Pipelines with Hugging Face

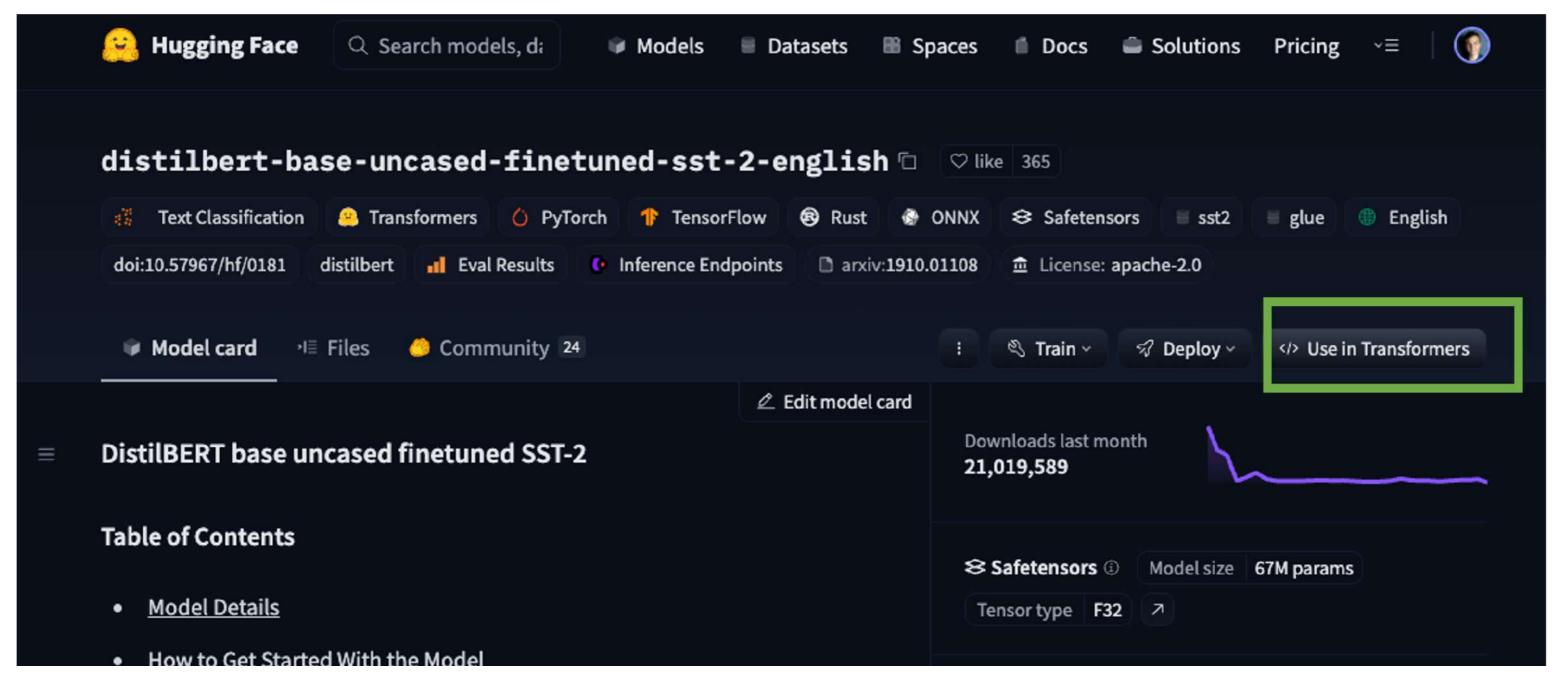
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Jacob H. Marquez
Lead Data Engineer



Use in transformers



¹ https://huggingface.co/distilbert-base-uncased-finetuned-sst-2-english



Use in transformers

```
Hugging Face
dist
       How to use from the Transformers © library
        # Use a pipeline as a high-level helper
                                                                                                      Copy
        from transformers import pipeline
        pipe = pipeline("text-classification", model="distilbert-base-uncased-finetuned-sst-2-english")
        # Load model directly
                                                                                                     Copy
Distill
        from transformers import AutoTokenizer, AutoModelForSequenceClassification
        tokenizer = AutoTokenizer.from_pretrained("distilbert-base-uncased-finetuned-sst-2-english")
Table
        model = AutoModelForSequenceClassification.from_pretrained("distilbert-base-uncased-finetuned-sst-2-e
       Quick Links
```

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Use in transformers

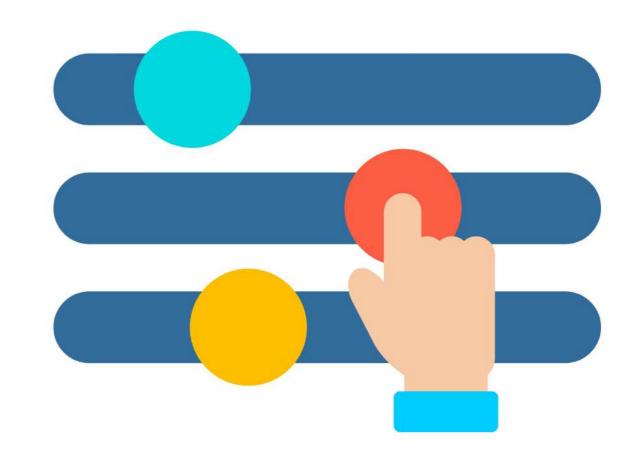
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Auto classes

- General class for using:
 - Models
 - Tokenizers
 - Configurations
 - Processors
 - Feature extractors
- Flexible and direct
- More control for ML tasks



¹ https://huggingface.co/docs/transformers/model_doc/auto



AutoModels

- Auto classes to directly download a model
- AutoModel class for each type of task

```
from transformers import AutoModelForSequenceClassification
model = AutoModelForSequenceClassification.from_pretraineed(
   "distilbert-base-uncased-finetuned-sst-2-english"
)
```

AutoTokenizers

- Prepare text input data
- Recommended to use the tokenizer paired with the model

```
from transformers import AutoTokenizer
tokenizer = AutoTokenizer.from_pretrained(
    "distilbert-base-uncased-finetuned-sst-2-english"
)
```

The pipeline module

- Contains all task-specific steps
- Best for quickly performing tasks
- Great for getting started

from transformers import pipeline

Task pipelines

```
from transformers import (
   SummarizationPipeline,
   TextClassifiationPipeline,
   AudioClassificationPipeline,
   ImageSegmentationPipeline
)
```

- Task-specific pipeline for each task
- Leverage Auto classes
- Download model and relevant processing

¹ https://huggingface.co/docs/transformers/main_classes/pipelines



Creating a pipeline

```
my_pipeline = pipeline(task="text-classification")

my_pipeline = pipeline(model="distilbert-base-uncased-finetuned-sst-2-english")

my_pipeline = pipeline(
   task="text-classification",
   model="distilbert-base-uncased-finetuned-sst-2-english"
)
```

Creating a pipeline

```
my_pipeline = pipeline(task="text-classification")
```

No model was supplied, default to distilbert-base-uncased-finetuned-sst-2-english



pipeline with Auto classes

```
from transformers import pipeline, AutoModelForSequenceClassification

model = AutoModelForSequenceClassification.from_pretrained(
   "distilbert-base-uncased-finetuned-sst-2-english"
)

my_pipeline = pipeline(model=model)
```

Using the pipeline

```
input = "Hi, welcome to this awesome course!"
my_pipeline(input)
```

```
[{'label': 'POSITIVE', 'score': 0.9998550415039062}]
```

Let's practice!

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NLP and tokenization

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Jacob H. Marquez
Lead Data Engineer



Hugging Face and NLP

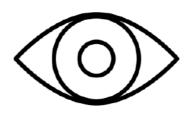


Natural Language Processing

Hugging Face and NLP



Natural Language Processing



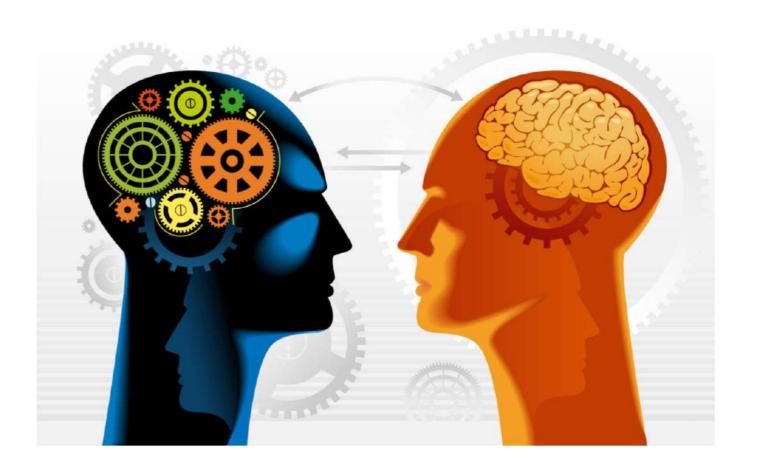
Computer Vision



Audio

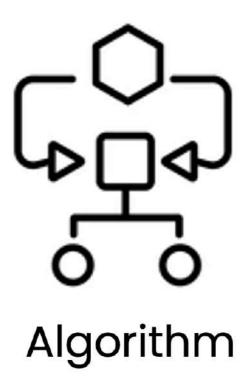
Natural language processing (NLP)

- Subfield of Al
- Enable computers to understand, interpret, and generate human language



Natural language processing

- Subfield of Al
- Enable computers to understand, interpret, and generate human language
- Algorithms, models, and transformers
- Understand words, semantics, contextual nuances



Tokenization

- Converting a sequence into smaller parts
- Numerical form

Sentence: "I am the instructor for this video."

Tokens

["I", "am", the", "instructor", ..., "video", "."]

Numerical Form

"I" = [0.232, 0.545, 0.876, ..., 0.385]

Tokenization

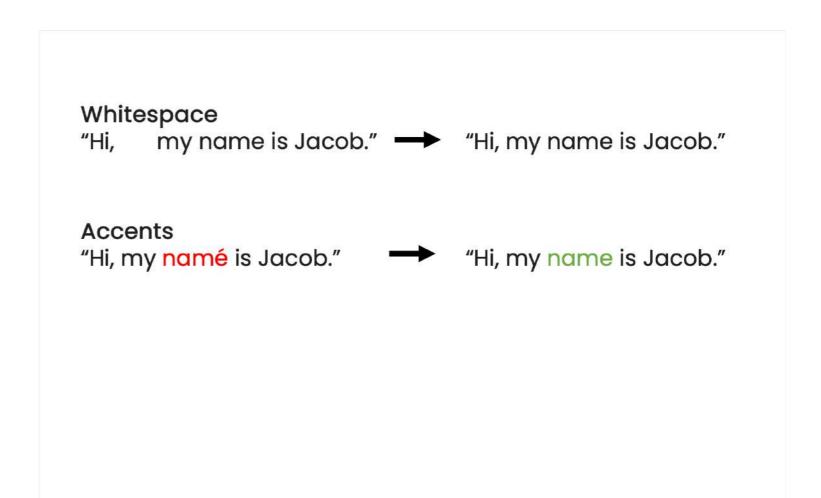
- Converting a sequence into smaller parts
- Numerical form
- Components:
 - Normalization: transforming and cleaning
 - Pre-tokenization: splitting into smaller tokens
 - Tokenization model

- Tokenization is one of the first steps of NLP
- Supports the model in building contextual knowledge

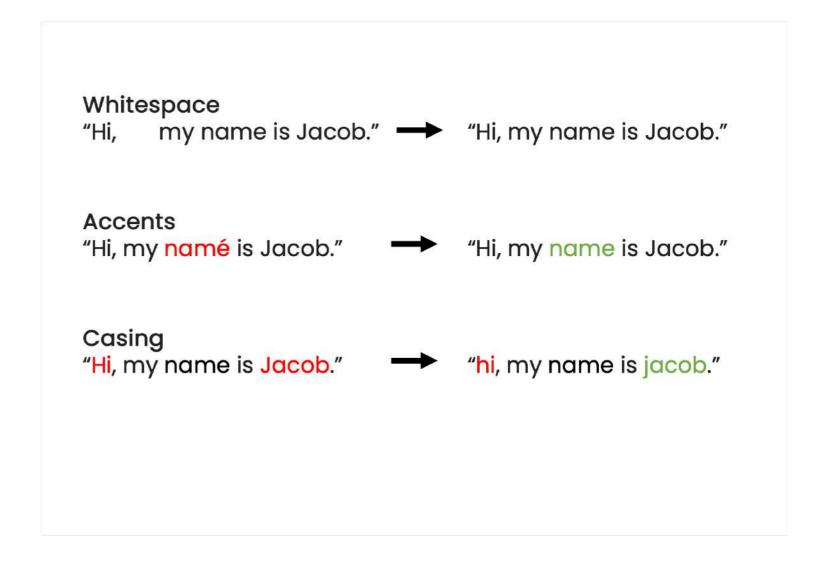
- Cleaning text
- Removing whitespaces

Whitespace

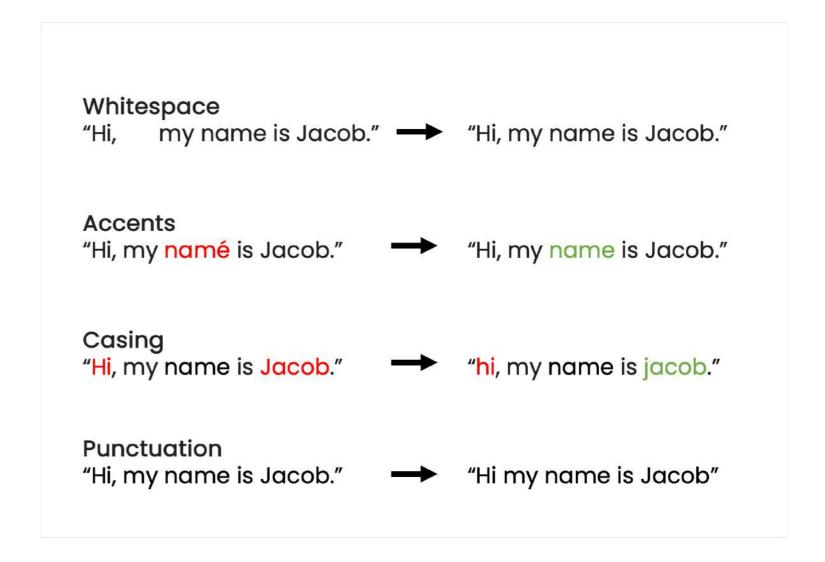
- Cleaning text
- Removing whitespaces
- Accents



- Cleaning text
- Removing whitespaces
- Accents
- Lowercasing



- Cleaning text
- Removing whitespaces
- Accents
- Lowercasing
- Punctuation



Pre-tokenization

- Input text split into smaller tokens
- Several types of pre-tokenization methods
- Split by whitespace
- Difficulty with languages that don't use spaces separate words

"hi, my name is jacob." — ["hi", "my", "name", "is", "jacob"]

Tokenizer models

- Byte-Pair Encoding, WordPiece, SentencePiece, Unigram
- Each has specific tokenization process
- Goal is to create vocabulary of characters
- Understand most common patterns

Using tokenizers

```
from transformers import AutoTokenizer
tokenizer = AutoTokenizer.from_pretrained("distilbert-base-uncased")
print(tokenizer.backend_tokenizer.normalizer.normalize_str("HOWDY, how aré yoü?"))
howdy how are you
from transformers import pipeline
```

my_pipeline = pipeline(model="distilbert-base-uncased")

Using tokenizers

```
from transformers import GPT2Tokenizer
input = "HOWDY, how aré yoü?"
gpt_tokenizer = GPT2Tokenizer.from_pretrained("gpt2")
gpt_tokens = gpt_tokenizer.tokenize(text=input)
print(gpt_tokens)
'Howdy', 'Ghow', 'Gare', 'Gyou', '?']
```

Let's practice!

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Text classification

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Jacob H. Marquez Lead Data Engineer



- Assign a set of predefined categories to text
- Sentiment analysis

Text
"I love pineapple on pizza!"

"I don't like pineapple on pizza!"

Category
Positive

Negative

¹ https://huggingface.co/tasks/text-classification



- Assign a set of predefined categories to text
- Sentiment analysis
- Question Natural Language Inference (QNLI)
 - Entailment means true
 - Not entailment means false
 - Neutral means no relationship

Question
"What famous palace
is located in London?"

"London contains four World Heritage Sites: the Tower of London; Kew Gardens; the site comprising the Palace of Westminster." Category
Not entailment (false)

¹ https://huggingface.co/tasks/text-classification



- Assign a set of predefined categories to text
- Sentiment analysis
- Question Natural Language Inference (QNLI)
- Topic modeling

Text
"The phone died very quick..."

Category
"battery"

- Assign a set of predefined categories to text
- Sentiment analysis
- Question Natural Language Inference (QNLI)
- Topic modeling
- Grammatical correctness

Text
"This course is great!"

**Course is gravy."

Category
Acceptable

Unacceptable

Challenges of text classification

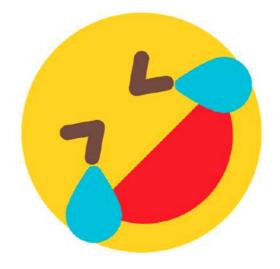


Ambiguity

Challenges of text classification



Ambiguity

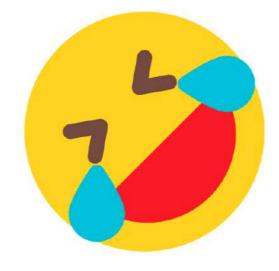


Sarcasm, Irony

Challenges of text classification



Ambiguity



Sarcasm, Irony



Multilingual

Getting started with text classification

```
from transformers import pipeline

classifier = pipeline(task="text-classification")

classifier('I love it!')

[{'label': 'POSITIVE', 'score': 0.9998656511306763}]
```

Grammatical correctness

```
classifier = pipeline(
    task="text-classification",
    model="abdulmatinomotoso/English_Grammar_Checker"
)
classifier("I write cheese strings.")
```

```
[{'label': 'LABEL_0', 'score': 0.95}]
```

QNLI

```
classifier = pipeline(
   task="text-classification",
   model="cross-encoder/qnli-electra-base"
)
classifier("Where is Seattle located?, Seattle is located in Washington state.")
```

```
[{'label': 'LABEL_0', 'score': 0.9978110194206238}]
```

Zero-shot classification

- Transfer learning
- Unseen labels can be determined
- Without specific training
- Helpful when lack of resources to train new model

```
task = "zero-shot-classification"
```

```
modelId = "facebook/bart-large-mnli"
```

```
classifier = pipeline(
  task=task,
  model=modelId
)
```

¹ https://huggingface.co/tasks/zero-shot-classification



Zero-shot classification

```
text = "Wikipedia earlier this month released its list of the 25 most viewed...."
candidate_labels = ['politics', 'science', 'technology']
output = classifier(text, candidate_labels)
print(output["labels"][0])
print(output["scores"][0])
technology
0.93600781
```



Let's practice!

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Summarization

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Jacob H. Marquez Lead Data Engineer



What is summarization?

Original Text

David G. Robinson is a data scientist at the Heap analytics company. He is a co-author of the tidytext R programming language package and the O'Reilly book, *Text Mining with R*. Robinson has previously worked as a chief data scientist at DataCamp and as a data scientist at Stack Overflow. He was also a data engineer at Flatiron Health in 2019.

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David G. Robinson is a data scientist. He is a co-author of the tidytext R package and the O'Reilly book.

Extractive versus Abstractive

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Summarized Text

David G. Robinson is a data scientist. He is a co-author of the tidytext R package and the O'Reilly book.

Extractive

- Pieces are extracted to curate representative information
- Use sentence scoring

Abstractive

- New text generated
- Deep understanding of text
- Transformer models useful here

Use cases



News Summaries

Use cases







Content Recommendations

Use cases



News Summaries



Content Recommendations



Language Translation

Summarization with pipeline

```
from transformers import pipeline
model = "sshleifer/distilbart-cnn-12-6"
summarizer = pipeline(task="summarization", model=model)
text = "This is my really large text about Data Science..."
summary_text = summarizer(text)
print(summary_text[0]['summary_text'])
"Data science is a field involving multiple interdisciplinary fields."
```

Adatacamp

Parameters for summarization

```
summarizer = pipeline(task="summarization", min_length=10, max_length=50)
```

- Put constraints around minimum and maximum number of words
- Ensure results are meaningful but not verbose
- Small storage capacity, enhance readability, improve quality

Example Error

```
Your max_length is set to 142, but your input_length is only 81. Since this is a summarization task, where outputs shorter than the input are typically wanted, you might consider decreasing max_length manually, e.g. summarizer('...', max_length=40)
```

Working with multiple inputs

```
list_of_text = [row["text"] for row in data]
```

- Inputs may not be the same length
- May cause inconsistencies and incorrect results

```
summaries = summarizer(list_of_text, truncation=True)
```

Use the maximum length size for a token specified by the model

Let's practice!

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