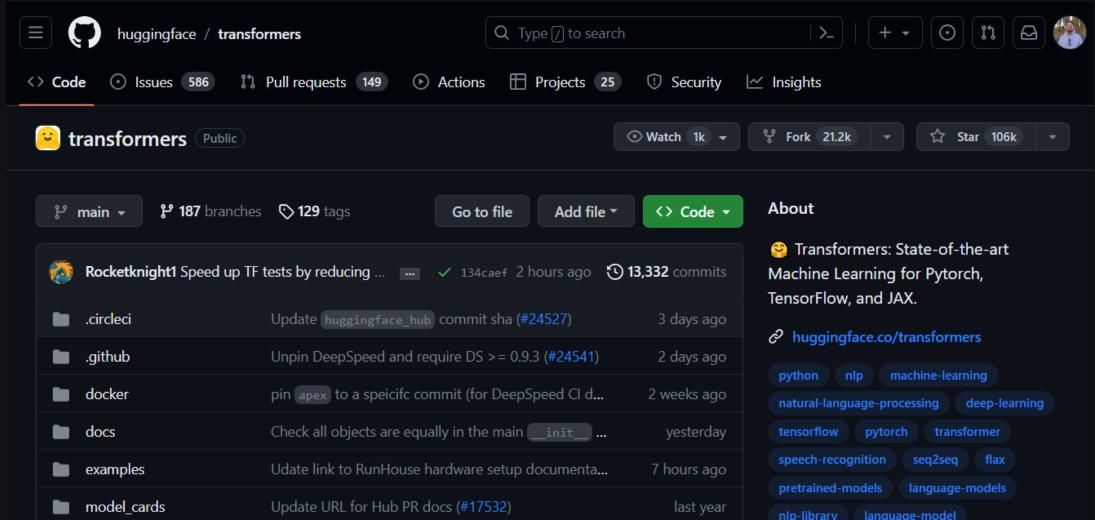




Source: https://pypi.org/project/transformers/



Source: https://github.com/huggingface/transformers

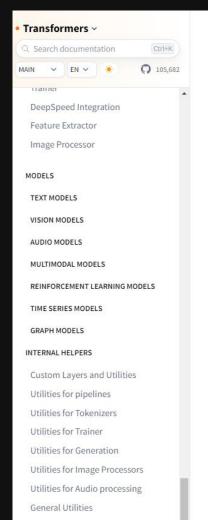


pip install transformers

Let's get back to Google Colab!



Source: https://huggingface.co/docs/transformers/main/tasks/asr



Automatic speech recognition



Open in Colab Dopen Studio Lab

Automatic speech recognition

Load MInDS-14 dataset

Preprocess

Evaluate Train

Inference

Automatic speech recognition (ASR) converts a speech signal to text, mapping a sequence of audio inputs to text outputs. Virtual assistants like Siri and Alexa use ASR models to help users everyday, and there are many other useful user-facing applications like live captioning and note-taking during meetings.

This guide will show you how to:

- Finetune <u>Wav2Vec2</u> on the <u>MInDS-14</u> dataset to transcribe audio to text.
- 2. Use your finetuned model for inference.

The task illustrated in this tutorial is supported by the following model architectures: Data2VecAudio, Hubert, M-CTC-T, SEW, SEW-D, UniSpeech, UniSpeechSat, Wav2Vec2, Wav2Vec2-Conformer, WavLM

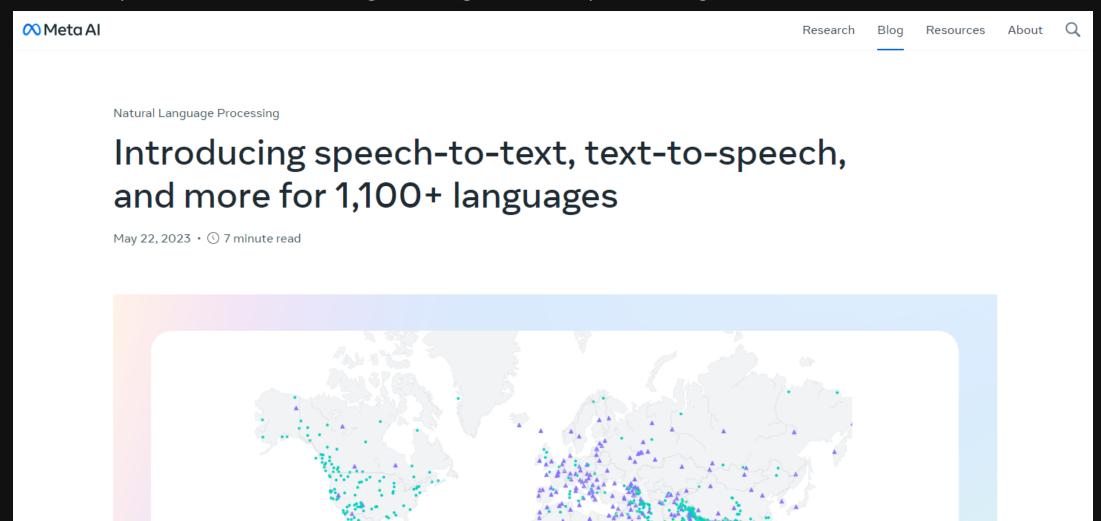
Source: https://huggingface.co/docs/transformers/main/tasks/asr

The task illustrated in this tutorial is supported by the following model architectures:

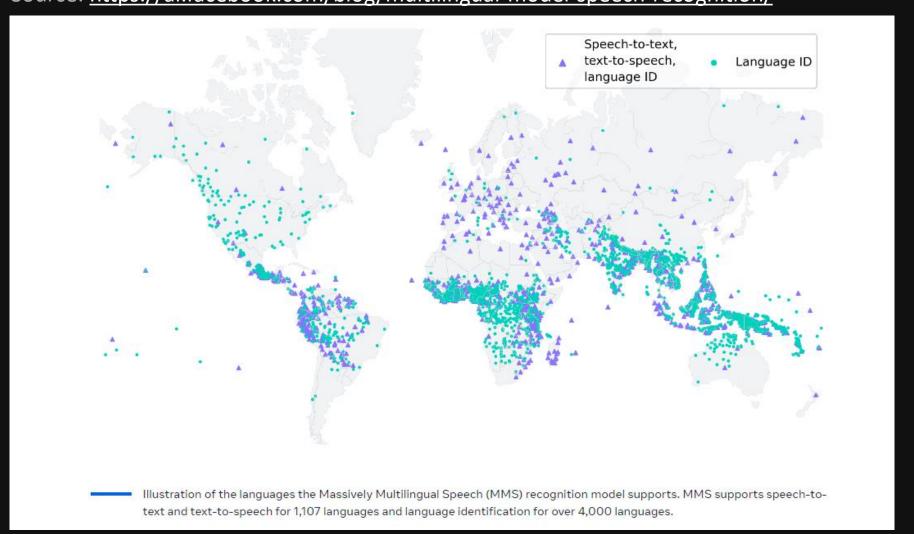
Data2VecAudio, Hubert, M-CTC-T, SEW, SEW-D, UniSpeech, UniSpeechSat, Wav2Vec2, Wav2Vec2-Conformer, WavLM



Source: https://ai.facebook.com/blog/multilingual-model-speech-recognition/



Source: https://ai.facebook.com/blog/multilingual-model-speech-recognition/



Source: https://ai.facebook.com/blog/multilingual-model-speech-recognition/

We trained multilingual speech recognition models on over 1,100 languages using a 1B parameter wav2vec 2.0 model. As the number of languages increases, performance does decrease, but only very slightly: Moving from 61 to 1,107 languages increases the character error rate by only about 0.4 percent but increases the language coverage by over 18 times.

Source: https://github.com/facebookresearch/fairseq/tree/main/examples/mms

ource. https://github.com/facebook/escarch/fairseq/tree/mair/examples/minis





MMS: Scaling Speech Technology to 1000+ languages

The Massively Multilingual Speech (MMS) project expands speech technology from about 100 languages to over 1,000 by building a single multilingual speech recognition model supporting over 1,100 languages (more than 10 times as many as before), language identification models able to identify over 4,000 languages (40 times more than before), pretrained models supporting over 1,400 languages, and text-to-speech models for over 1,100 languages. Our goal is to make it easier for people to access information and to use devices in their preferred language.

You can find details in the paper Scaling Speech Technology to 1000+ languages and the blog post.

An overview of the languages covered by MMS can be found here.

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README.md

Transformers

MMS has been added to Transformers. For more information, please refer to Transformers' MMS docs.

Click here to find all MMS checkpoints on the Hub.

Checkout the demo here Open in HF Spaces

Source: https://huggingface.co/docs/transformers/main/en/model-doc/mms

Automatic Speech Recognition (ASR)

The ASR model checkpoints can be found here: mms-1b-fl102, mms-1b-l1107, mms-1b-all. For best accuracy, use the mms-1b-all model.

Tips:

- All ASR models accept a float array corresponding to the raw waveform of the speech signal. The raw waveform should be pre-processed with <u>Wav2Vec2FeatureExtractor</u>.
- The models were trained using connectionist temporal classification (CTC) so the model output has to be decoded using Wav2Vec2CTCTokenizer.
- You can load different language adapter weights for different languages via <u>load_adapter()</u>. Language adapters only consists of roughly 2 million parameters and can therefore be efficiently loaded on the fly when needed.



Summary

Summary

1. Check out the different tools and choose the one that best suits your needs!

 After this lecture, you can start using speech recognition in your projects right away - just use the examples presented today.

Thank you! Feel free to ask questions now or contact me later!

How to contact me?

You can contact me via LinkedIn: https://www.linkedin.com/in/mpfmorawski/

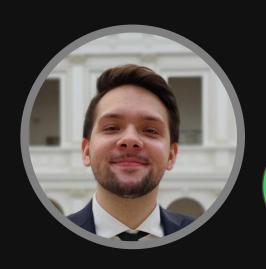




Or check out more on my GitHub: https://github.com/mpfmorawski







Morawski Maciej @mpfmorawski

Hope to see you soon!