

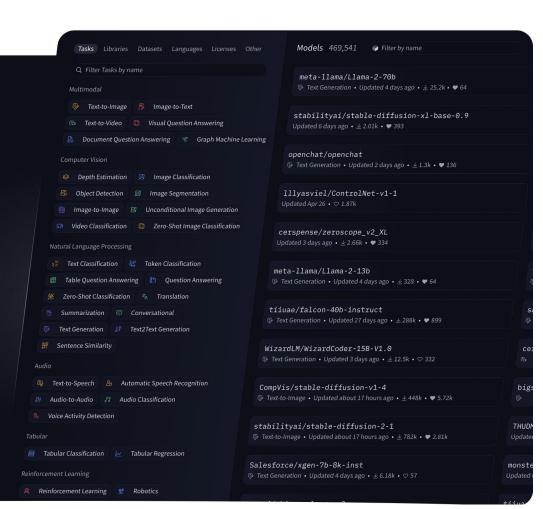
# The AI community building the future





# The AI community building the future.

The platform where the machine learning community collaborates on models, datasets, and applications.



## Hugging Face

Transformers library

#### Use it with



Use it for

Applying state of the art transformer models





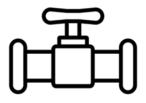


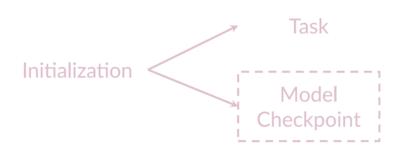


Fine-tuning pretrained transformer models

### Tasks

**Pipelines** 





Use Inputs for the task

**Sentiment Analysis** 

Sequence

**Question Answering** 

Context and questions

Fill-Mask

Sentence and position

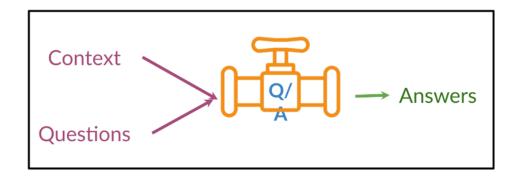
## Hugging Face: Using Transformers



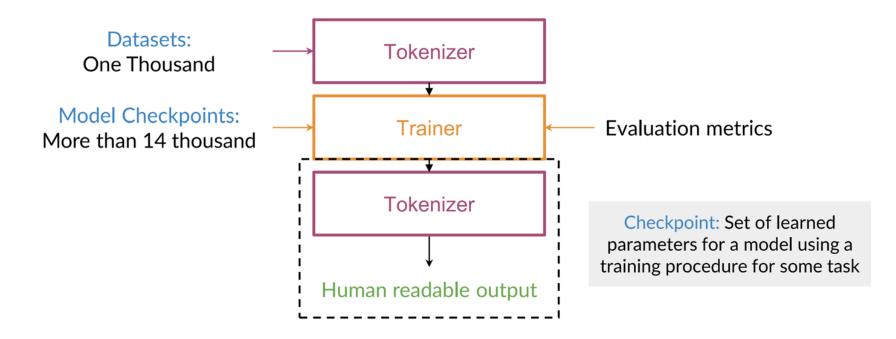


**Pipelines** 

- 1. **Pre-processing** your inputs
- 2. Running the model
- 3. **Post-processing** the outputs



## Hugging Face: Fine-Tuning Transformers



## Checkpoints



Huge number of model checkpoints that you can use in your pipelines.

But **beware**, not every checkpoint would be suitable for your task.

## Model Checkpoints

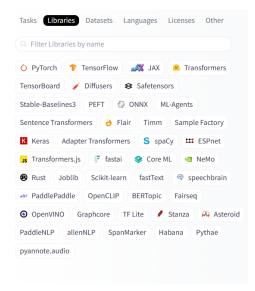
#### Model Checkpoints:

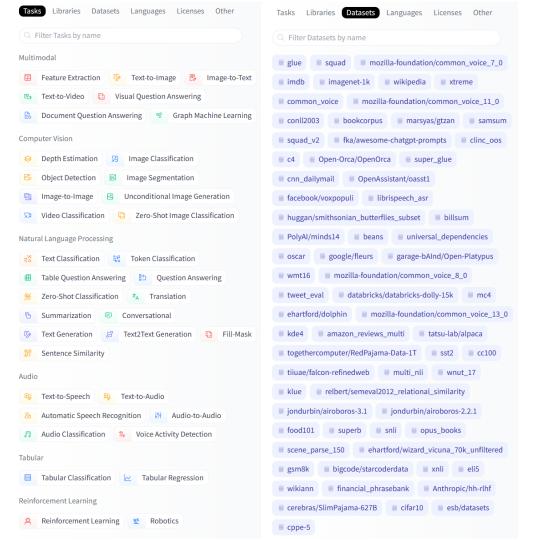
More than 15 thousand (and increasing)

Upload the architecture and weights with 1 line of code!

Model	Dataset	Name in 😜
DistilBERT	Stanford Question Answering Dataset (SQuAD)	distilbert-base-cased- distilled-squad
BERT	Wikipedia and Book Corpus	bert-base-cased
	···	<b></b>

#### Model Card and Datasets

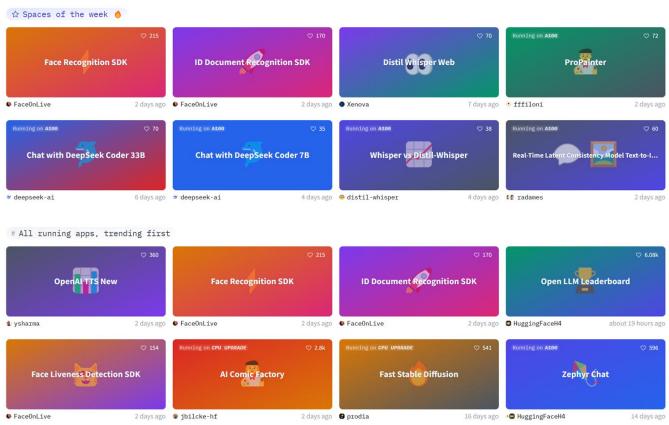




#### **Hugging Spaces**

Discover amazing ML apps made by the community!

**Create new Space** 



#### **Hugging Spaces APIs**

#### Pipelines for inference

The <u>pipeline()</u> makes it simple to use any model from the <u>Hub</u> for inference on any language, computer vision, speech, and multimodal tasks. Even if you don't have experience with a specific modality or aren't familiar with the underlying code behind the models, you can still use them for inference with the <u>pipeline()</u>! This tutorial will teach you to:

- Use a <u>pipeline()</u> for inference.
- Use a specific tokenizer or model.
- Use a <u>pipeline()</u> for audio, vision, and multimodal tasks.

```
from transformers import pipeline
transcriber = pipeline(task="automatic-speech-recognition")
transcriber("https://huggingface.co/datasets/Narsil/asr_dummy/resolve/main/mlk.flac")
Output:
{'text': 'I HAVE A DREAM BUT ONE DAY THIS NATION WILL RISE UP LIVE UP THE TRUE MEANING OF ITS TREES'}
```

#### Load pretrained instances with an AutoClass

With so many different Transformer architectures, it can be challenging to create one for your checkpoint. As a part of Transformers core philosophy to make the library easy, simple and flexible to use, an AutoClass automatically infers and loads the correct architecture from a given checkpoint. The from\_pretrained() method lets you quickly load a pretrained model for any architecture so you don't have to devote time and resources to train a model from scratch. Producing this type of checkpoint-agnostic code means if your code works for one checkpoint, it will work with another checkpoint - as long as it was trained for a similar task - even if the architecture is different.

Load pretrained instances with an Auto Class

AutoTokenizer

AutoImageProcessor

AutoFeatureExtractor

AutoProcessor

AutoModel

```
from transformers import AutoImageProcessor
image processor = AutoImageProcessor.from pretrained("google/vit-base-patch16-224")
 from transformers import AutoFeatureExtractor
 feature extractor = AutoFeatureExtractor.from pretrained(
 ... "ehcalabres/wav2vec2-lg-xlsr-en-speech-emotion-recognition")
from transformers import AutoProcessor
processor = AutoProcessor.from pretrained("microsoft/layoutlmv2-base-uncased")
from transformers import AutoModelForSequenceClassification
model = AutoModelForSequenceClassification.from pretrained("distilbert-base-uncased")
from transformers import AutoModelForTokenClassification
model = AutoModelForTokenClassification.from pretrained("distilbert-base-uncased")
```