

Assignment II: Machine Translation

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Find it on Github: <https://github.com/hkukend/DASC7606A-B-A2>


Optional Assignment:

- **This assignment is optional** due to the coming final exams
- **The final assignment grades** will be the maximum between Assignment 1 and 2

Objectives:

- **Use the Hugging Face Hub for datasets:** discover, download, and prepare translation datasets
- **Load pretrained models and tokenizers** from the community
- **Build training pipelines using** transformers, datasets, accelerate, and evaluate
- **Fine-tune a translation model** end-to-end for a chosen language pair (zh-sim->en)
- **Evaluate translation quality** using BLEU (via sacrebleu) and report results
- **Develop debugging skills** for identifying and fixing common deep learning issues in HF version

HuggingFace 🤗:

 **Hugging Face**

Models

Datasets

Spaces

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The AI community building the future.

The platform where the machine learning community collaborates on models, datasets, and applications.

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Filter Tasks by name

Multimodal

- Text-to-Image
- Image-to-Text
- Text-to-Video
- Visual Question Answering
- Document Question Answering
- Graph Machine Learning

Computer Vision

- Depth Estimation
- Image Classification
- Object Detection
- Image Segmentation
- Image-to-Image
- Unconditional Image Generation
- Video Classification
- Zero-Shot Image Classification

Natural Language Processing

- Text Classification
- Token Classification
- Table Question Answering
- Question Answering
- Zero-Shot Classification
- Translation
- Summarization
- Conversational
- Text Generation
- Text2Text Generation
- Sentence Similarity

Audio

- Text-to-Speech
- Automatic Speech Recognition
- Audio-to-Audio
- Audio Classification
- Voice Activity Detection

Tabular

- Tabular Classification
- Tabular Regression

Reinforcement Learning

- Reinforcement Learning
- Robotics

Models 469,541 Filter by name

meta-llama/Llama-2-70b

- Text Generation
- Updated 4 days ago
- ± 25.2k
- ♥ 64

stabilityai/stable-diffusion-xl-base-0.9

- Updated 6 days ago
- ± 2.01k
- ♥ 393

openchat/openchat

- Text Generation
- Updated 2 days ago
- ± 1.3k
- ♥ 136

lillyasviel/ControlNet-v1-1

- Updated Apr 26
- ♥ 1.87k

cerspense/zeroscope_v2_XL

- Updated 3 days ago
- ± 2.66k
- ♥ 334

meta-llama/Llama-2-13b

- Text Generation
- Updated 4 days ago
- ± 328
- ♥ 64

tiiuae/falcon-40b-instruct

- Text Generation
- Updated 27 days ago
- ± 288k
- ♥ 899

WizardLM/WizardCoder-15B-V1.0

- Text Generation
- Updated 3 days ago
- ± 12.5k
- ♥ 332

CompVis/stable-diffusion-v1-4

- Text-to-Image
- Updated about 17 hours ago
- ± 448k
- ♥ 5.72k

stabilityai/stable-diffusion-2-1

- Text-to-Image
- Updated about 17 hours ago
- ± 782k
- ♥ 2.81k

Salesforce/xgen-7b-8k-inst

- Text Generation
- Updated 4 days ago
- ± 6.18k
- ♥ 57

HuggingFace 🤗 Datasets:

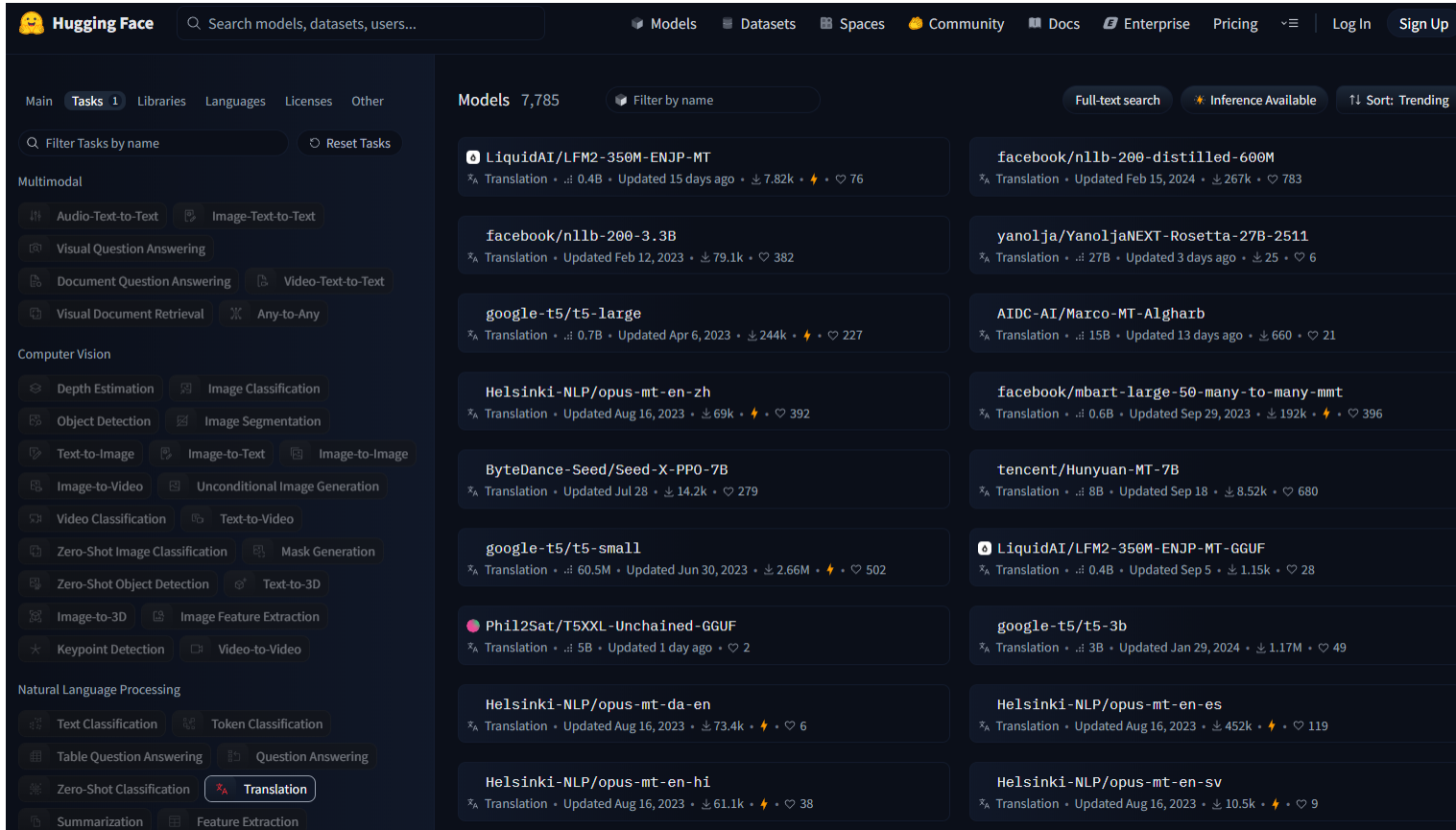
The screenshot shows the HuggingFace Datasets interface. The top navigation bar includes links for Models, Datasets, Spaces, Community, Docs, Enterprise, Pricing, Log In, and Sign Up. The main content area displays a grid of dataset cards, each with a title, viewer count, update date, size, and download count. The left sidebar contains filters for Tasks, Libraries, Languages, and Licenses, along with a search bar for filtering tasks by name. The dataset cards are organized into categories like Multimodal, Computer Vision, Natural Language Processing, and Audio.

Use one line to load datasets from HuggingFace platform:

```
>>> from datasets import load_dataset
>>> dataset = load_dataset("wmt19", "zh-en")
>>> dataset
DatasetDict({
  train: Dataset({
    features: ['translation'],
    num_rows: 25984574
  })
  validation: Dataset({
    features: ['translation'],
    num_rows: 3981
  })
})
```

HuggingFace Transformers:

Use one line to load tokenizers and pretrained models from HuggingFace platform:



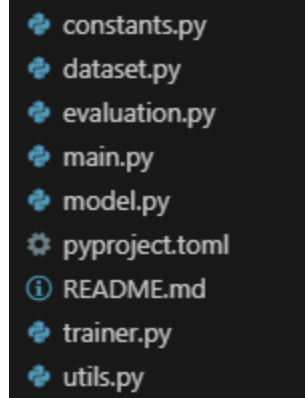
The screenshot shows the HuggingFace website interface. The top navigation bar includes the HuggingFace logo, a search bar, and links for Models, Datasets, Spaces, Community, Docs, Enterprise, Pricing, Log In, and Sign Up. The left sidebar has tabs for Main, Tasks, Libraries, Languages, Licenses, and Other. The main content area displays a grid of models, including LiquidAI/LFM2-350M-ENJP-MT, facebook/nllb-200-distilled-600M, facebook/nllb-200-3.3B, yanolja/YanoljaNEXT-Rosetta-27B-2511, google-t5/t5-large, AIDC-AI/Marco-MT-Algharb, Helsinki-NLP/opus-mt-en-zh, facebook/mbart-large-50-many-to-many-mmt, ByteDance-Seed/Seed-X-PP0-7B, tencent/Hunyuan-MT-7B, google-t5/t5-small, LiquidAI/LFM2-350M-ENJP-MT-GGUF, Phil2Sat/T5XXL-Unchained-GGUF, google-t5/t5-3b, Helsinki-NLP/opus-mt-da-en, Helsinki-NLP/opus-mt-en-es, Helsinki-NLP/opus-mt-en-hi, and Helsinki-NLP/opus-mt-en-sv. The bottom section shows 'Natural Language Processing' models, including Text Classification, Token Classification, Table Question Answering, Question Answering, Zero-Shot Classification, Translation, Summarization, and Feature Extraction.

```
>>> from transformers import AutoTokenizer, AutoModel
>>> tokenizer = AutoTokenizer.from_pretrained(pretrained_model_name_or_path = "Helsinki-NLP/opus-mt-en-zh")
/.../venv/lib/python3.13/site-packages/transformers/models/ MarianTokenizer from_pretrained:
on_marian.py:175: UserWarning: Recommended: pip install sacremoses.
warnings.warn("Recommended: pip install sacremoses.")
>>> tokenizer
MarianTokenizer(name_or_path='Helsinki-NLP/opus-mt-en-zh', vocab_size=65001, model_max_length=512, is_fast=False,
padding_side='right', truncation_side='right', special_tokens=['eos_token': '</s>', 'unk_token': '<unk>', 'pad_
oken': '<pad>'], clean_up_tokenization_spaces=False, added_tokens_decoder={
  0: AddedToken("</s>", rstrip=False, lstrip=False, single_word=False, normalized=False, special=True),
  1: AddedToken("<unk>", rstrip=False, lstrip=False, single_word=False, normalized=False, special=True),
  65000: AddedToken("<pad>", rstrip=False, lstrip=False, single_word=False, normalized=False, special=True)
})
>>>
>>> model = AutoModel.from_pretrained(pretrained_model_name_or_path = "Helsinki-NLP/opus-mt-en-zh")
>>> model
MarianModel(
  (shared): Embedding(65001, 512, padding_idx=65000)
  (encoder): MarianEncoder(
    (embed_tokens): Embedding(65001, 512, padding_idx=65000)
    (embed_positions): MarianSinusoidalPositionalEmbedding(512, 512)
    (layers): ModuleList(
      (0-5): 6 x MarianEncoderLayer(
        (self_attn): MarianAttention(
          (k_proj): Linear(in_features=512, out_features=512, bias=True)
          (v_proj): Linear(in_features=512, out_features=512, bias=True)
          (q_proj): Linear(in_features=512, out_features=512, bias=True)
          (out_proj): Linear(in_features=512, out_features=512, bias=True)
        )
        (self_attn_layer_norm): LayerNorm((512,), eps=1e-05, elementwise_affine=True)
        (activation_fn): SiLU()
        (fc1): Linear(in_features=512, out_features=2048, bias=True)
        (fc2): Linear(in_features=2048, out_features=512, bias=True)
        (final_layer_norm): LayerNorm((512,), eps=1e-05, elementwise_affine=True)
      )
    )
  )
  (decoder): MarianDecoder(
    (embed_tokens): Embedding(65001, 512, padding_idx=65000)
    (embed_positions): MarianSinusoidalPositionalEmbedding(512, 512)
    (layers): ModuleList(
      (0-5): 6 x MarianDecoderLayer(
        (self_attn): MarianAttention(
          (k_proj): Linear(in_features=512, out_features=512, bias=True)
          (v_proj): Linear(in_features=512, out_features=512, bias=True)
          (q_proj): Linear(in_features=512, out_features=512, bias=True)
          (out_proj): Linear(in_features=512, out_features=512, bias=True)
        )
        (self_attn_layer_norm): LayerNorm((512,), eps=1e-05, elementwise_affine=True)
      )
    )
  )
)
```

What we will do:

The workflow is organized into 5 files:

- **Dataset Sourcing & Preparation (dataset.py)**
Explore HF Datasets, load splits, perform filtering/mapping, and train/validation/test preparation
- **Model & Tokenizer Setup (model.py)**
Select a suitable base model from the Hub and initialize tokenizer and config
- **Training Pipeline (trainer.py)**
Configure TrainingArguments, data collators, metrics, logging, and checkpointing
- **Run & Monitor Training (main.py)**
Orchestrate end-to-end training and validation, with periodic evaluation
- **Evaluation & Reporting (evaluation.py)**
Compute BLEU on the held-out test set; save artifacts and summary



- constants.py
- dataset.py
- evaluation.py
- main.py
- model.py
- pyproject.toml
- README.md
- trainer.py
- utils.py

What we will do:

- **Goal:** Fine-tune a suitable HF model on a translation dataset and achieve competitive BLEU.
- **What you can modify:** You may change any files in the repo **except** `main.py`, part of `evaluation.py` and `utils.py`. The test set must not be modified.
- **What to improve:**
 - **Enriching Datasets:** Use a more varied dataset (no leakage of the test set into training)
 - **Base model choice:** Select an appropriate pretrained model for your language pair
 - **Training pipeline:** Tune hyperparameters (batch size, LR, epochs, schedulers, label smoothing, gradient accumulation)
 - **Data processing:** Tokenization lengths, filtering, cleaning, language codes, special tokens
 - **Advanced HF features:** Mixed precision (`fp16`/`bf16`), gradient checkpointing, LoRA/PEFT, better data collators, scheduler choices, early stopping

Example Accepted Datasets/Models

- Datasets: `wmt14`, `wmt16`, `wmt19`, `opus100`, `ted_talks_iwslt`, etc. (via HF Datasets)
- Models: MarianMT (Helsinki-NLP/opus-mt-xx-yy), mT5, MBART-50, M2M100, NLLB-200 (ensure your pair is supported), and even **LLMs**, etc.

- `constants.py`
- `dataset.py`
- `evaluation.py`
- `main.py`
- `model.py`
- `pyproject.toml`
- `README.md`
- `trainer.py`
- `utils.py`

Grading:

We will re-run `main.py` and evaluate on the fixed test set. BLEU (SacreBLEU) is the primary metric.

Important Considerations:

1. **Error-Free Execution:** Your code must run without errors (and avoid GPU OOM on the provided environment)
2. **Correct Data Usage:** Do not alter or leak the test set into training
3. **No Personal Pre-trained Model:** "Downloads last month" of loaded pre-trained model on HuggingFace should be **GREATER THAN 10**
4. **Reasonable Performance:** Achieve competitive BLEU given the chosen model and setup
5. **Runtime:** Complete in a reasonable time budget (≤ 12 hours with **ONE GPU on HKU GPU Farm**)

BLEU-based Grading (dummy thresholds, subject to adjustment):

- **BLEU ≥ 25 :** 100%
- **BLEU ≥ 24 :** 90%
- **BLEU ≥ 23 :** 80%
- **BLEU ≥ 22 :** 70%
- **BLEU ≥ 21 :** 60%
- **BLEU ≥ 20 :** 50%
- **BLEU < 20 / Fail to reproduce / Overtime:** 0%

Important dates:

- Assignment II Release: Nov. 06 (Thursday)
- Submission Deadline: Nov. 30 (Sunday) (23:59 GMT+8)

Late submission policy :

- All submissions later than the deadline will NOT be accepted

Questions!

If any more questions, please contact kendwj@hku.hk
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