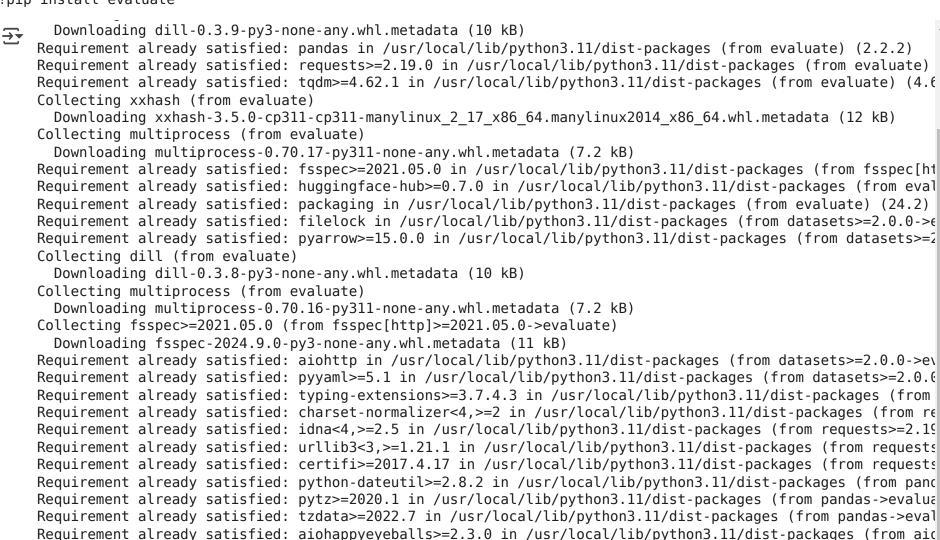
## !pip install evaluate



Requirement already satisfied: aiosignal>=1.1.2 in /usr/local/lib/python3.11/dist-packages (from aiohttp->c Requirement already satisfied: attrs>=17.3.0 in /usr/local/lib/python3.11/dist-packages (from aiohttp->data

```
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                                           - 143.5/143.5 kB 8.6 MB/s eta 0:00:00
Downloading xxhash-3.5.0-cp311-cp311-manylinux 2 17 x86 64.manylinux2014 x86 64.whl (194 kB)
                                          - 194.8/194.8 kB 9.3 MB/s eta 0:00:00
Installing collected packages: xxhash, fsspec, dill, multiprocess, datasets, evaluate
  Attempting uninstall: fsspec
   Found existing installation: fsspec 2024.10.0
   Uninstalling fsspec-2024.10.0:
      Successfully uninstalled fsspec-2024.10.0
ERROR: pip's dependency resolver does not currently take into account all the packages that are installed.
gcsfs 2024.10.0 requires fsspec==2024.10.0, but you have fsspec 2024.9.0 which is incompatible.
Successfully installed datasets. 3 2 A dill.A 3 8 evaluate.A 4 3 fssner.2A24 Q A multinrocess.A 7A 16 xxhask
```

pip install transformers datasets torch evaluate segeval

Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.11/dist-packages (from transform \* Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from transformers) (2.3 Requirement already satisfied: tokenizers<0.22,>=0.21 in /usr/local/lib/python3.11/dist-packages (from transpackages) Requirement already satisfied: safetensors>=0.4.1 in /usr/local/lib/python3.11/dist-packages (from transform) Requirement already satisfied: tgdm>=4.27 in /usr/local/lib/python3.11/dist-packages (from transformers) (4 Requirement already satisfied: pyarrow>=15.0.0 in /usr/local/lib/python3.11/dist-packages (from datasets) (

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Requirement already satisfied: nvidia-cufft-cu12==11.0.2.54 in /usr/local/lib/python3.11/dist-packages (from
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Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2
Building wheels for collected packages: segeval
  Building wheel for segeval (setup.py) ... done
  Created wheel for segeval: filename=segeval-1.2.2-py3-none-any.whl size=16161 sha256=0b0b084f8df9c6857c97
  Stored in directory: /root/.cache/pip/wheels/bc/92/f0/243288f899c2eacdfa8c5f9aede4c7la9bad0ee26a01dc5ead
Successfully built segeval
Installing collected packages: segeval
Successfully installed segeval-1.2.2
```

import evaluate
import numpy as np

```
from transformers import AutoTokenizer, AutoModelForTokenClassification, Trainer, TrainingArguments
from datasets import load dataset
# Token to authenticate with Hugging Face platform
HF TOKEN = "hf wQxqRdWrPuxXbpTpEEfYixEICoHxGedJwN"
# Load the dataset
dataset = load dataset("ncbi disease")
# Load BioBERT tokenizer and model
model name = "dmis-lab/biobert-base-cased-v1.1"
tokenizer = AutoTokenizer.from pretrained(model name, use auth token=HF TOKEN)
model = AutoModelForTokenClassification.from pretrained(model name, num labels=3)
# Preprocess the dataset: Tokenization and label alignment
def tokenize and align labels(examples):
    tokenized inputs = tokenizer(
        examples["tokens"], truncation=True, padding="max length", max length=128, is split into words=True
    labels = []
    for i, label in enumerate(examples["ner tags"]):
        word ids = tokenized inputs.word ids(batch index=i) # Map tokens to words
        label ids = [-100 if word id is None else label[word id] for word id in word ids]
        labels.append(label ids)
    tokenized inputs["labels"] = labels
    return tokenized inputs
# Apply tokenization and label alignment to the dataset
tokenized datasets = dataset.map(tokenize and align labels, batched=True)
# Split into training and testing datasets
train dataset = tokenized datasets["train"]
test dataset = tokenized datasets["test"]
# Load segeval metric for Named Entity Recognition
metric = evaluate.load("segeval")
def compute metrics(predictions):
    logits, labels = predictions
```

```
# Get the label names from the dataset
    label names = dataset["train"].features["ner tags"].feature.names
    # Convert logits to predictions
    predictions = logits.argmax(axis=-1)
    # Map predictions and references from indices to label names
    true predictions = [
        [label names[p] for p, l in zip(pred, label) if l != -100]
        for pred, label in zip(predictions, labels)
    true labels = [
        [label names[l] for p, l in zip(pred, label) if l != -100]
        for pred, label in zip(predictions, labels)
    # Compute the metrics using the string labels
    results = metric.compute(predictions=true predictions, references=true labels)
    return {
        "precision": results["overall precision"],
        "recall": results["overall recall"],
        "f1": results["overall f1"],
        "accuracy": results["overall accuracy"],
    }
# Training arguments
training args = TrainingArguments(
    output dir="./results",
    evaluation strategy="epoch",
    learning rate=5e-5,
    per device train batch size=8,
    per device eval batch size=8,
    num train epochs=3,
    weight decay=0.01,
    logging dir="./logs",
    logging steps=10,
```

```
# Initialize Trainer
trainer = Trainer(
    model=model,
    args=training_args,
    train_dataset=train_dataset,
    eval_dataset=test_dataset,
    tokenizer=tokenizer,
    compute_metrics=compute_metrics,
)

# Train the model
trainer.train()

# Evaluate the model
results = trainer.evaluate()
print("Evaluation Results:", results)
```



ist-packages/huggingface hub/utils/ auth.py:94: UserWarning: not exist in your Colab secrets.

gging Face Hub, create a token in your settings tab (<a href="https://huggingface.co/settin">https://huggingface.co/settin</a> nis secret in all of your notebooks.

tion is recommended but still optional to access public models or datasets.

9.70k/9.70k [00:00<00:00, 609kB/s]

5.83k/5.83k [00:00<00:00, 419kB/s]

ease contains custom code which must be executed to correctly load the dataset. Yo in future by passing the argument `trust remote code=True`.

om code? [y/N] y

1.14M/? [00:00<00:00, 18.9MB/s]

200k/? [00:00<00:00, 11.3MB/s]

206k/? [00:00<00:00, 12.1MB/s]

5433/5433 [00:01<00:00, 4133.83 examples/s]

924/924 [00:00<00:00, 3838.77 examples/s]

941/941 [00:00<00:00, 3686.59 examples/s]

ist-packages/transformers/models/auto/tokenization auto.py:810: FutureWarning: The

313/313 [00:00<00:00, 17.1kB/s]

213k/213k [00:00<00:00, 4.53MB/s]

436M/436M [00:03<00:00, 78.1MB/s]

nClassification were not initialized from the model checkpoint at dmis-lab/biobert nis model on a down-stream task to be able to use it for predictions and inference

5433/5433 [00:02<00:00, 2698.42 examples/s]

924/924 [00:00<00:00, 2557.11 examples/s]

941/941 [00:00<00:00, 2157.41 examples/s]

6.34k/6.34k [00:00<00:00, 459kB/s]

```
€6>:83: FutureWarning: `tokenizer` is deprecated and will be removed in version 5.
e` is currently set to the same value as `TrainingArguments.output dir`. If this w
i.wandb.ai to your netrc file: /root/.netrc
the SDK backend. Please refer to https://wandb.me/wandb-core for more information
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t/wandb/run-20250118 044825-j4d87802
ases (docs)
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ation Loss
              Traceback (most recent call last)
≥6> in <cell line: 0>()
    2 6 frames
ist-packages/torch/autograd/graph.py in _engine_run backward(t outputs, *args,
nooks = register logging hooks on whole graph(t outputs)
able. execution engine.run backward( # Calls into the C++ engine to run the
ts, *args, **kwargs
into the C++ engine to run the backward pass
```

```
import evaluate
import numpy as np
from transformers import AutoTokenizer, AutoModelForTokenClassification, Trainer, TrainingArguments
from datasets import load dataset
# Token to authenticate with Hugging Face platform
HF TOKEN = "hf wQxgRdWrPuxXbpTpEEfYixEICoHxGedJwN"
# Load the dataset
dataset = load dataset("ncbi disease")
# Reducing dataset to a smaller fraction
train dataset = dataset["train"].select(range(len(dataset["train"]) // 4))
test dataset = dataset["test"].select(range(len(dataset["test"]) // 4))
# Load BioBERT tokenizer and model
model name = "dmis-lab/biobert-base-cased-v1.1"
tokenizer = AutoTokenizer.from pretrained(model name, use auth token=HF TOKEN)
model = AutoModelForTokenClassification.from pretrained(model name, num labels=3)
# Preprocess the dataset: Tokenization and label alignment
def tokenize and align labels(examples):
    tokenized inputs = tokenizer(
        examples["tokens"], truncation=True, padding="max length", max length=128, is split into words=True
    labels = []
    for i, label in enumerate(examples["ner tags"]):
        word ids = tokenized inputs.word ids(batch index=i) # Map tokens to words
        label ids = [-100 if word id is None else label[word id] for word id in word ids]
        labels.append(label ids)
    tokenized inputs["labels"] = labels
    return tokenized inputs
# Apply tokenization and label alignment to the dataset
tokenized train = train dataset.map(tokenize and align labels, batched=True)
tokenized test = test dataset.map(tokenize and align labels, batched=True)
```

```
# Load segeval metric for Named Entity Recognition
metric = evaluate.load("segeval")
def compute metrics(predictions):
    logits, labels = predictions
    # Get the label names from the dataset
    label names = dataset["train"].features["ner tags"].feature.names
    # Convert logits to predictions
    predictions = logits.argmax(axis=-1)
    # Map predictions and references from indices to label names
    true predictions = [
        [label names[p] for p, l in zip(pred, label) if l != -100]
        for pred, label in zip(predictions, labels)
    true labels = [
        [label names[l] for p, l in zip(pred, label) if l != -100]
        for pred, label in zip(predictions, labels)
    # Compute the metrics using the string labels
    results = metric.compute(predictions=true predictions, references=true labels)
    return {
        "precision": results["overall precision"],
        "recall": results["overall recall"],
        "f1": results["overall f1"],
        "accuracy": results["overall accuracy"],
    }
# Training arguments with checkpoint saving
training args = TrainingArguments(
    output dir="./results", # Directory to save the model and checkpoints
    evaluation strategy="epoch",
    learning rate=5e-5,
    per device train batch size=8,
    per device eval batch size=8,
    num train epochs=2, # Reduced epochs to avoid long training times
```

```
weight decay=0.01,
    logging dir="./logs",
    logging steps=10,
    save strategy="epoch", # Save checkpoints at the end of every epoch
    save steps=100, # Save model every 100 steps
    load best model at end=True, # Load the best model when training ends
    metric for best model="f1", # Use F1 score to select the best model
# Initialize Trainer with checkpoint callback
trainer = Trainer(
    model=model.
    args=training args,
    train dataset=tokenized train,
    eval dataset=tokenized test,
    tokenizer=tokenizer,
    compute metrics=compute metrics,
# Train the model
trainer.train()
# Evaluate the model
results = trainer.evaluate()
print("Evaluation Results:", results)
```

/usr/local/lib/python3.11/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/settings.co 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% 9.70k/9.70k [00:00<00:00, 577kB/s]

ncbi disease.py: 100% 5.83k/5.83k [00:00<00:00, 334kB/s]

Downloading data: 1.14M/? [00:00<00:00, 23.9MB/s]

Downloading data: 200k/? [00:00<00:00, 6.34MB/s]

Downloading data: 206k/? [00:00<00:00, 6.84MB/s]

Generating train split: 100% 5433/5433 [00:01<00:00, 4362.32 examples/s]

Generating validation split: 100% 924/924 [00:00<00:00, 4006.63 examples/s]

941/941 [00:00<00:00, 3505.78 examples/s] Generating test split: 100%

/usr/local/lib/python3.11/dist-packages/transformers/models/auto/tokenization auto.py:810: FutureWarning: The warnings.warn(

config.json: 100% 313/313 [00:00<00:00, 14.5kB/s]

vocab.txt: 100% 213k/213k [00:00<00:00, 6.65MB/s]

pytorch model.bin: 100% 436M/436M [00:05<00:00, 23.9MB/s]

Some weights of BertForTokenClassification were not initialized from the model checkpoint at dmis-lab/biober You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference

Map: 100% 1358/1358 [00:00<00:00, 2449.43 examples/s]

Map: 100% 235/235 [00:00<00:00, 1327.48 examples/s]

Downloading builder script: 100% 6.34k/6.34k [00:00<00:00, 468kB/s]

/usr/local/lib/python3.11/dist-packages/transformers/training args.py:1575: FutureWarning: `evaluation strate warnings.warn(

<ipython-input-3-12214082883d>:87: FutureWarning: `tokenizer` is deprecated and will be removed in version 5 trainer = Trainer(

wandb: WARNING The `run name` is currently set to the same value as `TrainingArguments.output dir`. If this \ wandb: Logging into wandb.ai. (Learn how to deploy a W&B server locally: https://wandb.me/wandb-server)

wandh. You can find your ADT key in your brouger bare, bttms://wandh ai/authoriza

Wallub. Tou call Ithiu your AFT key th your browser here. https://wallub.at/authortze

wandb: Paste an API key from your profile and hit enter, or press ctrl+c to quit:wandb: Appending key for ap: wandb: Using wandb-core as the SDK backend. Please refer to <a href="https://wandb.me/wandb-core">https://wandb.me/wandb-core</a> for more information

Tracking run with wandb version 0.19.2

Run data is saved locally in /content/wandb/run-20250118 050403-vz2t8ou0

Syncing run ./results to Weights & Biases (docs)

View project at <a href="https://wandb.ai/mohamedlemine-telmoudy-ensam-rabat/huggingface">https://wandb.ai/mohamedlemine-telmoudy-ensam-rabat/huggingface</a>

View run at https://wandb.ai/mohamedlemine-telmoudy-ensam-rabat/huggingface/runs/vz2t8ou0

[290/340 1:02:02 < 10:46, 0.08 it/s, Epoch 1.70/2]

Epoch	Training Loss	<b>Validation Loss</b>	Precision	Recall	F1	Accuracy
1	0.061500	0.088064	0.753676	0.818363	0.784689	0.968949

[040/040 4.4 4.40 Facal 0/0]