Final Project Report-Team 6

October 23, 2023

[ ]:
```
PROJECT = "aai520-project" # @param {type:"string"}
GROUP = "aai520-group6" # @param {type:"string"}
MODELS_STR = "distilbert-base-uncased, bert-base-uncased, roberta-base" #␣
 ↪@param {type:"string"}
MODELS = [model.strip() for model in MODELS_STR.split(',')]
TASK = "question-answering" # @param {type:"string"}
DATASET = "squad_v2" # @param {type:"string"}
NUM_TRAIN_EPOCHS = 4  # @param {type:"integer"}
LEARNING_RATE = 2e-5  # @param
GRADIENT_ACCUMULATION_STEPS = 4  # @param {type:"integer"}
PER_DEVICE_TRAIN_BATCH_SIZE = 128  # @param {type:"integer"}
PER_DEVICE_EVAL_BATCH_SIZE = 128  # @param {type:"integer"}
EVALUATION_STRATEGY = 'steps'  # @param {type:"string"}
EVAL_STEPS = 100  # @param {type:"integer"}
SAVE_STRATEGY = "steps"  # @param {type:"string"}
SAVE_STEPS = 100  # @param {type:"integer"}
LOGGING_STEPS = 100  # @param {type:"integer"}
FP16 = True  # @param {type:"boolean"}
DATALOADER_NUM_WORKERS = 2  # @param {type:"integer"}
REPORT_TO = 'tensorboard'  # @param {type:"string"}
LOAD_BEST_MODEL_AT_END = True  # @param {type:"boolean"}
DISABLE_TQDM = False  # @param {type:"boolean"}
PUSH_TO_HUB = True  # @param {type:"boolean"}
OVERWRITE_OUTPUT_DIRECTORY = True  # @param {type:"boolean"}
```

[ ]:
```
#-------------------------------------------------------------------------------#
# CLEAN ENVIRONMENT
#-------------------------------------------------------------------------------#

import gc
gc.collect()

import torch
torch.cuda.empty_cache()
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")

#-------------------------------------------------------------------------------#
```

```
# MOUNT GOOGLE DRIVE
#-------------------------------------------------------------------------------#

from pathlib import Path
from google.colab import drive
drive.mount('/content/drive', force_remount=True)
DRIVE_PATH = Path("/content/drive/My Drive/Colab Notebooks")

#-------------------------------------------------------------------------------#
# HANDLE PATHS
#-------------------------------------------------------------------------------#

PROJECT_PATH = DRIVE_PATH / PROJECT
PROJECT_PATH.mkdir(parents=True, exist_ok=True)
CHECKPOINTS_DIR = PROJECT_PATH / "checkpoints"
CHECKPOINTS_DIR.mkdir(parents=True, exist_ok=True)
LOGS_DIR = PROJECT_PATH / "logs"
LOGS_DIR.mkdir(parents=True, exist_ok=True)
VISUALS_DIR = PROJECT_PATH / "visuals"
VISUALS_DIR.mkdir(parents=True, exist_ok=True)

#-------------------------------------------------------------------------------#
# SYMLINK FOR FASTER FILESYSTEM TRAVERSAL
#-------------------------------------------------------------------------------#

SHORTCUT = Path("/content/project")
if not SHORTCUT.exists():
    SHORTCUT.symlink_to(PROJECT_PATH)

#-------------------------------------------------------------------------------#
# INSTALL DEPENDENCIES
#-------------------------------------------------------------------------------#

!pip install \
    -qqq \
    --progress-bar=off \
    datasets \
    evaluate \
    huggingface_hub \
    python-dotenv \
    tensorboardcolab \
    transformers[torch] \
    txtai

#-------------------------------------------------------------------------------#
# IMPORT LIBRARIES & SETUP/CONFIGURE
#-------------------------------------------------------------------------------#
```

```
import json
import os

import evaluate
from datasets import load_dataset
from dotenv import load_dotenv
from evaluate.visualization import radar_plot
from huggingface_hub import EvalResult, ModelCard, ModelCardData, RepoCard,␣
 ↪login
from transformers.trainer_utils import get_last_checkpoint
from txtai.pipeline import HFTrainer



# LOGIN TO HUB
load_dotenv(dotenv_path=PROJECT_PATH / ".env")
login(token=os.getenv('HF_TOKEN'))

# IGNORE TRANSFORMER WARNINGS
os.environ['TRANSFORMERS_NO_ADVISORY_WARNINGS'] = 'true'

#-------------------------------------------------------------------------------#
# LOAD DATASET
#-------------------------------------------------------------------------------#

# DATASET
ds = load_dataset(DATASET)
ds.push_to_hub(repo_id=f"{GROUP}/{DATASET}")

#-------------------------------------------------------------------------------#
# ITERATE THROUGH ALL MODELS
#-------------------------------------------------------------------------------#

combined_results = {}

for pretrained_model in MODELS:

    # RENAME MODEL TO DISTINGUISH FINETUNED MODEL
    finetuned_model = pretrained_model.replace('base', 'finetuned')
    print(f'FINE-TUNING: {finetuned_model}')

    # CONSTRUCT STANDARD IDENTIFER FOR PUBLISHING
    IDENTIFIER = f"{GROUP}/{finetuned_model}-{DATASET}"

    # LOAD CHECKPOINTS
    model_checkpoints = CHECKPOINTS_DIR / finetuned_model
```

```
    model_checkpoints.mkdir(parents=True, exist_ok=True)
    last_checkpoint = get_last_checkpoint(model_checkpoints)

#-------------------------------------------------------------------------------#
# TRAIN MODEL
#-------------------------------------------------------------------------------#

    # use txtai for preconfigured preprocessing of SQuAD datasets
    # https://github.com/neuml/txtai
    trainer = HFTrainer()
    model, tokenizer = trainer(

        # SPECIFICATIONS
        base=last_checkpoint or pretrained_model,
        task=TASK,

        # DATASETS
        train=ds["train"],
        validation=ds["validation"],

        # HYPERPARAMETERS
        num_train_epochs=NUM_TRAIN_EPOCHS,
        learning_rate=LEARNING_RATE,
        gradient_accumulation_steps=GRADIENT_ACCUMULATION_STEPS,
        per_device_train_batch_size=PER_DEVICE_TRAIN_BATCH_SIZE,
        per_device_eval_batch_size=PER_DEVICE_EVAL_BATCH_SIZE,

        # CHECKPOINTS/OUTPUTS
        checkpoint=last_checkpoint,
        output_dir=str(model_checkpoints),

        # STRATEGIES
        evaluation_strategy=EVALUATION_STRATEGY,
        eval_steps=EVAL_STEPS,
        save_strategy=SAVE_STRATEGY,
        save_steps=SAVE_STEPS,
        logging_steps=LOGGING_STEPS,

        # OPTIMIZATIONS
        fp16=FP16,
        dataloader_num_workers=DATALOADER_NUM_WORKERS,

        # SAVING/LOGGING
        report_to=REPORT_TO,
        logging_dir=str(LOGS_DIR),
        load_best_model_at_end=LOAD_BEST_MODEL_AT_END,
        disable_tqdm=DISABLE_TQDM,
```

```
        # PUBLISH TO HUB
        push_to_hub=PUSH_TO_HUB,
        hub_model_id=IDENTIFIER,
        hub_token=os.getenv('HF_TOKEN')

    )

#--------------------------------------------------------------------------------#
# VALIDATE MODEL
#--------------------------------------------------------------------------------#

    print(f'VALIDATING: {finetuned_model}')
    evaluate.logging.set_verbosity_info()
    results = evaluate.evaluator("question-answering").compute(
        model_or_pipeline=IDENTIFIER,
        data=ds["validation"].select(range(2)),
        metric=DATASET,
        squad_v2_format=True,
    )

    for metric_type, metric_value  in results.items():
        metric_name = metric_type.upper() if metric_type == 'f1' else␣
 ↪metric_type.capitalize()
        evaluate.push_to_hub(
            model_id=IDENTIFIER,
            metric_value=metric_value,
            metric_type=metric_type,
            metric_name=metric_name,
            dataset_type=DATASET,
            dataset_name="SQuAD v2",
            dataset_split="validation",
            task_type="question-answering",
            task_name="Question Answering",
            overwrite=True
        )

    # SAVE METRICS
    metrics_file_path = model_checkpoints / 'metrics.json'
    with open(metrics_file_path, 'w') as f:
        json.dump(results, f, indent=4)

    combined_results[pretrained_model] = results

#--------------------------------------------------------------------------------#
# CREATE MODEL CARD FOR HUGGINGFACE
#--------------------------------------------------------------------------------#
```

```
    # ITERATE AND ADD ALL METRICS
    eval_results = [
        EvalResult(
            task_type=TASK,
            dataset_type=DATASET,
            dataset_name='SQuAD v2',
            metric_type=metric_name.replace('_', ' ').title(),
            metric_value=metric_value
        )
        for metric_name, metric_value in results.items()
    ]

    # CREATE MODEL CARD DATA
    model_card_data = ModelCardData(
        language='en',
        license='mit',
        model_name=finetuned_model,
        eval_results=eval_results,

        model_details=f"""
        ## Abstract
        This model, '{finetuned_model}', is a question-answering chatbot␣
↪trained on the SQuAD dataset, demonstrating competency in building␣
↪conversational AI using recent advances in natural language processing. It␣
↪utilizes a BERT model fine-tuned for extractive question answering.

        ## Data Collection and Preprocessing
        The model was trained on the Stanford Question Answering Dataset␣
↪(SQuAD), which contains over 100,000 question-answer pairs based on␣
↪Wikipedia articles. The data preprocessing involved tokenizing context␣
↪paragraphs and questions, truncating sequences to fit BERT's max length, and␣
↪adding special tokens to mark question and paragraph segments.

        ## Model Architecture and Training
        The architecture is based on the BERT transformer model, which was␣
↪pretrained on large unlabeled text corpora. For this project, the BERT base␣
↪model was fine-tuned on SQuAD for extractive question answering, with␣
↪additional output layers for predicting the start and end indices of the␣
↪answer span.

        ## SQuAD 2.0 Dataset
        SQuAD 2.0 combines the existing SQuAD data with over 50,000␣
↪unanswerable questions written adversarially by crowdworkers to look similar␣
↪to answerable ones. This version of the dataset challenges models to not␣
↪only produce answers when possible but also determine when no answer is␣
↪supported by the paragraph and abstain from answering.
```

```
        """,
        intended_use=f"""
        - Answering questions from the {DATASET} dataset.
        - Developing question-answering systems within the scope of the␣
↪{PROJECT}.
        - Research and experimentation in the NLP question-answering domain.
        """,
        limitations_and_bias=f"""
        The model inherits limitations and biases from the '{pretrained_model}'␣
↪model, as it was trained on the same foundational data.
        It may underperform on questions that are ambiguous or too far outside␣
↪the scope of the topics covered in the {DATASET} dataset.
        Additionally, the model may reflect societal biases present in its␣
↪training data.
        """,
        ethical_considerations=f"""
        This model should not be used for making critical decisions without␣
↪human oversight,
        as it can generate incorrect or biased answers, especially for topics␣
↪not covered in the training data.
        Users should also consider the ethical implications of using AI in␣
↪decision-making processes and the potential for perpetuating biases.
        """,
        evaluation=f"""
        The model was evaluated on the {DATASET} dataset using various metrics.␣
↪These metrics, along with their corresponding scores,
        are detailed in the 'eval_results' section. The evaluation process␣
↪ensured a comprehensive assessment of the model's performance
        in question-answering scenarios.
        """,
        training=f"""
        The model was trained over {NUM_TRAIN_EPOCHS} epochs with a learning␣
↪rate of {LEARNING_RATE}, using a batch size of {PER_DEVICE_TRAIN_BATCH_SIZE}.
        The training utilized a cross-entropy loss function and the AdamW␣
↪optimizer, with gradient accumulation over {GRADIENT_ACCUMULATION_STEPS}␣
↪steps.
        """,
        tips_and_tricks=f"""
        For optimal performance, questions should be clear, concise, and␣
↪grammatically correct.
        The model performs best on questions related to topics covered in the␣
↪{DATASET} dataset.
        It is advisable to pre-process text for consistency in encoding and␣
↪punctuation, and to manage expectations for questions on topics outside the␣
↪training data.
        """
```

```
    )

    # CREATE AND SAVE MODEL CARD
    model_card = ModelCard.from_template(model_card_data)
    model_card_path = model_checkpoints / 'README.md'
    model_card.save(model_card_path)

#-------------------------------------------------------------------------------#
# SAVE AND PUBLISH FINAL MODEL COMPONENTS
#-------------------------------------------------------------------------------#

    # MODEL
    model.save_pretrained(model_checkpoints)
    model.push_to_hub(repo_id=IDENTIFIER,repo_type="model")

    # TOKENIZER
    tokenizer.save_pretrained(model_checkpoints)
    tokenizer.push_to_hub(repo_id=IDENTIFIER, repo_type="model")

    # MODEL CARD
    model_card.push_to_hub(IDENTIFIER)

#-------------------------------------------------------------------------------#
# CLEAN UP BEFORE NEXT ITERATION
#-------------------------------------------------------------------------------#

    del model, tokenizer, trainer
    torch.cuda.empty_cache()
    gc.collect()
```