

```
from datasets import load_dataset
```

```
ds = load_dataset("nielsr/funsd")
```

```
/usr/local/lib/python3.12/dist-packages/huggingface_hub/utils/_auth.py:94: UserWarning:
The secret `HF_TOKEN` does not exist in your Colab secrets.
To authenticate with the Hugging Face Hub, create a token in your settings tab (https://huggingface.co/se
You will be able to reuse this secret in all of your notebooks.
Please note that authentication is recommended but still optional to access public models or datasets.
  warnings.warn(

README.md: 100%                               755/755 [00:00<00:00, 42.1kB/s]

data/train-00000-of-00001.parquet: 100%          12.3M/12.3M [00:02<00:00, 7.33MB/s]

data/test-00000-of-00001.parquet: 100%          4.38M/4.38M [00:02<00:00, 2.06MB/s]

Generating train split: 100%                    149/149 [00:00<00:00, 772.34 examples/s]

Generating test split: 100%                      50/50 [00:00<00:00, 942.38 examples/s]
```

```
pip install transformers datasets torch pillow pytesseract
```

```
Requirement already satisfied: transformers in /usr/local/lib/python3.12/dist-packages (4.57.3)
Requirement already satisfied: datasets in /usr/local/lib/python3.12/dist-packages (4.0.0)
Requirement already satisfied: torch in /usr/local/lib/python3.12/dist-packages (2.9.0+cpu)
Requirement already satisfied: pillow in /usr/local/lib/python3.12/dist-packages (11.3.0)
Collecting pytesseract
  Downloading pytesseract-0.3.13-py3-none-any.whl.metadata (11 kB)
Requirement already satisfied: filelock in /usr/local/lib/python3.12/dist-packages (from transformers) (3.16.1)
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```

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Requirement already satisfied: frozenlist>=1.1.1 in /usr/local/lib/python3.12/dist-packages (from aiohttp

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Requirement already satisfied: multidict<7.0,>=4.5 in /usr/local/lib/python3.12/dist-packages (from aiohttp)
Requirement already satisfied: propcache>=0.2.0 in /usr/local/lib/python3.12/dist-packages (from aiohttp)
Requirement already satisfied: yarl<2.0,>=1.17.0 in /usr/local/lib/python3.12/dist-packages (from aiohttp)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.12/dist-packages (from python-dateutil)
Downloading pytesseract-0.3.13-py3-none-any.whl (14 kB)
Installing collected packages: pytesseract
Successfully installed pytesseract-0.3.13
```

```
from datasets import load_dataset

dataset = load_dataset("nielsr/funsd")

train_dataset = dataset["train"]
test_dataset = dataset["test"]

print(train_dataset[0])
```

```
{'id': '0', 'words': ['R&D', ':', 'Suggestion:', 'Date:', 'Licensee', 'Yes', 'No', '597005708', 'R&D', 'C']}
```

```
LABELS = [  
    "0",  
    "B-QUESTION", "I-QUESTION",  
    "B-ANSWER", "I-ANSWER",  
    "B-HEADER", "I-HEADER"  
]  
  
label2id = {label: i for i, label in enumerate(LABELS)}  
id2label = {i: label for label, i in label2id.items()}
```

```
from transformers import LayoutLMv3Processor  
import torch  
processor = LayoutLMv3Processor.from_pretrained(  
    "microsoft/layoutlmv3-base",  
    apply_ocr=False  
)  
  
def preprocess(example):  
    image = example["image"].convert("RGB")  
    encoding = processor(  
        image,  
        example["words"],  
        boxes=example["bboxes"],  
        word_labels=example["ner_tags"],  
        truncation=True,  
        padding="max_length",  
        return_tensors="pt"  
    )  
  
    # The processor returns tensors with a batch dimension of 1 for a single input.
```

```

    # We need to remove this for each individual example before batching.
    for k, v in encoding.items():
        if isinstance(v, torch.Tensor) and v.ndim > 1 and v.shape[0] == 1:
            encoding[k] = v.squeeze(0)

    return encoding

encoded_train = train_dataset.map(
    preprocess,
    batched=False,
    remove_columns=train_dataset.column_names
)

encoded_test = test_dataset.map(
    preprocess,
    batched=False,
    remove_columns=test_dataset.column_names
)

```

Map: 100%

149/149 [00:05<00:00, 40.66 examples/s]

Map: 100%

50/50 [00:01<00:00, 20.76 examples/s]

```

from transformers import LayoutLMv3ForTokenClassification
from transformers import Trainer, TrainingArguments, DefaultDataCollator

model = LayoutLMv3ForTokenClassification.from_pretrained(
    "microsoft/layoutlmv3-base",
    num_labels=len(LABELS),
    id2label=id2label,
    label2id=label2id
)

```

```

)

training_args = TrainingArguments(
    output_dir="./funsd_model",
    learning_rate=2e-5,
    per_device_train_batch_size=2,
    per_device_eval_batch_size=2,
    num_train_epochs=3,
    eval_strategy="steps",
    logging_steps=100,
    save_steps=500,
    save_total_limit=2,
    report_to="none"
)

data_collator = DefaultDataCollator()

trainer = Trainer(
    model=model,
    args=training_args,
    train_dataset=encoded_train,
    eval_dataset=encoded_test,
    data_collator=data_collator # Using DefaultDataCollator for preprocessed data
)

trainer.train()

```

```

Some weights of LayoutLMv3ForTokenClassification were not initialized from the model checkpoint at micro
You should probably TRAIN this model on a down-stream task to be able to use it for predictions and infer
/usr/local/lib/python3.12/dist-packages/torch/utils/data/dataloader.py:668: UserWarning: 'pin_memory' arg
warnings.warn(warn_msg)
/usr/local/lib/python3.12/dist-packages/transformers/modeling_utils.py:1621: FutureWarning: The `device`
warnings.warn(

```

```
warnings.warn(\
```

 [225/225 1:36:49, Epoch 3/3]

Step	Training Loss	Validation Loss
------	---------------	-----------------

100	1.112700	0.671565
-----	----------	----------

200	0.661500	0.585828
-----	----------	----------

```
TrainOutput(global_step=225, training_loss=0.8458343972100152, metrics={'train_runtime': 5841.0771,  
'train_samples_per_second': 0.077, 'train_steps_per_second': 0.039, 'total_flos': 117831588424704.0,  
'train_loss': 0.8458343972100152, 'epoch': 3.0})
```

```
import torch

def infer(example):
    model.eval()

    # Convert the image to RGB before processing
    image_rgb = example["image"].convert("RGB")

    encoding = processor(
        image_rgb,
        example["words"],
        boxes=example["bboxes"],
        return_tensors="pt",
        truncation=True,
        padding="max_length"
    )

    with torch.no_grad():
        outputs = model(**encoding)

    predictions = outputs.logits.argmax(-1).squeeze().tolist()

    results = []
    for word, label_id in zip(example["words"], predictions):
        results.append({
            "word": word,
            "label": id2label[label_id]
        })

    return results
```



```
import torch
from PIL import Image
import numpy as np

def infer(example):
    model.eval()

    image = example["image"]

    # ensure image is RGB (3 channels)
    if isinstance(image, Image.Image):
        image = image.convert("RGB")
    elif isinstance(image, np.ndarray):
        if image.ndim == 2: # grayscale
            image = np.stack([image]*3, axis=-1)
        image = Image.fromarray(image).convert("RGB")

    encoding = processor(
        image,
        example["words"],
        boxes=example["bboxes"],
        return_tensors="pt",
        truncation=True,
        padding="max_length"
    )

    with torch.no_grad():
        outputs = model(**encoding)

    predictions = outputs.logits.argmax(-1).squeeze().tolist()
```

```
results = []
for word, label_id in zip(example["words"], predictions):
    results.append((word, id2label[label_id]))

return results
```

```
sample = test_dataset[0]
predictions = infer(sample)
```

```
for w, l in predictions:
    print(w, "->", l)
```

```
/usr/local/lib/python3.12/dist-packages/transformers/modeling_utils.py:1621: FutureWarning: The `device
warnings.warn(
T0: -> 0
DATE: -> B-ANSWER
3 -> B-ANSWER
Fax: -> B-ANSWER
NOTE: -> B-ANSWER
82092117 -> I-ANSWER
614 -> B-HEADER
-466 -> 0
-5087 -> 0
Dec -> 0
10 -> B-ANSWER
'98 -> I-ANSWER
17 -> 0
:46 -> 0
P. -> 0
01 -> 0
ATT. -> 0
```

GEN. -> 0
ADMIN. -> I-HEADER
OFFICE -> 0
Attorney -> I-HEADER
General -> I-HEADER
Betty -> 0
D. -> 0
Montgomery -> 0
CONFIDENTIAL -> 0
FACSIMILE -> 0
TRANSMISSION -> 0
COVER -> 0
SHEET -> 0
(614) -> 0
466- -> 0
5087 -> 0
FAX -> 0
NO. -> 0
George -> I-QUESTION
Baroody -> I-QUESTION
(336) -> I-QUESTION
335- -> I-QUESTION
7392 -> I-QUESTION
FAX -> I-QUESTION
NUMBER: -> B-HEADER
PHONE -> I-HEADER
NUMBER: -> B-HEADER
(336) -> I-HEADER
335- -> I-HEADER
7363 -> I-HEADER
NUMBER -> B-QUESTION
OF -> I-QUESTION
PAGES -> I-QUESTION
INCLUDING -> I-QUESTION
COVER -> I-QUESTION

```
SHEET: -> I-QUESTION  
June -> I-QUESTION  
Flynn -> I-QUESTION  
for -> I-QUESTION
```