

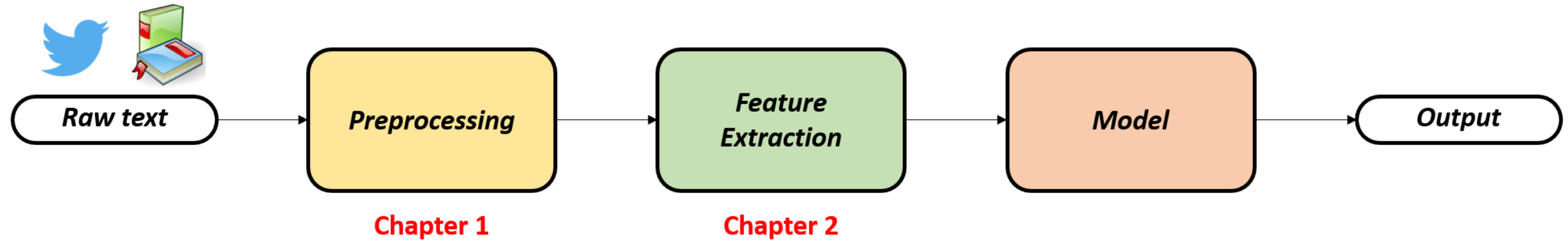
Hugging Face pipelines for sentiment analysis

NATURAL LANGUAGE PROCESSING (NLP) IN PYTHON

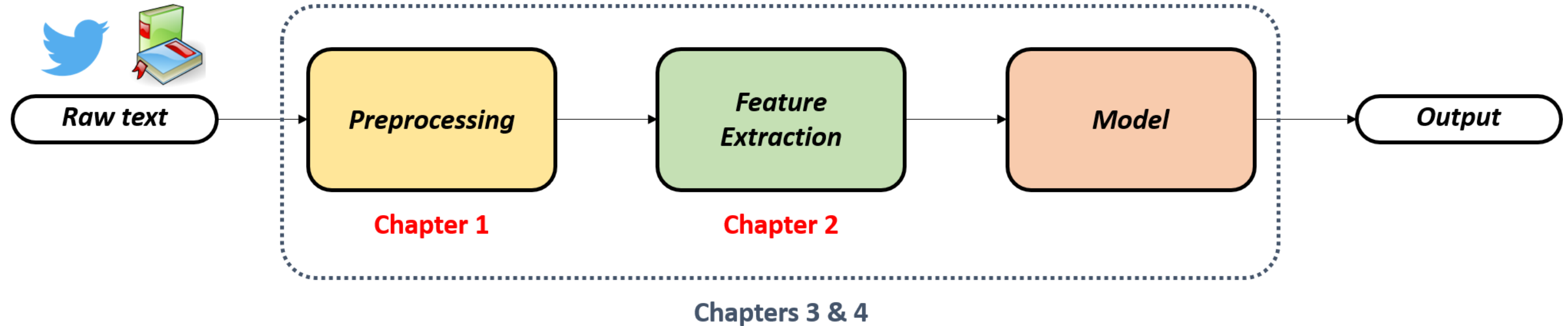


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Recap: NLP workflow



Hugging Face pipelines



Hugging Face Pipelines

- Ready-made workflow that handles all steps in a function call
- Defining a pipeline requires:
 - NLP task
 - Model to perform the task

Pipelines for sentiment analysis

- Text classification task
- Predicts if text expresses positive or negative emotion



Positive

Negative

Models for text classification

The screenshot shows the Hugging Face website's 'Models' section. The left sidebar contains navigation links for 'Tasks', 'Libraries', 'Datasets', 'Languages', 'Licenses', and 'Other'. Under 'Tasks', there are categories like 'Multimodal' and 'Computer Vision' with various sub-tasks. The main content area displays a list of models, each with its name, task, update time, size, and popularity metrics. The models are sorted by 'Trending'.

Hugging Face Search models, datasets, users...

Models 1,663,301 Filter by name Full-text search Sort: Trending

Model Name	Task	Updated	Size	Popularity
deepseek-ai/DeepSeek-Prover-V2-671B	Text Generation	Updated 6 days ago	2.03k	676
Qwen/Qwen3-235B-A22B	Text Generation	Updated 5 days ago	30.5k	694
nari-labs/Dia-1.6B	Text-to-Speech	Updated about 1 hour ago	113k	1.85k
Qwen/Qwen3-30B-A3B	Text Generation	Updated 6 days ago	67.6k	452
nvidia/parakeet-tdt-0.6b-v2	Automatic Speech Recognition	Updated 5 days ago	2.56k	242
JetBrains/Mellum-4b-base	Text Generation	Updated 6 days ago	654	230
XiaomiMiMo/MiMo-7B-RL	Text Generation	Updated about 18 hours ago	2.13k	218
microsoft/Phi-4-reasoning-plus	Text Generation	Updated 3 days ago	2.5k	197
Qwen/Qwen3-32B	Text Generation	Updated 7 days ago	87.4k	271
Qwen/Qwen2.5-Omni-3B	Any-to-Any	Updated 6 days ago	3.14k	163
Qwen/Qwen3-8B	Text Generation	Updated 7 days ago	91.3k	227
microsoft/Phi-4-reasoning	Text Generation	Updated 3 days ago	1.66k	143

¹ <https://huggingface.co/models>

Pipelines in code

```
from transformers import pipeline

classification_pipeline = pipeline(
    task="sentiment-analysis", # or text-classification
    model="distilbert/distilbert-base-uncased-finetuned-sst-2-english"
)

result = classification_pipeline("I really liked the movie!!")
print(result)
```

```
[{'label': 'POSITIVE', 'score': 0.9998093247413635}]
```


Sentiment analysis on a batch of texts

```
texts = ["I really liked the movie!!",  
         "Great job ruining my day.",  
         "This product exceeded my expectations.",  
         "Wow, just what I needed... another problem.",  
         "Absolutely fantastic experience!"]  
  
results = classification_pipeline(texts)  
print(results)
```

```
[{'label': 'POSITIVE', 'score': 0.9998093247413635},  
 {'label': 'NEGATIVE', 'score': 0.8666700124740601},  
 {'label': 'POSITIVE', 'score': 0.998874843120575},  
 {'label': 'POSITIVE', 'score': 0.98626708984375},  
 {'label': 'POSITIVE', 'score': 0.9998812675476074}]
```

Assessing sentiment analysis models

```
texts = ["I really liked the movie!!",  
         "Great job ruining my day.",  
         "This product exceeded my expectations.",  
         "Wow, just what I needed... another problem.",  
         "Absolutely fantastic experience!"]  
true_labels = ["POSITIVE", "NEGATIVE", "POSITIVE", "NEGATIVE", "POSITIVE"]  
results = classification_pipeline(texts)  
predicted_labels = [result['label'] for result in results]
```

```
from sklearn.metrics import accuracy_score  
accuracy = accuracy_score(true_labels, predicted_labels)  
print(f"Accuracy: {accuracy}")
```

Accuracy: 0.80

Let's practice!

NATURAL LANGUAGE PROCESSING (NLP) IN PYTHON

Zero-shot classification and QNLI

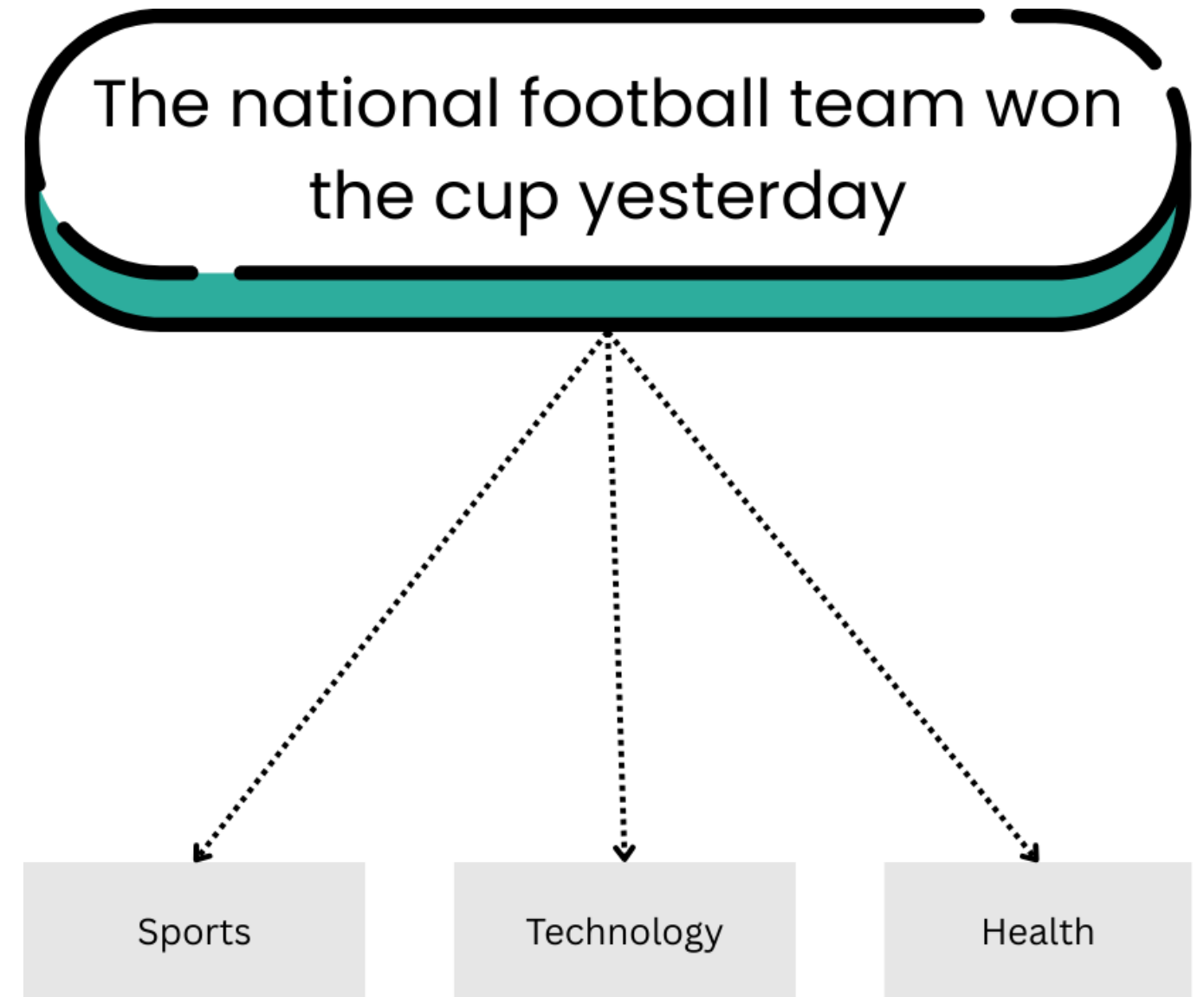
NATURAL LANGUAGE PROCESSING (NLP) IN PYTHON



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Zero-shot classification

- Allows model to assign text to labels it hasn't seen before
- Uses natural language prediction to get the output
- Useful for:
 - Content tagging
 - Customer support
 - Filtering news articles



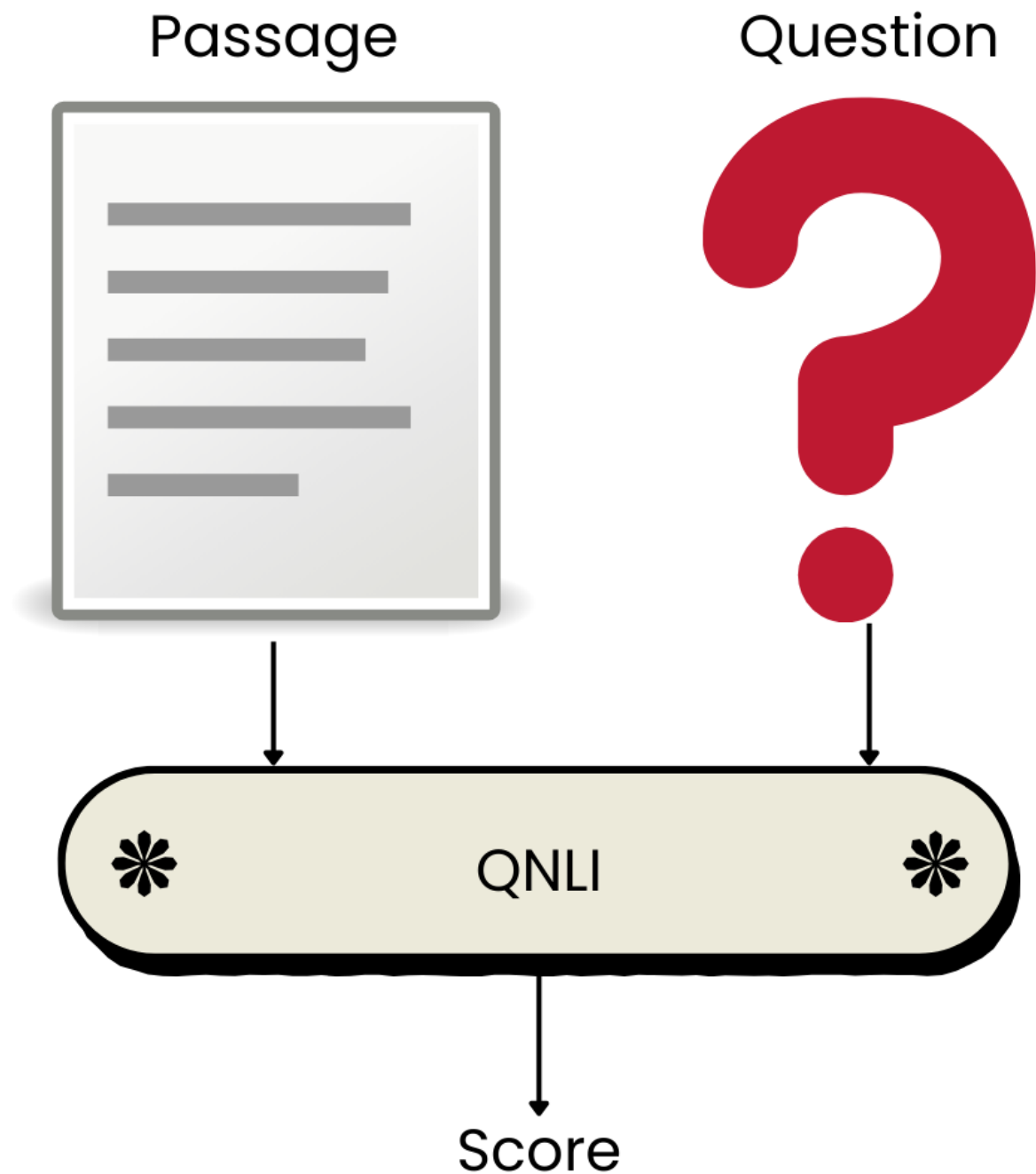
Zero-shot classification pipeline

```
from transformers import pipeline
zero_shot_classifier = pipeline(
    task="zero-shot-classification",
    model="MoritzLaurer/DeBERTa-v3-base-mnli-fever-anli"
)
text = "The national football team won the cup yesterday."
candidate_labels = ["sports", "technology", "health"]
result = zero_shot_classifier(text, candidate_labels)
print(result)
```

```
{'sequence': 'The national football team won the cup yesterday.',
 'labels': ['sports', 'technology', 'health'],
 'scores': [0.9948731064796448, 0.0029330444522202015, 0.002193822991102934]}
```

Question natural language inference (QNLI)

- Determines whether the answer to a question can be found in a passage
- Useful for:
 - Document search
 - Chatbots
 - Information retrieval



QNLI pipeline

```
from transformers import pipeline
qnli_pipeline = pipeline(
    task="text-classification",
    model="cross-encoder/qnli-electra-base"
)
passage = "Penguins are found primarily in the Southern Hemisphere."
question = "Where do penguins live?"
result = qnli_pipeline({"text": question, "text_pair": passage})
print(result)
```

```
{'label': 'LABEL_0', 'score': 0.9951545000076294}
```

QNLI pipeline

```
from transformers import pipeline
qnli_pipeline = pipeline(
    task="text-classification",
    model="cross-encoder/qnli-electra-base"
)
passage = "Penguins are found primarily in the Southern Hemisphere."
question = "What is the capital of Paris?"
result = qnli_pipeline({"text": question, "text_pair": passage})
print(result)
```

```
{'label': 'LABEL_0', 'score': 0.008907231502234936}
```


Let's practice!

NATURAL LANGUAGE PROCESSING (NLP) IN PYTHON

Question similarity and grammatical correctness

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Question similarity

- Identifies when two questions are paraphrases
- Useful for:
 - Deduplication
 - Clustering similar questions
 - Improving search accuracy
- Done with models trained on the Quora Question Pairs (QQP) dataset



QQP pipeline

```
from transformers import pipeline
qqp_pipeline = pipeline(
    task="text-classification",
    model="textattack/bert-base-uncased-QQP"
)
question1 = "How can I learn Python?"
question2 = "What is the best way to study Python?"
result = qqp_pipeline({"text": question1, "text_pair": question2})
print(result)
```

```
{'label': 'LABEL_1', 'score': 0.6853412985801697}
```

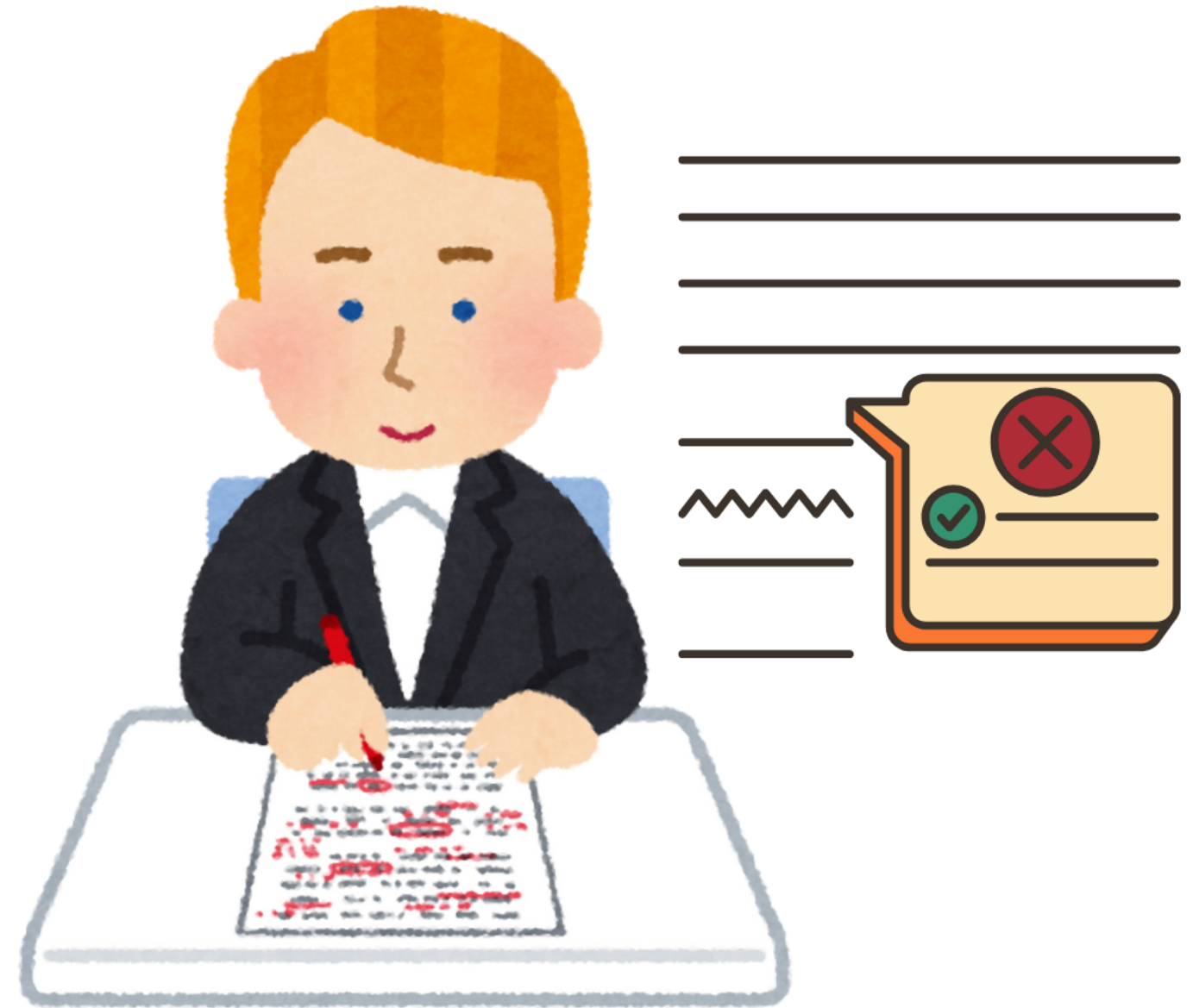
QQP pipeline

```
from transformers import pipeline
qqp_pipeline = pipeline(
    task="text-classification",
    model="textattack/bert-base-uncased-QQP"
)
question1 = "How can I learn Python?"
question2 = "What is the capital of France?"
result = qqp_pipeline({"text": question1, "text_pair": question2})
print(result)
```

```
{'label': 'LABEL_0', 'score': 0.9999338388442993}
```

Assessing grammatical correctness

- Assess how much a text is grammatically correct
- Useful for:
 - Educational tools
 - Grammar checkers
 - Writing assistants
- Done with models trained on the Corpus of Linguistic Acceptability (CoLA) dataset



CoLA pipeline

```
from transformers import pipeline
cola_classifier = pipeline(
    task="text-classification",
    model="textattack/distilbert-base-uncased-CoLA"
)
result = cola_classifier("The cat sat on the mat.")
print(result)
```

```
[{'label': 'LABEL_1', 'score': 0.9918296933174133}]
```


CoLA pipeline

```
from transformers import pipeline
cola_classifier = pipeline(
    task="text-classification",
    model="textattack/distilbert-base-uncased-CoLA"
)
result = cola_classifier("The cat on sat mat the.")
print(result)
```

```
[{'label': 'LABEL_0', 'score': 0.9628171324729919}]
```

Let's practice!

NATURAL LANGUAGE PROCESSING (NLP) IN PYTHON