

## DEPARTMENT OF COMPUTER SCIENCE & ENGINEERING



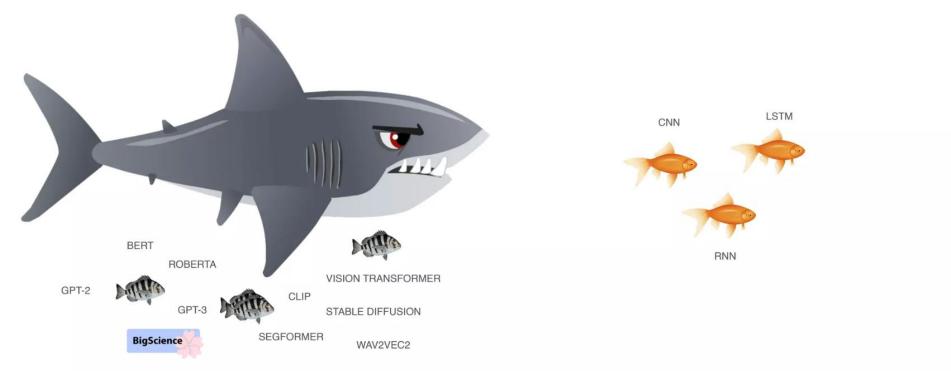
# Hugging Face Transformers Inference API

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## Transformers Are Eating Deep Learning





"Transformers are emerging as a general-purpose architecture for ML" <a href="https://www.stateof.ai/">https://www.stateof.ai/</a>

RNN and CNN usage down, Transformers usage up! https://www.kaggle.com/kaggle-survey-2021

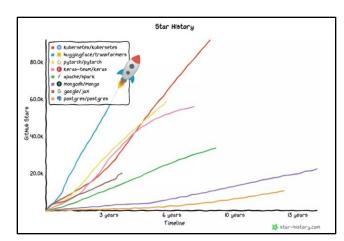




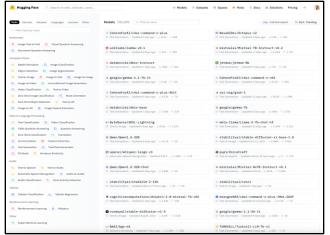
# Hugging Face

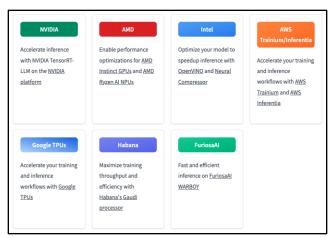
## Hugging Face

- Hugging Face Transformers:
  - Provide APIs and tools to easily download and train state-of-the-art pre-trained models.
- Hugging Face Hub:
  - The Github of machine learning models.
- Hugging Face Optimum:
  - Provides a set of performance optimization tools for deployment on various hardware.



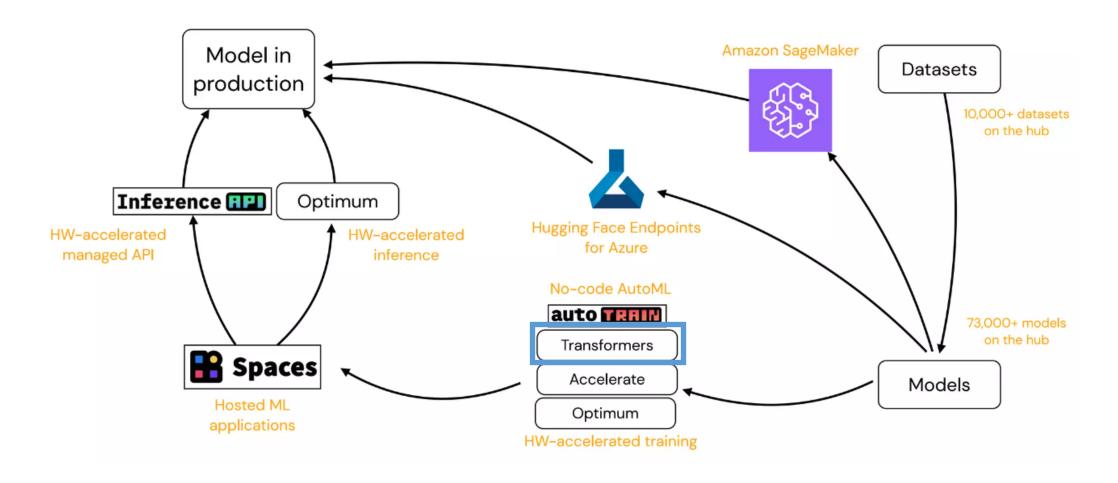














# Hugging Face Transformers

### **Transformers Overview**



- Hugging Face Transformers provides APIs and tools to download and train state-of-the-art pre-trained models easily.
- Transformers support frameworks of PyTorch, TensorFlow, and JAX for the implementation of those pre-trained models.
- Users can use a different framework at each stage of a model's life:
  - Train a model in three lines of code in one framework;
  - Load it for inference in another;
  - Models can also be exported to a format like ONNX and TorchScript for deployment in production environments.

## **Pipelines**



- The pipelines are the easiest way to use models for inference.
- These pipelines are objects that abstract most of the complex code from the library, offering a simple API dedicated to <u>a list of tasks</u>.
- Two categories of pipeline abstractions:
  - The **pipeline()** which is the most powerful object encapsulating all other pipelines.
  - *Task-specific pipelines* are available for audio, computer vision, natural language processing, and multimodal tasks.

Audio

Audio classification

Automatic speech recognition

Computer vision

Image classification

Object detection

Image segmentation

Depth estimation

Natural language processing

Text classification

Token classification

Question answering

Summarization

Translation

Language modeling

Multimodal

Document question answering





Task	Description	Modality	Pipeline identifier	
Text classification	assign a label to a given sequence of text	NLP	pipeline(task="sentiment-analysis")	
Text generation	generate text given a prompt	NLP	pipeline(task="text-generation")	
Summarization	generate a summary of a sequence of text or document	NLP	pipeline(task="summarization")	
Image classification	assign a label to an image	Computer vision	pipeline(task="image-classification")	
Image segmentation	assign a label to each individual pixel of an image (supports semantic, panoptic, and instance segmentation)	Computer vision	pipeline(task="image-segmentation")	
Object detection	predict the bounding boxes and classes of objects in an image	Computer vision	pipeline(task="object-detection")	
Audio classification	assign a label to some audio data	Audio	pipeline(task="audio-classification")	
Automatic speech recognition	transcribe speech into text	Audio	pipeline(task="automatic-speech- recognition")	
Visual question answering	answer a question about the image, given an image and a question	Multimodal	pipeline(task="vqa")	
Document question answering	answer a question about the document, given a document and a question	Multimodal	pipeline(task="document-question- answering")	
Image captioning	generate a caption for a given image	Multimodal	pipeline(task="image-to-text")	





https://huggingface.co/docs/transformers/v4.39.3/en/main\_classes/pipelines#transformers.pipeline

- Pipelines are made of:
  - A tokenizer in charge of mapping raw textual input to token.
  - A <u>model</u> to make predictions from the inputs.
  - Some (optional) post processing for enhancing model's output.





• Text Classification Example:

Code	Output		
from transformers import pipeline	[{'label': 'NEGATIVE', 'score': 0.9986492991447449}]		
<pre>pipe = pipeline(task="text-classification",</pre>	[{'label': 'POSITIVE', 'score': 0.9998718500137329}]		
<pre>result1 = pipe("The food in HKUST is terrible") print(result1)</pre>			
<pre>result2 = pipe("This course is a awesome") print(result2)</pre>			





• Translation Example:

Code	Output
from transformers import pipeline	<pre>[{'translation_text': "Hugging Face est un platform d'open-source pour l'apprentissage de machine."}]</pre>
<pre>text = "translate English to French: Hugging Face is a community-based open-source platform for machine learning." translator = pipeline(task="translation", model="google/flan-t5-small")</pre>	
<pre>text_in_chinese = translator(text) print(text_in_chinese)</pre>	





• Summarization Example:

Code	Output		
from transformers import pipeline	<pre>[{'summary_text': 'Summarization creates a shorter version'}]</pre>		
<pre>summarizer = pipeline(task="summarization", model="google/flan-t5-small", max_new_tokens=9)</pre>			
<pre>text = "Summarization creates a shorter version of a text from a longer one while trying to preserve most of the meaning of the original document."</pre>			
<pre>result = summarizer(text) print(result)</pre>			





• Language modeling Example:

Code	Output		
<pre>from transformers import pipeline generator = pipeline(task="text-generation", model="facebook/opt-350m")</pre>	[{'generated_text': 'Which university is best in Europe?\n\nThe University of Oxford is the best university in Europe.'}]		
<pre>prompt_text = "Which university is best in Europe?" result = generator(prompt_text) print(result)</pre>			

## Model API



- Model is the base class implement the common methods for loading/saving a model:
  - From a local file or directory or;
  - From a pretrained model configuration provided by the library (downloaded from HuggingFace's AWS S3 repository).
- Model class implements a methods which are common among all the models to, e.g.:
  - Resize the input token embeddings when new tokens are added to the vocabulary;
  - Prune the attention heads of the model.
- PyTorch: PreTrainedModel;
- TensorFlow: **TFPreTrainedModel**;
- Flax: FlaxPreTrainedModel.

#### Model API

```
y class transformers.PreTrainedModel 

Q
```

<source>

```
( config: PretrainedConfig, *inputs, **kwargs )
```

```
( repo_id: str, use_temp_dir: Optional = None, commit_message: Optional = None, private: Optional =
None, token: Union = None, max_shard_size: Union = '5GB', create_pr: bool = False, safe_serialization:
bool = True, revision: str = None, commit_description: str = None, tags: Optional = None,
**deprecated_kwargs )
```

#### **Parameters**

- repo\_id (str) The name of the repository you want to push your model to. It should contain your organization name
  when pushing to a given organization.
- use\_temp\_dir (boo1, optional) Whether or not to use a temporary directory to store the files saved before they are pushed to the Hub. Will default to True if there is no directory named like repo\_id, False otherwise.
- commit\_message (str, optional) Message to commit while pushing. Will default to "Upload model".
- private (boo1, optional) Whether or not the repository created should be private.
- token (bool or str, optional) The token to use as HTTP bearer authorization for remote files. If True, will use the token
  generated when running huggingface-cli login (stored in ~/.huggingface). Will default to True if repo\_url is not
  specified.
- max\_shard\_size (int or str, optional, defaults to "5GB") Only applicable for models. The maximum size for a checkpoint before being sharded. Checkpoints shard will then be each of size lower than this size. If expressed as a string, needs to be digits followed by a unit (like "5MB"). We default it to "5GB" so that users can easily load models on free-tier Google Colab instances without any CPU OOM issues.
- create\_pr (boo1, optional, defaults to False) Whether or not to create a PR with the uploaded files or directly commit.
- safe\_serialization (bool, optional, defaults to True) Whether or not to convert the model weights in safetensors format for safer serialization.
- revision (str, optional) Branch to push the uploaded files to.
- commit\_description (str, optional) The description of the commit that will be created
- tags (List[str], optional) List of tags to push on the Hub.



#### Tokenizer API



- A tokenizer is in charge of preparing the inputs for a model. The library contains tokenizers for all the models.
- The base class e.g., **PreTrainedTokenizer** implement the common methods for encoding string inputs as model inputs and instantiating/saving tokenizers either from a local file or directory or from a pretrained tokenizer provided by the library (downloaded from HuggingFace's AWS S3 repository).
- Tokenizer implements the main methods for using all the tokenizers:
  - <u>Tokenizing</u> (splitting strings in sub-word token strings), converting tokens strings to ids and back, and encoding/decoding (i.e., tokenizing and converting to integers).
  - <u>Adding new tokens</u> to the vocabulary in a way that is independent of the underlying structure (BPE, SentencePiece...).
  - <u>Managing special tokens</u> (like mask, beginning-of-sentence, etc.): adding them, assigning them to attributes in the tokenizer for easy access and making sure they are not split during tokenization.

#### **Tokenizer API**



Class attributes (overridden by derived classes)

- •vocab\_files\_names (Dict[str, str]) A dictionary with, as keys, the \_\_init\_\_ keyword name of each vocabulary file required by the model, and as associated values, the filename for saving the associated file (string).
- pretrained\_vocab\_files\_map (Dict[str, Dict[str, str]]) A dictionary of dictionaries, with the high-level keys being the \_\_init\_\_ keyword name of each vocabulary file required by the model, the low-level being the short-cutnames of the pretrained models with, as associated values, the url to the associated pretrained vocabulary file.
- •max\_model\_input\_sizes (Dict[str, Optional[int]]) A dictionary with, as keys, the short-cut-names of the pretrained models, and as associated values, the maximum length of the sequence inputs of this model, or None if the model has no maximum input size.
- pretrained\_init\_configuration (Dict[str, Dict[str, Any]]) A dictionary with, as keys, the short-cut-names of the pretrained models, and as associated values, a dictionary of specific arguments to pass to the \_\_init\_\_ method of the tokenizer class for this pretrained model when loading the tokenizer with the <u>from\_pretrained()</u> method.
- ${\color{red} \bullet model\_input\_names} \ (\texttt{List[str]}) \texttt{A list of inputs expected in the forward pass of the model}.$
- padding\_side (str) The default value for the side on which the model should have padding applied. Should be 'right' or 'left'.
- \*truncation\_side (str) The default value for the side on which the model should have truncation applied. Should be 'right' or 'left'.

### **Generation API**



- Each framework has a generate method for text generation implemented in their respective **GenerationMixin** class:
  - PyTorch generate() is implemented in GenerationMixin.
  - TensorFlow generate() is implemented in TFGenerationMixin.
  - Flax/JAX generate() is implemented in FlaxGenerationMixin.
- Regardless of the framework of choice, one can parameterize the generate method with a **GenerationConfig** class instance.





https://huggingface.co/docs/transformers/generation\_strategies#customize-text-generation

- Some of the commonly adjusted parameters include:
  - max\_new\_tokens: the maximum number of tokens to generate. In other words, the size of the output sequence, not including the tokens in the prompt. Or you can implement some **StoppingCriteria**.
  - num\_beams: by specifying a number of beams higher than 1, you are effectively switching from greedy search to beam search. This strategy evaluates several hypotheses at each time step and eventually chooses the hypothesis that has the overall highest probability for the entire sequence. This has the advantage of identifying high-probability sequences that start with a lower probability initial tokens and would've been ignored by the greedy search.
  - do\_sample: if set to True, this parameter enables decoding strategies such as multinomial sampling, beam-search multinomial sampling, Top-K sampling and Top-p sampling. All these strategies select the next token from the probability distribution over the entire vocabulary with various strategy-specific adjustments.
  - num\_return\_sequences: the number of sequence candidates to return for each input. This option is only available for the decoding strategies that support multiple sequence candidates, e.g. variations of beam search and sampling. Decoding strategies like greedy search and contrastive search return a single output sequence.

## Customize Text Generation: Beam Search



- A variant of breadth-first search (BFS) to expand the search space of greedy search.
  - Greedy search: in every generation step, keep the token with the highest probability.
- Beam search: keep top-k (num\_beams=k) tokens for generation.
  - For the first decode step, the k words with the highest current conditional probability are selected as the first word of the candidate output sequence.
  - For each subsequent decode step, based on the output sequence of the previous step, select k largest conditional probability in all combinations as the candidate output sequence by the current step.
  - It always keeps k candidates and picks the best of the candidates at the end.





• Three tokens: {I, L, U}

	Step 1	Step 2	Step 3
I	0.6	0.2	0.05
L	0.3	0.7	0.05
U	0.1	0.1	0.9

## Beam Search Example



Step 4-1 (ILL)

Step 4-1 (ILU)

• Three tokens: {I, L, U}

• k = 2

Step 2-1 (I)

0.7

Step 3-1 (IL)

0.2

0.2 0.12 0.1

0.042

0.084

Step 1

0.6

0.1 0.06

0.7

0.294

L

0.3

U

0.1

Step 2-2 (H)

0.5

0.15

0.42

0.2

Step 3-2 (LH)

0.03

0.2 0.06

0.4

0.06

0.3

0.09

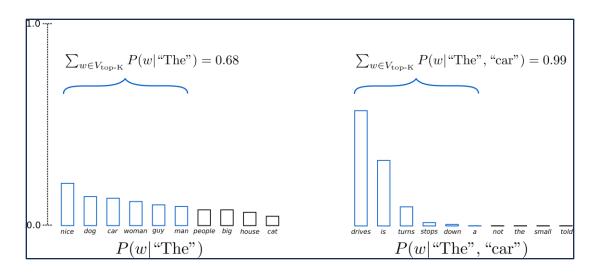
0.3

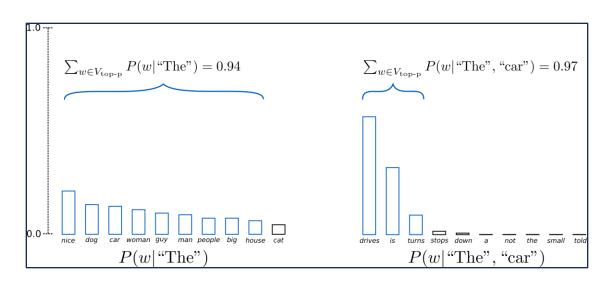
0.045

## Sampling



- <u>Sampling</u>: instead of determinstic selecting the largest tokens, we use a random number generator to sample tokens following the distribution computed by the LM.
- <u>Top-k sampling</u>: only the k (e.g., k = 6) most likely next words are filtered to be sampled.
- <u>Top-p sampling</u>: chooses from the smallest possible set of words whose cumulative probability exceeds the probability p (e.g., p = 0.92).





Top-k sampling

Top-p sampling



# An Inference Example



#### Define the Generate Function

#### Code

```
import argparse
import torch
from transformers import AutoTokenizer, AutoModelForCausalLM
from transformers.models.opt.modeling opt import *
def generate(task_info, device, model, tokenizer):
    contexts = task info["prompt seqs"]
    inputs = tokenizer(contexts, return tensors="pt").to(device)
    print(f"start_ids: length ({inputs.input ids.shape[0]}) ids: {inputs.input ids}")
    input length = inputs.input ids.shape[1]
    outputs = model.generate(
        **inputs, do sample=True, top_p=task_info['top_p'],
        temperature=1.0, top k=1,
        max new tokens=task info["output len"],
        return dict in generate=True,
        output scores=True, # return logit score
        output hidden states=False, # return embeddings
    print(f"[INFO] raw output: {outputs.keys()} {len(outputs)}, {outputs[0].shape},
({outputs[1][0].shape}, {outputs[1][1].shape}) {len(outputs[2])}")
    token = outputs.sequences[0, input_length:] # exclude context input from the output
    print(f"[INFO] raw token: {token}")
    output = tokenizer.decode(token)
    print(f"[INFO] \n[Context]\n{contexts}\n\n[Output]\n{output}\n")
```

#### Test the Model



#### Code

```
def test_model(args):
    print(f"<test model> initialization start")
    device = torch.device(args.get('device', 'cpu'))
   tokenizer = AutoTokenizer.from pretrained(args['hf model name'])
   model = AutoModelForCausalLM.from pretrained(args['hf model name'])
   model = model.to(device)
   torch.manual_seed(0)
   torch.cuda.empty_cache()
   task info = {
        "seed": 0.
        "prompt_seqs": None,
        "output len": 16,
        "beam width": 1,
        "top k": 50,
        "top p": 1,
        "beam_search_diversity_rate": 0,
        "len_penalty": 0,
        "repetition penalty": 1.0,
        "stop": args.get("stop", []),
        "logprobs": 5,
   print(f"<test model> initialization done")
   if args["interactive"]:
        while True:
            prompt data = input("Please enter the prompt input:\n")
            task_info["prompt_seqs"] = prompt_data
            generate(task info, device, model, tokenizer)
    else:
        generate(task info, device, model, tokenizer)
```





#### References



- <a href="https://www.youtube.com/watch?v=fckyXntHy1s&t=1070s">https://www.youtube.com/watch?v=fckyXntHy1s&t=1070s</a>
- <a href="https://huggingface.co/docs/transformers/index">https://huggingface.co/docs/transformers/index</a>
- <a href="https://huggingface.co/docs/transformers/generation\_strategies">https://huggingface.co/docs/transformers/generation\_strategies</a>
- <a href="https://huggingface.co/blog/how-to-generate">https://huggingface.co/blog/how-to-generate</a>